
NATIONAL CENTER FOR EDUCATION STATISTICS

Technical
Issues in
Large-Scale
Performance
Assessment

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Foreword

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Recently there has been a significant expansion in the use of performance assessment in large scale testing **programs**. This proliferation has given rise to a range of testing formats such as **constructed-responses, essays, experiments, portfolios, exhibitions, interviews, and direct observations**. Support for this effort comes primarily from **curriculum reformers and policymakers** who feel it only makes good sense to test what we **teach**, and test the way we **teach**. Many school **districts**, state testing **programs**, and national and international assessments have incorporated performance assessments into their **programs**. Efforts are continuing to use such large scale assessments to shed light on the thinking and learning processes of **students**, and to encourage teachers to focus their teaching based on the content and skills of the **test**.

Although, there has been significant support from curriculum and policy **stakeholders**, the technical feasibility of large scale **performance** assessments has remained a **question**. This report is intended to contribute to the debate by reviewing some of the technical issues that must be dealt with by any developer of large scale **performance assessments**. The report is also intended to surface **issues**, articulate **problems**, and where **possible**, give advice on how to **proceed**. The report is not intended to be a technical or users manual on how to solve technical **problems**. The report is being written at a time **in** which many of the technical problems of large scale **performance** assessments are just beginning to **surface**. As these problems are recognized and solved, the state-of-the-art is expected to change **rapidly**.

The report is divided into **five chapters: Validity, Generalizability, Comparability, Performance Standards and Fairness, Equity, and Bias**. Each represents a major technical topic that developers of large scale performance assessments should expect to **encounter**. The following is a brief **summary** of some of the main points in each **chapter**.

Validity. This chapter provides a comprehensive view of **validity**. Validity is defined as not only the evaluative or evidential **information** about score inferences but also **information** about actual as well as potential consequences of score **interpretations**. It is argued that validity is not a property of the test but instead is a property of inferences or interpretations we make from test **scores**. **Messick** argues that validity is an essential concept for all types of qualitative as well as quantitative **summaries**.

The authors argue that since "**performance** assessments promise potential benefits for teaching and **learning**, it is important to accrue evidence of such positive consequences as well as evidence that adverse consequences are **minimal**." The primary adverse consequence that should be investigated is the potential negative impact on individuals or groups derived from sources of invalidity such as construct underrepresentation or construct-irrelevant **variance**. In the former **case**, individuals may be scoring low because the assessment is missing something that best represents the **construct**. In the latter **case**, individuals may score low because the measurement process contains something irrelevant that interferes with the student's ability to demonstrate **proficiency**.

Generalizability. This chapter provides an overview of generalizability theory and integrates the literature on reliability of performance assessment with the conceptual framework of generalizability theory. Generalizability theory is viewed as the product of the marriage between classical test theory and analysis of variance methodology.

The methodological development in this chapter is quite Comprehensive and easy to understand. It should be required reading for everyone involved in the development of large scale performance assessments. In addition, the author makes several points that represent lessons learned from the literature. Some of these lessons are:

- The number of raters has very little effect on the error variance found in most generalizability studies. This is probably because the scoring rubrics are generally well defined and the raters are well trained. The main conclusion from this is that increasing the number of raters does not increase test reliability.
- The number of tasks is inversely related to the error variance in most generalizability studies. The fewer the tasks the larger the error variance. One approach to reducing error variance might be to narrow the domain of tasks so that each is a slight modification of the others. The other approach is to increase the number of tasks (say to 10 or more) to reduce the error. Although this may be a costly alternative, it often makes more sense than restricting the domain.

Comparability. In order to provide indicators of trends in academic achievement, large scale performance assessments must be comparable across administrations. With multiple-choice testing this is a relatively easy matter because testing conditions are highly standardized, and large numbers of unidimensional items are scored by computer. Performance assessments on the other hand tend to be less standardized, involve fewer tasks, are more multidimensional, and are scored by humans, not computers. These differences make it more difficult to ensure comparability from one administration to the next.

The authors make the point that the degree of comparability required in performance assessments depends on the kind of decision being made and the importance of the consequences attached to those decisions. As the stakes get higher, the requirements for comparability get higher. The chapter concludes with the observation that strict forms of comparability, such as equating and calibration, may not be possible with many large scale performance assessments. However, weaker forms of comparability, such as statistical and social moderation, are attainable. Finally, the authors observe that many of the problems leading to lack of comparability would be mitigated if more precise content specifications for performance exercises were available.

Performance Standards. This chapter deals with the setting of performance standards on large scale performance assessments. There is much in the literature on setting performance standards on multiple-choice-based large assessments but there is very little on setting standards for large scale performance-based assessments. The authors cite four generalizations from two decades of research on performance standards.

- In almost all cases performance standards are arbitrarily (although not capriciously) set on a performance continuum.
- Performance standards are method dependent.

- Those who set performance standards can't objectively evaluate the quality of their **standards**.
- Widely used performance standing setting methods presume an underlying **unidimensional** scale.

The last generalization is particularly out of sync with almost all **performance** assessments that are **often** explicitly **multidimensional**. In **addition**, the authors argue that new methods of setting **performance** standards are needed because most **performance** assessments are **often** based on only a few **tasks**, each potentially requiring a separate performance **standard**.

The chapter describes two new approaches to setting **performance** standards that do not make the above **unidimensionality assumptions**. Each of the approaches have been evaluated within the context of the National Board for Professional Teaching Standards (NBPTS). The two new approaches **are**:

- **Iterative, Judgmental Policy Capturing Procedure:** In this method panelists respond independently to graphic profiles of performance for hypothetical **students**. The panelists make **judgements** as to whether the overall performance (**or profile**) should be considered deficient (1), competent (2), accomplished (3), or highly accomplished (4). Various analytic model fitting methods were used to assign weights to each dimension of the **profile**. The essence of the **Iterative, Judgmental Policy Capturing** procedure is that the panelists standard setting policies are inferred from their reactions to the **profiles** of candidate performance presented to **them**.
- **Multi-Stage Dominant Profile Procedure:** In this method a variety of integrative procedures are used to get the panelists to formulate explicitly their standard setting **policies**. This involves more up front group discussion and **reflection**. The procedure is different from the previous one in that the **panelists'** standards are generated directly through discussion rather than inferred from panelist **ratings**.

The chapter concludes with several unresolved technical issues that need to be addressed in setting **performance standards**.

- Since **performance** standards usually involve an artificial dichotomization of a **performance criteria**, how do you minimize misclassification near the **cut-score**?
- Since performance standards are method **dependent**, how do you assess this source of error in your **procedures**?
- How large should the standard setting panel **be**?
- Who should compose the standard-setting panel (**e.g.**, experts or **stakeholder groups**)?
- In order to ground criterion-referenced performance standards in **reality**, how do you incorporate the use of **normative information**?
- How much training should standard setting panelists **receive**?
- How do you report the sources of error and any adjustments of the standard setting **recommendations**?

Fairness, Equity, and Bias. The authors define fairness as essentially the **same** thing as differential **validity**. **However**, they go beyond the narrow psychometric concern of differential validity (**bias**) to include concerns about the educational and social policy that forms the context

for the assessment (**equity**). It is possible for an assessment to be considered unbiased in a technical **sense**, yet be used in the service of a policy that does not promote **equity**.

The point is made that bias can creep into an assessment at various stages throughout the **development**, data collection and scoring of an **assessment**. In order to minimize **bias**, test developers need **to**: (1) make sure there is diversity among the developers of the content **framework**, test administrators and the scoring **panels**; (2) require a sound sensitivity review on all assessment materials for **sexist**, ethnically **insensitive**, or stereotypic assessment **stimuli**; and (3) conduct statistical differential item **functioning** studies on all items or performance **tasks**.

The authors conclude their chapter by discussing the various ways high-stakes large scale assessments may have unintended negative consequences for poor and minority **students**. The major culprit is that when the stakes are **high**, educators tend to focus resources on what is **tested**. This often leads to a narrowing of the curriculum but rise in test **scores**. **Unfortunately**, a rise in test scores does not necessarily mean an improvement in the overall quality of education for the general **population**. When the stakes are **high**, educators **will** do the same thing with performance-type assessments that they used to do with multiple-choice **testing**. For **example**, they might exclude more students with disabilities or limited English proficiency so they will not count in the aggregate **summaries**, target instruction to students near the cut-scores and ignore those at the bottom and **top**, or encourage low achieving students to drop **out**. Such practices tend to corrupt performance assessments as an indicator and disproportionately impact poor and minority **students**.

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Validity of Performance Assessments

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Validity is an overall evaluative judgment of the degree to which empirical evidence and theoretical rationales support the *adequacy* and *appropriateness* of *interpretations* and *actions* based on test scores or other modes of assessment (Messick, 1989). Validity is not a property of the test or assessment as such, but rather of the meaning of the test **scores**. These scores are a function not only of the items or stimulus **conditions**, but also of the *persons* responding as well as the *context* of the **assessment**. In **particular**, what needs to be valid is the meaning or interpretation of the scores as well as any implications for action that this meaning entails (Cronbach, 1971). The extent to which score meaning and action implications hold across persons or population groups and across settings or contexts is a persistent and perennial empirical **question**. This is the main reason that validity is an evolving property and validation a continuing **process**.

Because evidence is always **incomplete**, validation is essentially a matter of making the most reasonable case to guide both current use of the test and current research to advance understanding of what the test scores **mean**. In other **words**, validation is basically a matter of constructing a network of evidence supporting (**or challenging**) the intended purpose of the **testing**. This chapter addresses the **forms** of evidence that this network should reasonably encompass and highlights the need for persuasive arguments or rationales whenever pertinent evidence is **foregone**.

Introductory Groundwork

The principles of validity apply not just to interpretive and action inferences derived from test scores as ordinarily **conceived**, but also to inferences based on any means of observing or documenting consistent behaviors or **attributes**. **Thus**, the term "**score**" is used generically **here in** its broadest sense to mean any coding or summarization of observed consistencies or performance regularities on a **test, questionnaire, observation procedure**, or other assessment device such as work **samples, portfolios**, and realistic problem **simulations**.

The Value of Validity

This general usage subsumes qualitative as well as quantitative **summaries**. It **applies**, for **example**, to behavior **protocols**, to clinical **appraisals**, to computerized verbal score **reports**, and to behavioral or performance judgments or **ratings**. **Hence**, the principles of validity apply to all **assessments**. These include **performance** assessments **which**, because of their promise of positive consequences for teaching and **learning**, are becoming increasingly popular as purported instruments of standards-based education **reform**. **Indeed**, it is precisely because of these politically salient potential consequences that the validity of **performance** assessment needs to be systematically **addressed**, as do other basic measurement **issues** such as **reliability, comparability**,

and **fairness**. As applied to performance assessment and **standard-setting**, these issues taken together constitute the main concern of the present **report**.

These issues are critical for performance assessment because **validity, reliability, comparability**, and fairness are not just measurement **principles**, they are *social values* that have meaning and force outside of measurement whenever evaluative judgments and decisions are **made**. As a salient social **value**, validity assumes both a scientific and a political role that can by no means be fulfilled by a simple correlation coefficient between test scores and a purported criterion (*i.e.*, classical criterion **validity**) or by expert judgments that test content is relevant to the proposed test use (*i.e.*, traditional content **validity**).

Indeed, broadly **speaking**, validity is nothing less than an evaluative **summary** of both the evidence for and the actual as well as potential consequences of score interpretation and use (*i.e.*, construct validity conceived **comprehensively**). This comprehensive view of validity integrates considerations of **content, criteria**, and consequences into a construct framework for empirically testing rational hypotheses about score meaning and **utility**. **Fundamentally, then**, score validation is empirical evaluation of the meaning and consequences of **measurement**. As **such**, validation combines scientific inquiry with rational **argument** to justify (**or nullify**) score interpretation and **use**.

Conceptions of Performance Assessment

We next attempt to **clarify** the meaning of performance **assessment**, because different conceptions have distinctly different implications for **validation**. In **essence**, a performance assessment requires the student to execute a task or process and bring it to completion (**Wiggins, 1993**). That **is**, the student **performs, creates**, or produces something over a sufficient duration of time to **permit** evaluation of either the **process or the product**, or **both**. This is in contradistinction to the impoverished trace or storable record resulting when one merely marks a correct or preferred option on an answer sheet as in a multiple-choice **test**, which does not reflect the amount or kind of thinking or effort that may underlie the choice of **option**. The choice of an answer may reflect recognition or **recall**, to be **sure**, but also a worked-through solution or **guessing**.

Indeed, with respect to task **processing**, the boundary between multiple-choice (MC) tests and performance assessments is a **fuzzy** one because some students on many MC items and most students on **difficult** MC items execute the solution process as a means of selecting the appropriate option (**Traub, 1993**). A more critical distinction is that the selected option can only be appraised for correctness or goodness with respect to a single **criterion**. There is no **record**, as in the typical performance **assessment**, of an extended process or product that can be scored for multiple aspects of **quality**.

A further complication is that the contrast between MC items and open-ended performance tasks is not a **dichotomy**, but a continuum representing different degrees of structure versus openness in the allowable **responses**. This continuum is variously described as ranging from multiple-choice to student-constructed products or presentations (**Bennett, 1993**), for **example**, or from multiple-choice to demonstrations and portfolios (**Snow, 1993**). Successive intervening stages include items requiring reordering or **rearranging**, substitution or **correction**, simple

completion or cloze procedures, short essays or complex **completions**, problem exercises or **proofs**, teach-back procedures, and long **essays**.

Apart from multiple-choice, the remainder of the continuum is referred to as involving "**student-constructed responses**." However, not **all** student-constructed responses—notably those involving **rearranging**, **substitution**, and simple completion—are properly considered to be performance assessments because they do not yield a storable record of an extended process or **product**.

Prototypical performance assessments occur more toward the unstructured end of the response continuum and include such exemplars as portfolios of student work over **time**, exhibits or displays of knowledge and **skill**, open-ended tasks with no single correct approach or **answer**, and hands-on **experimentation**. The openness with respect to response possibilities enables students to exhibit skills that are difficult to tap within the **predefined** structures of **multiple-choice**, such as shaping or restructuring a **problem**, defining and **operationalizing variables**, manipulating **conditions**, and developing alternative problem **approaches**.

Evaluations of student achievement on such open-ended tasks usually rely on the professional judgment of the **assessor**, and some proponents view such subjectivity of scoring to be the hallmark of performance assessment (e.g., Frederiksen & Collins, 1989; Stiggins, 1991). However, this view appears too restrictive because some performance tasks can be objectively scored and some scoring judgments are amenable to expert-system computer algorithms (e.g., Bejar, 1991; Sebrechts, Bennett, & Rock, 1991).

A more likely hallmark of educational **performance** assessments is their nearly universal focus on higher-order thinking and problem-solving **skills**. According to Baker, O'Neil, and Linn (1993), "**virtually all** proponents of performance-based assessment intend it to measure aspects of higher-order thinking **processes**" (p. 1211). Indeed, performance assessments in education frequently attempt to tap the complex structuring of multiple skills and **knowledge**, including basic as well as higher-order **skills**, embedded in realistic or otherwise rich problem contexts that require extended or demanding **forms** of reasoning and **judgment**. In this **regard**, Wiggins (1993) views "**authentic**" **performance** assessments as tapping understanding or the application of good judgment in adapting knowledge to fashion performances effectively and **creatively**.

This mention of authentic assessments broaches a further **distinction**. Just as performance assessments are a more open-ended subset of student-constructed **responses**, so-called authentic assessments are a more realistic subset of performance **assessments**. In **particular**, authentic assessments pose engaging and worthy problems (**usually** involving multistage **tasks**) in realistic settings or close simulations so that the tasks and **processes**, as well as available time and **resources**, parallel those in the real **world**. The assessment challenge of complex **performance** tasks in general and authentic tasks in particular revolves around issues of **scoring**, **interpretation**, and **generalizable** import of key aspects of the complex **performance**, especially if the task is not completed **successfully**.

In **performance** assessment, one might start by clarifying the nature of the higher-order **competencies** or other constructs to be assessed and then select or construct tasks that would optimally reveal **them**. Or, **contrariwise**, one might start with an important task that is worthy of mastery in its own right and ask what **competencies** or other constructs this task **reveals**. This

contrast embodies a tension in performance assessment between construct-centered and task-centered approaches (Messick, 1994). However, what is critical in performance assessment is not what is operative in the task performance but what is captured in the test score and **interpretation**. Hence, the validity of the construct **interpretation** needs to be addressed sooner or later in either **approach**, as does the nature of convergent and **discriminant** evidence needed to sustain that **validity**.

Construct-Driven Versus Task-Driven Performance Assessment

The task-centered approach to performance assessment begins by **identifying** a worthy task and then determining what constructs can be scored and **how**. Often the mastery of such a worthy task functions as the target of the assessment in its own **right**, as opposed to serving as a vehicle for the assessment of **knowledge, skills**, or other **constructs**. This might **occur**, for **example**, in an arts contest or an Olympic figure-skating competition or a science **fair**. In such **cases**, **replicability** and **generalizability** are not at **issue**. All that counts is the quality of the performance or product submitted for **evaluation**, and the validation focus is on the judgment of **quality**. But note that in this usage of performance assessment as **target**, inferences are not to be made about the **competencies** or other attributes of the **performers**, that is, inferences from observed behavior to constructs such as knowledge and skill underlying that **behavior**.

Large-scale educational projects such as dissertations are often treated as targets in this **manner**, by crediting the complex accomplishment as meeting established standards with no requirement of **predictiveness** or domain **generalizability** (Baker et al., 1993). However, action implications of such complex assessments usually **presume**, with little or no specific **evidence**, that there is a global prediction of future **success**, that the knowledge and skills exhibited in the assessment **will** enable the student to accomplish a range of similar or related tasks in broader **settings**.

In **contrast**, such **presumptions** should be confronted by empirical evidence in the performance assessment of **competencies** or other **constructs**—that is, where the performance is the vehicle not the target of **assessment**. A major form of this evidence bears on **generalizability** and transfer **which**, as we shall **see**, represent critical aspects of construct **validity**. In **effect**, the meaning of the construct is tied to the range of tasks and situations that it generalizes and transfers **to**.

The task-centered approach to performance assessment is in danger of tailoring scoring criteria and rubrics to properties of the task and of representing any educed constructs in task-dependent ways that might limit **generalizability**. In **contrast**, the nature of the constructs in the construct-centered approach guides the selection or construction of relevant tasks as well as the rational development of construct-based scoring criteria and **rubrics**. **Focussing** on constructs also alerts one to the possibility of **construct-irrelevant** variance that might distort either the task performance or its **scoring**, or both (Messick, 1994). The task-centered approach is not completely devoid of **constructs**, of **course**, because task selection is often influenced by implicit construct notions or informal theories of learning and **performance**. The key issue is the extent to which the constructs guide scoring and interpretation and are explicitly linked to evidence supporting that interpretation as well as discounting plausible rival **interpretations**.

Sources of Invalidity

Construct-irrelevant variance is one of the two major threats to **validity**, the other being construct **underrepresentation**. A fundamental feature of construct validity is *construct representation*, whereby one attempts to **identify** through cognitive-process analysis the theoretical mechanisms underlying task **performance**, primarily by decomposing the task into requisite component processes and assembling them into a functional model (Embretson, 1983). Relying heavily on the cognitive psychology of information **processing**, construct representation refers to the relative dependence of task responses on the **processes, strategies**, and knowledge (including **metacognitive** or **self-knowledge**) that are implicated in task **performance**.

In the threat to validity known as "**construct underrepresentation**," the assessment is too narrow and fails to include important dimensions or facets of the **construct**. In the threat to validity known as "**construct-irrelevant variance**," the assessment is too **broad**, containing excess reliable variance that is irrelevant to the interpreted **construct**. Both **threats** are operative in **all assessment**. Hence a primary validation concern is the extent to which the **same** assessment might **underrepresent** the focal construct while **simultaneously** contaminating the scores with **construct-irrelevant variance**.

There are two basic kinds of **construct-irrelevant variance**. In the language of ability and achievement **testing**, these might be called "**construct-irrelevant difficulty**" and "**construct-irrelevant easiness**." In the **former**, aspects of the task that are extraneous to the focal construct make the task irrelevantly difficult for some individuals or **groups**. An **example** is the intrusion of undue reading-comprehension requirements in a test of subject-matter **knowledge**. In **general**, construct-irrelevant **difficulty** leads to construct scores that are invalidly low for those individuals adversely affected (**e.g.**, knowledge scores of poor readers or of **examinees** with limited English **proficiency**).

In **contrast**, construct-irrelevant easiness occurs when extraneous clues in item or task formats **permit** some individuals to respond correctly or appropriately in ways irrelevant to the construct being **assessed**. Another instance occurs when the specific test material is highly familiar to some **respondents**, as when the text of a reading comprehension passage is well-known to some readers or the musical score for a sight-reading exercise invokes a well-drilled rendition from some **performers**. Construct-irrelevant easiness leads to scores that are invalidly high for the affected individuals as reflections of the construct under **scrutiny**.

The concept of construct-irrelevant variance is important in all educational and psychological **measurement**, including **performance assessments**. This is especially true of richly contextualized assessments and authentic simulations of real-world **tasks**. This is the case **because**, "**paradoxically**, the complexity of context is made manageable by contextual **clues**" (Wiggins, 1993, p. 208). And it matters whether the contextual clues that are responded to are construct-relevant or represent construct-irrelevant difficulty or **easiness**.

However, what constitutes construct-irrelevant variance is a tricky and contentious issue (Messick, 1994). This is especially true of **performance assessments**, which typically invoke constructs that are higher-order and complex in the sense of **subsuming** or organizing multiple **processes**. For **example**, skill in **communicating** mathematical ideas might well be considered irrelevant variance in the assessment of mathematical knowledge (**although** not necessarily vice versa). But both communication skill and mathematical knowledge are considered relevant parts

of the higher-order construct of mathematical power according to the content standards developed by the National Council of Teachers of **Mathematics**. It all depends on how compelling the evidence and arguments are that the particular source of variance is a relevant part of the focal construct as opposed to affording a plausible rival hypothesis to account for the observed performance regularities and relationships with other **variables**.

Authenticity and Directness As Validity Standards

Two **terms** that appear **frequently**, and usually in **tandem**, in the literature of performance assessment are "**authentic**" and "**direct**" **assessment**. They are most often used in connection with assessments involving realistic simulations or criterion **samples**. If authenticity and directness are important to consider when evaluating the consequences of assessment for student **achievement**, they constitute tacit validity **standards**, so we need to address what the labels "**authentic**" and "**direct**" might mean in validity **terms**.

The major measurement concern of authenticity is that nothing important has been left out of the assessment of the focal **construct**. This is **tantamount** to the familiar validity standard of minimal construct underrepresentation (**Messick, 1994**). **However**, although authenticity implies minimal construct **underrepresentation**, the obverse does not **hold**. This is the case because minimal construct underrepresentation does not necessarily imply the close simulation of real-world problems and resources typically associated with authenticity in the current literature on performance **assessment**. In any **event**, convergent and **discriminant** evidence is needed to appraise the extent to which the ostensibly authentic tasks represent (**or underrepresent**) the constructs they are interpreted to **assess**.

The major measurement concern of directness is that nothing irrelevant has been added that distorts or interferes with construct **assessment**. This is tantamount to the familiar validity standard of minimal construct-irrelevant variance (**Messick, 1994**). **Incidentally**, the term "**direct assessment**" is a misnomer because it always promises too **much**. In education and **psychology**, "**all** measurements are indirect in one sense or **another**" (**Guilford, 1936, p. 3**). Measurement always **involves**, even if only **tacitly**, intervening processes of **judgment, comparison, or inference**. The key **issue, then**, is not directness per se but the **minimizing** of construct-irrelevant variance in performance assessment **scores**.

Aspects of Construct Validity

The validity issues of score **meaning, relevance, utility**, and social consequences are many faceted and **intertwined**. They are difficult if not impossible to disentangle **empirically**, which is why validity has come to be viewed as a unified concept (**APA, 1985; Messick, 1989**). For **example**, social consequences provide evidence contributing to score **meaning**, and utility is both validity evidence and a value **consequence**. The essence of unified validity is that the **appropriateness, meaningfulness, and usefulness** of score-based inferences are inseparable and that the integrative power derives from empirically grounded score **interpretation**.

However, to speak of validity as a unified concept does not imply that validity cannot be usefully differentiated conceptually into distinct aspects to underscore issues and nuances that might otherwise be downplayed or **overlooked**, such as the social consequences of performance assessments or the role of score meaning in applied **use**. The intent of these distinctions is to

provide a means of addressing **functional** aspects of validity that help disentangle some of the complexities inherent in appraising the **appropriateness, meaningfulness,** and usefulness of score inferences.

In **particular**, six distinguishable aspects of construct validity are highlighted as a means of addressing central issues implicit in the notion of validity as a **unified concept**. These are **content, substantive, structural, generalizability, external,** and consequential aspects of construct validity. In effect, these six aspects conjointly function as general validity criteria or standards for all educational and psychological measurement (Messick, 1989). However, these six aspects must not be viewed as separate and substitutable validity types—as the erstwhile trinity of **content, criterion,** and construct validities often **were—but** rather as interdependent and complementary **forms** of validity **evidence**. As general validity **criteria**, they can be specialized for apt application to **performance assessments—as** discussed **selectively,** for **example,** by Linn, Baker, and Dunbar (1991) and by Moss (1992)—**but** none should be **ignored**.

A brief characterization of these six aspects is presented **next**, followed by six sections discussing the validity issues and sources of evidence bearing on each **aspect**:

- The content aspect of construct validity includes evidence of content **relevance, representativeness,** and technical quality (Lennon, 1956; Messick, 1989).
- The substantive aspect refers to theoretical rationales for the observed consistencies in test **responses,** including process models of task performance (Embretson, 1983), along with empirical evidence that the theoretical processes are actually engaged by respondents in the assessment **tasks**.
- The structural aspect appraises the fidelity of the scoring structure to the structure of the construct domain at issue (Loevinger, 1957).
- The **generalizability** aspect examines the extent to which score properties and interpretations generalize to and across population **groups, settings,** and tasks (Cook & Campbell, 1979; Shulman, 1970), including validity generalization of test-criterion relationships (Hunter, Schmidt, & Jackson, 1982).
- The external aspect includes convergent and **discriminant** evidence from **multitrait-multimethod** comparisons (Campbell & Fiske, 1959), as well as evidence of criterion relevance and applied utility (Cronbach & Gleser, 1965).
- The consequential aspect appraises the value implications of score interpretation as a basis for action as well as the actual and potential consequences of test **use, especially** in regard to sources of invalidity related to issues of **bias,** fairness, and distributive justice (Messick, 1980, 1989).

Content Relevance and Representativeness

A key issue for the content aspect of construct validity is the specification of the boundaries of the construct domain to be assessed—that is, determining the knowledge, skills, and other attributes to be revealed by the assessment tasks. The boundaries and structure of the construct domain can be addressed by means of job analysis, task analysis, curriculum analysis, and domain theory (Messick, 1989). If concern is with the application of domain processes in

real-world **settings**, the techniques of job and task analysis should prove **useful** both in determining domain structure and in selecting relevant realistic assessment **tasks**. If concern is with the learning of domain **processes**, analyses of curricula and instruction should prove useful for determining the construct domain and for selecting assessment tasks attuned to the level of developing expertise of the learners. Such considerations are especially germane to authentic assessment because they bear on the issue of "**authentic to what?**"

Both job and curriculum analysis contribute to the development of domain **theory**, as does scientific inquiry into the nature of the domain processes and the ways in which they combine to produce effects or **outcomes**. A major goal of domain theory is to understand the construct-relevant sources of task **difficulty**, which then serves as a guide to the rational development and scoring of performance **tasks**. At whatever stage of its **development, then**, domain theory is a primary basis for specifying the boundaries and structure of the construct to be **assessed**.

It is also important to make explicit the relationship between the construct domain and the assessment **specifications**, by formulating what amounts to a test blueprint indicating whether the assessment is to include all components of the construct domain or only part of **them**. This is important because score inferences should be limited to what can be sustained by the assessment and not casually generalized to a broader construct **domain**.

Moreover, the description of the construct **domain**, as well as the assessment **specifications**, should distinguish them from other similar or related construct **domains**. **Ideally**, the descriptions should be clear enough so that test developers or other experts can judge whether a task refers to one construct domain or the **other**. In any **event**, as we **shall** see in the sections on the substantive and external aspects of construct **validity, discriminant** evidence needs to be produced showing that the focal construct is operative in task performance as opposed to similar or related **constructs**, that is, evidence to discount plausible rival **interpretations**.

However, it is not sufficient to select tasks that are relevant to the construct **domain**. In addition, the assessment should assemble tasks that are representative of the domain in some **sense**. The intent is to insure that **all** important parts of the construct domain are covered (**or** at least those subsets included in the assessment **specifications**). This is usually described as selecting tasks that sample domain processes in terms of their functional **importance**, or what **Brunswik (1956)** called ecological **sampling**. Functional importance can be considered in terms of what people **actually** do in the performance **domain**, as in job **analyses**, but also in terms of what characterizes and differentiates expertise in the **domain**, which would usually emphasize different tasks and **processes**.

Both the content relevance and **representativeness** of assessment tasks are traditionally appraised by expert professional **judgment**, documentation of which serves to address the content aspect of construct **validity**. **However**, as we shall see in the next section on the substantive **aspect**, such expert judgment is not sufficient because it is not just domain content that needs to be represented in assessment tasks but domain **processes**.

In standards-based education **reform**, two types of assessment standards have been **distinguished**. One type is called "**content standards**," which refers to the kinds of things a student should know and be able to do in a subject **area**. The other type is called "**performance standards**," which refers to the level of competence a student should attain at key stages of

developing expertise in the knowledge and skills specified by the content **standards**. Performance standards also **circumscribe**, either explicitly or **tacitly**, the form or forms of **performance** that are appropriate to be evaluated against the **standards**.

From the discussion thus **far**, it should be clear that not only the assessment tasks but also the content standards themselves should be relevant and representative of the construct **domain**. That **is**, the content standards should be consistent with domain theory and be reflective of the structure of the construct **domain**. This is the issue of the construct validity of content **standards**. There is also a related issue of the construct validity of **performance standards**. That **is**, increasing achievement levels or **performance standards** (**as** well as the tasks that benchmark these **levels**) should reflect increases in complexity of the construct under scrutiny and not increasing sources of **construct-irrelevant** difficulty (Messick, 1996). More extensive coverage of these and other issues related to standards-based assessment will appear in the subsequent chapter on **standard-setting**.

Substantive Theories, Process Models, and Process Engagement

The substantive aspect of construct validity emphasizes the role of substantive theories and process modeling in **identifying** the domain processes to be revealed in assessment tasks (Embretson, 1983; Messick, 1989). Two important points are **involved**: One is the need for tasks **providing** appropriate **sampling** of domain processes in addition to traditional coverage of domain **content**; the other is the need to move beyond traditional professional judgment of content to accrue empirical evidence that the ostensibly sampled processes are actually engaged by respondents in task **performance**.

Thus, the substantive aspect adds to the content aspect of construct validity the need for empirical evidence of response consistencies or performance regularities reflective of domain processes (Loevinger, 1957). Such evidence may derive from a variety of **sources**, for **example**, from "think-aloud" protocols or eye-movement records during task **performance**, from correlation patterns among part **scores**, from consistencies in response times for task **segments**, or from mathematical or computer modeling of task processes (Messick, 1989, pp. 53-55; Snow & Lohman, 1989). In **sum**, the issue of domain coverage refers not just to the content **representativeness** of the construct measure but also to the process representation of the construct and the degree to which these processes are reflected in construct **measurement**.

Another issue is the extent to which the assessed task processes correspond to domain **processes**, as opposed to being distorted by sources of irrelevant method **variance**. This depends on whether the assessment tasks **mimic** or simulate the domain conditions with **sufficient** comprehensiveness and fidelity to engage the domain processes with minimal **distortion**, which is a primary aim of authentic **assessment**. The point here is that empirical evidence is needed to assure that the higher-order thinking processes that authentic assessments aspire to address are actually operative in task **performance**. For **example**, for some individuals the task **performance** might not reflect problem **solving**, but rather a memorized **solution**. As another **instance**, a verbal reasoning task might be failed by some respondents because of inadequate verbal knowledge rather than poor inductive **inference**. The test user is in the best position to evaluate the meaning of individual scores under the specific applied **circumstances**, that **is**, to appraise the extent to which the intended score meaning **might** have been eroded by **contaminating** variables operating locally.

The core concept bridging the content and substantive aspects of construct validity is **representativeness**. This becomes clear once one recognizes that the term "representative" has two distinct meanings, both of which are applicable to performance assessment. One is in the cognitive psychologist's sense of representation or modeling (Suppes, Pavel, & Falmagne, 1994); the other is in the Brunswikian sense of ecological sampling (Brunswik, 1956; Snow, 1974). The choice of tasks or contexts in assessment is a representative **sampling issue**, which is central to the content aspect of construct **validity**. The comprehensiveness and fidelity of simulating the construct's realistic engagement in performance is a representation **issue**, which is central to the substantive **aspect**. Both issues are important in performance **assessment**; they are critical to the very meaning of authentic assessment.

Scoring Models As Reflective of Task and Domain Structure

According to the structural aspect of construct **validity**, scoring models should be rationally consistent with what is known about the structural relations inherent in behavioral manifestations of the construct in question (Loevinger, 1957; Peak, 1953). That is, the theory of the construct domain should guide not only the selection or construction of relevant assessment **tasks**, but also the rational development of construct-based scoring criteria and **rubrics**.

Ideally, the manner in which behavioral instances are combined to produce a score should rest on knowledge of how the processes underlying those behaviors combine dynamically to produce **effects**. **Thus**, the internal structure of the assessment (**i.e.**, interrelations **among** the scored aspects of task and **subtask performance**) should be consistent with what is known about the internal structure of the construct domain (Messick, 1989). This property of construct-based rational scoring models is called "**structural fidelity**" (Loevinger, 1957).

To the extent that different assessments (**i.e.**, those involving different tasks or different settings or **both**) are geared to the same construct **domain**, using the **same** scoring model as well as scoring criteria and **rubrics**, then the resultant scores are likely to be comparable or can be rendered comparable using equating **procedures**. To the degree that the different assessments do not adhere to the **same specifications**, then score comparability is jeopardized but can be variously approximated using **calibration**, **projection**, and moderation procedures (Mislevy, 1992).

Score comparability is clearly important for normative or accountability purposes whenever individuals or groups are being **ranked**. **However**, score comparability is also important even when individuals are not being directly **compared**, but are held to a common **standard**. Score comparability of some type is needed to sustain the claim that two individual performances in some sense meet the **same local, regional, national, or international standard**. These issues are addressed more **fully** in the subsequent chapter on **comparability**.

Generalizability and the Boundaries of Score Meaning

The concern that a performance assessment should provide representative coverage of the content and processes of the construct domain is meant to insure that the score interpretation not be limited to the sample of assessed tasks but be **generalizable** to the construct domain more **broadly**. Evidence of such **generalizability** depends on the degree of correlation of the assessed tasks with other tasks representing the construct or aspects of the **construct**. For **example**, how

well one can generalize from a sample of writing about a particular topic in a particular genre to skill in writing about other topics in the same or different genre depends on the pattern of correlations among different topic and genre **scores**. This issue of **generalizability** of score inferences across tasks and contexts goes to the very heart of score **meaning**. **Indeed**, setting the boundaries of score meaning is precisely what **generalizability** evidence is meant to **address**.

However, because of the extensive time required for the typical performance **task**, there is a conflict in performance assessment between time-intensive depth of **examination** and the breadth of domain coverage needed for **generalizability** of construct **interpretation**. This conflict is **usually addressed by** means of a variety of **trade-offs**. For **example**, one suggestion is to increase the **number** of **performance** assessments for each student or to increase the number of tasks in each **assessment**. Here the trade-off is between breadth of coverage and **nonassessment** instructional activities that might instead have filled the extended testing **time**.

Another suggestion is to use a matrix-sampling design with different performance tasks administered to different samples of **students**. Here the gain in breadth of coverage comes at the expense of individual student scores **or**, at **least**, of comparable individual **scores**. **Nonetheless**, matrix **sampling** is especially useful when the accountability concern focuses on some aggregate level of inference such as the **school, district, state, or nation**. Another approach is to develop assessments that represent a mix of efficient structured exercises broadly tapping multiple aspects of the construct and time-intensive open-ended tasks tapping integral aspects **in depth** (Messick, 1994), which involves a trade-off between the **number** of performance tasks and the **number** of brief structured **exercises**. The internal structure of interrelations among the briefer exercises and performance tasks bears on the substantive and especially the structural aspect of construct **validity**. Such structures undergird and guide decisions as to how responses should be aggregated into composite or multiple scores to represent the **construct**.

This conflict between depth and breadth of coverage is **often** viewed as entailing a trade-off between validity and reliability (**or generalizability**). It might better be depicted as a trade-off between the valid description of the specifics of a complex task and the power of construct **interpretation**. In any **event**, as Wiggins (1993) **stipulates**, such a conflict signals a design problem that needs to be carefully negotiated in performance **assessment**.

In addition to **generalizability** across **tasks**, the limits of score meaning are also affected by the degree of **generalizability** across time or occasions and across observers or raters of the task **performance**. Such sources of measurement error associated with the **sampling of tasks, occasions**, and scorers underlie traditional reliability concerns; they are examined in more detail in the subsequent chapter on **generalizability**.

Convergent and Discriminant Correlations with External Variables

The external aspect of construct validity refers to the extent to which the assessment **scores'** relationships with other measures and **nonassessment** behaviors reflect the expected **high, low**, and interactive relations implicit in the theory of the construct being **assessed**. **Thus**, the meaning of the scores is substantiated externally by appraising the degree to which empirical relationships with other **measures**, or the lack **thereof**, is consistent with that **meaning**. That **is**, the constructs represented in the assessment should rationally account for **the** external pattern of **correlations**.

The external aspect emphasizes two intertwined sets of relationships for the assessment scores: one between the task scores and different methods for measuring both the same and distinct **constructs**, and the other between measures of the focal construct and exemplars of different constructs predicted to be variously related to it on **theoretical grounds**. **Theoretically**, expected empirical consistencies in the first set include both convergent and **discriminant correlation patterns**, the convergent pattern indicating a correspondence between measures of the same construct and the **discriminant** pattern indicating a distinctness from measures of other constructs. These patterns are often displayed in what is called a **multitrait-multimethod** matrix (Campbell & Fiske, 1959).

Convergent evidence signifies that the measure in question is coherently related to other measures of the **same** construct as well as to other variables that it should relate to on theoretical grounds. **Discriminant** evidence signifies that the measure is not unduly related to exemplars of other distinct **constructs**. **Discriminant** evidence is particularly critical for discounting plausible rival alternatives to the focal construct **interpretation**. Both convergent and **discriminant evidence** are basic to construct **validation**.

Theoretically, expected consistencies in the second set of relationships mentioned above indicate a **lawful** relatedness between measures of different **constructs**. This lawful relatedness has been referred to as "**nomological validity**" by Campbell (1960) and as "**nomothetic span**" by Embretson (1983). The basic notion of **nomological** validity is that the theory of the construct being addressed provides a rational basis for deriving empirically testable links between the assessment task scores and measures of other **constructs**. Corroborative evidence helps to validate both the assessment and the construct **theory**. The assessment gains credence to the extent that the score correlations reflect theoretical implications of the **construct**, while the construct theory gains credence to the extent that score data jibe with its **predictions**.

Among the relationships **falling** within the purview of **nomological** validity or **nomothetic span** are those between the assessment scores and criterion measures pertinent to **selection, placement, licensure, program evaluation**, or other accountability purposes in applied **settings**. Once **again**, the construct theory points to the relevance of potential relationships between the assessment scores and criterion **measures**, and empirical evidence of such links attests to the utility of the scores for the applied **purpose**.

The issue of utility is evaluated in terms of the benefits or desired outcomes of the assessment relative to its costs (Cronbach & Gleser, 1965; Messick, 1989). **Thus**, although the cost of **performance** assessments in **terms** of time and resources is an important **consideration**, the choice **among** alternative assessment approaches should not be determined solely by cost or **efficiency**. **Rather**, such decisions should weigh both the costs and the benefits of the **assessment**, that is, its utility for the applied **purpose**.

Consequences as Validity Evidence

Because performance assessments promise potential benefits for teaching and **learning**, it is important to accrue evidence of such positive consequences as well as evidence that adverse consequences are **minimal**. In this **connection**, the consequential aspect of construct validity includes evidence and rationales for evaluating the intended and unintended consequences of score interpretation and use in both the **short- and long-term**. Particularly prominent is the

evaluation of any adverse consequences for individuals and **groups**, especially gender and racial/ethnic **groups**, that are associated with bias in scoring and interpretation or with unfairness in test **use**. **However**, this **form** of evidence should not be viewed in isolation as a separate type of **validity**, say, of "**consequential validity**." **Rather**, because the values served in the intended and unintended outcomes of test interpretation and use both derive from and contribute to the meaning of the test **scores**, appraisal of social consequences of the testing is also seen to be subsumed as an aspect of construct validity (Messick, 1964, 1975, 1980).

The primary measurement concern with respect to adverse consequences is that any negative impact on individuals or groups should not derive from any source of test invalidity such as construct underrepresentation or construct-irrelevant variance (Messick, 1989). That **is**, low scores should not occur because the assessment is missing something relevant to the focal construct **that**, if **present**, would have permitted the affected students to display their **competence**. **Moreover**, low scores should not occur because the measurement contains something irrelevant that interferes with the affected **students'** demonstration of **competence**.

However, reducing adverse impact associated with sources of test invalidity does not mean that there would necessarily be less adverse impact associated with the valid description of existing group **differences**. For **example**, a possible unintended consequence of performance assessment in education is increased adverse impact for gender and racial/ethnic groups because of short-term misalignments in their educational experiences **vis-à-vis** authentic testing and **teaching**. If **found**, one should monitor the situation to see how short-term it is likely to be and what resources are needed to redress the new imbalance. Positive and negative consequences of **assessment**, whether intended or **unintended**, are discussed in more depth in the subsequent chapter on **fairness**.

Aspects of Validity Specialized for Performance Assessment

Some proponents of **performance** assessment have proposed specialized validity criteria tailored for performance tasks (Frederiksen & Collins, 1989; Linn et al, 1991). In effect, these specialized criteria emphasize selected issues in some but not **all** of the six general validity aspects just described as they are applied to performance tasks (Messick, 1994; Moss, 1992). **However**, a few of these specialized criteria highlight different perspectives that warrant further **comment**. Especially important because it is at the heart of authentic assessment in education is what Linn and his colleagues (1991) call "**meaningfulness**" and what Frederiksen and Collins (1989) call "**transparency**." The concern here is that if the assessment itself is **to** be a worthwhile educational experience serving to motivate and direct **learning**, then the problems and tasks posed should be meaningful to the students and communicate clearly what is expected of **them**. That **is**, not only should students know what knowledge and skills are being **assessed**, but the criteria and standards of good **performance** should be **clear** to **them**, in terms of both how the performance is to be scored and what steps might be taken to improve **performance**. In this **sense**, the criteria and standards of successful **performance** are transparent or **demystified** and hence should be more readily internalized by students as self-directive goals (Baron, 1991; Wiggins, 1993).

Evidence needs to be **accrued**, of **course**, that the **performance** tasks are meaningful and that the **performance** standards are understood and facilitate **learning**, because the **meaningfulness** or transparency of performance assessments cannot be taken for **granted**. Such evidence is also

pertinent to the substantive and consequential aspects of construct **validity**. Moreover, there are important instances where transparency may be **counterproductive**, namely, where novelty occurs in the task performance that is not **amenable** to the transparent standards of **goodness**. Indeed, the very salience of the transparent standards might **hamper** the generation of **novelty**. In such **cases**, the challenge is to transform the standards or to develop new **ones**. That is, transparency and creativity may be in **conflict**.

A concept closely related to meaningfulness is what **Frederiksen** and Collins (1989) call "systemic validity." Their point is that as instruments of education **reform**, **performance** tests should "induce curricular and instructional changes in educational systems (and learning strategy changes in students) that foster the development of the cognitive traits that the tests are designed to measure" (p. 27). However, because interpretation of such teaching and learning consequences as reflective of test validity (or invalidity) assumes that **all** other aspects of the educational system are working well or are **controlled**, the use of the label "systemic validity" is **problematic**. It might better be called "systemic facilitation" because in practice the issue is not just the systemic validity of the tests **but, rather**, the validity of the system as a whole for improving teaching and **learning**. In any **event**, the concept of systemic validity is a specialized instance of the consequential aspect of construct validity because it focuses on one type of testing consequence—indeed, on one type of systemic consequence—among many (**Messick**, 1989, p. 85).

Validity As Integrative Summary

These six aspects of construct validity apply to **all** educational and psychological **measurement**, including performance **assessments**. Taken **together**, along with aspects of validity specialized for performance assessments such as meaningfulness or **transparency**, they provide a **way** of addressing the multiple and interrelated validity questions that need to be answered in justifying score interpretation and **use**. They are highlighted because most score-based interpretations and action **inferences**, as well as the elaborated rationales or arguments that attempt to legitimize them (**Kane**, 1992), either invoke these properties or assume **them**, explicitly or **tacitly**. That **is**, most score interpretations refer to relevant content and operative **processes**, presumed to be reflected in scores that concatenate responses in domain-appropriate ways and are **generalizable** across a range of **tasks**, **settings**, and **occasions**. Furthermore, score-based interpretations and actions are typically extrapolated beyond the test context on the basis of presumed or documented relationships with **nontest** behaviors and anticipated outcomes or **consequences**.

The challenge in test validation is to link these inferences to convergent evidence supporting them as well as to **discriminant** evidence discounting plausible rival **inferences**. Evidence pertinent to **all** of these aspects needs to be integrated into an overall validity judgment to sustain score inferences and their action **implications**, or else provide compelling reasons why **not**, which is what is meant by validity as a unified **concept**.

Overview

The traditional conception of validity divided it into three separate and substitutable **types**—namely, **content**, **criterion**, and construct **validities**. This view is fragmented and

incomplete, especially in failing to take into account evidence of the value implications of score meaning as a basis for action and of the social consequences of score **use**. The new unified concept of validity interrelates these issues as fundamental aspects of a more comprehensive theory of construct validity addressing both score meaning and social values in both test interpretation and test **use**. That **is**, unified validity integrates **considerations** of **content**, **criteria**, and consequences into a construct framework for empirically testing rational hypotheses about score meaning and theoretically relevant **relationships**, including those of both an applied and a scientific **nature**. Six distinguishable aspects of construct validity are highlighted as a means of addressing central issues implicit in the notion of validity as a unified **concept**. These are **content**, **substantive**, **structural**, **generalizability**, external, and consequential aspects of construct **validity**. These six aspects are not separate and substitutable validity **types**, as in the traditional validity **conception**, but rather are interdependent and complementary **forms** of evidence in the unified view of **validity**. In **effect**, these six aspects together **function** as general validity criteria or standards for **all** educational and psychological **measurement**, including performance **assessments**.

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Generalizability of Performance Assessments

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Historically, in psychology and education, reliability issues have been addressed principally using classical test theory, which postulates that an observed score can be decomposed into a "true" score and a single, undifferentiated random error term, E (see Feldt & Brennan, 1989). Generalizability theory liberalizes classical theory by providing models and methods that allow an investigator to disentangle multiple sources of error that contribute to E . This is accomplished in part through the application of certain analysis of variance (ANOVA) methods.

The purposes of this chapter are: (a) to provide an overview of those aspects of the conceptual framework and methods of generalizability theory that are particularly relevant to performance assessments; (b) to integrate the current literature on the reliability of performance assessments into the framework of generalizability theory; and (c) to aid researchers and practitioners in the assessment of the generalizability of performance assessments. Reliability/generalizability coefficients are considered as well as error variances and standard errors of measurement. In addition, the generalizability of both individual scores and group mean scores is discussed.

Generalizability Theory: Basic Concepts

In a sense, classical test theory and ANOVA can be viewed as the parents of generalizability theory. However, generalizability theory has a unique conceptual framework. Among the concepts in this framework are *universes of admissible observations* and *G (Generalizability) studies*, as well as *universes of generalization* and *D (Decision) studies*.

An extensive, in depth explication of the concepts and methods of generalizability theory is provided by Cronbach, Gleser, Nanda, and Rajaratnam (1972). Brennan (1992a) provides a somewhat less detailed treatment. Overviews of essential features of generalizability theory are provided by Feldt and Brennan (1989), and Shavelson and Webb (1991). An introduction is provided by Brennan (1992b). Brennan and Johnson (1995) use generalizability theory to treat some of the issues covered in

this paper. Recently, in their consideration of generalizability analysis for educational assessments, Cronbach, Linn, Brennan, and Haertel (1995) have covered topics that partly overlap those treated in this paper.

In this section, the concepts and methods of generalizability theory are briefly explained and illustrated using an example from the performance testing literature reported by Shavelson, Baxter, and Gao (1993), who state that:

The California Assessment Program (CAP) conducted a voluntary statewide science assessment in 1989-1990. .. Students were posed five independent tasks. More specifically, students rotated through a series of five self-contained stations at timed intervals (about 15 mins.). At one station, students were asked to complete a problem solving task (determine which of these materials may serve as a conductor). At the next station, students were asked to develop a classification system for leaves and then to explain any adjustments necessary to include a new mystery leaf in the system. At yet another, students were asked to conduct tests with rocks and then use the results to determine the identity of an unknown rock. At the fourth station, students were asked to estimate and measure various characteristics of water (e.g., temperature, volume). And at the fifth station, students were asked to conduct a series of tests on samples of lake water to discover why fish are dying (e.g., is the water too acidic?). At each station, students were provided with the necessary materials and asked to respond to a series of questions in a specified format (e.g., fill in a table).

A predetermined scoring rubric developed by teams of teachers in California was used to evaluate the quality of students' written responses (California State Department of Education, 1990) to each of the tasks. Each rubric was used to score performance on a scale from 0 to 4 (0 = no attempt, 1 = serious flaws, 2 = satisfactory, 3 = competent, 4 = outstanding). All tasks were scored by three raters. (p. 222)

Universe of Admissible Observations and G Study Considerations

For the CAP example, the universe of admissible observations (UAO) consists of two facets: tasks (t) and raters (r). Since, in principle, any task could be evaluated by any rater, these facets are crossed in the UAO, and this crossing is denoted $t \times r$. Persons (p) or students are not viewed as part of the UAO. Rather, they constitute the population.

As reported by Shavelson et al. (1993), the G study design for the CAP example consisted of taking a sample of five tasks from the UAO, administering them to a sample of

persons, and then having three raters evaluate all products/results produced by all persons. This is a verbal description of a fully crossed G Study $p \times t \times r$ design.

Strictly speaking, for the CAP example, the G study is a random effects G study because the authors assumed that the potential set of tasks and raters in the UAO were both indefinitely large, with the actual tasks and raters in the G study viewed as samples from the UAO.

G study variance components. The principal results of a G study are estimated variance components for each of the effects in a G study design. These estimates are obtained using analysis of variance procedures (see Brennan, 1992a, for details). For the CAP example, the estimated variance components are reported in the second column of Table 1. For example, the estimated variance component for persons is $\hat{\sigma}_p^2 = .298$, and the estimated variance component for the interaction of persons and tasks is $\hat{\sigma}_{pt}^2 = .493$.

The variance component for persons can be interpreted in the following manner. Suppose an investigator could obtain each person's mean (or expected value) over all tasks and raters in the UAO. The variance of these scores would be σ_p^2 , which is estimated to be $\hat{\sigma}_p^2 = .298$ for the CAP data. Similarly, $\hat{\sigma}_r^2 = .003$ is the estimated variance of rater mean scores, where each mean (or expected value) is over all persons in the population and all tasks in the UAO. The estimated variance of task mean scores in the UAO is $\hat{\sigma}_t^2 = .092$, which suggests that tasks differ somewhat in difficulty.

Interaction variance components are somewhat more difficult to interpret. Consider, for example, $\hat{\sigma}_{pt}^2$ in the CAP example. The fact that $\hat{\sigma}_{pt}^2 = .493$ is considerably greater than zero suggests that there is a considerably different rank ordering of person mean scores for each of the various tasks in the UAO. By contrast, the fact that $\hat{\sigma}_{pr}^2 = 0$ means that the various raters rank order persons similarly. Also, $\hat{\sigma}_{rt}^2 = .002$ suggests that the various raters

Table 1
CAP Generalizability Analyses^a

		D Study Estimated Variance Components		
			$n'_t = 5$ $n'_r = 3$	$n'_t = 10$ $n'_r = 1$
	G Study $\hat{\sigma}^2$			
Persons (p)	0.298	$\hat{\sigma}_p^2$	0.298	0.298
Tasks (t)	0.092	$\hat{\sigma}_T^2 = \hat{\sigma}_t^2 / n'_t$	0.018	0.009
Raters (r)	0.003	$\hat{\sigma}_R^2 = \hat{\sigma}_r^2 / n'_r$	0.001	0.003
pt	0.493	$\hat{\sigma}_{pT}^2 = \hat{\sigma}_{pt}^2 / n'_t$	0.099	0.049
pr	0.000	$\hat{\sigma}_{pR}^2 = \hat{\sigma}_{pr}^2 / n'_r$	0.000	0.000
tr	0.002	$\hat{\sigma}_{TR}^2 = \hat{\sigma}_{tr}^2 / n'_t n'_r$	0.000	0.000
ptr,e	0.148	$\hat{\sigma}_{pTR,e}^2 = \hat{\sigma}_{ptr}^2 / n'_t n'_r$	0.010	0.015
		$\hat{\sigma}_\tau^2 = \sigma \hat{\gamma}_p^2$	=	0.30
		$\hat{\sigma}_\delta^2 = \hat{\sigma}_{pT}^2 + \hat{\sigma}_{pR}^2 + \hat{\sigma}_{pTR,e}^2$	=	0.11
		$\hat{\sigma}_\Delta^2 = \hat{\sigma}_\delta^2 + \hat{\sigma}_T^2 + \hat{\sigma}_R^2 + \hat{\sigma}_{TR}^2$	=	0.13
		$\hat{\rho}^2 = \hat{\sigma}_p^2 / [\hat{\sigma}_p^2 + \hat{\sigma}_\delta^2]$	=	0.73
		$\hat{\Phi} = \hat{\sigma}_p^2 / [\hat{\sigma}_p^2 + \hat{\sigma}_\Delta^2]$	=	0.70

^aG study variance components were provided by Xiaohong Gao.

rank order the difficulty of the tasks similarly. The last variance component, $\hat{\sigma}_{ptr,e}^2 = .148$, is a residual variance component that includes the triple-order interaction and all other unexplained sources of variation.

The fact that $\hat{\sigma}_r^2$, $\hat{\sigma}_{pr}^2$, and $\hat{\sigma}_{rt}^2$ are all close to zero suggests that the rater facet does not contribute much to variability in observed scores. By contrast, $\hat{\sigma}_t^2$ and especially $\hat{\sigma}_{pt}^2$ are quite large suggesting that the task facet contributes greatly to score variability.

The G study variance components provide a decomposition of the variance over p, t , and r of single person-task-rater scores:

$$\sigma_{X_{ptr}}^2 = \sigma_p^2 + \sigma_t^2 + \sigma_r^2 + \sigma_{pt}^2 + \sigma_{pr}^2 + \sigma_{tr}^2 + \sigma_{ptr,e}^2, \quad (1)$$

which is usually called "total variance" in the generalizability theory literature, because it is analogous to "total" sums of squares in analysis of variance (see Cronbach et al., 1972). That is, in generalizability theory, the phrase "total variance" refers to the sum of the G study variance components. From the second column of Table 1 it is evident that the largest contributors to total variance are persons and person-task interactions.

Other examples. The CAP assessment results are typical, in a sense, of generalizability results for many programs that involve performance assessments. For Example, Figure 1 reports the percent of total variance accounted for by each of the seven variance components in the $p \times t \times r$ design for CAP and five other performance assessment programs. In examining Figure 1 the reader is cautioned not to attach undue importance to the numerical values for the percents, which are for single person-task-rater scores. In particular, these percents should not be interpreted as percents for average (or total) scores for which decisions might be made. For purposes of this paper, what is important is that there are similarities in the profiles of the percents for the various studies. (Note that the magnitudes of the actual variance components across studies would not be comparable because, among other things, the studies involve different scoring metrics.)

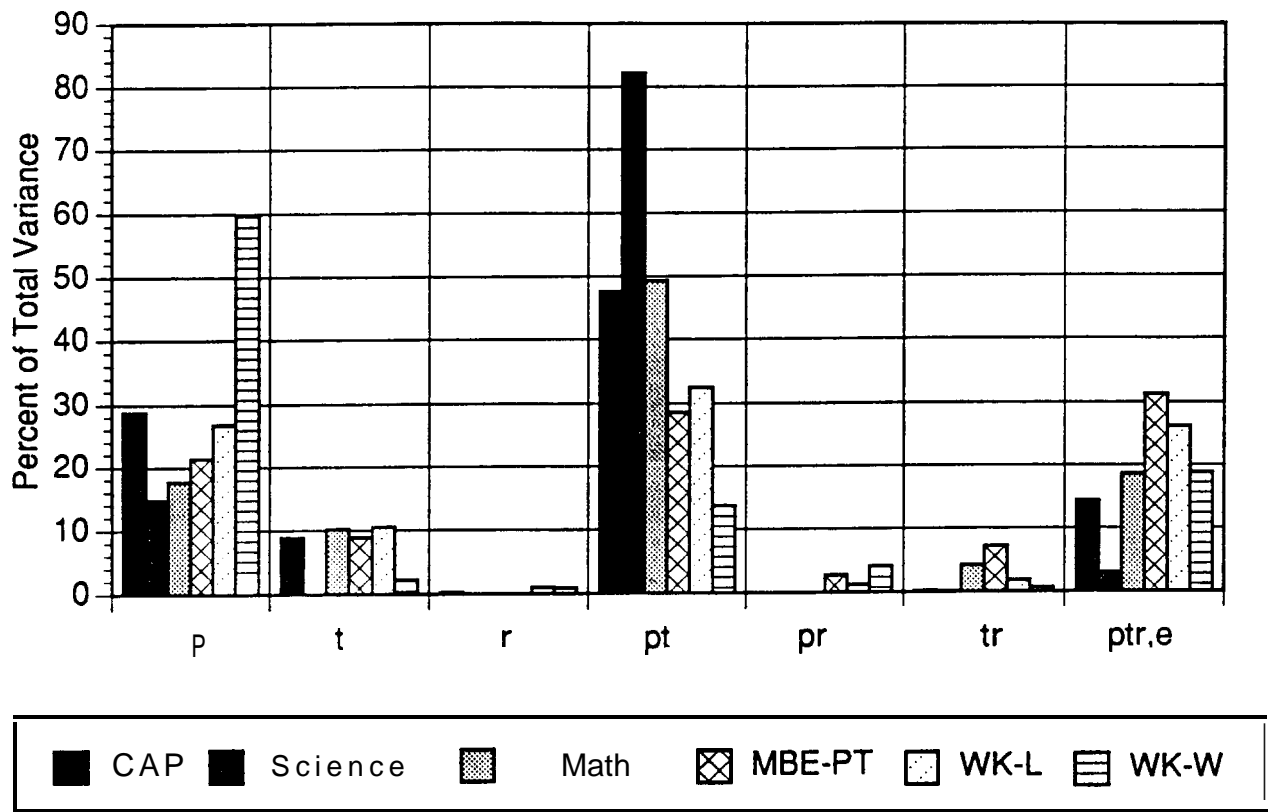


Figure 1. Percents of total variance for variance components for the $p \times t \times r$ design for six different assessments.

The science and math assessments are discussed by Shavelson et al. (1993) in the same article in which CAP is treated. The MBE-PT results are from a generalizability analysis of performance tasks for possible inclusion in the Multistate Bar Examination (MBE) used by many states to admit candidates to the practice of law (see Gamache & Brennan, 1994). The WK-L and WK-W are for Listening (L) and Writing (W) tests for a new program called Work Keys (WK) being developed by American College Testing (see Brennan, Gao, & Colton, 1995).

For all programs the rater facet contributes relatively little to total variance for single person-task-rater observations, as evidenced by the fact that $\hat{\sigma}_r^2$, $\hat{\sigma}_{pr}^2$, and $\hat{\sigma}_{tr}^2$ are all relatively small. It is also evident that for all programs $\hat{\sigma}_t^2$ is small suggesting that tasks are quite similar in average difficulty.

Perhaps most importantly, however, it is clear from Figure 1 that $\hat{\sigma}_{pt}^2$ is larger than $\hat{\sigma}_p^2$ for all assessments except WK-W and, indeed, $\hat{\sigma}_{pt}^2$ is the largest of all the variance components for CAP, Science, Math, and WK-L. The relatively large magnitude of $\hat{\sigma}_{pt}^2$ suggests that there is only a limited degree of across-task generalizability. A similar conclusion has been reported for other programs by Baker (1992), Dunbar, Koretz, & Hoover (1991), Lane, Stone, Ankenmann, and Liu (1992), Linn (1993), Linn and Burton (1994), van der Vleuton and Swanson (1990), and Welch (1991) among others.¹

For the $p \times t \times r$ design considered thus far, each rater evaluates all tasks performed by all persons. Sometimes, of course, such a design is not logistically or administratively feasible. A common alternative is a design in which different raters evaluate each task. For such a design, raters are nested within task, the design is denoted $p \times (r:t)$, and there are five variance components: σ_p^2 , σ_{τ}^2 , σ_{pt}^2 , and $\sigma_{pr:t,e}^2$. For example, this design is employed

operationally in the writing assessment component of the EXPLORE Program (ACT,1994).

For EXPLORE, the average over forms of the estimated variance components is:

$\hat{\sigma}_p^2 = .47$, $\hat{\sigma}_t^2 = .05$, $\hat{\sigma}_{r:t}^2 = .00$, $\hat{\sigma}_{pt}^2 = .52$, and $\hat{\sigma}_{pr:t,e}^2 = .27$. Again $\hat{\sigma}_{pt}^2$ is the largest variance component, and the rater facet contributes little to total variance.

Infinite Universe of Generalization and D Study Considerations

Effectively, Equation 1 states that G study variance components provide a decomposition of the total observed score variance for single person-task-rater scores. Let n_p , n_t , and n_r be the G study sample sizes for persons, tasks, and raters, respectively. Then, for the CAP example, the total observed score variance is the variance of the $n_p n_t n_r$ observed scores, each of which could be 0,1,2,3, or 4. In practice, of course, decisions about persons (the objects of measurement) will not likely be made based on persons' scores for a single task evaluated by a single rater. Rather, decisions will be made based on average scores over multiple tasks and/or raters. Indeed, a typical D (Decision) study consideration is to identify values of n'_t and n'_r (which need not equal n_t and n_r , respectively) that result in acceptably small error variance and/or acceptably large reliability-like coefficients.

Another D study consideration is the specification of a universe of generalization (UG), which is the universe to which a decision maker wants to generalize. In this section it will be assumed that the UG mirrors the UAO in the sense that the task and rater facets are both infinite. Strictly speaking, this means that the UG is a universe of randomly parallel forms of the measurement procedure, where each such form consists of a different sample of tasks and a different sample of raters. (In a subsequent section, a restricted UG is considered that is smaller than the infinite UG.) A person's universe score is his or her mean (or expected) score over all randomly parallel forms of the measurement procedure in the UG. As such, universe score is analogous to true score in classical theory.

An additional D study consideration is the design structure of the D study. In this section, it will be assumed that the D study design mirrors the G study design in the sense

that both are fully crossed, meaning that all persons respond to the same tasks, and the responses/products of all persons to all tasks are evaluated by the same raters.

D study variance components. G study variance components are used to estimate D study variance components for average scores over n'_t tasks and n'_r raters. The process is very simple. Let σ_α^2 be the G study variance component for α (e.g., if σ_{pt}^2 , $a = pt$), and let $\sigma_{\alpha'}^2$ be the corresponding D study variance component. If α contains t , then $\sigma_{\alpha'}^2 = \sigma_\alpha^2 / n'_t$; if α contains r , then $\sigma_{\alpha'}^2 = \sigma_\alpha^2 / n'_r$; and if α contains both t and r , then $\sigma_{\alpha'}^2 = \sigma_\alpha^2 / (n'_t n'_r)$. The resulting equations for the estimated D study variance components are provided in the third column of Table 1. Note that D study variance components are designated using upper-case subscripts for R and T to emphasize that these variance components are for mean scores over n'_r raters and n'_t tasks.

The D study variance component σ_p^2 is the variance of **persons'** universe scores for the infinite **UG**. As such, it is called universe score **variance**, which is analogous to true score variance in classical **theory**.

D study variance components are used to estimate relative and absolute error **variances**, as well as two reliability-like coefficients called **generalizability coefficients** and **dependability coefficients**.

Absolute error variance. Absolute error is simply the difference between a person's observed and universe score. The variance over persons of absolute errors is:

$$\begin{aligned}\sigma_\Delta^2 &= \sigma_T^2 + \sigma_R^2 + \sigma_{pT}^2 + \sigma_{pR}^2 + \sigma_{TR}^2 + \sigma_{pTR,e}^2 \\ &= \frac{\sigma_t^2}{n'_t} + \frac{\sigma_r^2}{n'_r} + \frac{\sigma_{pt}^2}{n'_t} + \frac{\sigma_{pr}^2}{n'_r} + \frac{\sigma_{tr}^2}{n'_t n'_r} + \frac{\sigma_{ptr,e}^2}{n'_t n'_r} .\end{aligned}\quad (2)$$

That is, absolute error variance is the sum of all the D study variance components except for universe score **variance**. For the CAP data with $n'_t = 5$ and $n'_r = 3$, Table 1 reports that $\hat{\sigma}_\Delta^2 = .13$. The square root is $\hat{\sigma}_\Delta = .36$, which is the A - type, or **absolute**, standard error of measurement (SEM). Consequently, adding and subtracting .36 to **persons'** observed scores

over five tasks and three raters provides approximate 68% confidence intervals for persons' universe scores.

Relative error variance. Relative error is the difference between a person's observed deviation score and his or her universe deviation score. The variance over persons of relative errors is:

$$\begin{aligned}\sigma_{\delta}^2 &= \sigma_{pT}^2 + \sigma_{pR}^2 + \sigma_{pTR,e}^2 \\ &= \frac{\sigma_{pt}^2}{n'_t} + \frac{\sigma_{pr}^2}{n'_r} + \frac{\sigma_{ptr,e}^2}{n'_t n'_r}\end{aligned}\quad (3)$$

That is, relative error variance is the sum of the D study variance components that include the index p and at least one other index. The square root of relative error variance is analogous to the SEM in classical test theory. For the CAP data with $n'_t = 5$ and $n'_r = 3$, relative error variance is $\hat{\sigma}_{\delta}^2 = .11$, and the square root is $\hat{\sigma}_{\delta} = .33$, which is the δ -type, or relative, SEM. Note that $\sigma_{\delta}^2 \leq \sigma_{\Delta}^2$ because σ_{δ}^2 does not contain σ_T^2 , σ_R^2 , or σ_{TR}^2 .

Generalizability coefficient. A generalizability coefficient is defined as:

$$\rho^2 = \frac{\sigma_{\tau}^2}{\sigma_{\tau}^2 + \sigma_{\delta}^2}, \quad (4)$$

where σ_{τ}^2 is a generic notation for universe score variance. That is, a generalizability coefficient is the ratio of universe score variance to itself plus relative error variance. As such, a generalizability coefficient is analogous to a reliability coefficient in classical theory. For the case considered in this section, $\sigma_{\tau}^2 = \sigma_p^2$, and σ_{δ}^2 is given by Equation 3.

Therefore,

$$\rho^2 = \frac{\sigma_p^2}{\sigma_p^2 + \frac{\sigma_{pt}^2}{n'_t} + \frac{\sigma_{pr}^2}{n'_r} + \frac{\sigma_{ptr,e}^2}{n'_t n'_r}}. \quad (5)$$

For the CAP data with $n'_i = 5$ and $n'_r = 3$, $\hat{\rho}^2 = .73$, as indicated in Table 1.

Dependability coefficient. A dependability coefficient is defined as:

$$\Phi = \frac{\sigma_{\tau}^2}{\sigma_{\tau}^2 + \sigma_{\Delta}^2} . \quad (6)$$

That is, a dependability coefficient is the ratio of universe score variance to itself plus absolute **error variance**. For the case considered in this section, $\sigma_{\tau}^2 = \sigma_p^2$, and σ_{Δ}^2 is given by Equation 2 Therefore,

$$\Phi = \frac{\sigma_p^2}{\sigma_p^2 + \frac{\sigma_t^2}{n'_i} + \frac{\sigma_r^2}{n'_r} + \frac{\sigma_{pt}^2}{n'_i} + \frac{\sigma_{pr}^2}{n'_r} + \frac{\sigma_{tr}^2}{n'_i n'_r} + \frac{\sigma_{ptr,e}^2}{n'_i n'_r}} . \quad (7)$$

The only difference between ρ^2 and Φ is that ρ^2 uses σ_{δ}^2 as error **variance**, whereas Φ uses σ_{Δ}^2 as error **variance**. Since $\sigma_{\Delta}^2 \geq \sigma_{\delta}^2$, it follows that $\Phi \leq \rho^2$. From Table 1, for $n'_i = 5$ and $n'_r = 3$, $\hat{\Phi} = .70$ which is less than $\hat{\rho}^2 = .73$.

Consequences of different sample sizes. Figure 2 provides $\hat{\rho}^2$, $\hat{\Phi}$, $\hat{\sigma}_{\delta}$, and $\hat{\sigma}_{\Delta}$ for the CAP example with n'_i ranging from 1 to 12 and n'_r ranging from 1 to 3. These results might be employed to examine the consequences of using various numbers of tasks and **raters**.

Perhaps the most **striking** result in Figure 2 is that the number of raters has very little influence on the magnitude of **SEM's** and **coefficients**. This is a direct result of the previously noted fact that variance components involving the rater facet are quite **small**, presumably because the scoring rubrics are well defined and the raters are well **trained**.

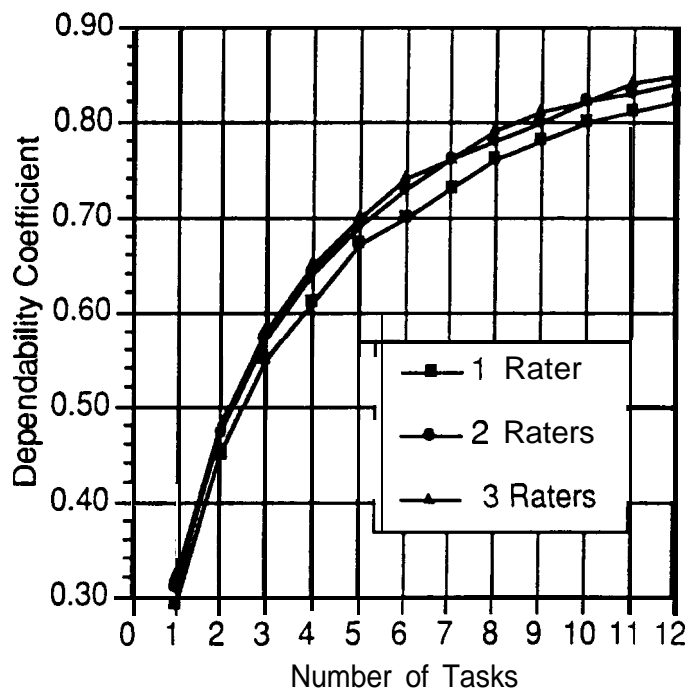
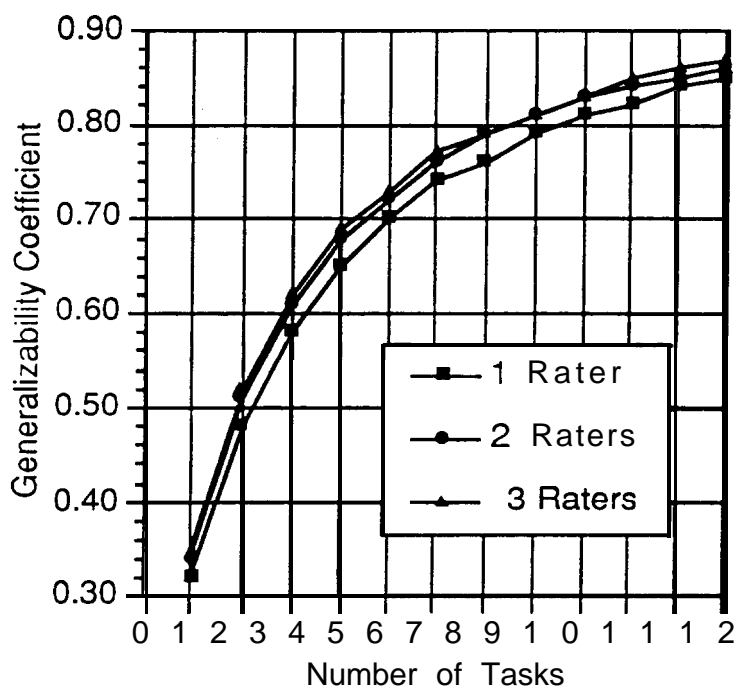
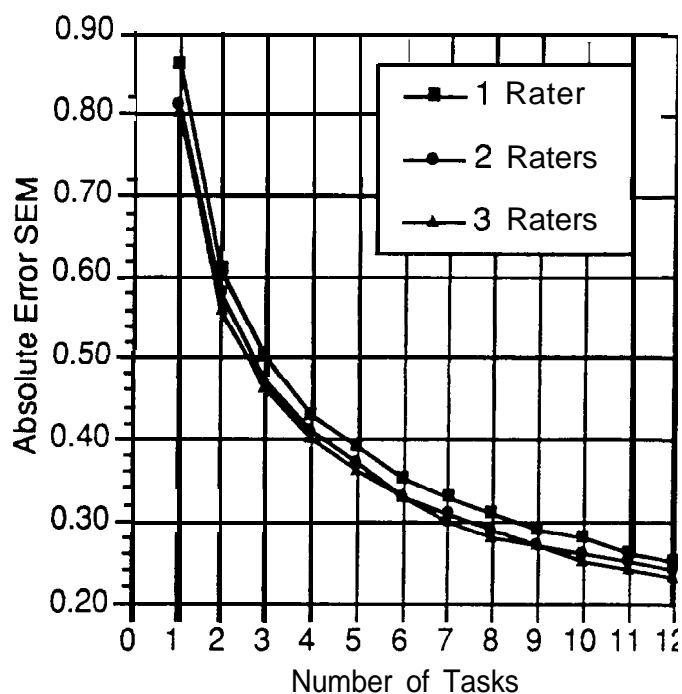
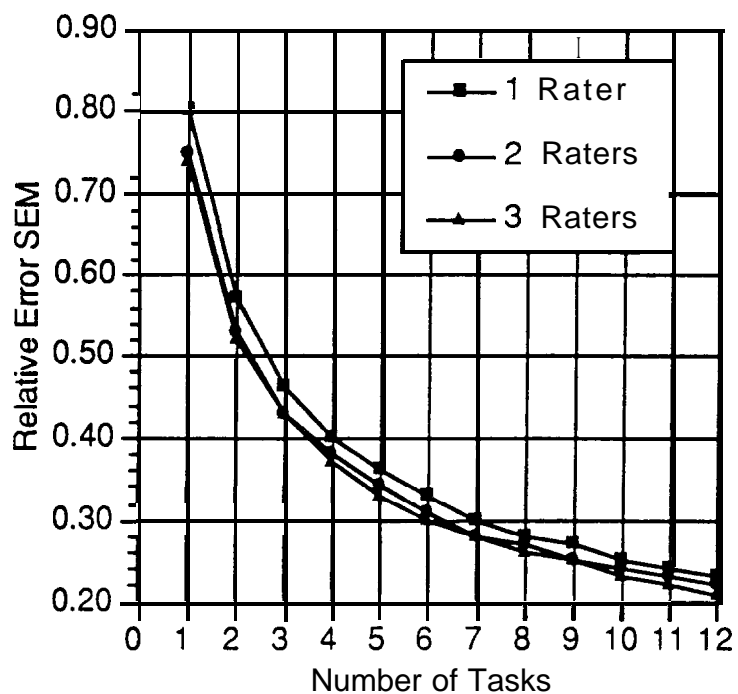


Figure 2. Generalizability Results for CAP

Consequently, there seems to be no compelling psychometric reason to employ more than one rater.

Another obvious fact from Figure 2 is that increasing the number of tasks has a dramatic effect on lowering SEM's and increasing the values of coefficients. With one rater, nine tasks are required for $\hat{\rho}^2 \geq .80$, and ten tasks are needed for $\hat{\Phi} \geq .80$ (see also the last column of Table 1). Developing, administering, and scoring that many tasks obviously would not be a trivial undertaking.

On the other hand, in some circumstances, an investigator might argue that it is not too sensible to choose n'_i on the basis of the resulting magnitude of $\hat{\rho}^2$ or $\hat{\Phi}$, because both depend on universe score variance. Such an investigator might be satisfied with a somewhat low value for a coefficient provided individual persons were measured accurately enough. If so, the investigator would be more interested in the magnitude of SEM's than coefficients. To consider this possibility, recall that CAP scores for a single task range from 0 to 4, which means that average scores over n'_i tasks have the same range, but with fractional scores frequently occurring. Assuming normally distributed observed scores, if the investigator wanted to be 95% certain that persons' observed and universe scores differed by no more than two points, only three tasks are required with one rater. However, to be 95% certain that observed and universe scores differ by no more than one point, 12 tasks are required with one rater, and about 10 tasks are required with three raters!²

When $n'_r = 1$, Figure 3 provides 68% and 95% Δ -type confidence intervals for the CAP data for examinees with a universe score of $\tau = 2$, assuming normally distributed observed scores about τ . For example, when $n'_i = 3$, there is a 68% probability that examinees with a true score of 2 will obtain observed scores between 1.5 and 2.5.

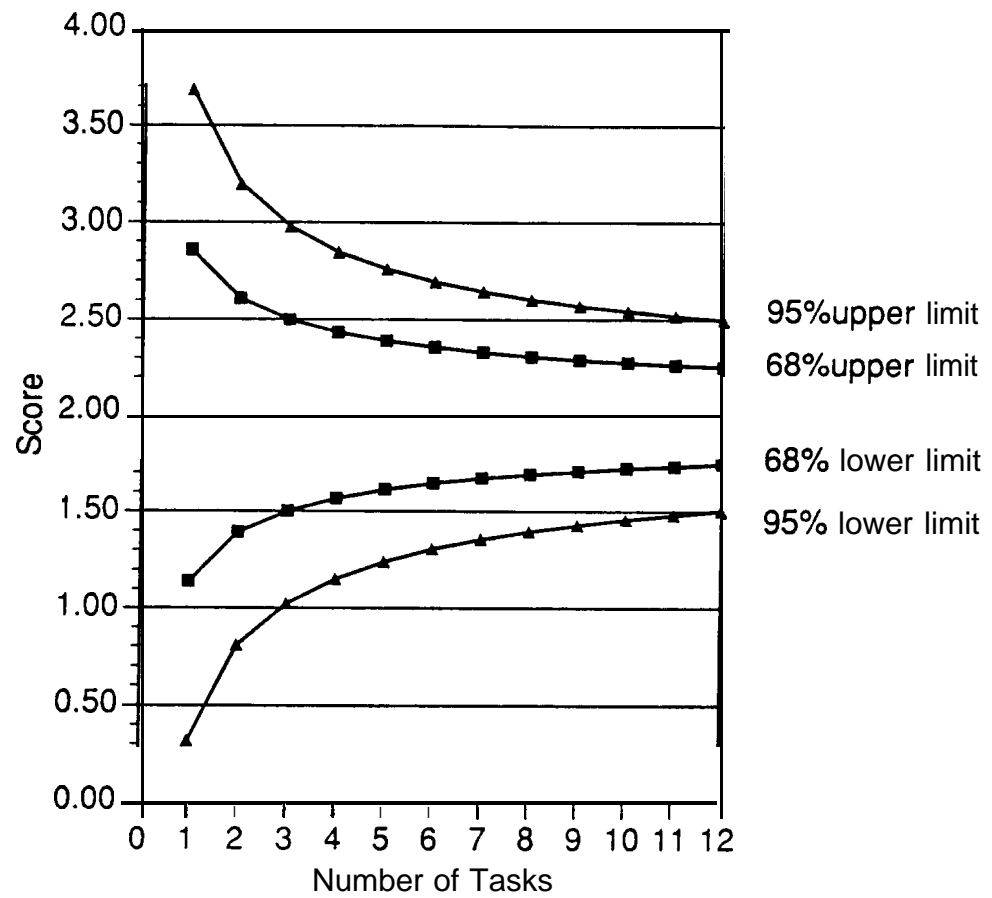


Figure 3. Confidence intervals for CAP data based on absolute error SEM's using a single rater and various numbers of tasks, assuming a universe score of 2.

The limits of the confidence intervals in Figure 3 can also be used as minimum passing and maximum failing scores, as discussed by Linn(1994). For **example**, from Figure 3, with $n'_i = 8$, a minimum passing score of 2.6 and a maximum failing score of 1.4 are required for 95% confidence that correct decisions will be made when the standard is set at 2.

Contributions of Different Facets

This section gives additional consideration to the influence of **rater**, **task**, and other facets on the measurement precision of performance **assessments**. In doing **so**, each facet is treated somewhat **separately**. This is done to isolate issues associated with the various **facets**, and to facilitate relating traditional results to the types of **generalizability** results discussed in the previous **section**. In **practice**, of **course**, **generalizability** analyses involving many facets simultaneously are greatly to be **preferred**.

Generalizing Over Raters

The examples in the previous section suggest that for current performance assessments the rater facet does not **contribute** much to variability in observed **scores**. While these results have been found in a wide range of current performance **assessments**, it should not be assumed that such results are necessarily **inevitable**. For **example**, Linn (1993) reports that the classic studies of Starch and Elliott (1912,1913) on the grading of high school work in English and mathematics demonstrated an extraordinary range of grades assigned to written essays and extended responses in **geometry**. **Indeed**, the Starch and Elliott studies provided considerable support to the increased use of objective testing in the early part of this **century**. **Undoubtedly**, lack of trained raters and agreed-upon scoring rubrics **contributed** to the Starch and Elliott **results**.

Until **recently**, for any measurement procedure involving subjective **scoring**, probably the most **frequently** discussed measurement issue was inter-rater **reliability**. **Indeed**, if inter-rater reliability was **high**, it was often assumed that there were no other reliability issues of **consequence**. Although this narrow perspective no longer

predominates, inter-rater reliability is still a very important issue. Indeed, high inter-rater reliability is viewed by most investigators as a necessary, although not sufficient, condition for adopting a performance test.

The phrase "inter-rater reliability" may seem to have a self-evident interpretation. Actually, however, there are at least two general measurement perspectives on inter-rater reliability. One perspective typically involves inter-rater reliability coefficients. The other perspective considers error variances or standard errors. Both perspectives involve observed differences in ratings, but the two perspectives are not isomorphic.

Inter-rater reliability coefficients. In the performance testing literature, two general conclusions about inter-rater reliability seem to predominate. First, when tasks are the same for all students and scoring procedures are well-defined, inter-rater reliability tends to be quite high. Second, when different students respond to different tasks, choose their own tasks (e.g., select their own essay topics), or produce unique products, then inter-rater reliability tends to be relatively low. This appears to be especially true for portfolio assessments (see Gearhart, Herman, Baker, & Whittaker, 1992, and Koretz, Klein, McCaffrey, & Stecher, 1993). Another way to state these conclusions is to say that when tasks are standardized inter-rater reliability tends to be high, and when tasks are not standardized it tends to be low.

These conclusions are to be expected based on the manner in which variance components enter the standardized and non-standardized inter-rater reliability estimates. The standardized estimate is typically obtained by correlating the ratings of two different raters to the responses of a group of persons to the same task.³ In terms of the variance components introduced in the first section, this correlation is approximately equal to the generalizability coefficient

$$\rho^2 = \frac{\sigma_p^2 + \sigma_{pt}^2}{\sigma_p^2 + \sigma_{pt}^2 + \sigma_{pr}^2 + \sigma_{ptr,e}^2}. \quad (8)$$

The denominator of Equation 8 is identical to the denominator of ρ^2 in Equation 5 when $n'_t = n'_r = 1$. For the case considered in this section, $n'_t = 1$ because only one task is involved in the correlation, and $n'_r = 1$ because a correlation between two raters gives an estimate of reliability for a single rater.

The numerators of the generalizability coefficients in Equations 5 and 8 differ in that Equation 8 includes not only σ_p^2 but also σ_{pt}^2 . When all persons respond to the same single task, effectively the single task is hidden and fixed for all persons.

Consequently, the universe of generalization is a restricted universe in which raters constitute the only random facet. Statistically this leads to σ_{pt}^2 being included with σ_p^2 in the numerator of Equation 8. Because σ_{pt}^2 is usually quite large in performance testing, the numerator of ρ^2 in Equation 8 is likely to be large resulting in a relatively high value of ρ^2 .

Effectively, σ_{pt}^2 is part of universe score variance in Equation 8, whereas almost always σ_{pt}^2 is more properly viewed as part of error variance (see Equations 2 and 3). For this reason (and another reason considered later), almost always inter-rater reliability coefficients for standardized situations are too big relative to more appropriate estimates of generalizability for the reported scores on a performance test (see Equation 5).

The non-standardized estimate of inter-rater reliability is typically obtained by correlating two ratings of a different task or product for each person. In such cases, the design has tasks or products nested within persons and crossed with raters, $[(t:p) \times r]$, and the correlation is approximately equal to the generalizability coefficient

$$\rho^2 = \frac{\sigma_p^2}{\sigma_p^2 + \sigma_{t:p}^2 + \sigma_{pr}^2 + \sigma_{tr:p,e}^2}, \quad (9)$$

where $\sigma_{t:p}^2 = \sigma_t^2 + \sigma_{pt}^2$ and $\sigma_{tr:p,e}^2 = \sigma_{tr}^2 + \sigma_{ptr,e}^2$. The non-standardized coefficient in Equation 9 will be less than the standardized coefficient in Equation 8 for two reasons: (a) the numerator of the non-standardized coefficient is smaller than the numerator of the

standardized coefficient by σ_{pt}^2 ; and (b) the denominator of the non-standardized coefficient is larger than the denominator of the standardized coefficient by $\sigma_t^2 + \sigma_{tr}^2$, which contributes to the relative error variance σ_g^2 for the non-standardized coefficient.

The variance component σ_{pt}^2 does not appear in the numerator of the non-standardized coefficient in Equation 9 because tasks vary over persons, and therefore person-task interaction does not contribute to universe score variance. Rather σ_{pt}^2 is included in $\sigma_{t:p}^2$ and contributes to error variance. The different role that σ_{pt}^2 plays in Equations 8 and 9 is often the principal reason that the non-standardized inter-rater reliability coefficient is smaller than the standardized coefficient. In addition, Equation 9 will tend to be smaller than Equation 8 if tasks are unequal in average difficulty (i.e., σ_t^2 is relatively large), and/or raters are differentially capable of rating different tasks (i.e., σ_{tr}^2 is relatively large). Either or both of these conditions are probably more likely to characterize portfolio assessments than other types of performance assessments.

The generalizability coefficients in Equations 8 and 9 approximate the inter-rater reliability coefficients most frequently reported in the literature. Note, however, that they are for making decisions based on only one rating of only one task. This is evident from the fact that the variance components in Equations 8 and 9 containing r and/or t are all divided by one. Frequently, inter-rater reliability coefficients are incorrectly interpreted as estimates of reliability when persons' scores are the sum or average of two ratings. Such estimates are easily obtained from Equations 8 and 9 by halving variance components that contain r . The resulting estimates of generalizability for two ratings are necessarily larger than those for a single rating.

Standard errors and error variances. In the performance testing literature, issues of rater reliability are most frequently discussed in terms of inter-rater reliability coefficients. Sometimes, however, such issues are treated (and perhaps more appropriately treated) from the perspective of differences in the actual ratings (e.g., a plot of two ratings for each person). It is intuitively clear that such differences reflect error in

some **sense**. Indeed, it can be shown that, when data are available for two ratings and decisions will be based on the mean of $k' = 1$ or 2 ratings, the A-type standard error for a given person is

$$\hat{\sigma}_{\Delta_p} = |X_{p1} - X_{p2}| / \sqrt{2k'}. \quad (10)$$

This standard error is the absolute value of the difference between the two ratings for the person divided by $\sqrt{2}$ if $k' = 1$, or divided by 2 if $k' = 2$. This equation is appropriate whether or not each person responds to the same **task(s)** -- i.e., whether or not tasks are **standardized**. The average over persons of individual error variances is the overall A-type error **variance**:

$$\hat{\sigma}_{\Delta}^2 = \sum_p \hat{\sigma}_{\Delta_p}^2 / n_p. \quad (11)$$

If the two ratings are based on the same **task**, then task is fixed and generalization is over the rater **facet**, only. If the two ratings are based on different **tasks**, then generalization is over both tasks and **raters**. In **general**, of **course**, standard errors for generalizing over **raters**, only, are likely to be smaller than standard errors for generalizing over both tasks and **raters**. For **example**, Gamache and Brennan (1994) report that $\hat{\sigma}_{\Delta_p}$ for generalizing over **raters**, only, was on average about 60% as large as $\hat{\sigma}_{\Delta_p}$ for generalizing over both raters and **tasks**. Their results are for experimental performance tasks for a law **examination**.

Adjudication and other issues. In performance **testing**, it is relatively common practice to obtain a third rating if two ratings differ by more than one rating scale **point**. There are various ways in which this third rating might be **used**, but very often the net effect is that the final two ratings for a person differ by no more than one **point**. Under these **circumstances**, for $k' = 2$, $\hat{\sigma}_{\Delta_p}$ is either .5 or 0. Furthermore, if p_1 is the

proportion of persons with a difference score of 1, the overall value of $\hat{\sigma}_{\Delta}$ will be $.5\sqrt{p_1}$, which is likely to be quite small. For example, if $p_1 = .36$ then $\hat{\sigma}_{\Delta} = .30$ of a rating scale point.

In short, whenever some process is used to adjudicate ratings, standard errors of measurement are likely to be relatively small, and in this sense the ratings will appear quite **reliable**. Hence, an adjudication process is beneficial only if it does not distort the intended construct being **rated**. Such distortion could **occur**, for **example**, if the adjudicators were systematically influenced by the original **ratings**, or if the adjudicators were experts untrained in the scoring **rubric**.

Carefully constructed scoring **rubrics**, an intensive training session for **raters**, and an adjudication process usually produce ratings with small error **variance**. However, small error variance does not guarantee that inter-rater reliability coefficients will be **high**. This follows from the fact that universe score **variance**, σ_p^2 , is in the numerator of any inter-rater reliability **coefficient**, but is absent from error **variance**. Consequently, if σ_p^2 is small relative to error **variance**, then an inter-rater reliability coefficient could be small even if error variance is small. This "thought experiment" illustrates that an **inter-rater** reliability coefficient encapsulates information about the magnitude of true differences among person (universe score **variance**) relative to **errors**.

If true differences among persons are of no consequence -- as might be the case in a fully criterion-referenced context -- then inter-rater reliability coefficients may be of little **value**. Real world testing contexts are seldom so clear **cut**, however. In short, error variances and inter-rater reliability coefficients capture **overlapping**, but still **different**, types of **information**. **Often, therefore**, it is sensible to report **both**. In any **case**, it is always advisable to report estimated variance components (see AERA, APA, NCME, 1985, p.19).

Rater vs. Task Reliability

The use of well-constructed scoring rubrics with well-trained raters can substantially reduce errors attributable to **raters**. However, virtually **all** currently available research on performance testing suggests that generalizing over tasks is an error-prone **activity**, no matter how **well** the tasks are **designed**, primarily because σ_{pt}^2 tends to **be** relatively **large**, and secondarily because tasks tend to be somewhat different in difficulty (σ_t^2 is greater than 0).

Dunbar et al. (1991) reviewed a number of studies of direct assessments of **performance**, primarily in the area of **writing**. To compare the influence of raters and tasks on reliability they computed average reliability due to raters and what they called "**score reliability**."⁴ In the notation of this **chapter**, the Dunbar et al. (1991) reliability due to raters is approximated by Equation 8 (provided all persons are rated by the same **raters**), and

$$\text{score reliability} \equiv \frac{\sigma_p^2}{\sigma_p^2 + \sigma_{pt}^2}. \quad (12)$$

Since Equation 12 involves σ_{pt}^2 , in the **terminology** of this chapter it seems more sensible to refer to this coefficient as "**task reliability** ." It is equivalent to the **generalizability** coefficient in Equation 5 for a single task and an infinite number of **raters**. Another interpretation of Equation 12 is that it is a **generalizability** coefficient under the assumption that raters are perfectly consistent in **all respects**, and the only source of error is person-task **interactions**.

Although comparing Equations 8 and 12 is **awkward**, these equations do give some sense of the relative influence of raters and tasks on **generalizability**. Note also that multiplying the right-hand sides of Equations 8 and 12 gives the **generalizability** coefficient in Equation 5 for a single task and a single **rater**. (See Kane, 1982, pp. 145-146, for a discussion of a similar result in **terms** of the reliability-validity **paradox**.) Table

2, with slight modifications, is from Dunbar et al.(1991). This table summarizes results from the studies they examined.

Several observations are evident from Table 2. Most importantly, for all studies task reliability is relatively small suggesting that person-task interactions are a considerable source of error. It is also evident from Table 2 that there is considerable variability in rater reliability. In particular, rater reliability tends to be lower for the older studies. Commenting on this, Dunbar et al.(1991) state:

Lindquist (1927), for example, provided very general scoring rubrics for a 10-point holistic scale yielding a mean rater reliability of .33, whereas Hieronymus and Hoover (1987) developed very specific rubrics for a 4-point scale and obtained a coefficient of .91 from ratings made by a group of professional readers. The data in Hildebrand (1991) were collected by the same procedures with the same instrumentation used by Hieronymus and Hoover, but the setting was that of a field experiment rather than a controlled standardization of a set of scoring protocols. This contrast, a controlled standardization versus a field experiment, demonstrates the effect that administrative conditions can have on the reliability of raters.(p.293).

It appears, then, that often methods can be found to increase rater reliability. However, comparable methods do not exist for increasing task reliability. Of course, tasks should be developed as carefully as possible, and if this is not done then it is likely that reliability will be adversely affected. However, well-constructed tasks do not ensure high reliability, primarily because there is considerable variability in examinee performance on different tasks -- even for tasks in the same domain. That is, σ_{pt}^2 tends to be relatively large.

Of course, in principal, σ_{pt}^2 can be reduced by narrowing the domain of tasks. So, for example, an investigator could define the tasks in a domain in such a way that each of them is simply a slight modification of the others. Doing so may well decrease σ_{pt}^2 and, therefore, increase task reliability. However, restricting the domain of tasks in this way leads to a narrowing of the universe of generalization and, in this sense, a decrease in

Table 2^aReliability Studies of **Direct** Assessments of Performance

Data source	Measurement context	Average reliability	
		Raters	Tasks
Lindquist (1927)	Assess effectiveness of a laboratory method for instruction in writing for college students	.33	.26
Coffman (1966)	Determine the validity of objective tests for predicting composite essay scores	.39	.26
Swartz, Patience, & Whitney (1985)	Develop an assessment of writing skill for awarding high school equivalency diplomas	.74	.60
Applebee, Langer, & Mullis (1986)	Characterize national trends in writing skill among 9-, 13- & 17-year olds in NAEP	.78	NR
Breland, Camp, Jones, Morris, & Rock (1987)	Evaluate the use of essay tests to predict performance in college-level writing courses	.59	.41
Hieronimus & Hoover (1987)	Develop performance-based measure of writing skill in grades 3 through 8	.91	.46
Hildebrand (1991)	Measure the effects of revision strategies on scores in direct writing assessment	.67	NR
Purves (1992)	Develop writing tasks and scoring protocols for international comparisons of achievement	NR	.42
Welch (1991)	Assess generalizability of essay scores on test for second-year college students	.76	.44 ^b

Note. The reliability estimates reported here are simple averages of all the coefficients reported in the original. They are adjusted to reflect an assessment based on 1 reader and 1 sample of performance via the Spearman-Brown formula. NR means "not reported."

^aFrom Dunbar, Koretz, and Hoover (1991), with minor modifications.

^bComputed using Equation 12, which gives a different result from that reported by Dunbar, Koretz, and Hoover (1991).

validity (see Kane,1982). This is an example of the so-called "reliability-validity paradox. " It is generally not advisable to take steps to increase reliability that lead to a decrease in validity.

One of the most important considerations in the development of a performance test is a careful specification of the task facet in the universe of **generalization**. At a **minimum**, an investigator should be able to defend the set of tasks in a performance test as a reasonable representation of the domain of tasks that might have been **used**. **Otherwise**, there is little basis for claiming that performance on the particular tasks in a performance test can be generalized to a larger universe of **tasks**. (See Shavelson et al., 1993, p. 216, for an example of a specification of a domain of **tasks**.)

The importance of accurate specification of a subject matter domain for **performance** assessments has been illustrated by **Shavelson, Gao, and Baxter (1996)**. For the domain of elementary science they demonstrated that an inappropriately broad specification of the domain leads to overestimating task variability and underestimating **generalizability**. **Conversely**, an inappropriately narrow specification of the domain will likely lead to underestimating task variability and overestimating **generalizability**. **Note, however**, that **generalizability** theory per se does not **tell** an investigator how narrow or wide a domain should **be**. It is the investigator's responsibility to clearly specify the domain and defend that **specification**.

Other Facets

Shavelson et al.(1993) provide the following perspective on relevant facets for performance **assessments**:

... we view a performance assessment as a sample of student **performance** drawn from a complex universe defined by a combination of **all possible tasks, occasions, raters, and measurement methods**. We view the task facet to be representative of the content in a subject-matter **domain**. The occasion facet includes **all possible occasions** on which a decision maker would be equally willing to accept a score on the performance **assessment**. We view the rater facet as including **all possible individuals** who could be trained to score performance **reliably**. These

three facets are, traditionally, thought of as sources of unreliability in a measurement.

In addition, we incorporate a method facet into our definition of the universe of generalization. This formulation moves us beyond reliability into a sampling theory of validity (cf. Kane, 1982). Specifically, we view the method facet to be all possible methods (e.g., short answer, computer simulation) that a decision maker would be equally willing to interpret as bearing on student achievement.

Occasion as a facet. From a classical perspective, sampling variability due to occasions most closely corresponds to the notion of test-retest reliability. Also, variability due to occasions is incorporated in traditional notions of intra-rater reliability, which reflects variability in ratings for the same raters on two occasions. Ideally, from the perspective of minimizing error variance, an investigator would like examinee performance-test products to be minimally changed over occasions during which no instruction occurred. Similarly, an investigator would like ratings for the same raters to be stable over occasions.

For at least two reasons, there are very few studies in the performance testing literature that incorporate more than one occasion. First, doing so is logistically difficult and quite costly. Second, in operational settings collecting data on two occasions is usually not an intended part of the testing process. Even so, it is highly desirable that at least small-scale G studies be conducted that involve occasion as a facet in order to examine the extent to which an investigator can legitimately claim that scores obtained on one occasion are generalizable to scores that might be obtained on different, but similar, occasions.

Ruiz-Primo, Baxter, and Shavelson (1993) and Shavelson et al. (1993) examined the stability of several elementary science performance assessments. Their results suggest that variance attributable to the interaction of persons, tasks, and occasions (σ_{pio}^2) was very large -- indeed, many times larger than universe score variance and also larger than σ_{pt}^2 . However, variance attributable to persons and occasions (σ_{po}^2) was quite small. This means that, over an infinite number of tasks, there was little person-occasion interaction in their data, but for any single task persons were rank ordered

differently on different occasions. The obvious remedy for large σ_{pto}^2 is to use a large number of tasks and/or occasions in making decisions about **examinees**, but this may not be feasible in **practice**.

Another study of the stability of science performance assessments has been reported by Tamir (1974), who found that reliability was on the order of .35 for a design that involved equivalent **problems**, different **raters**, and two **occasions**. Also, Carey (1991) and Mayberry and Hiatt (1990) studied the stability of military job **performance**, and found retest reliabilities of about .70.

Usually, when occasion is considered as a facet in the performance testing **literature**, it is associated with the time when **examinee** performance occurs or products are **created**. **Actually, however**, there is a second occasion facet that could influence the **generalizability** of scores on performance **assessments**. This facet involves the **occasion(s)** on which the ratings are **obtained**. This second occasion facet would be important if **judges'** rating were not stable over **time**. For **example**, Wainer (1993) comments **that**:

During the course of the 1988 NAEP writing **assessment**, some 1984 essays were included in the mix for the purpose of assessing **change**. The difference in the scores of these **essays**, from one assessment to the **next**, was so large that it was deemed wise to determine change through the very expensive restoring of a large sample of 1984 essays by the 1988 judges (Johnson & Zwick, 1988). No mere statistical adjustment was apparently sufficient (p.15).

Actually, this is a very complicated situation because change in occasion is confounded with change in judges (**and** perhaps subtle changes in rubrics or training **procedures**).

It would be imprudent to use the results of the studies referenced in the previous paragraphs as a basis for sweeping conclusions about the extent to which scores on performance tasks are invariant over **occasions**. **However**, given these results and results with other modes of **testing**, it certainly would be surprising if scores on performance tasks did not exhibit at least some variability over occasions of testing and/or **rating**.

Method as a facet. Cronbach et al. (1972), Brennan (1992a), and Kane (1982) have all observed that generalizability theory often blurs distinctions between reliability and **validity**. This is indeed the case if method or mode of testing (**e.g., performance tasks, multiple-choice items, short-answer questions, etc.**) is incorporated as a facet in the universe of **generalization**.

If results are invariant over mode of **testing**, then there is evidence of convergent **validity**, and a supportable argument can be made that different modes of testing provide exchangeable **information**. If results are not invariant over mode of **testing**, then different modes provide different types of information about student **performance**.

Shavelson et al. (1993) examined mode of testing for two science **performance** tasks and four methods (**observations of actual performance, notebook reports of steps employed, computer administration of tasks, and short-answer questions**). They concluded that not **all** methods **converge**, and "**certain** methods may measure different aspects of achievement (p. 229). " Although this is **only one study**, it seems plausible that their conclusions will generalize to other **settings**. This does not **mean, however**, that one mode is preferable to **another**. Such a conclusion can be drawn only through a joint consideration of psychometric **properties, content, context, logistical, and cost considerations**.

Scoring rubrics/procedures. Wainer (1993, p.15) suggests that performance assessments yield acceptable levels of accuracy only when scoring rubrics are rigidly **defined**. This may be one reason why most empirical analyses of performance assessments effectively consider scoring rubrics as fixed in the sense that only one rubric is **used**. In **principal, however**, there may **be** many rubrics that could **be used**. This is another **example** of blurred distinctions between reliability and **validity**. **If** two or more rubrics are in **principal** equally **acceptable**, then the issue is primarily in the realm of **reliability**. **However**, if the acceptable rubrics are not equally **preferable**, then the matter

is largely one of validity. For example, the "ideal" rubric may be so costly to implement that a simpler rubric is adopted for operational use.

Put in the language of **generalizability theory**, the issue is the extent to which an investigator can generalize over **rubrics**, or the extent to which **examinee** scores are in some sense invariant over **rubrics**. If scores vary depending on the rubric, then it is not likely that scores can be interpreted meaningfully without a clear understanding of the specific rubric **employed**. This is one reason why the interpretation of performance test scores is often more demanding than for traditional modes of **testing**. Decision makers must understand not only what is being tested but also the standards and procedures used to assign **scores**.

Generalizability of Group Mean Scores

In many contexts performance assessments are used to make decisions about groups (**e.g., classes, schools, or districts**) rather than **individuals**. In such **cases**, the objects of measurement are groups rather than **persons**, and the scores of interest are group mean scores (see Kane & Brennan, 1977).

Complete Desire

Suppose pupils (p) are nested within groups (g), and all pupils respond to the same tasks (t). This is a description of the $(p:g) \times t$ design. Of course, the pupils' products must be scored by raters, but the rater facet will be suppressed in this section to simplify discussion. As noted previously, the literature suggests that doing so is not too problematic if raters are well trained to use carefully constructed **rubrics**.

For the $(p:g) \times t$ design there are five variance components: σ^2 , σ_{gt}^2 , and $\sigma_{pt:g,e}^2$. Since the objects of measurement are **groups**, the facets in the universe are pupils and **tasks**, both of which are assumed to be **random, here**. Under these **circumstances**, universe score variance is

$$\sigma_{\tau}^2 = \sigma_g^2. \quad (13)$$

That is, universe score variance is variance attributable to group means.

Letting $n'_{p:g}$ be the D study number of pupils within each group, and letting n'_t be the D study number of tasks, relative error variance is '

$$\sigma_{\delta}^2 = \frac{\sigma_{p:g}^2}{n'_{p:g}} + \frac{\sigma_{gt}^2}{n'_t} + \frac{\sigma_{pt:g,e}^2}{n'_{p:g}n'_t}, \quad (14)$$

and absolute error variance is

$$\sigma_{\Delta}^2 = \sigma_{\delta}^2 + \frac{\sigma_t^2}{n'_t}. \quad (15)$$

Note, in particular, that when group means are the objects of measurement, variability due to persons is part of both relative and absolute error variance. Consequently, both error variances decrease not only when the number of tasks increases, but also when there is an increase in the number of pupils within each group.

Using Equations 13 and 14 a generalizability coefficient is still given by Equation 4, and using Equations 13 and 15 a dependability coefficient is still given by Equation 6.

Gao, Brennan, and Shavelson (1994) report the following estimated variance components for the CAP science performance assessments described in the first section of this chapter:

$$\hat{\sigma}_g^2 = .09, \hat{\sigma}_{p:g}^2 = .21, \hat{\sigma}_t^2 = .06, \hat{\sigma}_{gt}^2 = .07 \text{ and } \hat{\sigma}_{pt:g,e}^2 = .52. \quad (16)$$

The G study that led to these estimates was based on a random sample of 15 students from each of 40 California public schools.

The pupil-by-task variance component is the largest, by far, which is consistent with other research cited earlier. (In this case, however, note that pupil-by-task variance is confounded with residual error.) The second largest variance component is for pupils

within schools, which contributes to error variance. Large variation among pupils within schools leads to uncertainty about school universe scores. Note, in particular, that $\hat{\sigma}_{p:g}^2 = .21$ is much larger than universe score variance among schools, $\hat{\sigma}_g^2 = .09$.

Variation attributable to tasks is relatively small suggesting that tasks differ somewhat in difficulty. Variation attributable to school-by-task interaction is of the same magnitude suggesting that task difficulty varies somewhat by school.

Figure 4 provides estimates of the absolute SEM and the generalizability coefficient for 1 to 12 tasks and 20, 40, and 60 pupils within a school. From Figure 4 it is clear that both n'_i and $n'_{p:g}$ are influential in determining $\hat{\sigma}_\Delta$ and $\hat{\rho}^2$. Also, there are trade-offs between increasing $n'_{p:g}$ and n'_i . For example, when $n'_{p:g} = 20$, doubling the number of tasks from 6 to 12 leads to about as much improvement in the generalizability coefficient as doubling the number of persons from 20 to 40 when $n'_i = 6$. Note, also, that it appears that measurement precision does not improve much by using more than about 50 persons within a school.

Percent passing intervals. Another use that can be made of the variance components for a $(p:g) \times t$ design is to estimate confidence intervals for the percent of examinees from a school who will exceed a particular score. A procedure for doing so has been described by Linn and Burton (1994, pp. 7-8).⁵ Using the Gao et al. (1994) data, Figure 5 provides the Linn-Burton approximate 68% confidence intervals for a school with a universe score of 2.25, given a passing score of 2. Under these circumstances, the true pass rate is about 71%, but observed pass rates may exhibit great variability y , even for relatively large values of $n'_{p:g}$ and n'_i .

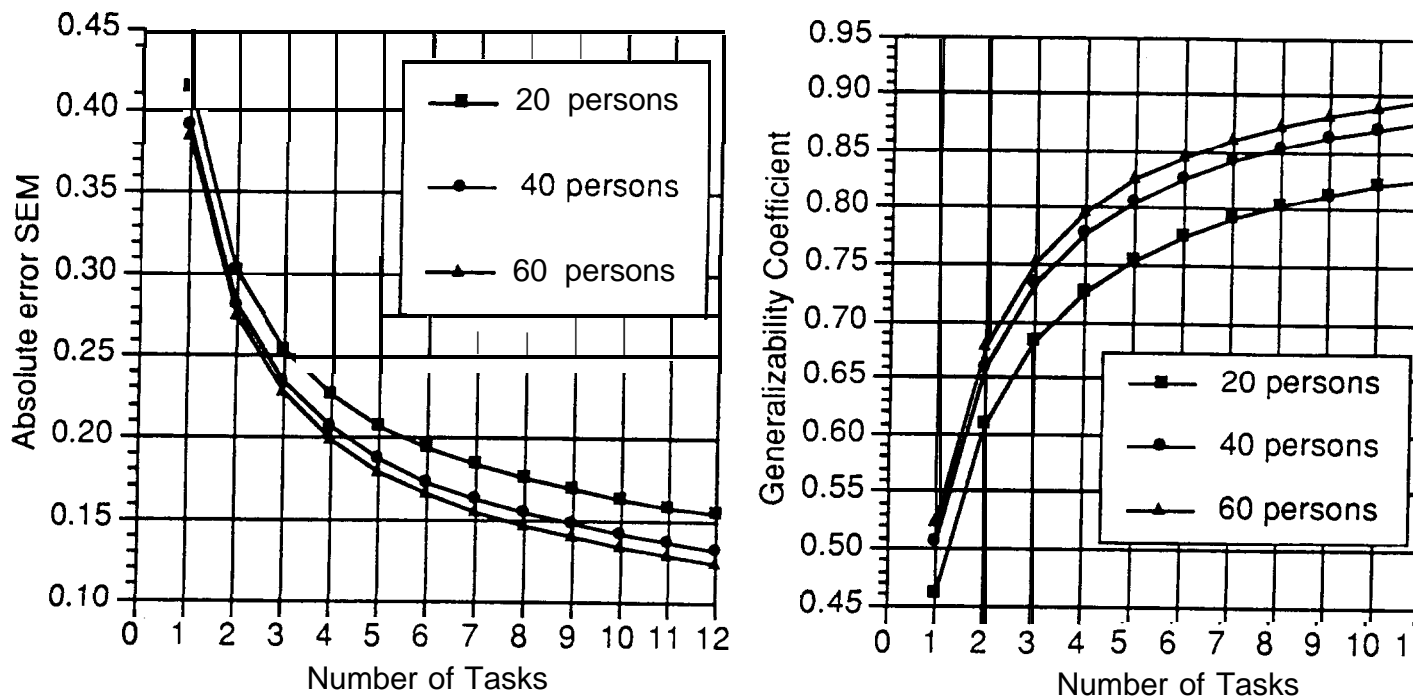


Figure 4. Results for CAP data from Gao, Brennan, and Shavelson (1994).

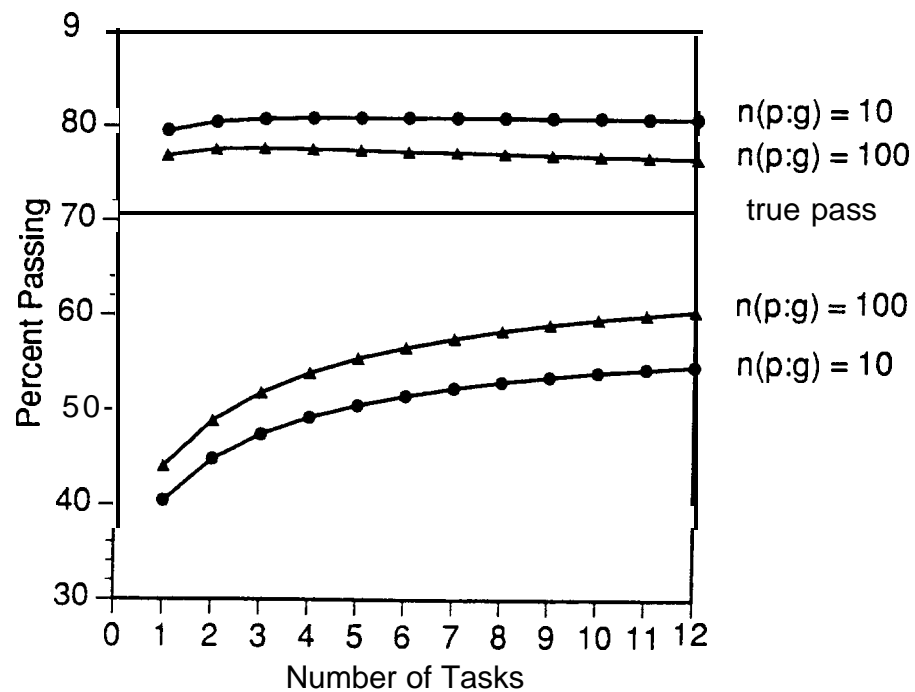


Figure 5. 68% confidence intervals for percent of examinees who will pass in a school with a universe score of 2.25 assuming a passing score of 2. Results are based on application of the Linn-Burton (1994) procedure using variance components in Gao, Brennan, and Shavelson (1994).

For example, when $n'_i = 5$, a 68% confidence interval for percent passing covers a 23 percentage point range from 55% to 78%. This means **that**, over replications (**e.g.**, different times of testing) 68% of the **time**, the observed pass rate (**given** a passing score of 2) for a school with a universe score of 2.25 would likely range from 55% to 78%. Not only is this a very broad **range**, but also the interval is not symmetric about the true passing rate of 71%. For these **data**, it appears that very large numbers of tasks and persons would be required to obtain narrow **intervals**.

Violations of "conventional wisdom." Not infrequently, investigators assume that reliability for **groups** is necessarily greater than reliability for **persons**, and/or error variance for groups is necessarily less than error variance for **persons**. Using generalizability theory, Brennan (1995) has shown that this "conventional wisdom" is not necessarily **true**. In **particular**, violations of this conventional wisdom with respect to **generalizability** coefficients are quite **likely**, especially when variability due to persons within groups is relatively **large**, and/or the number of persons within groups is relatively **small**.

Brennan (1995) shows **that**, when pupils (**over groups**) are the objects of measurement, a generalizability coefficient is

$$\rho^2 = \frac{\sigma_g^2 + \sigma_{p:g}^2}{\sigma_g^2 + \sigma_{p:g}^2 + \frac{\sigma_{gt}^2}{n'_i} + \frac{\sigma_{pt:g,e}^2}{n'_i}}. \quad (17)$$

In this **case**, universe score variance involves not only variability due to groups but also variability due to persons within **groups**.

Consider, again, the Gao et al. (1994) estimated variance components in Equation 16. Using these estimates in Equation 17 with $n'_i = 5$ gives $\hat{\rho}^2 = .72$. Using Equations 13, 14, and 4, it can be shown that when schools are the objects of measurement $\hat{\rho}^2 < .72$ if $n'_{p:g} < 15$. In other **words**, when $n'_i = 5$ and $n'_{p:g} < 15$, school reliability is less than pupil **reliability**, which violates the conventional **wisdom**.

Suppose $n'_i = 5$ and $n'_{p:g} = 10$. In this case, as noted above, school reliability is less than pupil reliability. However, school error variance (both relative and absolute) is less than pupil error variance, in accordance with the conventional wisdom. This paradox is easily resolved by noting that when scores are aggregated both universe score variance and error variance are likely to **decrease**, but not necessarily at the same **rate**. Hence, for at least some purposes it can be misleading to consider either **generalizability** coefficients or error variances in **isolation**. This is an important consideration given the current trend towards reporting aggregated scores on performance **assessments**.

Matrix-Sampling Designs

Historically, matrix sampling designs have been used primarily to estimate moments (especially the mean) for a distribution of **examinee scores**. More recently, such designs have been used in **NAEP** to estimate entire distributions of **examinee scores**.

Gao, Shavelson, and Baxter (in press) and **Gao et al. (1994)** have described how **matrix** sampling designs can also be used in performance testing to estimate error variances and coefficients when **schools (or other groups of examinees)** are the objects of **measurement**. In this approach, each sample of pupils within a school is randomly split into k sub-samples with each sub-sample taking a different set of **tasks**. The principal advantage of this design is that a pupil needs to take only a small number of **tasks**, but data are collected for a **large** number of **tasks**. This can lead to substantial reductions in test-taking **time** without negatively affecting measurement **precision**, provided relatively large numbers of pupils are **available**. For **example, Gao et al. (in press)** found that a **matrix** sampling design with one hour of testing time per pupil gave a **level** of **generalizability** equal to that of a $(p:g) \times t$ complete design with **2.5** hours of testing **time**.

The disadvantages of this design are that large numbers of pupils are **required**, and many tasks need to be **developed**. In many **circumstances, however**, these disadvantages may be an acceptable price to pay for the substantial reduction in per-pupil testing time and the increased content coverage inherent in having many tasks **administered**.

Concluding Comment

As this chapter illustrates, from at least some **perspectives**, scores on performance assessments are less **generalizable** than scores on **more** traditional **tests**. At the **same time, however**, the apparent realism (what some call the "authentic" nature) of **performance** assessments is intensely appealing to many **people**. This appeal has led some researchers and practitioners to down-play the importance of **reliability/generalizability** considerations in the evaluation of performance **assessments**. That is unfortunate from a technical *viewpoint and often unnecessary from a practical perspective*, as long as excessive claims are not made for performance **assessments**.

It is undeniably clear **that**, in most **cases**, the realism of performance **assessments**, as currently **conceptualized**, is purchased at the price of some limitations on **generalizability**. That does not render such assessments undesirable *per se*. It does **suggest, however**, that decision **makers** need to be cognizant of reasonable restrictions imposed by budget **limitations**, student **time**, and rater availability -- restrictions that directly or indirectly limit **generalizability**.

How might the dependability of scores on performance assessments be **increased**? When scores are reported for groups **only**, multiple matrix sampling procedures may **help**. **Also**, for either group-level or individual-level **scores**, it **may be** advisable to supplement **performance** assessments with more traditional modes of **testing**. Another possibility may be create "**small**" performance assessments that require less administration **time**, thereby permitting a student to respond to a larger number of **assessments**. When only a few **time-intensive, "large"** performance assessments are **used**, it is probable that in most circumstances student-level scores will not be very **dependable**.

Footnotes

¹Strictly speaking, the magnitude of σ_{pt}^2 is influenced by the occasion (*o*) on which the data were collected. That is σ_{pt}^2 reflects not only *pt* interactions but also *pto* interactions. A similar type of "occasion confounding" influences the other variance components containing *p*, at least theoretically. By the same line of reasoning, variance components containing *r* may be influenced by the occasion on which ratings were obtained.

²Assuming normally distributed observed scores, a 95% confidence interval covers a width of approximately four SEM's. Consequently, for the difference between the upper and lower limits to be two points, $\hat{\sigma}_{\Delta} = 2 / 4 = .50$. For the difference to be one point, $\hat{\sigma}_{\Delta} = 1 / 4 = .25$. These values can be used in Figure 1 to obtain the required numbers of tasks and raters.

³Throughout this section it is assumed that the same two raters are used for all persons. If different raters were used for each person the reported equations would be different, but the basic conclusions would be unaffected. Also, the discussion could be couched in terms of any number of raters. Two raters are assumed here because that is the most common circumstance.

⁴Dunbar et al.(1991) employed a procedure discussed by Gulliksen(1950, pp. 212-214) to estimate reliability due to scores -- what Gulliksen called "content reliability. " Gulliksen (1950) described his coefficient as the "reliability of an essay test corrected for attenuation due to the inaccuracy of reading (p. 214). " It can be shown that Gulliksen's content reliability is $\sigma_p^2 / [\sigma_p^2 + \sigma_{pt}^2]$, which is called score reliability by Dunbar et al. (1991).

⁵The Linn and Burton (1994) procedure makes heavy use of normality assumptions. For this and other technical reasons, the results should be interpreted cautiously. Still, such results are likely to aid decision makers in qualifying any statements about percents of passing examinees.

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Comparability

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Large-scale assessment programs are moving away from **objective**, multiple-choice tests to include more "**authentic**", "**direct**", or "**performance-based**" assessments. Using these newly rediscovered assessment **formats**, students demonstrate what they know or can do by engaging in tasks that should be of interest to **them**. Scores might even be based on records of actual classroom **performance**, in the form of student **portfolios**, **projects**, or **exhibitions**. This shift in testing method and rationale raises significant technical and normative **issues**, several of which center on the problems of assuring that measurements taken at different **times**, in different **places**, or using different performance exercises can be validly **compared**. In this **chapter**, we attempt to sort out some threats to the comparability of **measurements**, with special attention to issues that arise with **performance assessment**. Research and anecdotal evidence are presented when available to inform the likely magnitude of potential problems or to suggest ways to **respond**, but much of this discussion is necessarily **theoretical**. Practice is evolving rapidly and to date there are more questions than **answers**.

With multiple-choice **tests**, comparability is less **problematical**. Administration conditions are highly **standardized**; test stimuli are limited to **self-contained**, written **materials**; students work in isolation from one **another**; and their mode of responding is tightly **controlled**. Scoring is an almost perfectly **objective**, mechanical **process**. Even with these **tests**, of **course**, comparability has been **questioned**. Are scores from a district's first year with a new test comparable to scores the following **year**? (see, for **example**, Linn, Graue & Sanders, 1990). Are scores for language-minority students comparable in meaning to scores from native speakers of standard English? Are scores on different forms of the test **comparable**? How about scores obtained under high-stakes versus low-stakes testing conditions (Madaus, 1988)? How comparable are scores with the same name from tests by different **publishers**?

With performance **testing**, similar questions and some new ones **arise**; and finding satisfactory answers poses greater technical challenges than with more objective **tests**. This is mainly because most **performance** tests are less standardized than multiple-choice tests and because they almost always contain substantially fewer **items**. Standardization is weaker because administration and scoring are more complicated and more difficult to **control**. Many aspects of testing and scoring may be **left** to the on-the-spot judgment of **administrators**, **scorers**, or even the **examinees themselves**. **Indeed**, incorporation of student choice or provision for **local** adaptation of performance assessments may be touted as a **virtue**.

The number of independent items on performance tests is usually smaller because each item is more complex and takes longer to **complete**. **Also**, the separate questions students respond to within a given performance task are **often interdependent**. **Again**, this is seen as an advantage of these tests relative to multiple-choice tests that cover a large number of **independent**, **decontextualized** facts or **ideas**. **But**, as Yen (1993) has noted, this interdependence can result in

an exaggerated impression of the reliability of the measurement. Furthermore, the use of more items makes scores more comparable because the effects of students' individual reactions to specific items tend to average out when more items are used. One student may have special knowledge that makes this item easy, or another may have a particular misconception that makes that item difficult. These sorts of random influences matter less when averaged over more items. Similarly, a larger number of items increases comparability across test forms, because the particular features of each item matter less when more items are put together. With performance testing, therefore, because scores are likely to be based on fewer independent items, the idiosyncrasies of items and examinees have greater influence on the overall score.

The first major section of this chapter addresses the comparability of scores obtained using a single performance task. This is followed by a section on comparability across performance tasks, focusing especially on tasks intended to be interchangeable or to measure the same areas of knowledge and skill. The third major section takes up comparability at the level of tests including more than one performance exercise. In each section, different threats to comparability are discussed.

A question that must surely arise is what degree of comparability is necessary for operational use of performance assessments. Of course, no simple answer is possible. Different aspects of comparability will be more or less relevant in a given situation. As with any psychometric desiderata, the stringency of comparability requirements will depend on the kind of decision being made (e.g., "absolute" decisions about status with respect to a cutting score versus "relative" decisions about the rank ordering of students or schools); the importance of the consequences attached to those decisions; the level of aggregation at which scores will be reported and used (individuals versus aggregates like classrooms, schools, or states); the relative costs of mistakenly passing versus mistakenly failing an individual; the quality of other relevant, available information and how it is combined with performance test information; and the ease with which faulty decisions can be detected and revised. Still other factors include the cost of additional testing; the size, importance, and duration of the testing program; and the time and money available for further research and development. The psychometric quality of similar performance assessments elsewhere will also inform expectations. Finally, it is difficult to escape the impression that the level of reliability typical of multiple-choice objective tests represents some kind of standard to which tests using other formats may be held.

Comparability of Scores Obtained Using a Single Performance Task

In this section we address the comparability of scores obtained using just one performance exercise with different students or with the same student on more than one occasion. The section begins with comparability of administration conditions, then turns to comparability in scoring, and finally takes up comparability across student populations, examining some student characteristics that may influence the meaning of scores.

Administration Conditions. Imagine a classroom where students are taking a multiple-choice test. Each works alone, silently attending to his or her own paper. If space permits, students may be seated at every other desk. All have received identical instructions, read from a script provided for the test administrator. They work from identical sets of printed questions, recording their responses on identical answer sheets. Rules about what student questions the teacher may answer (and how they are to be answered), whether calculators may be used, and similar matters

are clearly **specified**. The test is accurately **timed**. All these administration conditions can be replicated any time the test is **administered**.

It is more difficult to predict what the same classroom might look like during a performance **assessment**. With some performance **tasks**, the testing session **might** look much the **same**. For **others**, the scene would be entirely **different**. Students might be working in **groups**; might be using **nontext** equipment or **manipulables**; might be free to consult whatever reference materials happened to be available in the **classroom**; might be **free** to ask the teacher questions the task designers never **anticipated**. The logistics of equipment setup and cleanup might compromise the accuracy of **timing**. The number of students in the **class**, size of the **room**, and configuration of desks or other facilities (**e.g., sink, electrical outlets**) might all **affect performance** and therefore **compromise** comparability over **classrooms**. Even if scripts are **provided**, the demands made on the test administrator to maintain order and provide logistical support **may be** considerably greater than with written **examinations**. **Consequently**, comparability across test administrators may be **diminished**.

The argument has been made that if **performance** assessments are to drive classroom **instruction**, and if group activity is a desirable feature of classroom **instruction**, then performance assessments ought to involve group **activity**. **Moreover**, the **argument goes**, ability to work in groups is itself a valued learning outcome and should be **assessed**. Regardless of the **rationale**, group activity complicates the comparability of performance **assessments**. At the individual student **level**, it **may be difficult** to disentangle the contributions of each student to a **common product**. Even if each student turns in his/her own **work**, the scores of those placed with more or less able peers may not be **comparable**. Comparability is also compromised at the aggregate **level**. A teacher's decision to distribute the most able students across all the groups **formed** or to place them in a single group **will** affect average group-level **performance**. Most **important**, the **competencies** of individuals cannot be inferred from group-level performances without making strong and probably untenable assumptions about the nature of group processes (**Webb, 1995**).

With portfolio-based **assessments**, **administration** problems **multiply**. The portfolio usually consists of some required and some optional entries representing the student's best **work**, culled from up to a year or more of classroom **instruction**. In this **context**, rules about appropriate versus inappropriate collaboration or coaching are hard to **specify** and harder to **enforce**. A major **determinant** of the quality of portfolios from a given classroom is likely to be the amount of time and effort the teacher devotes to portfolio-relevant **assignments**. In **addition**, the conditions under which students create their portfolios may vary substantially from one classroom to **another**. Research papers written by students with access to well-stocked school libraries versus an incomplete set of encyclopedias are clearly not comparable unless the conditions under which they were created can somehow be taken into **account**—a problem for which there is as yet no **solution**.

The context in which a given **performance** task is administered also encompasses the perceived consequences of **success or failure**. In some state testing **programs**, for **example**, improvement or decline relative to the previous year's test scores is rewarded or **penalized**. A **similar** issue may arise for the National Assessment of Educational Progress (**NAEP**) if it should ever change from a low-stakes national testing program to an aggregate of potentially high-stakes state-level testing **programs**.

A final requirement for comparability across administrations relates to test **security**. In most large-scale assessment **programs**, comparability over time is of paramount **importance**. Trend lines are always more **informative** than isolated results for a single year. Most designs for maintaining comparability over time call for keeping at least some items secure so that they can be **readministered** under comparable (**secure**) **conditions**. With performance **exercises**, maintaining security may be considerably more difficult than with the **simpler, briefer**, more numerous items used in the **past**. Just by virtue of their **novelty**, **performance** exercises may be more **memorable**. Beyond **that**, the use of a smaller number of such **tasks**, requirements for special **equipment**, and even the virtue of greater student interest and involvement may all work against test **security**, encouraging students to remember what they have done and talk about it with their **peers**. For low-stakes testing applications in which the focus is on overall trends as opposed to measurement of **individuals**, one solution **might** be to administer an exercise to **different, randomly equivalent**, groups of **examinees** in successive **years**.

A counter argument has been advanced (e.g., Frederiksen & Collins, 1989) that performance assessments ought to be **public**, that students and teachers should know exactly what standards of excellence **are**, and what is **expected**. In **performance** assessments like diving or a skater's compulsory **figures**, the argument is **compelling**. But in **schools**, performing an assessment task is usually interpreted less as an end in itself than as an indicator of some broader capability with respect to a construct domain the task is intended to **represent**. When the intended inference is to some broader **domain**, there is a serious risk that teaching directly to a specific task will change the skills it requires and undermine its validity (Madaus, 1988). **Nonetheless**, at least in **theory**, a solution to the problem of **noncomparability** between secure and compromised administrations might be to publicize a large number of **performance exercises**, from which those on the test are to be **selected**, well in advance of the **first administration**.

Comparability in Scoring. In large-scale testing **programs**, scoring usually follows a model that has evolved over the past decade or more out of experience with **performance** assessments of **students' writing**. A scoring rubric is developed and anchor papers are chosen to **exemplify** each performance **level**, or sometimes the boundaries between successive **levels**. Raters are trained in the use of the rubric and must attain a criterion level of accuracy in scoring a set of benchmark **papers**. During operational **scoring**, raters are organized into small groups ("**tables**"), each with a more experienced "**table leader**" who can answer specific questions as they **arise**. Some previously scored papers are seeded throughout those being scored for the first time so that each **rater's** accuracy can be monitored **continually**. This helps to keep **raters'** standards from drifting over the course of a scoring **session**. If an individual rater appears to be performing below **standard**, the table leader may "**read behind**" that **rater**, monitoring his or her **performance** more closely until the problem is **corrected**. By these **methods**, with some care and **effort**, relatively high levels of scorer consistency can usually be attained (**see, for example, Dunbar, Koretz, & Hoover, 1991; Shavelson, Baxter, & Gao, 1993**). Even with rigorous training and frequent calibration **checks, however**, scorers still contribute a source of uncertainty and thereby increase the overall measurement **error**. **Moreover**, open-ended responses are more complex and time consuming to score than selected **responses; interrater** reliability must be examined for each new **assessment**.

Performance assessment scoring may be viewed as a social process in which a group of individuals negotiates meanings and comes to consensus about the interpretation of scoring **rubrics**. Rules for handling unusual responses may be formulated on-the-fly and **communicated**

verbally to members of the **group**. If the entire scoring process is replicated **later**, the emergent consensus may not be quite the **same**. This especially threatens the comparability of scores from year to **year**. If it is possible to seed some papers from the previous year along with new papers to be **scored**, any **drift** in standards **may be estimated**. But if writing prompts or other aspects of assessments have changed from year to **year**, it may not be possible to have raters make equivalent judgments of responses to different **tasks**, nor to conceal the year each response represents.

Photocopying of **students' papers**, sometimes entailed by the logistics of **rescoring**, may also affect **comparability**. A study of score reliability for student **performance** data from the Kentucky Instructional Results **Information System (KIRIS)** found a significant effect of scoring photocopies versus originals at one but not another grade level (**Richard Hill**, personal communication, January 14, 1994).

Finally, perhaps making a virtue of **necessity**, most state-level performance assessment systems rely on classroom teachers for much of the work of scoring and in some **designs**, teachers may score their own **students' work**. It is reasonable to assume that in some **cases**, **teachers' scorings** of their own **students' work** may not be comparable to scores assigned by other **raters**, especially if teachers are aware of the identity of the **students**.¹

*Comparability Across Student **Populations**.* Every test is designed with some target population in **mind**. Decisions about **content, format, layout, timing**, and instructions are **all** conditioned by the **age, language**, and culture of the intended **examinees**. This is in part because the curriculum's intended learning outcomes differ according to **students' ages** and other **characteristics**, but also because any assessment task requires skills beyond those it is intended to **measure**. Successful performance typically depends on **examinees' understanding** that they should show their best **work**; their willingness to do **so**; their ability to understand the task **requirements**; and their mastery of the communication skills necessary to produce storable **responses**. These and other attitudes or capabilities not explicitly part of what is to be measured but nonetheless necessary for **successful test performance** **may be** referred to as the test's ancillary or enabling **skill requirements**. If some **examinees** are deficient in a **test's ancillary abilities**, then it is biased against **them**. They **will** not score as well as other **examinees** equally proficient with respect to the knowledge or skills the test was designed to **assess**. Variation in scores of otherwise equally capable **examinees** due to differences in their ancillary abilities gives rise to *construct-irrelevant variance* in test **scores**.

The ancillary skill requirements of performance assessments are likely to exceed those of more conventional **tests, although**, as with any **test**, they may be minimized through careful test **design**. The materials and instructions provided are more complex and varied and the required modes of responding are more **demanding**. Perhaps the most obvious threat to the validity of most **performance** assessments is their dependence on reading and **writing**. Scores for native speakers of standard English may not be comparable to scores of students for whom standard English is a second language or **dialect**. Consider a performance assessment in **mathematics, say**, that calls for students to explain their approach to a **problem**. Even if **communicating** about mathematics is explicitly included in the knowledge and skills to be **assessed**, scores are likely to be interpreted as reflecting primarily mathematics **proficiency**. **Clearly**, when the **examinees** are language-minority **students**, that interpretation must be made **with** considerable **caution**. At the

very least, the relative contributions of different abilities to the construct measured will vary from one language group to another.

In addition to language, comparability across student groups may be limited by differences in motivation (or understanding of and compliance with the demand characteristics of the testing situation). The rhetoric of performance assessment sometime's suggests that students will almost inevitably find these new forms of testing activities stimulating and engaging. There is an often-repeated story of a fifth grader in California asking, after a performance test in science, "Can we take the test again tomorrow?" But children do not always take to the activities adults think they should, and other anecdotes suggest that students' favorite assessments may be those requiring the least effort, especially the least writing. Low motivation may distort the scores of all students and may be especially problematical for students who must struggle with writing.

Developers of performance assessments must also remain alert to the impact of a range of disabilities that might be irrelevant in more traditional testing situations. Difficulties with fine motor coordination, physical handicaps, or color blindness may threaten the comparability of scores for individual students.

Finally, a major determinant of performance test scores may be students' previous exposure to similar tasks, given either as performance assessments or in the course of regular classroom instruction. (Students are less likely to be familiar with novel testing formats.) If performance assessments become routine in the business of teaching and testing, then differences in exposure to novel task formats may no longer threaten score comparability. But at least for the next several years, students' uneven experience with performance testing may be a consideration. Initial dramatic year-to-year improvements in scores may have more to do with students' acquisition of ancillary skills than of target skills.

Comparability Across Performance Tasks

The previous section described threats to comparability that may arise even when the assessment task itself is held constant. In this section, we add a layer of complexity, focusing on the comparability of scores from different performance tasks. Following this discussion, the third major section of the chapter examines comparability of scores from entire tests. It will be seen there that aggregating across a number of performance exercises can mitigate noncomparabilities at the level of individual tasks, although careful test design, empirical studies, and usually, statistical adjustments will still be required to assure adequate comparability in large-scale testing programs.

The comparability of entire tests depends strongly on the comparability of single items, but item-level comparability is also of interest in its own right. Consider the case of a state-level testing program that administers different sets of items to different students in order to improve school-level achievement estimates, but which also produces individual-level scores. Unless students' scores are based solely on items administered to all of them in common, some degree of comparability must be assumed across the items given to different students. (There is a dilemma here. The more comparable the matrix-sampled items are, the less matrix sampling improves content coverage.) Item-level comparability is also implied when benchmark proficiency levels are illustrated with items drawn from an exercise pool. The interpretability of proficiency scales is only enhanced to the extent that the illustrative exercises are truly representative of all those at the corresponding proficiency level.

Measurement intents, ancillary requirements, and error variance. In analyzing the comparability of single performance tasks, it is helpful to think of scores as having three distinct components. First is the *intent* of the measurement—the construct-relevant knowledge, skills, or dispositions the task is designed to measure². Second is the set of ancillary requirements—additional construct-irrelevant knowledge, skills, or dispositions required for task success, including but not limited to familiarity with task instructions, motivation to show one's best work, and test: wiseness. Language proficiency and other communication skills may fall within a measurement's intent or within its ancillary requirements, as discussed in the previous section. Finally, scores inevitably include a component of *error*—A complex mixture of both random and idiosyncratic influences on scores. The comparability of performance tasks is a function of the similarities and differences among their intents, their ancillary requirements, and their error structures. Thus, it is useful to distinguish these components even though it may not be possible to separate them or to determine in which of these three respects two assessment: tasks actually differ.

This section next turns to comparability when all three of these components are the same and then when error structures, ancillary abilities, or both diverge. The comparability of performance tasks with different intents is then briefly considered. The section concludes with a brief treatment of additional issues raised by assessment designs permitting student or teacher choice.

Comparability among tasks with the same intent, the same ancillary requirements, and the same error structure. If tasks require the same ancillary and intended abilities and if the scores they yield are equally accurate for students at any given level of those abilities, then the scores they yield should be comparable. Such tasks satisfy the requirements for equating, the strongest form of linkage identified by Mislevy (1992a) and Linn (1993). In principle, given any two such tasks, say X and Y, it should be possible to find a single equating function that could be used to transform scores on task X to equivalent Y-scores or conversely. Note that even in this ideal case, it would not necessarily be appropriate to use the tasks interchangeably without applying some statistical adjustment. One task's scoring rubric might be more stringent than the other's, for example, in which case the raw scores from the two tasks would not be on the same scale. Note also that statistical equating methods might be difficult to apply if scores on the two tasks took on only a few discrete values (e.g. 1 - 6).

The most important question about this ideal case of common intents, ancillary requirements, and error structures is the degree to which it can be approximated in practice. An ongoing NSF-sponsored project being carried out by the RAND Corporation in collaboration with several other institutions is examining the comparability of science performance assessments designed to have varying degrees of parallelism. In one comparison, two tenth-grade science performance assessments were constructed, called *Rate of Cooling (ROC)* and *Radiation (RAD)*. In ROC, students compare three fabrics to see which is the best insulator; and in RAD they compare three paint colors to see which absorbs the least radiant heat energy. Each experiment included three trials, one for each material, in which temperature was measured at specified time intervals. Temperature changes were recorded and graphed, and parallel or identical questions were posed about temperature change, heat energy, and the concept of rate of change. Despite the formal similarity of the two tasks, observations during pilot studies suggested significant differences in the accuracy of students' measurement, the time required for the two tasks, the difficulty of manipulating the equipment, and the kinds of errors students made. This study

highlights the importance of attention to ancillary requirements of tasks as well as to the abilities they are intended to measure.

Shavelson, Gao, and Baxter (in press) report another study that shows how different levels of task comparability affect **generalizability**. The study underscores the importance of clearly specifying the domain of generalization. Three experimental tasks involving the choice behavior of sow bugs were administered to a sample of **fifth** and sixth grade students. Task 1 required students to design an experiment to determine if the bugs would choose a light or dark environment. Task 2 called for an experiment to determine if they would choose a wet or dry environment. Task 3 asked students to determine what combination of conditions from the first two experiments (wet and **dark**, wet and **light**, dry and **dark**, or dry and **light**) the bugs would choose.

The **generalizability** coefficient for a single experiment was .51 when the data from all three tasks were analyzed together. When the third task was excluded from the analysis, however, the **generalizability** coefficient increased to .62, showing that the first two tasks were more nearly equivalent to each other than either was to the third **experiment**. The third experiment involved the crossing of two factors rather than the consideration of a single **factor**, and as Shavelson, Gao, and Baxter (in press) noted, such designs were not part of the elementary science curriculum. Here, the first two tasks more nearly approximated the ideal of common **intent**s, ancillary **requirements**, and error structures than did the set of all three **tasks**.

Few **generalizability** studies have clearly distinguished between tasks intended to be interchangeable and those intended to measure different aspects of the subject-matter **domain**. Results such as those presented by Shavelson, Gao, and Baxter illustrate the importance of this distinction for the design of performance-based assessments and for the evaluation of their comparability. They also suggest that greater use might be made of **multivariate generalizability** models in analyzing performance test data.

Comparability among tasks with the same intent, the same ancillary requirements, but different error structures. If two tasks measure the **same** abilities but with different degrees of precision, they satisfy the requirements for **calibration**, Mislevy's (1992a) and Linn's (1993) second strongest form of **linking**. Large-scale testing programs rarely examine differences in precision at the level of single **tasks**, but it is clear from a consideration of the calibration model that such differences could be **important**. Consider, for example, a writing assessment in which responses to two different writing **prompts**, X and Y, are scored using prompt-specific rubrics derived from the same "generic" scoring **rubric**. Suppose the correlation between students' scores on the two tasks is .64. It would be typical to treat the two tasks as parallel (in particular, equally reliable) and to assume that each correlates .80 with the same underlying true **score**. But suppose prompt X is a little more ambiguous than Y, or its rubric is not as **clear**, or that anchor papers for X are not as well **chosen**. The correlation of .64 between the two tasks could equally well imply that Tasks X and Y correlate .750 and .853 with the true **score**, respectively. Following the logic of Linn's (1993) discussion of **calibration**, if scoring rubrics are constructed so that comparable proportions of students are classified at each score **level**, the most able students would then have a better chance of receiving the highest classification using Task Y and weaker students would have a better chance using Task X. Likewise, schools with the highest-achieving students would have a relative advantage on Task Y and conversely.

Evidence about differences in error variances across tasks could be obtained in several ways. In the example **above**, differences between X and Y due to the anchor papers or rubrics used in scoring could be detected by comparing the **interrater** reliabilities for the two **tasks**. Differences due to the prompts themselves could be estimated from the correlations of Tasks X and Y with those on a third **prompt**, say **Z**. Further evidence might be obtained by comparing patterns of correlations of Task X versus Y with other **variables**, although any differences might reflect discrepancies in the **tasks'** ability requirements as well as their error **structures**.

Comparability among tasks with the same intent, different ancillary requirements, and the same error structures. If assessment tasks differ in their ancillary **requirements**, then scores on those tasks will be comparable **only** if each is given to students in **full** command of the ancillary abilities it **requires**. **Conversely**, any single task that relies on ancillary abilities that one group of students possesses and another does not will be biased in favor of the **first group**. In **particular**, tasks with different ancillary requirements may be needed for comparisons across language groups or groups of students with different instructional **histories**. An example **may be helpful**. Consider the case of cross-national assessments where tests must be administered in several **languages**. The designers of such tests try to avoid knowledge or conventions that **may be** unfamiliar in some **countries**, unless that knowledge is specifically what an item is intended to **measure**. Items are translated and back-translated to assure that the meanings of questions posed in different languages are as near to identical as **possible**. The goal of these efforts is a set of tests differing only in certain of their ancillary **requirements**. **Ideally, all** of the students taking a given test would have **full** command of the language in which it was **written**, so that apart from random **error**, score differences would reflect only what the tests were intended to **measure**. If a mathematics test written in **French, say**, were given to students in France and the United **States**, it would be biased against **U.S.** students who lacked (**ancillary**) language **abilities**. **Likewise**, a test written in English would be biased against the French **students**. Matching the ancillary requirements of the test forms to those of the student populations **greatly** improves **comparability**, although the absolute level of comparability attained is difficult to **ascertain**.

An analogous problem arises in trying to measure complex reasoning abilities among students who have studied different topics in a given content **area**. Thinking must be about **something**. If the intent is to measure complex thinking **processes**, the content knowledge to which those processes are applied may be regarded as ancillary to the intent of the **measurement**. Take as an example the idea of a food **chain**. In one **classroom**, children who have studied the ecology of a meadow might construct a food chain with **foxes, mice**, and grass **seeds**. In another **classroom**, children who have studied the ocean might construct a food chain with baleen **whales, krill**, and **phytoplankton**. Some biology students may be able to discuss the relation of structure and **function** in the crayfish and others in the **frog**. Some students **may be** able to discuss number patterns in **Diophantine** equations and others in continued **fractions**. In each **case**, even if the reasoning processes are the **same**, those processes are applied to different knowledge **structures**. An assessment tied to any particular curriculum **will** be biased against students who have studied some other **curriculum**.

One response would be to attempt to construct a "**curriculum-fair**" test, with items sampled from a lot of different **curricula**, but that is not likely to be **practical**, especially not with performance **tests**. Another approach would be to build items equally unfamiliar to all **students**, but that is likely to change the nature of what is **measured**, resulting in a test that depends more on '**G**,' or general mental **ability**. **Moreover**, the logic and rhetoric of performance testing call for

closer ties to the **curriculum**. Even if **irrelevant, decontextualized** items were fair to **everyone**, they would be rejected on other **grounds**. The most popular **solution** is to offer students and teachers some degree of choice among alternative **performance** tasks intended to measure the **same thing**, but as discussed **below**, such student or teacher choice brings its own **complications**.

As difficult as it is to approach comparability **among** tests written in different **languages**, that problem seems almost trivial compared to building comparable tests of higher-order thinking that reflect different **curricula**. At **present**, the best that can be hoped for appears to be linkage at the level of Mislevy's (1992a) or Linn's (1993) **moderation**. Either human judges will have to reach consensus that alternative assessments reflect the same abstract **performance** standards ("**social moderation**") or **adjustments** will have to be imposed based on the correlations of different assessments with some **common** anchor test ("**statistical moderation**"). As Linn and Mislevy both **discuss**, linkages established in this way **will** require frequent **reexamination**. They are unlikely to remain stable over time or to be consistent across different groups of **examinees**.

Comparability among tasks with the same intent, different ancillary requirements, and different error structures. Even tasks intended to be interchangeable are at best only approximately **parallel**. It will be some time before our cumulative experience with **performance** assessment is **sufficient** to develop "**rules of thumb**" about the **functional** exchangeability of tasks constructed in different **ways**. Work like that of Shavelson, Gao, and Baxter, discussed **earlier**, is illustrative of the kinds of studies that will be **required**.

A more extreme case of assessments with the same intent but different ancillary requirements and different reliabilities is a comparison of performance **assessments, simulations**, and written examinations by Shavelson, Baxter, and Pine (1992). In this **study**, direct observations of students performing hands-on science tasks were compared with a scoring of their **notebooks**, with their responses to computer **simulations**, with short-answer **questions**, and with multiple-choice **questions**, all designed to measure the same **content**. The authors found that in some **cases**, computer **simulations** were fairly good surrogates for hands-on **tasks**, but that paper-and-pencil tests did not appear to measure the same **capabilities**. Correlations between hands-on and written task versions were only **moderate**. Hands-on **performance** was better for students who had received more innovative science instruction than for those who had **not**, whereas paper-and-pencil test scores did not differ for the two instructional **groups**. **However**, paper-and-pencil tests correlated more highly with standardized achievement tests than did the hands-on **tests**. Taken **together**, these results suggest that hands-on versus paper-and-pencil tests measure different **things**. In **particular**, hands-on tasks measured some abilities developed especially through innovative science instruction (**presumably** part of the intent of the **measurement**), whereas paper-and-pencil test scores were more influenced by (**presumably** **ancillary**) ability requirements that they had in common **with** the standardized achievement **tests**.

Comparability among tasks with different measurement intents. If comparability is viewed as a purely technical **problem**, then the best that can be hoped for by way of linkage among tasks with different intents is projection or moderation (Linn, 1993; Mislevy, 1992a). Comparability among single performance exercises designed to measure different abilities is problematical at **best**, which may be one reason why **generalizability** studies of **performance** exercises tend to show such large person-by-task **interactions**³. As noted **earlier**, ongoing research at the RAND Corporation and elsewhere should help to inform the levels of **generalizability** that can be achieved with tighter control over performance exercise **content**. One strong implication of this

analysis is that complex assessment tasks incorporating student or teacher choice should be designed to assure that whatever specific materials or questions are **selected**, some common set of target skills is **engaged**.

An alternative perspective on comparability among tasks **with** different measurement intents would focus not on the adequacy of generalization to a common **universe**, but rather on the **value** placed on the disparate performances represented by the separate tasks **themselves**. The performance of student A on task X and of student B on task Y might be considered comparable if they represent **distinct, nonexchangeable**, even **incommensurable**, attainments that are nonetheless regarded in some sense as being of equal **worth**. This normative component is rarely made explicit (**Wiley & Haertel, in press**). Even though the rhetoric of educational reform **often** stresses **individualization**, the theory and practice of educational measurement is almost exclusively concerned with the comparison of different **students'** attainments with respect to a common construct domain or intended learning **outcome**. In **particular**, some testing programs (**e.g., portfolio-based assessments**) offer teachers or students significant latitude in determining what assessment tasks they **will** undertake or what work they will **submit** for **evaluation**. When common scoring rubrics are applied to the resulting range of different **performances**, they effectively assign scores representing the value placed on a student's **attainment**, as opposed to describing the attainment itself (**i.e., the score alone tells how good the student's work was, but not what the student did**). There is a need for further development of both technical and philosophical bases for making such judgments explicit and for reaching consensus about the meaning of such **scores**.

Student or teacher choice. In classroom **testing**, it has long been popular to include sets of questions among which students choose a specified number to **answer**, omitting the **remainder**. This practice is generally discouraged by measurement specialists because it muddies the generalization from the sample of tasks administered to the domain they **represent**. If students are free to **choose**, they **will** probably avoid questions to which they do not know the **answer**. Thus, one cannot **infer** that a student who answers 80 percent of the attempted questions correctly would have about an 80 percent chance on another question randomly chosen from the **domain**. Choice also reduces **comparability** because the scorer must judge the relative quality of **students'** answers to different questions and because students **will** differ in their ability to choose the subset of questions they can answer to best **advantage**. Thus, test **wiseness** looms larger among the constructs actually **measured**. Despite these **disadvantages**, as explained in the last **section**, comparability could in principle be improved if students or teachers could select tasks requiring ancillary skills that matched their own capabilities from among a set of tasks having the same measurement **intent**. Thus, for **example**, some students might choose a food-chain assessment based on the ecology of a **meadow**, and others a food-chain assessment based on the ecology of the **ocean**.

Performance assessments in some large-scale testing programs have given students or teachers choices among alternative **texts**. Students may be offered a choice of what to **read**, to increase their motivation or to provide a context for studying their ability to offer reasoned justifications for such **choices**. Teachers might be offered choices in the materials used with younger children so that they can make allowances for differences in their interests or levels of reading **ability**.

The NAEP began to experiment with student choice in the 1992 Reading Assessment. Students at grades 8 and 12 were given paperback anthologies containing seven stories of about 1,000 words each, selected from various sources of authentic young-adult publications, by different authors and featuring a range of cultures. During the 50-minute testing period, students were to select and read one story and then answer twelve questions about why they chose the story they did, how they liked it, and about plot, theme, character, setting, and any personal relevance the story might have had to themselves. Clearly, allowing students to choose a story reduces comparability because the stories might not be equally difficult or the questions asked might not apply equally well to all of them. Most problematical, student choice in this context complicates generalization about the entire eighth or twelfth grade population's performance on any specific story. Further research in subsequent NAEP Reading Assessments should help to gauge the magnitude of these potential effects. NAEP Readers are to be given to nationally representative samples under a condition in which a specific story is assigned as well as under the choice condition. It should thus be possible to learn how students perform on a story they have chosen versus one randomly assigned.

In the matrix sampling situation where students respond to different matrixed tasks, it is reasonable to assume that each task would be answered similarly if administered to a different, randomly equivalent sample. Hence, the data may be treated statistically as resembling a set of responses to a long test (formed from all of the alternate test forms together), with data missing at random. Where students choose which task to complete, the problem is complicated because the missing data are no longer random. Consequently, one may no longer assume that on average, each task would be answered in the same way by those examinees who did not respond to it as by those who did. As discussed by Allen, Holland, and Thayer (1993) and Wainer, Wang, and Thissen (1994), equating tests where examinees choose which problems to attempt depends on strong, usually untestable assumptions. Furthermore, recent experimental evidence presented by Wang, Wainer, and Thissen (1995) raises serious doubts about the viability of those assumptions.

The 1991-92 administration of the Standard Assessment Tasks (SATs) to 7-year-old children in England illustrates some potential problems when teachers are offered choices. A task was included at Level 2 that asked children to read aloud from one of 20 specified books (Gipps, 1993). The teacher could select any of the 20 books for any given child. As a result, some children were asked to read from a book that was new to them while others were asked to read from a book with which they were quite familiar. As noted by Gipps, "obviously for some children the task was much easier since they already knew the story and had practiced reading it" (p.11). Reading aloud from an unfamiliar book and reading aloud from a book after having practiced the reading may both be skills worthy of assessment, but tasks calling for one or the other should not be considered interchangeable. Such variability in assessment conditions seriously undermines the comparability of children's performances.

In current student portfolio systems, even "required" entries offer considerable latitude for student choice. One required entry from the 1991-92 Kentucky Instructional Results Information System (KIRIS) writing portfolio, for example, was "A personal response to a cultural event, public exhibit, sports event, media presentation, or to a book, current issue, math problem, or scientific phenomenon". Such broad specifications reflect the tension between instructional and accountability purposes for portfolios. Considerable value is placed on giving students the opportunity to choose what they consider their best work, but comparability is necessarily

compromised. Less standardized entries are more difficult to compare and opportunities for savvy students or teachers to make strategic choices are increased.

Comparability of Tests Including Performance Exercises

This last major section moves from **the** level of individual exercises to entire tests or exercise pools. Even though current assessment **programs** still rely primarily upon objective written **questions**, the discussion focuses on a hypothetical assessment comprising **performance** assessments **only**. The section begins by discussing the importance of clear content specifications for any **assessment**, and then argues that in order to assure comparability over **time**, assessments using **performance** exercises may require more detailed and more **fine-grained** content specifications than are typical of current assessment **programs**.

The importance of content specifications in large-scale assessments. A major goal of most large-scale assessment programs is to provide scores that are comparable over **time**. In **principle**, such comparability requires that the content assessed remain the **same**. Mislevy (1992b, p.200) for **example**, acknowledges that despite the appropriate and successful use of **unidimensional** statistical models to scale **NAEP data**, the **NAEP** exercise pools are not really **unidimensional**. He likens the exercise pool to the standard market basket of goods used to calculate the **Consumer Price Index**. If the **NAEP's "market basket of skills"** is **changed**, then the meaning of **NAEP** scale scores is **changed**.

That being **said**, for large-scale assessments that rely almost exclusively on **brief**, objective test **questions**, close attention to the precise content assessed does not seem to be very **important**. A large number of items are **used**, so that each specific item has only a **small** influence on aggregate score **distributions**. **Moreover**, most items are retained from one assessment cycle to the **next**, so that even if items added to the pool are somewhat different from those **removed**, the overall content mix changes only **slowly**. **Finally**, in assessments using item-response theory (**IRT**) **methods**, the difficulty of each new item is automatically accounted for as it is added to the **pool**, and items measuring abilities less highly correlated with the major dimension assessed by the entire pool **may be** assigned lower discriminations and thereby receive less **weight**.

IRT equating and gradual replacement of exercises **notwithstanding**, as ideas about **curriculum** and instruction have evolved and **curriculum** frameworks have been **revised**, there have at times been significant shifts in the kinds of content **tested**. Turning once again to **NAEP**, Baldwin (1989) and Pandey (1989) documented substantial shifts over time in the proportions of reading and mathematics exercises classified into different content **categories**. Even trends over adjacent time intervals have been based on quite different content **mixes**.

Minor shifts in content from year to year are tolerated in any assessment **program**, as some exercises are retired due to technical **flaws**, because they have been published as illustrative items or because some aspect of their content has become **dated**. The new items created to replace such retired items rarely measure the identical **skills**. **Nonetheless**, major shifts in content may require the introduction of a new scale and the start of a new trend **line**. The **shifts** in the **NAEP** content frameworks and **assessments**, for **example**, were judged to be too great for the 1990 mathematics assessments and the 1992 reading assessment to continue the same scale and **trends**. As was stated in the 1992 reading assessment **report**, "**the changes in the 1992 reading framework and assessment activities preclude any comparisons between the results of this report and those for previous NAEP reading assessments**" (Mullis, Campbell, & Farstrup, 1993, pp.

2-3). Comparisons to prior assessments are accomplished by "readministering the long-term reading assessment" (p. 3) to special trend **samples**. Thus, for a period of time after a major shift in the exercise **pool**, NAEP actually maintains two trend lines reflecting the old and the new **frameworks** and assessment activities.

Content specifications for assessments based on performance exercises. With performance exercises, careful and detailed specifications become much more **important**. Not only the measurement intents but the exercise **format**, ancillary **content**, and other characteristics may need to be **specified**, possibly down to the level of individual **exercises**. This is because relatively few performance exercises can be used and each is generally designed to measure something different from all the **others**. Detailed content specifications also take on greater importance with performance exercises because on the **whole**, they are more heterogeneous in format (**and** therefore ancillary **requirements**) than objective written test **questions**. More detailed specifications are called for because there is more to **specify**. **Timing**, **materials**, specific administration instructions and **conditions**, and other aspects of performance exercises are more variable than for multiple-choice **questions**.

It is no surprise that in **general**, the performance exercises included in large-scale assessments each measure a distinct learning **outcome**. Because of their cost in **development**, administration **time**, scoring **time**, and other **resources**, relatively few performance tasks can be included in an **assessment**. At the same **time**, most assessments must sample broad content **domains**, defined by **ambitious**, inclusive documents **like** the NAEP content **frameworks** or state curriculum **frameworks** often covering a year or more of **instruction**. Every exercise has to contribute as much as **possible**. The information gained by adding a task covering some new curriculum element is greater than that gained by adding redundant measures of content-and-process categories already **sampled**.

The facts that performance exercises are fewer in number and more distinctive imply that they must be sampled with greater care than multiple-choice items if the "**market basket**" of knowledge and skills assessed is to remain stable over **time**. Rather than treating a successive year's tasks as another random **sample** from a broad content **domain**, new exercises could be constructed such that each was closely parallel to the specific exercise it was intended to **replace**. Such a detailed sampling plan would permit the more **fine-grained** psychometric analysis described **below**.

Evacuating comparability of performance-based assessments. Statistical analyses of tests are largely divorced from content **analyses**. Although factor analysis may be used to establish that a heterogeneous collection of items is "**sufficiently unidimensional**" to justify IRT **scaling**, it is common to ignore finer content distinctions when items are **scaled**. In test **design**, item selection is **often** aimed primarily at assuring sufficient precision **throughout** the measurement **range**, with content **representativeness** being treated merely as a **constraint**, (**i.e.**, a certain number of items must come from each content **category**). Analytical methods that were appropriate for tests comprising many brief items are proving inadequate for tests comprising a smaller number of more costly **items**, more **carefully chosen**. New statistical methods may be invented (**e.g.**, Haertel & Wiley, 1993), but in the **meantime**, it **may be** possible to do a better job with current statistical **tools**.

Consider a test or exercise pool comprising ten performance **exercises**. All ten might be administered to each **examinee** to obtain an individual **score**, or the ten exercises might be

administered to distinct random **samples** of students in a school in order to obtain a school-level estimate of overall **achievement**. Under the analysis **proposed**, **detailed** specifications would be prepared for each of the ten separate **exercises**, each specification describing a **stratum** that the corresponding exercise **represented**. Then **another**, corresponding set of **ten** exercises would be **written**, closely parallel to the first **ten**. There would now be two exercises in each **stratum**. The purpose of this finer stratification of the exercise domain and the additional exercise construction would be to **unconfound** random and task-specific variance from that attributable to the distinct measurement intents of different exercises (**each** different intent corresponding to its own **stratum**). It **is**, of **course**, an unanswered empirical question what degree of error reduction would be **possible**. The answer would depend on the relative magnitudes of within-stratum and between-stratum **variation**. (See Cronbach, Linn, Brennan, & Haertel, 1995, for further discussion).

In a series of **ten** separate **generalizability studies**, each pair of parallel exercises would be administered to a sample of **examinees** to assess within-stratum **generalizability**, and the results of these analyses would be used to **estimate** the precision of a composite score treating exercise within stratum as **random**, but stratum as **fixed**. Error variances calculated in this way **might** be considerably less than the those calculated under the assumption that the ten tasks are **all** randomly **sampled** from a **single**, undifferentiated **pool**.

This **model** is not without its **difficulties**. To begin **with**, most performance exercises appear to be **unique**. It would be no easy task to define what each was measuring in a way that was sufficiently specific to realize the benefits of tighter control on assessment design and at the same time **sufficiently** general to **permit** the creation of one or more parallel **forms**. And **after** the specifications were **written**, the work of creating additional exercises and carrying out **all** those G studies would still **remain**. Perhaps most **serious**, in a typical **assessment**, the ten carefully specified strata would encompass only a small fraction of the content the assessment was intended to **represent**. The idea of fielding assessments year **after** year that **amounted** to minor variations on the **same** ten exercises would seem **unappealing**. If the assessment were **high-stakes**, instruction would soon focus on those ten **strata**, and the validity of the assessment as a measure of the **original**, broader domain would **erode**.

In **fact**, there would be no need to limit the evolution of the test through **time**. **Rather**, it would be accepted that part of the work of developing a **performance** exercise was at the same time to **specify** its "**shell**," that **is**, its stratum definition, **and**, at least for purposes of **analysis**, to construct a second exercise conforming to the **same shell**. Comparisons over **time** would be based only on common **shells**, but during each assessment cycle some new shells could be **introduced**. Data could be analyzed both to give precise estimates of status and trends with respect to a particular "**market** basket of **skills**" and also to obtain more honest **estimates** of **generalizability** to broader universes (**see Kane, 1982, for further relevant discussion**). Over **time**, the cumulation of studies focusing on domains in different content **areas**, defined in different ways and at different levels of **specificity**, would lead to steady improvements in the technology of performance **assessment**.

Summary

Comparability of **performance** exercises is a complex **topic**, encompassing multiple concerns of greater or lesser importance in any given **situation**. This chapter's discussion is somewhat

theoretical due to the paucity of empirical studies on performance tasks constructed following **specific, replicable procedures**. Although most of the specific threats to comparability of performance assessments also pertain to **brief, objective**, written test **questions**, their magnitude is greater for **performance** exercises because fewer such items are used in a typical test and because standardization is often considerably weaker for these types of **items**. In this **chapter**, we first examined the comparability of scores obtained using a single assessment **task**. Even holding the task **constant**, comparability may be limited due to uncontrolled variation in administration **conditions**, inclusion of group activities in the **assessment**, or task definitions that incorporate student or teacher choice concerning the specific tasks attempted or materials **used**. **Also**, test security may be more difficult to maintain for performance **exercises**, threatening comparability if tasks are reused with the **same** students or with different classes in the **same** school during several periods of the school **day**. Additional issues include consistency in scoring and the use of **performance** exercises with different student **populations**, especially students from different language **groups**, with handicapping **conditions**, or with different instructional **histories**.

Occasions **often** arise when scores on two or more different performance exercises must be **compared**, most notably when students or teachers are offered a degree of choice in the performance exercises **selected**. The chapter next turned to comparability across performance **exercises**, considering first the case of two exercises intended to measure the **same knowledge, skills**, or dispositions (**same** measurement **intent**); requiring the same **additional**, enabling skills (**common** ancillary **requirements**) to respond to the **performance** task **itself**; and having equal levels of accuracy (**common** error **structures**). Two such exercises **satisfy** the requirements for statistical **equating**, and present minimal problems of **comparability**.

The next case considered was that of **common** intents and ancillary requirements but different error **structures**. Tasks with this degree of similarity satisfy the requirements for statistical **calibration**, a weaker **form** of linkage than **equating**. For some **purposes**, one would want scoring rubrics for two such tasks to **classify examinees** in such a way that scores on the less reliable task had higher **variance**, and for other **purposes**, one would want the variances to be the **same**. **Information** on reliability at the level of specific tasks is rarely **reported**, but would not be especially difficult to **obtain**, and would be helpful in making decisions about category boundaries in scoring rubrics and other aspects of data **analysis**.

The case of **common intents**, different ancillary **requirements**, and common error structures arises when performance exercises are translated into different languages (**as** in cross-national **assessments**) or when higher-order thinking is compared for students studying different curricula by having each group reason about the material they have **studied**. As a practical **matter**, it is likely that the highest level of comparability attainable across curriculum-specific assessments will be that of statistical or social **moderation**.

Tasks intended to measure the **same** learning outcomes but having different ancillary requirements and error structures might include **performance** tests versus paper-and-pencil "**surrogates**." Here **again**, linkage at the level of statistical or social moderation is probably the best that can be **attained**. The same is true a **fortiori** of tasks with different **intents**.

The third major section of the chapter turned to comparability at the level of entire **tests**. **Here**, many of the concerns of the previous chapter are somewhat mitigated because the specific qualities of each item count for less in a composite **score**. It is **argued** that imprecision could be reduced and that the remaining imprecision could be evaluated more accurately if content

specifications for **performance** exercises were prepared in greater detail and tests were constructed as purposively weighted collections of tasks sampled from narrow content **domains**. Some practical **implications** of such test design strategy were **considered**. It **may be** hoped that over **time**, the cumulation of studies on methods of domain definition and performance task construction will lead to better assessment technology to improve educational **practice**.

Notes

¹ This may be viewed as a threat to comparability, but Moss, Beck, Ebbs, Matson, Muchmore, Steele, Taylor, and Herter (1992) offer another interpretation. They argue that classroom teachers are privileged **observers**, able to bring a richer interpretive context to bear in evaluating their own **students' work**: "The central interpretation will be the classroom **teacher's interpretation**, and it will be based not only on the portfolios but also on extensive knowledge of the students, their goals, and their instructional opportunities... This approach . . . acknowledges the singular value of the teacher's knowledge base in making **interpretations**, which cannot be duplicated by outside **readers**" (Moss, et al., 1992, p.19). It might be added by way of **rejoinder**, however, that privileged observers can also be biased **observers**.

² In using the term "**intent**," we do not mean to **imply** that intentions are **sufficient**. To disentangle these components and **determine** the extent to which each is reflected in actual test scores would require supportive evidence from empirical **research**.

³ Large person-by-task interactions may also be reported because variance components representing the person-by-task interaction and the person-by-task-by-occasion interaction are **confounded**. The **latter**, three-way interaction represents the instability across a single **individual's (hypothetical)** repeated trials on the same **task**. The total of these two variance components is often labeled "**person-by-task interaction**."

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Setting Performance Standards for Performance Assessments: Some Fundamental Issues, Current Practice, and Technical Dilemmas

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Since the turn of the **century**, with the pioneering work of **Binet**, **human** cognitive abilities have been broadly assessed by administering **selected-response**, pencil-and-paper tests that purport to measure various **achievements**, **aptitudes**, and knowledge that are held to be indicative of success in school and of **schooling**, prerequisite to effective performance in academic **pursuits**, essential to success in a wide variety of **careers**, and predictive of **performance** on the **job**. Selected-response tests that require the darkening of small bubbles with **number 2** pencils have grown to be a cultural universal in **schools**, when seeking **employment**, and when seeking professional **licensure** and **certification**.

Just in the past few **years**, the validity of **selected-response**, pencil-and-paper measures of school achievement and skill-based knowledge has been widely **challenged**. The use of selected-response tests of **students'** subject-matter achievement in every state in the Union has exposed the "**Lake Wobegon Effect**," wherein the average achievement of every **state**, and of most large school **systems**, was found to be above the national average (**Cannell, 1987**). Studies of the consequential validity of standardized tests of student achievement (**cf., Smith, 1989** and **Shepard, 1990**), have shown that such tests drive curricula and instruction in the nation's schools. **Lorrie Shepard (1990)**, **Lauren Resnick (1990)** and others have demonstrated that currently-used standardized achievement tests are grounded in the cognitive psychology of the 1970's, where it was **assumed (falsely)** that hierarchical learning of basic skills was prerequisite to **students'** acquisition of the higher-order analytic skills that are the real objective of formal education.

A new day has arrived in the testing of **students**. At the national **level**, through the New Standards Project (**Resnick & Resnick, 1992**) and the National Assessment of Educational Progress (**NAEP; cf., Applebee, Langer, Mullis, Latham & Gentile, 1993**), and in a number of states (**e.g., California, Vermont, Maryland; cf. Aschbacher, 1991**), test-like devices that attempt to measure **students'** skills and abilities to **perform** various tasks directly are under development or in **use**. These "**performance assessment**" measures are **changing the** landscape of student **assessment**. Rather than presenting students with printed multiple-choice questions that require

selection of a correct option from among those presented, performance assessments require students to construct responses to a wide range of **problems**. Performance assessments have been termed “authentic assessments,” since they often provide tasks that require problem-solving skills, and present challenges that are thought to model realistic applications that students will face later in life.

Introduction

In the enthusiasm that surrounds the new assessment **methodologies**, it must be realized that the demands of sound educational and psychological measurement are **ever-present**. Assessments of **students’** abilities must **satisfy** professional measurement **standards**, as exemplified in the 1985 *Standards for Educational and Psychological Testing*, regardless of the testing and measurement method **used**. If the potential of **performance** assessments of students is to be **realized**, such assessments must be demonstrated to yield measurements that (1) are sufficiently reliable to support the selection or classification of individuals or the evaluation of aggregates of students at **local, state, or national levels**, (2) validly support inferences concerning the **achievements, aptitudes, and performance capabilities** of those **assessed**, (3) **fairly**, and in an unbiased way, reflect the abilities of those assessed without regard to **gender, race, ethnic group membership, or economic circumstance**, and (4) when **performance** standards must be **set**, support the classification of **examinees** into decision-relevant categories (**e.g., worthy of admission, certification, licensure, graduation, etc.**) or their labeling as “**basic**,” “**proficient**,” “**advanced**,” etc.

Unfortunately, much of the available methodology for assessing the psychometric quality of measurement instruments was developed with selected-response tests in **mind**. In large **part**, the applicability of this methodology to performance assessment measures is either obviously limited, questionable, or **untested**. For **example**, currently-used methods for establishing performance standards on tests either demand **pencil-and-paper**, multiple-choice test items (Nedelsky, 1954) or presume the use of selected-response achievement test items (Angoff, 1971; Ebel, 1972; Jaeger, 1982).

This chapter is concerned with one component of the psychometrics of performance **assessment**: methods for setting performance standards on large-scale **performance** assessments of **students**. This topic is essentially unexplored in the educational and psychological testing literature although some practice-based experience has been accumulated in conjunction with NAEP, the teacher assessment program of the National Board for Professional Teaching **Standards**, and some statewide assessments of **students**. Even though methodology for setting performance standards on **performance** assessments is **embryonic**, we attempt in this chapter to define critical **issues**, review what is known and being **done**, and define a research agenda that can guide **future inquiry**. As the reader of this chapter will learn, the state of the art is far from a state of **grace**. Much work remains to be **done**.

Performance standard-setting is a judgmental **process**. Its results are therefore subjective by **definition**. Of **necessity**, the outcomes of any **performance** standard-setting process are **arbitrary**¹ but they need not be **capricious**². Effective standard setting requires that qualified persons make reasoned judgments in response to questions that are understandable and within their ability to **judge**, following well-structured efforts to inform them about the nature and consequences of the alternatives they **face**.

Many of the conclusions derived from two decades of research on performance-standard setting for selected-response tests can be expected to generalize to performance **assessments**. **First**, performance standards rarely occur naturally (**one** relatively rare exception is in some performance assessments in the **military**, where correct completion of a **task**; e.g., cleaning a rifle **correctly**, is directly observable and can be scored "**pass**" or "**fail**"). Far more **often**, the boundary between performances judged to warrant classification as a passing performance or eligibility for valued **rewards**, and those that do **not**, result from a process of negotiation or **arbitration**. Performances judged acceptable or worthy will most often differ in **magnitude**, rather than **kind**, from performances judged to be unacceptable or **inadequate**. It is this **conclusion**, as much as any **other**, that has led some to label all standard-setting processes as capricious (cf., Glass, 1978).

Second, performance standards are **method-dependent**. Specification of performance that warrants **graduation**, **certification**, classification as **proficient**, etc. depends to a substantial degree on the method used to elicit judgments from those who set performance standards (**for** a summary of literature on this **point**, see Jaeger, 1989).

Third, those who set performance standards cannot be assumed to be trustworthy judges of the quality of the methods they have **used**. Virtually **all** surveys of panelists who set performance standards have concluded that such panelists are "**somewhat confident**" to "**very confident**" about the quality of their resulting **standards**, regardless of method they have used and subsequent empirical evidence concerning the coherence and consistency of the performance standards they have **set**.

Finally, widely-used performance-standard-setting methods presume the existence of an underlying interval scale of performance on the test or assessment for which standards are to be **set**. This conclusion follows from the first (**above**), from the typical practice of averaging the standard-setting **recommendations** of members of standard-setting **panels**, and from the use of **parametric** statistical procedures in evaluating the **precision**, **consistency**, and coherence of **performance standards**. As **will** be noted **below**, performance assessment exercises are typically multidimensional in their measurement **characteristics**, are rarely **exchangeable**, and **are**, **therefore**, far less **likely** to **satisfy** the local independence assumptions of **unidimensional** scaling **models**. Such exercises violate the assumptions that underlie typical **performance-standard-setting methods**.

The Structure of This Chapter

The balance of this chapter consists of three major sections and a **summary**. In the next **section**, under the heading "**Some Fundamental Issues in Performance Standard-Setting**," we describe typical institutional and individual applications of **performance** standard-setting so as to delineate the contexts in which new methods of setting performance standards for performance assessments would be **used**. We **also** build the case that the performance-standard-setting methodology of the past is not applicable to a wide range of performance assessments in these **contexts**.

In the third section of the chapter we provide prominent examples of early efforts to establish performance standards for performance **assessments**. These examples include the National Assessment of Educational **Progress**, the assessment of highly accomplished teachers by the National Board for Professional Teaching **Standards**, and a variety **of** statewide assessments of **students**.

The fourth section of the chapter, under the heading "Some Technical Dilemmas," addresses such issues as the artificial polytomization of the performance continua to which standard-setting methods must be **applied**, the method-dependent nature of performance **assessments**, and a host of other technical issues that must be addressed in setting performance standards on performance **assessments**.

In our summary and conclusion we indicate important methodological and procedural issues that are yet to be addressed by the known methodology of setting performance standards for large-scale performance **assessments**, thus providing a rudimentary road map to needed **research**.

Some Fundamental Issues in Performance Standard-Setting

Among the various **reform** movements affecting our Nation's **schools**, two thrusts in particular have major implications for the tests used in student **assessments**. **First**, the education **community**, including teachers and curriculum specialists as well as politicians and educational policy makers are diligently working to develop standards-for discipline-oriented **curricula**, for **assessment**, and for the **delivery** of educational **services**. Greatest attention to date has focused on the curriculum standards **specifying** what students must know and be able to do to be considered proficient in various subject-matter **areas**. The first such standards were released by the National Council of Teachers of Mathematics in **1989**. Since that **time**, at least a half-dozen other professional organizations and **collaboratives** have made substantial progress toward **similar** documents for other academic **disciplines**. **Currently**, curriculum standards in **civics**, **geography**, **history**, **foreign languages**, **science**, **social studies**, and the arts have been **released**. **However**, few of these documents contain formal **performance standards**, although some provide a few illustrations.

Simultaneously, there has been growing concern that assessments should directly address more complex types of learning than those measured by conventional tests (**National Council on Education Standards and Testing, 1992**). In **addition**, assessments should require applications of knowledge and skills to "**real-world**" situations faced by individuals at **work**, in their own **lives**, and as citizens of a **community**. These more "**authentic**" forms of assessment require task performance in contexts that reflect the realities of everyday **life**, rather than simply providing isolated bits of information or knowledge (**Wiggins, 1993**). Taken **together**, these two **reform** thrusts signal the need for assessments that emphasize how well students can **perform** particular tasks **vis-à-vis** subject-matter **standards**. **Further**, the more that such **performance** assessments focus on the complexities of problem-solving and reasoning rather than declarative **knowledge**, **comprehension**, or routine application **skills**, the more consistent they **will** be with the goals set forth in the current curriculum **standards**.

If performance assessments are to be used in accountability-based strategies for promoting systemic educational **reform**, the issue of designing methods for setting **performance** standards on performance assessments must be **addressed**. A **number** of national and local assessment enterprises are beginning to examine this **issue**, notwithstanding the inherent complexities associated with developing curriculum **standards**, with developing performance assessments that accurately measure proficiency on those **standards**, and with establishing clear and concrete understandings of what evidence students should provide to show that they have met the **standards**.

How Performance Standards Are Used

Before tackling the **numerous** challenging technical and procedural issues implicit in this dramatic change in educational testing **approaches**, it is important to understand the various purposes of educational testing and potential uses for the **information** provided by **performance standards**. For this **discussion**, we have identified two broad classes of **applications**: those pertaining to *institutions* and those pertaining to *individuals*.

The two applications can differ considerably regarding the **amount** of detail required for an **individual**. The needs for information about institutional quality generally are more global in **nature**, because such data typically are used for making relatively long-term as compared to day-to-day **decisions**. Sampling of individuals can be used in assessing **institutions**, given that it is not **necessary** to obtain broad-based information about the performance of each **individual**. By assessing different aspects of **curriculum** standards for various representative groups of **students**, information can be aggregated across the different groups to develop a picture of educational performance for an entire institution or **system**.

In **contrast**, **information** for making decisions about individuals often needs to be comprehensive and timely because it may directly influence an important and relatively-immediate decision about selection or placement of the individual into an educational program or **institution**.

There are similarities in the ways that institutional and individual performance assessment results can be **used**. Both types of assessments can be used for formative or **summative purposes**. Feedback about success with particular aspects of curriculum can be used on an ongoing basis to make adjustments in an institutional program or pedagogical **approach**, or to develop specific remedial steps for individual **students**. Beyond such formative **uses**, assessments or tests also can provide the basis for final decisions about **institutions, programs, or individuals**. For **example**, schools or school systems may participate in **summative** evaluations to receive accreditation and students may need to pass a test to graduate from high **school**, or become licensed in a **profession**.

Institutional Applications of Performance Standard-Setting

Description for Public Information. Foremost among the uses of **performance** standard-setting is the public's right to know about the quality of our educational system and its **products**. Society as a whole has a vested interest in knowing whether today's graduates have the knowledge and understanding necessary to contribute to an **informed** citizenry and whether they can be relied upon to set enlightened policy for the **future**. Employers deserve to know whether students have the knowledge and skills required to meet daily production needs or for our nation to compete in a global **economy**. Parents want to know that their children are receiving a high-quality **education**, and students themselves should be interested in whether they are becoming well-prepared for adult **life**.

Public Accountability. Beyond the basic right to **know**, those responsible for funding and providing oversight for various aspects of the nation's education system need information to monitor **schools' performance**. Whether for the **nation**, a **school-district**, or a school **building**, student achievement results can be used to provide guidance on how resources invested in education might be **augmented** or used **differently**. Information **about students' competencies** can

also be used in formulating approaches to improving education. With the publication of the National Commission on Excellence in Education report, *A Nation at Risk*, and general discontent from employers and parents about the effectiveness of America's schools, the call for accountability information has been growing for more than a decade. In fact, this widespread concern has contributed directly to the movement toward a standards-based education system, including assessments to monitor students' progress in attaining performance standards. The role of assessments increasingly includes helping to define and guide instruction to make it more effective. The content of assessments impresses on students, school staff, and parents the importance of tested subject matter and associated expectations for learning.

Program Evaluation. One particular kind of accountability information concerns assessments about special programs designed to address the needs of particular groups of students. For example, special programs have been developed for students with limited proficiency in English, students determined to be at increased risk of academic failure due to disadvantaged socioeconomic backgrounds (e.g., ESEA, Chapter 1), or to provide special opportunities for high-achieving and talented students. Information needs about particular programs relate to decisions about sets of practices or groups of students within the education system, school district, or school.

Individual Applications of Performance Standard-Setting

Selection. One of the better known uses of educational testing in the United States is for selection for college entrance, with many high school students taking either the SAT or ACT as part of the college admission process. However, students may also take tests to be admitted to special programs or schools, such as those designed for students with special aptitudes in science or the performing arts. Although neither the ACT nor the SAT program recommends the use of fixed performance standards for college admission, the practice is, unfortunately, widespread.

Classification. Sometimes related to selection, the major purposes of classification assessments are to match students' performance levels against pre-set criteria to assign students or apportion them to different treatments, curricula, or programs. For example, a student might receive compensatory education, be placed in an advanced mathematics program, or be retained at the same grade level based, in part, on test results. Such assessments can document accurately each student's progress at the end of an extended period of instruction. These data can be used to profile students' strengths and weaknesses in particular areas, serving a diagnostic as well as a placement function.

Certification. Certification serves a public function, enabling individuals to demonstrate to others that a certain standard of performance has been met or that a certain set of skills has been mastered. That is, the individual who has been certified will be known to have completed a particular course of study and to have demonstrated a specified criterion of performance. The public then expects the individual to be able to perform particular tasks or functions. Although the Advanced Placement program involves testing for the receipt of credit for particular first-year college courses and some school districts test students to certify levels of minimal competence as part of receiving a high school diploma, certification is used less as part of the public education system in the United States than in some other countries (e.g., the French Baccalaureate, the Abitur in Germany, and the British Public Examinations). Test-based certification, however, is widely used as a professional endorsement of candidates'

employability in the United States. For **example**, licensing or certification occurs for **nurses**, **accountants**, **architects**, **lawyers**, and in many other occupations and **professions**. Also, the National Board for Professional Teaching Standards is developing procedures for advanced teacher **certification**. **Currently**, efforts are underway to develop standards for many more **occupations**, and employers have suggested certification as part of the public education **process**.

In *Learning a Living*, the **Commission on Achieving Necessary Skills (SCANS)** for effective job performance recommended that students should develop a cumulative **résumé** containing **information** about courses **taken**, projects **completed**, and assessment **results**. As students met the standards set for specific **skills**, that mastery would be noted on their **résumé**. When students had met the standards across courses and SCANS **competencies**, their **résumé** would show that they had been awarded the Certificate of Initial Mastery (CIM).

Why New Methods Are Needed for Setting Performance Standards on Performance Assessments

Because standard-setting **permeates** many aspects of **life**, including the occupational licensing and certification processes referenced **above**, the difficulties associated with applying such procedures to educational uses may not be **apparent**. For **example**, standards are very much part of athletics (**e.g.**, **performance** ratings for the **Olympics**, life-saving badges in **swimming**, certification for the **ski-patrol**, or a black-belt in the martial **arts**), and they are central to guiding aspects of daily life such as transportation (**e.g.**, the **safety** of **aircraft**) and regulation of food and drugs (**e.g.**, classifications of over-the-counter **drugs**, standards for safe fish or **meat**, and even classifications of the "hotness" of canned chili **peppers**.)

Although, as is discussed in the next **section**, experience with performance standard-setting can be gained from these efforts as well as initial attempts in educational **contexts**, many challenges **remain**. **Generally**, problems arise from attempts to make decisions based on very few tasks that often measure very diverse and discrete **accomplishments**. Consensus has not been reached on how to best aggregate **performance information** across those tasks to make an overall judgment either for individuals or **institutions**.

Making Judgments Based on Few Tasks. This issue relates to the breadth of material covered by a content standard and the number of performances required to make confident judgments about **students'** mastery of the material (**i.e.**, to **classify a performance**, based on a performance **standard**). For **example**, to **determine** whether students can use addition of whole numbers in appropriate contexts they could be given several problems to solve involving various applications of whole-number **addition**. Perhaps after successfully completing **10** or so such problems (**provided** they presented diverse **stimuli within** the definition of whole-number **addition**), some **confidence** could be achieved in generalizing to the domain and judging the **students'** proficiency in applying whole-number addition to real-world **problems**.

Unfortunately, however, performance standards eventually will need to address far larger domains, such as **geometry**, **algebra**, or even mathematics as a **whole**. To **date**, the work in setting curriculum standards has defined each domain as a vast and diverse network of **knowledge**, **skills**, and **understandings**. Also, especially as students grow **older**, tasks need to be far more **complex**, encompassing **students'** abilities to select **among** a repertoire of **approaches**, apply an effective **strategy**, and reach an appropriate solution (**such as**, actually building a machine that **works**). These two **goals**—the need for large numbers of tasks in order to assess

students reliably and the need for broad **tasks**, so as to **sample** a domain adequately—are difficult to **reconcile**, in that a broader array of performance tasks would yield far greater confidence in determining **students'** success in reaching a performance **standard**. Yet, the complexities of performance **tasks**, in terms of resources and **time**, preclude administering large numbers of them to individual **students**.

Performance tasks consume more resources to **develop**, **administer**, and **evaluate**, and require more energy and time on the part of the students participating in **assessments**. While educators may be more receptive to devoting fiscal resources as well as **students'** and **teachers'** time for participating in realistic **assessments**, there are reasonable **limits**. Performance assessments generally contain relatively **few**, complex **exercises**. Educators are often appropriately reticent to make high-stakes decisions about **students'** educational careers based on a limited number of data **points**.

Each Performance Typically Entails Degrees of Success. Also, assuming for the moment a curriculum standard and agreement about a performance task measuring it, students will undoubtedly demonstrate differing degrees of success in accomplishing the **task**. For **example**, drawing on **national**, **state**, and school-district experiences across the past several decades in moving from multiple-choice to direct writing **assessment**, **students'** essays generally are not evaluated as "**right**" or "**wrong**." Usually, a rating scale is employed whereby each essay is classified as more or less **successful** according to categories on an established **scale**. Summarized across **students**, performance can be described at various levels of the rating **scale**. Sometimes, there is a cut-point on the scale indicating that minimal competence has been **reached**, or that **performances** on the essays demonstrate higher-level writing **proficiency**, or **both**. Often, however, performance results are reported in terms of a **scale**, without making judgments about the **performance** in relation to a performance **standard**. Such **polytomous** results often are complicated to **aggregate**. First, decisions may need to be made for each task about what constitutes "**minimal**" or "**excellent**" **work**, as well as what level of response is required to meet the standard being **measured**. It is interesting that work across **states**, conducted by The New Standards Project, showed considerable agreement in ranking **students'** **work**, but far less unanimity in where cut-points fell (Linn, et al., 1991).

Historically, often because of resource **limits**, writing assessments only contained one task and student classification decisions were therefore difficult to **make**. More recently, however, research is showing that **generalizability among** performance tasks is very **tenuous**, and this underscores the importance of including a number of **tasks**. As previously **stressed**, one of the problems with **performance** assessments is including enough tasks for adequate domain **coverage**. The aim is to increase the number of tasks so as to increase the likelihood of accurately measuring performance across a domain of **interest**. This, then, leads to the problem of aggregating performance **information** across **tasks**. Because tasks measure different **accomplishments**, it may be likely that **examinees'** scores across tasks **will be diverse**. While a profile can lead to better understanding of student performance for diagnostic **purposes**, it would complicate making an absolute judgment about **students'** overall **performances**.

To be judged proficient in a **domain**, do students need to be proficient on each performance **task**? Or **perhaps**, is it sufficient to be proficient on most but not **all** of the **tasks**? Or should the overall judgment be based on an average across **tasks**? Each of these options requires a different approach to **performance standard-setting**.

Lack of Score Transferability. Finally, given the complexities of the tasks used in performance assessments and the difficulty of evaluating **students'** performances on the **tasks**, it is not surprising that developing equivalence across tasks is a **problem**. Not only **will** an individual's performances differ from task to **task**, but some tasks simply will be more difficult than **others**. One reason for having **performance** standards is to attempt to establish comparability among the **performances** of students in different **jurisdictions**. That **is, theoretically**, parents of students in one state and school district could compare their children's performances to **students'** achievements in another state or **country**, or employers and colleges could make comparisons across **students**. **Further**, the nation and school systems could monitor trends across time in attaining **curriculum standards**. **Yet**, a score on one task often means little in predicting the score on a **second**, seemingly **related**, but different **task**, either for individuals or for groups of **students**.

Developing comparable standards across performance assessments appears to be the most problematic venture of **all**. There **is, first**, the problem of designing appropriate performance tasks in terms of content **standards**, and **identifying** student work on the tasks that exemplifies success in meeting the **standards**. Implementing such assessments under consistent conditions and evaluating resulting performances reliably pose enormous operational **challenges**. **Subsequently**, difficulties arise in generalizing from performances on a few tasks to performance in an entire domain of content and in aggregating results across tasks in defensible **ways**.

Prominent Examples of Current Practice

Although the modern **metamorphosis** of performance assessment has only recently come on the **scene**, a number of such programs are operated in contexts that require the establishment of performance **standards**. Here we review a few prominent **examples** of **performance** assessment programs and describe methods that have been used to set performance **standards**. In providing these **descriptions**, we are not necessarily endorsing the practices they **represent**. As we have already **noted**, the problem of setting performance standards for complex performance assessments is **challenging**, and current solutions may not withstand careful **scrutiny**. Nonetheless, it is **useful** to see where we are before setting out on a new **journey**.

How Performance Standards Are Set for NAEP

The 1992 NAEP national and **trial** state assessments in **reading, mathematics**, and writing included a variety of exercise **formats**. Standard-setting activities were modified to accommodate various scoring procedures used in NAEP, and these depended on item **format**. Table 1 **summarizes** exercise format types by subject and **grade**. Multiple-choice (MC) items as well as most short constructed-response (SCR) items were scored **dichotomously**, that is, **'1 = right'** and **'0 = wrong'**. **However**, extended constructed-response (ECR) items were scored using rubrics appropriate to the item or **prompt**, on an ordered **scale**, generally from **'1'** to **'4'** or **'1'** to **'6'**, with higher scores representing progressively better **performance**.

In the case of dichotomously-scored exercises NAEP employed a **modified Angoff** (Angoff, 1971) procedure for establishing **cut-scores**. Since the policy **framework** for standard setting required three standards for each grade **level**, **Basic, Proficient, and Advanced**³, a broadly-representative group of judges was asked to estimate the probability of a correct response at the three levels for each of the dichotomously-scored **items**. The probability estimates were averaged over items and **judges**, and then multiplied by 100, to yield an estimated probability in the percent-correct metric for each achievement **level**.

Table 1—Number of Items, by Exercise Format in the 1992 NAEP Assessments by Grade

Grade	Math		Reading		Writing		
	MC	SCR	ECR	MC	SCR	ECR	ECR (prompts)
4	119	59	5	42	35	8	9
8	118	65	6	56	63	16	11
12	115	64	6	59	67	19	11

Sources:— NAEP 1992 Mathematics Report Card for the Nation and the States; NAEP 1992 Reading Report Card for the Nation and the States.

Legend

MC = multiple choice items (dichotomously scored)

SCR = short constructed response items (dichotomously scored)

ECR = extended constructed response items (polytomously scored)

The **Angoff** method is an approach to standard setting employed in a variety of settings and for a variety of **purposes**. According to **Berk (1986)**, the method is superior to most competing **procedures**, and is generally straightforward in the tasks required of **judges** and the interpretability of **results**. Although the **Angoff** method rarely has been used to set multiple test **standards**, it could be modified to provide such a **result**. The **Angoff** method has been employed almost universally to set standards for individual rather than aggregated test results, but could be modified to **accommodate** the latter **task**. Both of these '**modifications**' were introduced into the NAEP approach to meeting the National Assessment Governing Board's policy requirements of three test **standards**, and to accommodate the particular characteristics of the **NAEP assessment**, which yields only aggregated **scores**.

For **polytomously-scored exercises**, NAEP employed an *Exemplar Response* method of setting performance standards in 1992. Judges were provided a random selection of **examinee** responses to each **ECR question**, stratified by each score **point**. So that judges would not be unduly influenced by the score **scale**, the scores for each paper were not shared with the judges during the initial round of **ratings**. Judges were required to select from the distribution of **responses**, two papers that exemplified minimally acceptable performance for each **performance level** (i.e., *Basic*, *Proficient*, and *Advanced*). The scores associated with these papers were averaged over judges and equated to test-score **functions** for the set of **items**.

There are **few**, if **any**, methodologies described in the literature for establishing standards on tests composed of **polytomously-scored exercises**. In considering the options for NAEP, three methods were **considered**: (1) set standards on the scoring rubric (i.e., have standard-setting panelists select test standards from among the values of the **polytomous score scale**); (2) estimate distributions on the score **scale**; and (3) select exemplar **responses**. Setting standards on the scoring rubric would have limited the range of choices for **standard-setting panelists**, and would

have linked the standards to what students *can do* and less to what students *should do* (the latter were an integral part of the NAGB's policy definitions). There was also a concern that, if selecting among score-scale values, panelists would be less likely to take differences in prompt-difficulty into account. The second option, estimating distributions on the score scale, was not used because it was feared that this task might be too difficult for panelists to complete. Therefore, the most viable option seemed to be the *exemplar response method*, which preserved the desirable features of (1) being less normative in its orientation, (2) not being bounded by the values on the score scale, and (3) minimizing potential systematic bias that clearly could become an issue if panelists' selected standards from among score-scale values. In reading and mathematics, judges' estimates from the dichotomous and polytomous exercises were combined using an information weighting procedure (Luecht, 1993b) before mapping the final estimates onto the NAEP scale. In the NAEP writing assessment, the judges' proposed standards were mapped directly to the NAEP scale.

The State of Technical Debate

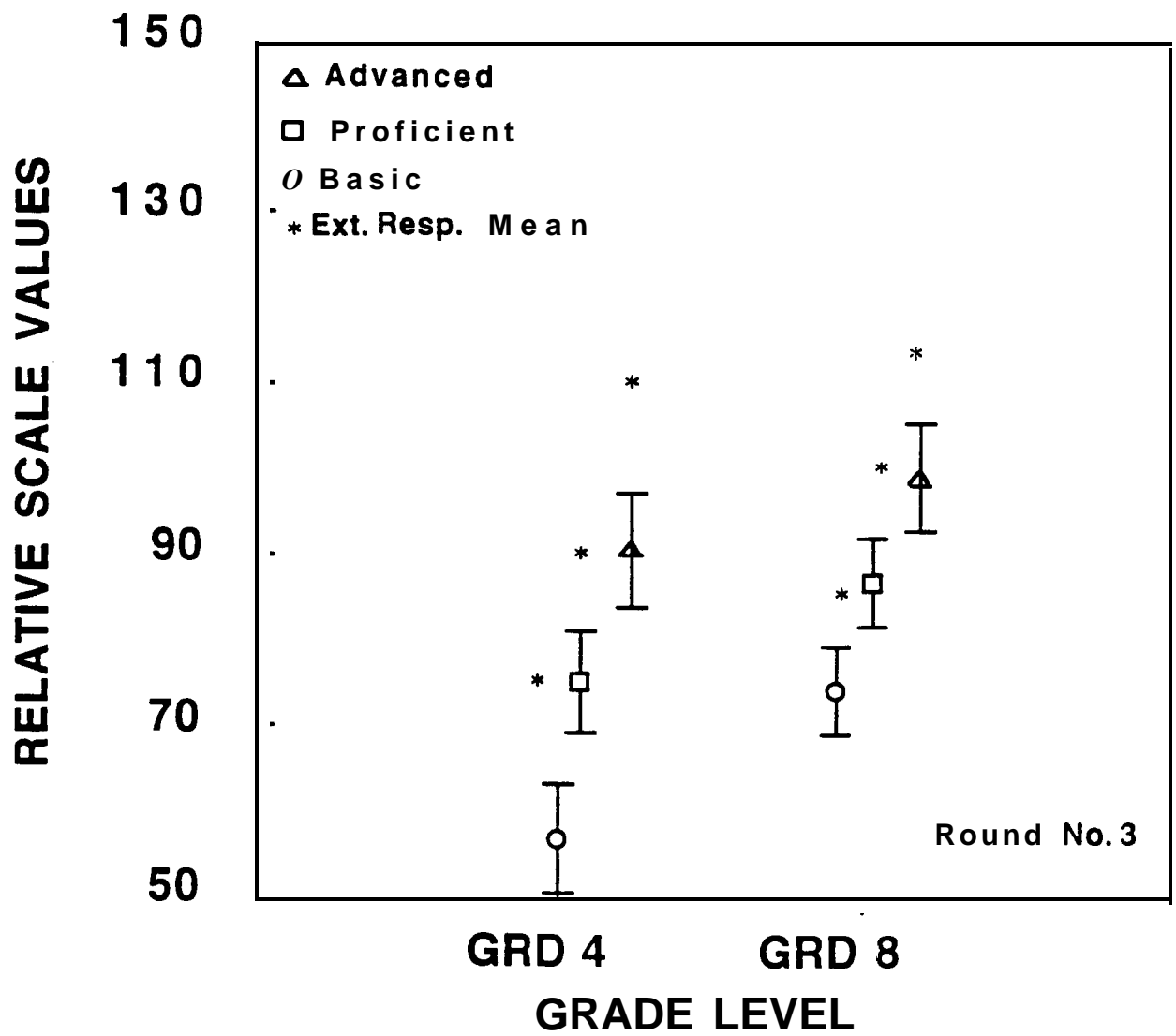
A host of technical issues came to light during the NAEP standard-setting process. The remainder of this section will attempt to describe several of these issues and offer some plausible explanations.

Discrepancy in Cut-Scores. Early in 1992, a standard-setting pilot study was conducted to explore various aspects of the process, including selection and training of panelists, programming and on-site data analysis, impact of feedback on panelists, achieving inter-rater and intra-rater consistency, and the standard-setting task itself. The 1992 reading assessment was selected for the pilot study because it employed three item formats, multiple-choice, short constructed-response, and extended constructed-response exercises. The first two were dichotomously scored, the third was scored using a rating scale from '1' to '4' (polytomous scoring). In order to examine the standard-setting approaches used with both scoring procedures, a separate standard was estimated for each. Figure 1 displays the results for the reading pilot study.

It is clear that performance standards for the polytomous exercises were set substantially higher than were standards for the dichotomous exercises. There was no empirical evidence to suggest why this was the case. And since it came to light only after the standard-setting meetings were over, it was not possible to interview panelists to ascertain why this happened. However, several plausible explanations were offered by the technical advisors to the project.

First, panelists were given a limited set of exemplar responses (16) from which to make their selections of papers that best represented performances at the thresholds of *Basic*, *Proficient*, and *Advanced*. In addition, the papers represented a rectangular distribution of responses, with roughly equal numbers at each score point. It is possible that there were no papers in the set at the threshold of each level, which had the effect of forcing the selection of higher test standards. A second plausible explanation revolves around the guessing factor. Guessing was virtually impossible on the polytomous exercises, and panelists were not instructed to take guessing into account in making their probability estimates on the dichotomous items (thus reducing their proposed standard on the dichotomous items). A third explanation offered was method effect. The task required of panelists on the polytomous exercises was very different from that required in the modified Angoff procedure applied to the dichotomous items. A fourth hypothesis was

Figure 1



that examinees simply don't try as hard to do well on the **polytomous** exercises as they might on the dichotomous **items**. Students are less motivated to respond when an exercise presents a production task. Therefore, **panelists'** estimates were not 'off target', but instead, **examinees** were not behaving the same way toward the **polytomous** exercises as the dichotomous **items**. Finally, the observed differences may be due to the possibility that the **polytomous** exercises (or at least as they are perceived by **panelists**) are assessing a different dimension of reading skills (or math or writing **skills**), in which case the multidimensional nature of the data is not being taken into account through the **unidimensional** IRT scaling used in NAEP.

Although these hypotheses remain untested due to lack of **data**, the concern is serious since NAEP continues to evolve and move toward more 'authentic' item **types**. The results suggest that more research is needed before adopting any particular methodology for setting standards on **polytomous exercises**. In 1992, this discrepancy in performance standards had minimal impact on the overall standards adopted by NAEP in **mathematics**, since the item pool contained a limited number of extended constructed-response **exercises**. The impact was slightly greater for **reading**, since about 15 percent of the item pool consisted of **ECR items**. In **writing**, however, the impact was so great that the achievement levels reported here were not officially adopted by the National Assessment Governing Board (Applebee, et al., 1994).

Alternative Methods for Polytomous Exercises. Luecht, (1993c) has done some work in the area of **polytomous-item** standard-setting **simulations**. A series of simulations was conducted to compare three methodologies for setting standards on **polytomous exercises**: *Poly-Angoff1*, *Poly-Angoff2*, and the *Exemplar Response* approach that had been used on the 1992 NAEP. Although other methods could have been **devised**, these three were selected for **further** research because they appeared to hold some potential for improving on the modified **Angoff** procedure used with dichotomous **items**. The simulations were also used to demonstrate the effect of various mapping procedures (discussed below).

The *Poly-Angoff1* method asked panelists to estimate the percent of **examinees** at the *Basic* level who would score at each score point on a **polytomous-item** score scale (i.e., from '1' to '6'). The *Poly-Angoff2* method generated **performance** standards by summing the *Poly-Angoff1* percentage ratings for any score greater than or equal to '2', since a score of '1' was always **incorrect**, and a score of '2' or higher was always at least partially **correct**. (This method was also pilot tested during the standard-setting meetings in 1992.) According to Luecht (1993c), the *Poly-Angoff1* ratings were found to be slightly more robust than those resulting from the *Poly-Angoff2* method. The *Poly-Angoff2* method, the *Exemplar Response* method, and a *hybrid* method that combined elements of both were explored during the 1994 standard-setting pilot tests for NAEP.

Mapping Panelists' Ratings to the NAEP Scale. In reading and **mathematics**, mapping **panelists'** estimates (in the percent-correct **metric**) to the NAEP scale was a fairly straightforward **procedure**. After the **panelists'** estimates were aggregated across **exercises**, the means were numerically mapped to the NAEP test characteristic curve (TCC). This **procedure**, called the **TCC method**, resulted in a NAEP scale score for each of the nine performance standards **mapped**. The TCC procedure was also employed in determining the writing cut scores in 1992. These are shown in table 2.

For purposes of **comparison**, and at the urging of technical advisors to the NAEP standard-setting **project**, a second method was also used to determine the writing cut **scores**. This

Table 2—NAEP Writing Cut Scores Based on Test Characteristic Curve Methodology by Grade

Grade	Basic	Proficient	Advanced
4	187.0	282.4	354.0
8	222.2	299.9	371.4
12	246.4	332.9	416.3

Source: ACT Writing Report, 1993.

Table 3—NAEP Writing Cut Scores Based on Plausible Values Methodology by Grade

Grade	Basic	Proficient	Advanced
4	203.3	240.2	264.0
8	240.5	275.6	297.6
12	266.4	298.7	319.0

Source: ACT Writing Report, 1993.

procedure capitalized on the plausible values (PV) technology used in NAEP to generate unbiased group estimates of **proficiency**. In the **PV method**, **examinee** scores corresponding to the *exemplar responses (papers)* selected by the panelists were identified and aggregated to determine the performance **standards**. Since individual NAEP scores were not **computed**, the *plausible values* for each **examinee** became the score **proxy**. These values were left **unweighted** to avoid other assumptions involved in using weighted **data**. The **PVs** were summed over responses and judges to arrive at the **performance** standards that are shown in **table 3**.

The question to be asked here **is**, which method is appropriate for generating **performance standards**? There are conceptual differences between the two **methods**. It might be suggested that the Test Characteristic Curve (**TCC**) method operates from a definition of *Advanced*, for **example**, as *consistently Advanced*, since the performance standard is based on the total test score **function**. On the other **hand**, one might feel that the **PV** method operates from a definition of *Advanced*, as *typically Advanced*, but *not necessarily always Advanced*, since the performance standard is based on the plausible values of individual **examinees**, who **may or may not** achieve the same score on all exercises presented to **them**. This distinction suggests a policy question, "Is the *Advanced* standard defined as *always Advanced*, or is it defined as *typically*

Advanced?" Perhaps the same distinction could be made at the *Proficient* and *Basic* levels as well. Would this suggest lower performance standards instead of higher ones?

From a technical perspective, there is evidence to suggest that the apparent differences in the performance standards resulting from the TCC and PV methods shown in tables 2 and 3 are largely a function of *regression bias*, *overpredicting* in some cases, and *underpredicting* in others (Luecht, 1993a). However, consensus has not been reached on this point, and more research is needed.

Finally, it must be mentioned that a National Academy of Education Panel on the NAEP Trial State Assessment has suggested more-fundamental flaws in the methods used to set performance standards on NAEP. The Panel holds that item- or exercise-based procedures for setting performance standards produce invalid results, and that methods based on judgments of the quality and adequacy of entire test booklets are needed (The National Academy of Education, 1993). Commissioned critiques of the National Academy of Education report by Cizek (1993) and Kane (1993) dissent from these conclusions as well as the policy recommendations contained in the report. Here again, debate continues.

How Performance Standards Are Set by the National Board for Professional Teaching Standards

The National Board for Professional Teaching Standards (NBPTS) is developing performance assessments that are used to identify classroom teachers who have achieved advanced levels of accomplishment in their profession. These teachers are eligible to receive National Board Certification. The performance exercises used by NBPTS elicit complex responses from teachers, including written evaluations of the videotaped classroom performance of another teacher; submission of selected samples of the work of the teachers' students, together with reflective analyses of the quality of that work and the link between the teachers' instruction and the students' performances; videotapes of the teachers engaged in specified activities with their students, coupled to descriptions of plans and objectives and self-evaluative analyses of instructional success; written responses to problems involving appropriate structuring of curriculum; and demonstrations of effective use of instructional resources, among others.

Teachers' responses to exercises are scored by trained assessors using highly-structured rubrics and scales that yield scores ranging from "1-" to "4+" for each exercise. Training of assessors is extensive, and includes demonstrations of scoring competence and consistency.

Two approaches to performance standard-setting have been evaluated by NBPTS. The first, termed *Iterative, Judgmental Policy Capturing*, elicits panelists' implicit standard-setting policies by fitting mathematical models to their responses to profiles of the performances of hypothetical candidates for certification on a set of exercises. The second, termed the *Multi-stage Dominant Profile Procedure*, elicits panelists' explicit standard-setting policies. These procedures are described in turn.

The Iterative, Judgmental Policy Capturing Procedure. In this procedure, standard-setting panelists receive extensive training on the nature of the exercises that compose an assessment and on the meaning of each possible score level associated with the scoring of each exercise. In distinct contrast to the standard-setting methods described earlier for NAEP, the methods used

by NBPTS do not assume that exercises are scored on a **unidimensional scale**. Thus the score scales used with different exercises can be **unique**.

During a pilot study, a panel of 20 middle-school teachers applied the **Iterative, Judgmental Policy Capturing Procedure (JPC)** to performance exercises contained in the **NBPTS Early Adolescence Generalist teacher assessment**. Panelists applied Judgmental Policy Capturing to 200 profiles composed of the scores of hypothetical candidates for National Board Certification on six of the exercises that compose the **Early Adolescence Generalist assessment**. Panelists responded independently to graphical profiles of the performances of hypothetical teachers by specifying for each profile, whether the overall **performance** of a teacher with that profile should be considered “**Deficient**” (1), “**Competent**” (2), “**Accomplished**” (3), or “**Highly Accomplished**” (4). The National Board (NBPTS) has asserted that only “**highly accomplished**” teachers will receive its **certification**. A sample profile is shown in figure 2.

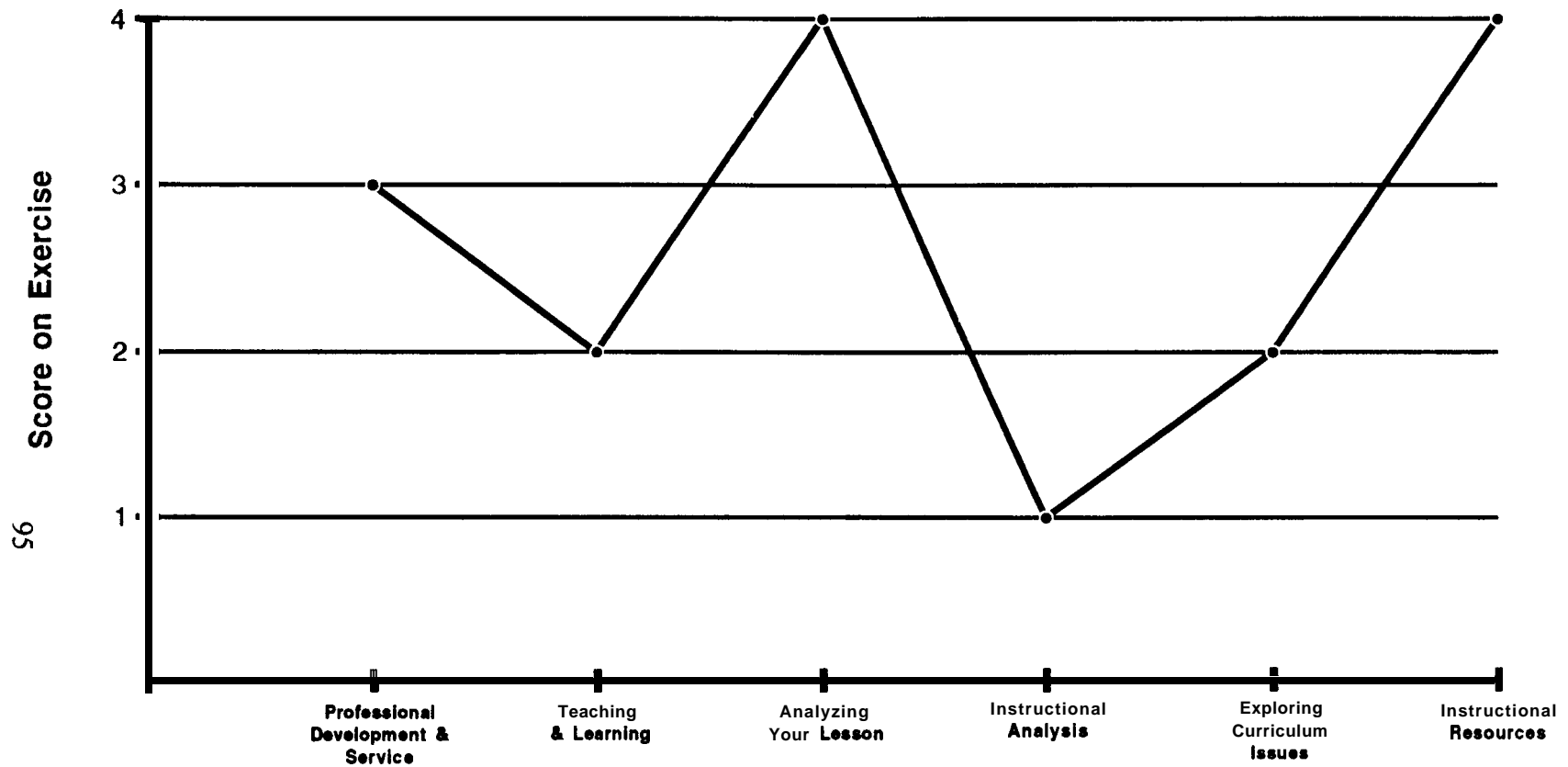
The **relative** weights each panelist applied to the six exercises in **specifying** the overall performances of hypothetical candidates were estimated using each of two analytic **models**. The **first**, a compensatory analytic **model**, was derived using ordinary least squares multiple regression. The **second**, a conjunctive analytic **model**, was derived using a multiplicative equation which, following a logarithmic **transformation**, also was evaluated through the use of ordinary least squares multiple regression. Thus two analytic models of the “**captured policy**” each panelist used in **specifying** performance standards and the magnitudes of relative weights that should be applied to the six exercises of the **Early Adolescence Generalist assessment** were **created**. Presumably, one could use a panelist’s captured policy to predict the overall score (s)he would give a candidate for National Board Certification who earned any profile of scores on the six **Early Adolescence Generalist exercises**.

After each panelist had responded independently to a set of 200 profiles and models of panelists’ policies had been **created**, the distributive weight each panelist applied to each of the six exercises was **estimated**. The distributive weights derived for each panelist summed to unity and could, therefore, be interpreted as proportions. For **example**, if a panelist assigned a distributive weight of 0.4 to the “**Teaching and Learning**” Exercise and a distributive weight of 0.2 to the “**Analyzing Your Lesson**” Exercise, we would conclude that (s)he placed twice as much weight on “**Teaching and Learning**” as on “**Analyzing Your Lesson**. ”

Distributions of the distributive weights each panelist applied to each of the six exercises were compiled and **graphed**. These graphs were provided to each **panelist**, with their own profile of distributive weights superimposed in graphical **form**, so they could see the weights their fellow panelists applied to each **exercise**, and **determine** whether their own distributive weights were similar **to**, or substantially different **from**, those applied by **others**. This form of **normative** feedback was followed by instruction on how to read and interpret the **graphs**, and an opportunity for panelists to discuss with their colleagues the relative importance they attached to each of the six exercises and their rationales for doing **so**. Initial discussion took place in groups of three or four panelists and was followed by a discussion by the entire **panel**.

After panelists had reviewed their own patterns of distributive weights and engaged in discussion with their **colleagues**, they once again completed a Judgmental Policy Capturing exercise. They again responded to the original set of 200 profiles, **specifying** for each, an overall performance rating on a **1-to-4 scale**, for a teacher with the indicated profile of performances on six exercises of the **NBPTS Early Adolescence Generalist assessment**.

Figure 2. Simulated Profile of a Candidate's Performances on Six Early Adolescence Generalist Assessment Exercises



Overall Evaluation of Candidate →	A ① Deficient	B ② Competent	C ③ Accomplished	D ④ Highly Accomplished

Fill the appropriate bubble (A, B, C or D) on the bubble sheet to indicate your judgment.

The model-fitting analyses described earlier were applied to the judgments panelists recorded in their second round of Judgmental Policy Capturing. The results of this new model fitting were used to estimate a new set of distributive weights for each panelist **and**, in **addition**, a table for each panelist that indicated the outcome of applying that panelist's captured policy to 114 hypothetical profiles of candidate **performance**. For each profile of performances on the six exercises, a panelist was able to see whether the application of her/his policy would result in the candidate being certified by the National Board or being denied **certification**.

Prior to a final round of Judgmental Policy Capturing, the results of which were analyzed as described **earlier**, panelists were once again given an opportunity to compare their profiles of distributive weights to graphically-presented distributions of the weights applied by their fellow **panelists**. They were also given an opportunity to discuss with fellow **panelists**, their rationales for applying greater weight to some exercises than to **others**.

Since an objective of the standard-setting study was to produce equations that would represent the judgments of all **panelists**, weighted averages of individual **panelists'** regression weights were computed to form a **single**, integrative **equation**. **Again**, separate equations were produced using each of two analytic **models**.

Some Results. Distributions of distributive weights that resulted from applying a compensatory analytic model to **panelists'** judgments elicited during the third round of Judgmental Policy Capturing are shown in figure 3. Extreme **values**, the first and ninth **deciles**, and the quartiles of the distributions are portrayed in box-and-whisker **charts**. Results obtained using a conjunctive analytic **model** were so similar to those obtained with a compensatory model that little would be gained from displaying **them**. They **are**, **therefore**, not **shown**.

The results of Judgmental Policy Capturing indicate that panelists assigned great weight to candidates' performances on the "Teaching and Learning" exercise in setting an overall performance standard (see figure 3). In contrast to traditional standard-setting procedures, which set a **unidimensional** test standard that is based on a simple or transformed sum of candidates' performances on all assessment exercises, JPC acknowledges the multidimensional nature of candidates' performances. It bases an assessment standard on profiles of candidates' performances.

A critical issue in the use of Judgmental Policy Capturing concerns the reasonableness of representing the captured policies of a group of panelists through a single policy that represents the central tendency of their **distribution**. At issue is the assumption that the policies of individual panelists can be regarded as randomly-differing observations drawn from a single **distribution**. To explore this **issue**, the vectors of weights that composed the **20 panelists'** policies were subjected to **nonmetric** multidimensional scaling analyses (Schiffman, Reynolds & Young, 1981). In **addition**, the matrix of correlations among overall performance scores assigned by the panelists to the 200 profiles of exercise scores were factor **analyzed**, using panelists as variables (Cattell, 1966).

The multidimensional scaling results did not reveal the presence of definable clusters of **panelists**, suggesting **instead**, a single distribution of distributive weights across all **panelists**. The results of factor analyzing the **20-by-20** matrix of correlations among **panelists'** overall ratings of profiles of exercise scores clearly demonstrated that their policies could be considered random

Figure 3

DISTRIBUTIONS OF RELATIVE WEIGHTS ASSIGNED TO EA GENERALIST EXERCISES BY YOU AND BY THE ENTIRE STANDARD-SETTING PANEL

JUDGE 20

PERCENT OF VARIANCE EXPLAINED = 71%

THE RELATIVE WEIGHTS YOU ASSIGNED TO EXERCISES

0.11 Professional Development and Service (PDS)

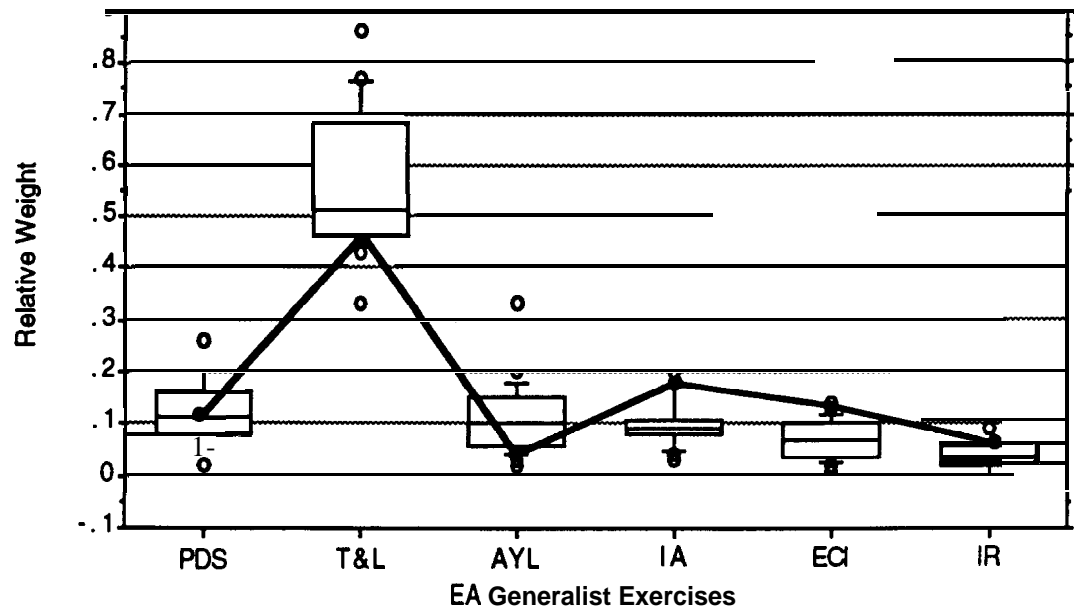
0.46 Teaching and Learning (T&L)

0.05 Analyzing Your Lesson (AYL)

0.18 Instructional Analysis (IA)

0.14 Exploring Curriculum Issues (ECI)

0.07 Instructional Resources (IR)



observations from a single distribution. A principal components analysis resulted in single factor that accounted for approximately 82 percent of the variance among panelists' policies.

These results were used to justify the derivation of a single standard-setting policy for the entire panel of 20 teachers. Analytically, the compensatory policy equations derived for individual panelists were combined through a weighted averaging procedure. The weights used were the inverses of estimated variances of estimates of corresponding regression coefficients. Through this procedure, more-precise estimates of regression coefficients were given greater weight than were less-precise estimates. Following the computation of a weighted average regression question, a linear transformation was applied so that a profile containing scores of four on each of the six exercises of the Early Adolescence Generalist assessment would yield an estimated performance score of four and a profile containing scores of one on each of the six exercises of the Early Adolescence Generalist assessment would yield an estimated performance score of one. This transformation was consistent with the desired result that perfect score performance on all exercises, should, with certainty, lead to National Board Certification.

In light of the scale definition noted earlier, the transformed JPC model is interpreted as follows: Any candidate with a predicted overall score, following transformation, that equals or exceeds 3.5 would receive National Board Certification. Candidates with lower predicted overall scores would not be certified. When this rule was applied to the Round 3 results of judgmental policy capturing, it was found that candidates who scored well on the "Teaching and Learning" Exercise would be certified, even though they performed less well on several other exercises in the assessment. For example, a candidate who earned scores of four on the "Teaching and Learning" Exercise, and the "Professional Development and Service" Exercise, but scores of three on the other four exercises would be certified, as would candidates who earned scores of four on the "Teaching and Learning" Exercise, the "Professional Development and Service" Exercise and the "Instructional Analysis" Exercise, scores of three on the "Exploring Curriculum Issues" Exercise and the "Instructional Resources" exercise, and a score of two on the "Analyzing Your Lesson" Exercise.

The pilot study demonstrated the feasibility of using Judgmental Policy Capturing as a method for setting performance standards on assessments composed of complex, multiply-scored performance-based exercises. The results of the study indicated (1) that the Judgmental Policy Capturing procedure and associated instructions used could be applied readily by panelists following a short period of instruction, (2) that panelists could provide judgments in response to as many as 200 six-dimensional profiles in a reasonably short time without undue fatigue (on average, panelists required 41 minutes to respond to 200 profiles), (3) that most panelists could achieve high levels of consistency in their responses to profiles, as reflected in coefficients of multiple determination between 0.70 and 0.87 (median of 0.79), and (4) that iterative use of Judgmental Policy Capturing reduces the variability among the captured policies of standard-setting panelists and leads to a reasonable degree of consensus around a collective policy. The results of the study also confirm that panelists applied differential importance weights to the exercises that composed the NBPTS Early Adolescence Generalist assessment package. Thus differential weighting of exercises in determining a performance standard (at least on the NBPTS assessment) appears to be essential.

The Multi-stage Dominant Profile Procedure. The Multi-Stage Dominant Profile procedure consists of four stages: individual policy creation, revision of individual policies following

discussion and **reconsideration**, application of individual policies to profiles of candidate performance (**administered** through a tailored Judgmental Policy Capturing **procedure**), and reaction to **summarized** policies created through detailed analyses of the application of individual policies to profiles of performance on an **assessment**. Whereas the Judgmental Policy Capturing procedure requires panelists to react to profiles of candidate **performance** that are presented to them so that their standard-setting policies can be **inferred**, the Multi-stage Dominant Profile procedure requires panelists to formulate and state the “**bottom-line**” standard-setting policies they would apply in screening candidates for some valued **reward**, such as National Board **Certification**. Panelists’ policy statements are then refined through **discussion**, and by analyzing their application to **profiles** of performance that challenge or **confirm panelists’** initial **policies**. A more-detailed description of the procedure **follows**. This procedure was applied in a recent pilot study conducted on behalf of the National Board for Professional Teaching **Standards**. Twenty middle-school teachers applied the procedure to six performance assessment exercises from the National Board’s Early Adolescence Generalist teacher **assessment**.

Policy Creation. During the first stage of the Multi-Stage Dominant Profile **procedure**, individual panelists stated policies that defined the lowest levels of performance on the six exercises **that**, in their **view**, warranted National Board **Certification**. Depending on the nature of their **policies**, panelists constructed one or more profiles of **performance that**, in their **view**, characterized the **performance** of a “**just barely certifiable**” Early Adolescence Generalist **teacher**. Following extensive instruction on the exercises for which **performance** standards were to be **set**, the meaning of each possible score on each **exercise**, and the nature of compensatory and conjunctive standard-setting **policies**, panelists were given ten blank profile **forms** on which to convey their “**bottom-line**” **profiles**. Panelists were told that only one profile would be necessary to convey a strictly conjunctive standard-setting **policy**, but that many “**bottom-line**” profiles could **exemplify** a compensatory **policy**. Panelists were asked to use as many profile **forms** as they needed to illustrate their standard-setting **policy**, and to write a paragraph that explained their **policy**.

Discussion and Reflection. Following their initial **specification**, panelists were invited to describe and discuss their standard-setting **policies**. Each panelist was provided with an opportunity to present her/his policy to the standard-setting panel and to describe the rationale underlying the presented **policy**. Stated policies were summarized on an overhead projector so that **all** could note how their own standard-setting policies compared to those of their **colleagues**. Panelists were then asked to reconsider their initial standard-setting policies in light of the discussion and what they had learned about their fellow **panelists’ recommendations** and **rationales**.

Panelists then worked **independently**, just as they had **originally**, to **specify** and illustrate their own standard-setting **policies**. Panelists completed as many illustrative profile forms as they needed (**ten** blank forms were given to each **panelist**, and no panelist used **all ten**) and again wrote one or two paragraphs describing their standard-setting **policies**.

Construction of Confirmatory and “Challenge” Profiles, and Summary Policies. Immediately following the standard-setting **session**, **panelists’** policies were analyzed to **identify** areas of agreement and **disagreement**. Many panelists provided one or more illustrative **profiles**, and all prepared brief descriptive statements of their **policies**. Content analysis of **panelists’** written policy statements and their accompanying profiles revealed several sophisticated policies

for determining candidates' qualifications for National Board Certification. These policies reflected what panelists had learned during their engagement in the Judgmental Policy Capturing procedure.

Using analyses of candidates' stated and illustrated policies, 105 profiles of performance on the six exercises that composed the NBPTS Early Adolescence Generalist assessment were constructed. These performance profiles either conformed to the "bottom-line" standard-setting policies that one or more panelists had specified or "challenged" those policies in the sense of conforming to all but one or two policy specifications. For example, if a panelist had specified in a "bottom-line" policy that, to be certified, a candidate could earn no more than two scores of two, must earn a score of four on a particular exercise, and must earn scores of three or more on the remaining exercises, a profile that contained three scores of two but otherwise conformed to the policy specifications was constructed. These profiles were used to examine the consistency with which panelists applied their "bottom-line" policies to the profiles of performance of hypothetical candidates. Panelists were sent the 105 profiles one week following the standard-setting session and were asked to state for each profile, whether a candidate with the indicated profile of performance should, or should not, receive National Board Certification as an Early Adolescence Generalist teacher.

Panelists were also sent three policy statements that represented the standard-setting policies of major subgroups of the standard-setting panel. They were asked, for each statement, whether or not they would accept the adoption of the statement as the National Board's standard-setting policy for certifying Early Adolescence Generalist teachers.

Some **Results**. Three standard-setting policies reflected the judgments of almost all panelists. These policies combined compensatory and conjunctive elements, in that panelists specified threshold levels of performance for some exercises, but allowed high scores on other exercises to compensate for lower-levels of performance on the balance of the assessment. When asked whether they would accept the adoption of these standard-setting policies by the National Board, 15 of 20 panelists endorsed a single policy, and four more endorsed minor variants of the majority policy. The final panelist created an entirely new policy that was at substantial variance with the judgments held by the balance of the panel.

Analyses of the responses of panelists to the 105 profiles of candidate performance created to confirm or challenge their policies revealed some variation in the consistency of panelists' application of their stated policies. However, on average, panelists were 77 percent consistent.

The results of the pilot study of standard-setting were very encouraging. The Judgmental Policy Capturing procedure was regarded by panelists as an important and helpful learning experience and as a useful strategy for eliciting their standard-setting policies. Following the Judgmental Policy Capturing procedure with the Multi-Stage Dominant Profile procedure permitted panelists to create explicit policies that accurately conveyed their judgments and to test and revise those policies when confronted with the ideas of their fellow panelists and with explicit profiles of candidate performance. The convergence of 75 percent of the panelists around a single standard-setting policy was a particularly encouraging result.

How Standards Were Set for the Maryland Statewide Assessment

Educators in Maryland have developed descriptions of what students should know and be able to do by the year 2000⁴. These descriptions together are called the Maryland Learning Outcomes. Content areas include **reading, mathematics, science, social studies**, writing and language usage.

The State of Maryland also developed a new **test**, the Maryland School Performance Assessment Program (MSPAP) to assess the ability of students to master the Maryland Learning Outcomes. The MSPAP is one element in Maryland's School **Performance Program reform initiative**. The test is designed to measure **students'** ability to apply what they have learned to real-world **problems**, and to relate and use knowledge from different subject **areas**.

Administered in grades **3, 5, and 8**, the MSPAP measures critical thinking and problem-solving **abilities**. MSPAP standards are set for satisfactory and excellent levels of performance. By the 1996 MSPAP test, **70** percent of students in each school are expected to meet satisfactory **standards**.

The Standard-Setting Process. MSPAP standards were set through a process that relied on the professional judgment of expert **educators**. The standards were reviewed and refined by a council that included representatives of **business, educational organizations, and the legislature**. Opportunity for **comment** was provided for the public at **large**.

In May 1993, a MSPAP Standards **Committee** was convened to set proficiency levels for satisfactory and excellent **performance**. The Standards Committee consisted of **17** content area and assessment experts from **11** school systems within Maryland and the Maryland State Department of **Education**. The Committee reviewed background material on the **MSPAP assessments**, including the learning outcomes that are assessed in **MSPAP; examples** of actual test **questions**, scoring **criteria**, and student **responses**; and 1992 MSPAP school performance results.

MSPAP scores fall into **five** levels of **proficiency**, with Level **1** the **highest**. The Standards Committee selected proficiency Level **3** or higher as the standard for satisfactory **performance**, and proficiency Level **2** or higher as the standard for excellent performance in each of the content areas/grade **levels**. The Standards **Committee** recommended a percentage range of students that must meet these levels by 1996. These ranges **are: satisfactory—70** percent to **80** percent; **excellent—20** to **30** percent.

The standards proposed by the Standards Committee were reviewed by the **MSPAP Standards Council**, including **12** persons from local boards of **education**, the Maryland State Teachers' Association, the Maryland Business **Roundtable**, the business **community, students, parents, and the State Legislature**. This council set specific goals for **schools**, within the ranges **recommended** by the Standards **Committee**, as to the percentage of students expected to meet the satisfactory and excellent levels of **performance**. The Standards Council recommended that by 1996 at least **70** percent of students should **perform** at the satisfactory level and **25** percent should **perform** at the excellent **level**. By 2000, **95** percent of students should perform at the satisfactory level in each content **area**, and **50** percent should perform at the excellent **level**. These standards will be reviewed **after** the 1996 test to see if adjustments are **needed**.

The State Superintendent of Schools reviewed the **recommendations** of the Standards Council and briefed local **superintendents**. The State Board of Education reviewed the Superintendent's **recommendations** toward the end of **May**. Regional meetings for the public were held in June and **July, 1993**, and the State Board of Education included the new **MSPAP** standards in the regulations covering public school standards in **July, 1993**.

Use of MSPAP Results. The main purpose of **MSPAP** is to assess school **performance**, not the performance of individual **students**. Individual **students' scores**, however, are available to be used in conjunction with other **information** to **inform** parents of their child's performance in particular content **areas**.

How Standards Were Set for the Kentucky Instructional Results Information System

In 1990, Kentucky's General Assembly adopted the Kentucky Education Reform Act (KERA)⁵. KERA established goals for the educational system and provided a procedure for assessing progress toward the **goals**.

One of the six education goals is:

"Develop students' ability to:

1. Use basic **communication** and mathematics **skills**;
2. Apply core concepts and principles from **mathematics**, the **sciences**, the **arts**, the **humanities**, social **studies**, and practical living to situations they **will** encounter throughout their **lives**;
3. Become self-sufficient **individuals**;
4. Become responsible members of a **family**, work **group**, or **community**;
5. Think and solve problems in a variety of **situations**; and
6. Connect and integrate experiences and knowledge from all **subject-matter fields**."

A Council on School Performance Standards was created to **further** define these cognitive **goals**. The Council developed **75** performance **goals**, or "**valued outcomes**," which the State Board of Education **approved**.

Kentucky proceeded to develop a new assessment **program** to measure **students' performance** on the valued **outcomes**. The assessment program was to include multiple-choice and open-ended **items**, performance tasks that can be **administered** in one class **period**, and longer tasks or projects that can be placed in a student's **portfolio**. The valued outcomes were to be the foundation for new models of school **curricula**. **NAEP** frameworks were also used to supplement the valued outcomes where **needed**. **KERA** stipulated that a transitional assessment would occur at Grades **4, 8, and 12** in 1991-92, with a **full-scale**, primarily **performance-based** assessment to be implemented soon **thereafter**. Information on development of the **tests**, transition to performance-based **assessment**, and the mechanics of test administration is available in a *Technical Report*.

Setting Performance Standards. It was decided to set four levels of performance **standards**, termed "novice," "apprentice," "**proficient**," and "**distinguished**." Considerable thought went

into deciding the appropriate number of levels as well as their **names**. Too many levels would make differences between adjacent categories somewhat trivial and would be hard to **communicate**. Too few levels would result in setting the standard for minimum competency at a very low **threshold**, removing the motivation for **improvement**.

Standard setting for writing portfolios was completed first by a Writing Advisory Committee consisting of approximately **10** teachers per grade level (**Grades 4, 8, and 12**), staff from the Kentucky Writing Project, university **faculty**, and members of the Kentucky Department of Education. The **committee** reviewed student **portfolios**, selected sample portfolios that exemplified the **standards**, and completed the scoring guide to **operationalize** the **standards**. These were incorporated into a set of training materials distributed to teachers **statewide**.

The first year's test was transitional in **nature**, containing common questions (**given** to all **examinees**) and matrix-sampled questions (**each examinee** is tested on only some of the **items**). Standards were set first on the common **items**; standards for the **matrix-sampled** questions were derived from data describing the relationship between the two types of **questions**.

For the **common items**, standard-setting was done for performance (**open-ended**) items **only**, and was based on examining actual student **work**. Scores for performance items ranged from 0 for no response or an irrelevant **response**, to 4 for a "**complete, insightful, strongly supported response**, demonstrating in-depth understanding of relevant concepts and processes."

Scoring guides for performance items were translated into **standards**. Standard-setting panelists were first trained on the writing scoring **guides**. **After training**, they were provided with several **examples** of scored writing **samples**. The committee members were then given several papers that had already been assigned **scores**. For each **question**, they independently decided which score reflected each level of **performance**. They then debated their decisions as a **group**, and reconsidered their decisions until consensus was **reached**. They arrived at a decision that scores of 0 and 1 "**generally indicated Novice work**," 2 was **Novice**, 3 was **Proficient**, and 4 was **Distinguished**.

Committee members were given several papers at the upper and lower ends of the range for each of the above score **values**, to discuss whether these papers cast doubt on their score-to-standard translation **scheme**. Their decision rules remained unchanged following this **discussion**. Finally, the **committees** were given **information** about the percentage of students attaining each **performance level**, so they could see that very small percentages of students achieved the Proficient or Distinguished **levels**. In spite of **this**, they decided not to change their original decision **rules**.

Standards for the matrix-sampled questions were derived from those established on the **common items** by using a regression **approach**. The methodology is **fully** explained in the *Technical Report*. The **procedure** resulted in **60** predicted **performance-level scores**. Cutoffs for each performance level were then **determined** so that the distribution of students classified into the various **performance** levels on the matrix-sampled items was as similar as possible to the distribution of students classified on the common **items**. Once cutoff points were **established**, each student was assigned to a **performance** level based on his/her score on the matrix-sampled **items**.

Standards were set for **performance** events by assuming that the distribution of scores across levels would be as close as possible to those obtained for the common **items**. Regression analysis

was not used for these events, however. Students with the highest scores on the performance events were placed in the Distinguished category until the percent of students at this level approximately equaled the percentage of students classified as Distinguished on the common items. Students with lower scores on performance events were subsequently placed in the Proficient, Apprentice, and Novice categories in approximately the same percentages as with the common items.

Use of Assessment Results. Assessment results were used to determine the percentage of students in each school whose scores warranted placement in each of the four categories—Novice, Apprentice, Proficient, and Distinguished. Schools received rewards or sanctions based upon their improvement or failure to improve over time. The 1991–92 test established each school’s baseline, from which targets for future improvement were determined. Test results collected during the 1992–93 and 1993–94 school years were used to measure each school’s progress toward its target and to establish a new baseline for future performance targets.

Some Technical Dilemmas and Procedural Issues

Although constructive work on setting performance standards for assessments containing performance exercises is ongoing, and some progress has clearly been made, many dilemmas and procedural concerns remain. In this section we review some troubling issues and unresolved dilemmas, thus, by implication, defining a research agenda for the coming years.

Artificial Dichotomization (Polytomization) of Performance Continua

The central problem in setting cut-points or standards for levels of performance is that they establish artificial dichotomies on a continuum of proficiency (Shepard, 1984). Proficiency in a population does not occur at discrete, easily recognizable levels. It is, rather, a matter of degree. The problem is how to treat the gray areas around the cut-points, since a certain proportion of examinees just above or just below a cut-point will almost inevitably be misclassified due to measurement error.

Most judgmental methods of setting performance standards begin with the concept of a minimally competent examinee to assist in establishing a passing threshold for a test. Judges must either decide which of the response categories in a multiple-choice item the minimally competent examinee would reject (Nedelsky, 1954), which items he/she would answer correctly (Jaeger, 1982), or the probability, for each item (or parcel of similar items), that he/she would give the correct answer (Angoff, 1971). These methods result in an informed but arbitrary (i.e., determined by judgment) performance standard. Examinees achieving a score above this standard are classified as competent; those below are classified as incompetent. The problem arises in the cases of false positives and false negatives (i.e., those persons who are misclassified on the basis of the test versus their actual abilities). Taking measurement error into account in establishing a performance standard can partially diminish the likelihood of committing one kind of error or the other.

Other methods combine judgment with empirical data on actual test performance. The borderline-group method furnishes judges with information on the actual performance of a group of examinees believed to be on the borderline of minimal competence. The median score of this group is adopted as the standard. In the contrasting-groups method, groups of “masters” and “non masters” are identified and given the test. Frequency polygons are developed for the scores

of each group. The standard for passing the test is set at the score where the frequency polygons cross. If it is found that an unreasonably high percentage of examinees fail the test, the threshold can be lowered. The degree to which the threshold is lowered can be based on how high the stakes are for a test. For example, if the test is used to qualify examinees for promotion to the next grade, or for professional licensing, the cost to examinees who fail even though they are truly qualified is high. The passing threshold should be lowered accordingly, but not so much that it is no longer a standard of minimally acceptable performance.

Ways to minimize the misclassification problem include:

- Consideration of the examinee population, not only the test items, in setting standards;
- Supplementing judgments with empirical data (e.g., whether the test actually distinguishes the competent from the incompetent);
- Use of mathematical models for adjusting cut points to minimize classification errors. Such models include the binomial error model developed by Subkoviak, Huynh's beta-binomial model, Bayesian decision-theoretic models, and linear loss models;
- Pilot tests of examinations, with revision of performance standards as appropriate; and
- Occasional review of performance standards as conditions change (e.g., when curriculum changes).

Shepard recommends seeking insight from several models of standard-setting. Evidence from different approaches, combined with normative evidence (actual performance), should be used to establish a range of plausible performance standards.

Performance Standards Tend To Be Method-Dependent

It is widely agreed that different methods of setting performance standards yield different results. The problem seems thus far to be intractable. The inconsistencies are inherent in the way judgments are made according to the various methods, as well as in the composition of panels of judges. In a review of research, Jaeger (cited in Phillips, 1993) summarized the results of 32 comparisons culled from 12 separate studies. He found that almost six times as many students would fail when using one standard-setting method rather than another. Inconsistencies arose even when different methods were used under seemingly identical conditions.

As already suggested, the results of several methods should be used in any given study, and these should be tempered with statistical techniques, use of empirical data, and knowledge of the examinees. Consideration of the use to which the performance standard will be put should also have weight in setting the cut-point(s). Shepard comments, "It is important to see that it is the nature of the problem rather than lack of effort which prevents us from finding the preferred model for standard setting" (Shepard, 1984, p. 187).

How Large Should a Standard-Setting Panel Be?

Although we hold no belief that performance standards are population parameters waiting to be discovered, we recognize that any standard-setting panel is composed of a sample of individuals, selected from a population of persons who might have been invited to serve. We also recognize that a second panel, selected using the same procedures as the first and presented with identical instructions and stimulus materials, would likely recommend a somewhat different

performance standard. Examinees whose performance is in the region bounded by the performance standards recommended by two different panels are victims of the process used to select a panel. They would have received some valued reward (admission, certification, classification as “advanced” learners, etc.) had their performances been judged against the performance standard specified by one panel, but denied those rewards if judged against the performance standard of the second.

Determining the number of panelists who should recommend a performance standard is not unlike other sample size problems. The larger the number of panelists, the more stable will be the resulting performance standard across samples of panelists. Since all popular standard-setting procedures have as a final step, computing the central tendency of a distribution of recommendations provided by individual panelists, the standard error of the final performance standard will vary inversely as the square root of the number of panelists. A seemingly reasonable criterion for determining the number of panelists to be sampled would be to produce a standard error of the mean recommended standard that was small, compared to the standard error of measurement of the assessment for which a standard was sought. Since variation in the recommended performance standard and measurement error variance contribute additively to errors of examinee classification, the size of the standard-setting panel should be such that the standard error of the mean (or median) recommended performance standard does not add appreciably to the overall standard error. Jaeger (1990) noted that if the standard error of the mean recommended performance standard is no more than a fourth as large as the standard error of measurement of the assessment, the overall standard deviation of errors due to sampling of panelists and the unreliability of the assessment would only be 3 percent larger than the standard error of measurement alone. By this criterion, random error due to sampling of panelists would contribute minimally to errors of classification of examinees. To apply this criterion in practice, both the standard error of measurement of the assessment for which a standard is sought, and the standard deviation of panelists’ recommended performance standards must be known. Standard errors of measurement are readily estimable. As experience with setting performance standards on assessments composed of performance exercises is gained, typical standard errors of recommended performance standards should become available as well.

Who Should Compose a Standard-Setting Panel?

Most professionals in the performance testing arena recommend that a standard-setting panel should be representative of a broad range of interested and informed groups. Broad representation avoids the selective bias that might result if only one concerned group (e.g., only members of the business community or only school administrators) constituted a standard-setting panel. It is important to identify the relevant groups from which judges should be chosen, and to sample the judges from these groups in such a way that they are representative of the relevant population. However, drawing on research concerning the nature of expertise (Chi, Glaser & Farr, 1988; Minsky & Papert, 1974; Chi, Glaser & Rees, 1982; Glaser, 1987), Jaeger (1991) argued that judges who recommend or set performance standards should be experts in the domain assessed by the test and in the roles sought by successful examinees. He recommended the use of standard-setting panels that were composed of experts, and not necessarily representative of stakeholder populations.

The issue here is quite important. Democratic values support the argument that a performance standards should be set through a process of negotiation that involves all

constituencies having an interest in the qualifications of those who would “pass” the test or assessment on the basis of the performance standard set. In the case of achievement testing, those constituencies would likely include parents, students, teachers, school administrators, potential employers of students, civic leaders, university admissions officers, and so on.

The principal **counterargument** is based on the contention that standard-setting panels should be restricted to those with solid understanding of (1) the test or assessment for which a standard is to be set, (2) the cognitive demands imposed by the **assessment**, and (3) the capabilities of examinees, given the opportunities they have been provided to learn the material tested. Lacking firm understanding of these **issues**, a standard-setting panel could readily set performance standards that were **unrealistic**, unfair to students and **educators**, and unlikely to influence instructional **improvement**. The 1985 *Standards for Educational and Psychological Testing* require that the qualifications of standard-setting panelists be documented (Standard 6.9). This requirement suggests that legitimate interest alone should not determine the composition of standard-setting **panels**; expertise should be required as **well**.

There are some standard-setting initiatives where the term “**expert**” has been broadly defined to include not only those who have expertise in the content area (**typically** gained through formal academic **preparation**), but **additionally**, members of the non-academic community who use the skills and knowledge of the content area in their trade or **profession**. When performance standards were set for the National Assessment of Educational **Progress**, “**expertise**” was defined very **broadly**. The National Assessment Governing Board regularly has **included**, and now by statute (P. L.. 103-382, 1994) must include on its standard-setting **panels**, school **administrators**, parents, and concerned members of the general **public**.

The Appropriate Uses of Normative Information

The desire for criterion-referenced performance standards. The nation’s governors and two Presidents have reached a consensus that there should be high expectations of what students should know and be able to **do**, and clear indicators of how **students’** actual performances measure up to these **expectations**. In other **words**, it is **no** longer enough to find out what students know and can **do**; new public policy demands an assessment of whether **students’** performance is “good enough.” **Therefore**, criteria are needed for deciding what is “good enough,” in the sense that students know what they should and are able to do what is expected of **them**.

In many applications of performance **assessment**, attempts are made to set absolute standards of **performance**, quite apart from current norm distributions of **performance**. The work of the National Board for Professional Teaching Standards is a case in **point**. The National Board has defined standards of “**accomplished performance**” in the teaching profession for more than a dozen teaching fields and will eventually define such standards for more than **30** teaching **fields**. These standards represent the National Board’s vision of accomplished teaching performance and are not grounded in typical current **practice**. In no sense are these performance standards normatively based.

The need to ground performance standards in reality. The dilemma arises when it is recognized that standard-setting panelists must make use of normative anchors if they are to set performance standards at levels that are practical and **defensible**. Much of our judgment in life is normatively **grounded**, and it is difficult to establish absolute standards apart from our normative **experience**. For **example**, we regard men who are over **78-inches** tall as “**very tall**” not because of

any absolute sense of the accumulation of inches of height that warrant that **appellation**, but because, in our normative **experience**, very few men achieve that **height**.

Normative data are essential in the standard-setting process for ensuring that performance standards will be appropriate to the population for which they are being **set**. Setting a performance standard in a **vacuum**, without reference to actual **performance**, could easily lead to thresholds that result in large classification **errors**. This would be particularly **harmful** in the case of high-stakes tests in which large numbers of **examinees** could be disqualified even though they are competent in the skills being **tested**. Even in group **assessments**, **however**, the standards would lack credibility if they did not reflect actual **abilities**. Such a situation could **arise**, for **example**, on a mathematics assessment if it was found that **examinees** who scored at an “advanced” level were actually poor mathematics **students**.

It was noted earlier that different methods of deriving **performance** standards yield **different**, sometimes widely divergent **standards**. A standard-setting process that used several methods as well as incorporating **normative** feedback to judges should result in more stable and realistic performance **standards**. Such **normative** data would help judges decide whether a particular standard actually separates **examinees** according to relevant skill **levels**, as **desired**. Judges should be given an opportunity to revise standards in light of empirical **consequences**. This iterative process might be repeated several times with more or different types of **feedback**. As **Shepard observed**, “The standard we are groping to express is a psychological construct in the **judges’** minds rather than in the methods . . . **Normative** expectations are always the starting **point**, because norms are the source of validity evidence and ultimately **determine** our psychological absolutes” (Shepard, 1984, p. 188).

So the **dilemma** of wanting to set absolute standards coupled with the reality of knowing from experience that totally unrealistic standards may result **remains**. It seems **best**, in light of research on the effects of **normative** feedback on performance standards (cf., Busch & Jaeger, 1990) to temper the desire for absolutes with the reality that normative feedback to standard-setting panelists **provides**.

Essential Training of Standard-Setting Panelists

Defining minimally acceptable performance. Methods of setting **performance** standards often require panelists to formulate some idea of a **minimally** acceptable **examinee**. Then for each test item or **exercise**, panelists must estimate whether that hypothetical **examinee** would give the correct **answer**, or must estimate the percentage of such **examinees** who would answer the item **correctly**. In setting multi-level **standards**, panelists must repeat this process for each performance **level**. Some methods use an iterative process in which panelists are given empirical **information** on the difficulty of each item or exercise and/or share each **others’** ratings in the previous **round**.

These classical approaches to standard-setting are most **useful** when items are dichotomously scored rather than being scored along a **polytomous scale**. As noted **earlier**, **performance** assessments in which **essays**, **portfolios**, and extended constructed responses are used do not lend themselves to classical standard-setting **methods**. **Further**, some assessments now include both multiple-choice and **performance items**, resulting in differential standards being **set**. The picture is **further** complicated by the **multidimensionality** of performance **items**, in which more than one skill or area of knowledge is **integrated**. The task becomes overwhelming for **judges**, who must

keep the test frameworks in mind while minimizing the role of personal experience and intuition in their judgments.

The difficulty in employing methods centered on defining minimally acceptable performance underscores the advisability of incorporating empirical evidence of the validity of **standards**. The contrasting-groups and borderline-group methods described above may be **helpful** in validating standards that are derived through other methods (**in** the sense of providing convergent **evidence**). Bootstrapping is another recently-developed **method**, in which the performance standard is set at the point at which it has been empirically **determined** that **examinees** who score above the standard **will**, with high **probability**, succeed in the task being **measured**. Benchmarking is a variant of this **approach**. In **benchmarking**, the standard-setter attempts to find examples of high **performance** as an aid in setting the **standard**, **as**, for **example**, the scores attained by high-performing math students in other **countries**.

Training on the nature of the standard-setting task. It is very important that panelists be adequately trained in the process of setting standards that they are expected to **use**. They must understand **fully**, and **internalize**, the test or assessment framework before rating items or exercises as to what they measure and their levels of **difficulty**. Panelists must have clear definitions of the knowledge and skills expected of minimally acceptable **examinees**, or of **examinees** at each of multiple performance **standards**. Panelists must also understand fully the content domain to be **assessed**. Because performance standards are based on **judgment**, a final decision necessarily **will** be **arbitrary**. With the requisite **training**, **however**, the performance standard will not be **capricious**.

Sources of Error and "Adjustment" of Standard-Setting Recommendations

If a performance standard is set to the value recommended by a standard-setting **panel**, half of the **examinees** whose true performance level is precisely equal to the standard will fail the **assessment**, and half the **examinees** whose true performance level is minutely below the standard will pass the **assessment**. This is true because errors of measurement influence the observed performances of **examinees** and **sampling** errors in the selection of panelists influence the performance **standard**. These sources of error are independent of each **other**.

It is possible to adjust **recommended performance** standards for either or both of these sources of **error**, but doing so requires conscious consideration of the relative costs of false-positive and false-negative classification **errors**. An adjustment for both sources of error can be made by calculating the standard error of measurement of the assessment and the standard error of the mean recommended **performance standard**. These standard errors must then be squared and **summed**, and the square root of the sum of squares must be **calculated**. We will call this result the "**total standard error**."

If the **recommended** performance standard were to be increased by one total standard **error**, the percent of **examinees** with true **performance** just below the original **performance** standard who would **pass**, due to these **errors**, would be reduced from 50 percent to 16 percent (**assuming normally-distributed errors**). Thus the rate of false-positive errors would be reduced **substantially**. **However**, the rate of false-negative errors for those whose true **performance** was equal to the original standard would be increased to **84 percent**. Lowering the original performance standard by one total standard error would have precisely the opposite **effect**. The

rate of false-negative errors would be diminished **substantially**, but the rate of false-positive errors would be similarly **increased**.

Whether performance standards should be adjusted to compensate for errors of measurement and sampling **errors**, and if so, in which **direction**, is a policy **matter**. Alerting policy-makers to the existence of the errors and to potential adjustments of performance standards that could be used to address the errors is the responsibility of professionals who recommend **performance standards**.

Summary

The need to establish performance standards is unlikely to diminish in the coming years. Education policy at the federal and state levels is firmly grounded in the notion that students and teachers should be held to high standards and that they should **demonstrate** their achievement of such **standards**. As long as policy-makers express the need for standardized evidence of student and teacher performance in terms of **pre-established** expectations and define public information needs **similarly**, those who establish performance standards **will** busily ply their **trade**.

In this chapter we have described the myriad ways performance standards are used and have addressed the need for new methods of establishing such standards for performance assessments of students and **teachers**. We have reviewed some strategies employed by those who have heretofore encountered the problem of setting performance standards on assessments that incorporate performance **exercises**—a literature that is principally fugitive and an area of inquiry that is **embryonic**. **Finally**, we have concluded with a review of technical dilemmas and procedural issues that plague all attempts to set performance **standards**, but are particularly important when standards are set for performance **assessments**. We hope that this latter section **will** be regarded as a challenge to those who would advance this important line of **inquiry**.

Notes

¹The word arbitrary is used here in accordance with the second meaning provided in the Oxford English Dictionary (19701): "Relating to, or dependent on the discretion of the **arbiter**, arbitrator, or other legally-recognized **authority**; **discretionary**, not fixed (p. 107)." Even though judges used to recommend test standards rarely have legal authority to do so, those who set standards of performance typically have statutory authority to establish education policy. In the case of the National Assessment of Educational **Progress**, for **example**, the National Assessment Governing Board has exercised authority provided in the NAEP authorization to adopt performance standards that were **recommended** by a standard-setting panel and its **contractor**.

²The word capricious also used in accordance with the second meaning provided in the Oxford English Dictionary (1971): "**Full of**, subject **to**, or characterized by **caprice**; guided by whim or fancy rather than by judgment or settled **purpose**; **whimsical**, **humoursome** (p.335)." Although the words arbitrary and capricious are sometimes used **interchangeably**, their connotative meanings are quite **different**. Reflective consideration and a well-defined judgment process often underlies **arbitrariness**, but never **capriciousness**. This distinction is central to the contrast made **here**.

³**Proficient** is defined as competency over challenging subject **matter**, *Basic* as partial **mastery**, and *Advanced* as superior **performance**.

⁴The following information about **Maryland's MSPAP** is taken **from** a Memorandum of July 27, 1993 from Nancy S. Grasmick, State Superintendent of **Schools**, to Members of the State Board of **Education**, and **appendices**.

⁵**Information** about Kentucky's performance assessment standards is derived from Kentucky Instructional Results Information **System**. 1991-1992 Technical **Report**, Kentucky Department of Education, Frankfort, KY, January 1993.

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Fairness in Large-Scale Performance Assessment¹

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Defining Fairness

In the context of **assessment**, fairness is integrally related to the concept of **validity**. Concerns about fairness **are**, in large **part**, concerns about the "differential validity" (Cole and Moss, 1989) of an assessment for any person or group of concern. **However**, concerns about fairness extend beyond the **consensual** boundaries of **validity**, which encompass the soundness of an assessment-based interpretation or **action**, to include concerns about the soundness of the educational or social policy of which the assessment is a **part**. In this **chapter**, we consider these two broad aspects of fairness: those focusing on the validity of an **assessment**, which we characterize as concerns about *bias*, and those focusing on the soundness of the educational or social policy to which the assessment **contributes**, which we characterize as concerns about *equity*.

Fairness is a concern with respect to **all** persons who are **assessed**, whether considered as individuals or as members of some definable **group**. Much of the relevant research has focused on fairness with respect to groups defined by social **background**: typically **race/ethnicity**, primary language, **gender**, and handicapping **conditions**, although any relevant grouping might be studied, for **example**, based on **age**, curriculum **experienced**, type of school **attended**, and so on. No validity research can warrant the validity of an interpretation or action for a particular individual unless it is **carefully** evaluated in the context of other relevant information about that individual. The *1985 Standards for Educational and Psychological Testing* and the *Code of Fair Testing Practice*, jointly prepared by AERA, APA, and NCME, delineate obligations for tests users and for assessment development professionals to those who are **assessed**. Our focus here is primarily on studying issues of bias and equity across groups differing in social **background**.

Bias, like **validity**, refers to the soundness of an interpretation or action based on a test score—not to the assessment instrument **itself**. An assessment may be biased for some purposes but not for **others**. **So**, bias must always be evaluated in the context of the way in which an assessment instrument is interpreted and **used**. Validity refers to the extent to which the available evidence supports the intended interpretation and action over alternative interpretations and actions given the intended **purpose**. Bias refers to the extent to which there is evidence of *differential validity* for any relevant subgroup of persons **assessed**. For every validity issue raised in the chapter by Messick and colleagues, there is a related bias issue that may be raised with respect to a group of concern. An assessment-based interpretation or action is biased when it is not equally valid for **all** relevant subgroups.

Concerns about *equity* spill over the consensual bounds of validity and bias to include questions about the fairness of the educational system in which the assessment was **used**. It is possible for an assessment to be considered unbiased in a technical sense—in the sense that the intended interpretation is equally valid across various groups of concern—and yet be used in service of a policy that fails to promote **equity**. **Clearly**, equal access to a high quality education is a minimum requirement of a fair educational **system**, **although**, as we will discuss **below**, the means of documenting and eliminating inequalities in the opportunity to learn are complex and **problematic**. In **addition**, differences in educational outcomes are also relevant indicators of fairness—there should be evidence that students are benefiting from their **education**. **Ideally**, in a fair **system**, the average achievement is high and social distribution of achievement is equivalent for **all** groups of concern (Lee and Bryk, 1989). The same might be said for the range of other valued outcomes of **schooling**, such as **self-esteem**, **citizenship**, or access to employment or higher **education**. Where differences exist between socially-defined **groups**, there is an on-going effort to understand the reasons for the difference and to implement policies that move in the direction of closing the gap and raising the overall level of **success**. The question for assessment evaluators is whether an assessment is contributing to or detracting from the fairness of the educational system of which it is a **part**.

The fairness of an assessment system must always be evaluated in light of the purpose it is intended to serve and the way in which the results are interpreted and **used**. An assessment may be "**fair**" for some **purposes**, but patently unfair for **others**. For **instance**, if an assessment is used for instructional **planning**, evidence that students have had the opportunity to learn the material assessed is not **necessary**, although it may be **useful**; **however**, if the same assessment is used to inform decision about promotion or certification for **graduation**, such information is **crucial**. **Moreover**, simple differences in **summary** statistics between socially defined groups are **not**, in and of **themselves**, indicators of bias or **inequity**; they **do**, **however**, highlight the necessity of a careful search for additional **evidence**, in light of the purpose of **assessment**, to explain the source of the differences and to inform decisions about the appropriateness of the assessment system and the policy that surrounds it. Our intent is to offer readers a framework for studying fairness—collecting and evaluating evidence to **inform** conclusions about bias and **equity**—**and**, where existing research **permits**, to offer suggestions for designing and using assessments in ways that enhance **fairness**.

In the section entitled "**Fairness and Bias**," we offer suggestions for minimizing bias during assessment **development**, **administration**, and **evaluation**. In the section entitled "**Fairness and Equity**," we cover three interrelated sets of **issues**. **First**, we **summarize** evidence on differences in **performance** between groups differing in social background and offer recommendations for responsible reporting of such **differences**. **Second**, we highlight the importance of understanding differential performance in light of the differences in antecedent **conditions**, especially the opportunity to **learn**. **Here**, we emphasize the legal and ethical importance of ensuring opportunity to learn when students are held accountable for their performance as in cases of decisions about promotion or **graduation**. We close with a consideration of the potential for differential consequences of performance assessments across socially defined **groups**, highlighting the need for on-going critical dialogue among **all stakeholders** in the assessment **process**.

Fairness and Bias

As mentioned earlier, the term *bias* will be used here to refer to the extent to which a test score and test use are valid for **all** intended individuals and **groups**. As **such**, if an assessment results in scores that systematically underestimate the status of members of a particular group on the construct in **question**, then the **test is** biased against members of that **group**. An assessment is also biased if it purports to measure a single construct across **groups**, but in fact measures different constructs in different **groups**. To take a familiar extreme **example**, an **assessment**, in English, of quantitative reasoning administered to two groups of students who differ significantly in English **proficiency**, may be a relatively pure measure of quantitative reasoning in English-proficient **group**, but probably reflects a mixture of knowledge of English and quantitative reasoning in the less English-proficient **group**.

The term *bias* will also be used here to apply to situations where an assessment is differentially valid in selecting individuals for favored treatment (**e.g.**, employment or college admissions) and in predicting some **future criterion**. We should note in passing that bias in selection and prediction is by no means a straightforward **matter**, and competing models of selection and prediction bias are still being debated (**see Hartigan & Wigdor, 1989; and Journal of Educational Measurement, Special Issue on Bias in Selection, Vol.13, 1976**).

Finally, it is worth repeating that what counts as “**bias**” ultimately depends upon the **purpose**, intended **interpretations**, and context of the **assessment**. Assessment exercises on NAEP, for **example**, have no consequences for **individuals, schools, or districts (at least not yet!)** and are very different from assessments for individual certification or **placement**, and lead to different concerns of the what constitutes bias and the nature of that **bias**.

With these **caveats**, we may **ask**, what specific factors affect bias in **performance** assessment and what procedures can be undertaken to minimize their **effects**? Biases in performance assessment may be conceived as falling under two broad **rubrics**. The **first**, biases emanating from *internal sources*, encompasses the actual content and administration of the assessment itself. The **second, external sources of bias**, refer to the relationship between the assessment and some outside criterion of practical or theoretical **interest**.

Internal Sources of Bias in Performance Assessment

In virtually **all** aspects of the assessment **situation**, internal biases can enter the **assessment**. Biases may be present in the specification of the content **framework**, in the development of actual tasks and **exercises**, in the selection and training of **scorers**, and in the actual scoring **rubrics**. **Moreover**, performance assessments do not follow a compensatory **model**. Poorly conceived exercises that draw heavily upon **construct-irrelevant** abilities cannot be rescued by high quality scorer **training**. A well-thought-out and comprehensive content framework can be negated by exercises that do not reflect the **framework**, or by an inadequate scoring **scheme**. Performance assessment is a good illustration of the old saw that a chain is only as strong as its weakest link.

Diversity of development and scoring panels. First, diversity **among** the experts selected to specify the content **framework**, to develop and evaluate candidate **exercises**, and to devise valid and reliable scoring schemes is **critical**. This is especially true for those subject areas where differences in the skill required for successful performance vary as a **function** of **context**. The

conduct of an arts assessment is a good **example**. Individuals of different racial, ethnic, or gender groups will probably vary substantially in their opinions about what range of content should be included for an arts assessment; what performance tasks should be **included**; and how the scoring should be developed and **applied**. The problems associated with a lack of diversity among experts are of particular concern with **performance assessments**, as there is far more room for subjective judgment with open-ended and extended response questions than with traditional multiple choice or true or false questions. Hence, it is particularly critical that a **full** array of opinions from experts of different cultures, ethnic, and social backgrounds be reflected in the process of test development for **performance assessment**.

Specifying the content framework. The first source of possible bias in assessment is in the specification of the content **framework**. If the initial conceptualization of what constitutes understanding of geometric **concepts**, for **example**, is **flawed**, then the tasks devised to assess that understanding will in all probability be similarly **flawed**. Even if the specification of the content framework is sufficiently rich and diverse to encompass a wide variety of original and creative demonstrations of **understanding**, the exercises designed to reflect that understanding may contain sources of irrelevant difficulty that disadvantage members of subgroups of the population. To continue with the above **example**, the understanding of a concept in **geometry**, and the ability to use that concept in problem **solving**, can be distinguished from the **ability** to write **reflectively** about **one's understanding**. Much of performance **assessment**, however, is dominated by precisely this kind of expressive and reflective **ability**. (To be **sure**, such abilities are important in their own **right**, but to the extent possible they should be distinguished from other **abilities**.) A given teacher can **often** write eloquently about what she or he wants students to learn and about "**constructivist approaches**" to **teaching**, but the actual behavior before the classroom may belie this **eloquence**. The opposite is also **true**. In a research project co-directed by the first author involving the development of expertise in **aircraft engine mechanics**, the recognized shop aces were often unable to explain an assembly or tear-down procedures away from the shop. They insisted upon going to the job site and **demonstrating** the procedure in question.

In some subject matters, specification of the content **framework** would appear to be a relatively **straightforward**, if **time-consuming**, **matter**. The disciplinary consensus achieved in the National Council of Teachers of Mathematics Standards is a case in **point**. These standards represent the work of an incredibly diverse set of individuals from a variety of **philosophical**, **social**, and ethnic backgrounds who teach in widely different cultural, **social**, and economic **contexts**. Yet consensus on what constitutes desired mathematical **competencies** in students and desired practices in teaching mathematics was **achieved**. Whether a similar consensus **will** be achieved in other disciplines remains to be **seen**. History and social **studies**, to take what are perhaps the most obvious **examples**, are largely "**constructive**" **enterprises**, involving matters of interpretation that will **vary** as a function of **ethnicity**, **gender**, and social **class**. Consensus regarding what constitutes the very "stuff" of history and social studies is **problematical**, not to mention what constitutes evidence of "**understanding**" social and historical **phenomena**. To **repeat**, it is important to ensure that a diversity of viewpoints be represented in the initial specification of the content **framework**.

Selection **of tasks and exercises**. The unidimensionality of most paper-and-pencil measures of cognitive ability constitute both a strength and, ironically, an essential weakness. On the one hand, unidimensionality facilitates test use and the interpretation of test results. Whether the test

model underlying performance is classically-based or based upon **item-response-theory**, persons with the same "observed score" or "**theta**" are presumed to be approximately equal in the ability being measured. In performance assessment, however, the situation is **fundamentally different**. **Generalizability** studies in a wide variety of contexts indicate that the variance components due to exercises and the **exercise-by-examinee** interaction constitute significant portions of the total variance in **performance**. The situation is exacerbated by the fact that the number of exercises are usually much smaller in typical performance assessments than in standardized multiple-choice group tests. Thus, even if a "**general factor**" in complex performance could **emerge**, it is unlikely that it would be reliable given the number of exercises **involved**.

The **multidimensionality** of **performance** assessments stems not only from the inherent complexity of such **assessments**, but also from the very real effects of **context**. Posing a task one way, as opposed to **another**, or in one context as opposed to **another**, can result in surprisingly different **results**. Horizontally presented addition problems turn out to be significantly more difficult for young children than the **same** problems presented **vertically**. This finding is no doubt traceable in part to **instruction**, but it highlights the fact that early knowledge is **precarious**, and seemingly **trivial**, surface differences in exercise formats can have substantial effects on **performance**.

One implication of the relatively large variance components due to exercises and the exercise-by-candidate interaction in performance assessment is that *which* exercises are included **in** an **assessment**, and the wording and context of those **exercises**, becomes important and has perhaps more implications for fairness and equity than would be the case in standardized group tests.

Sensitivity Review. Assessment stimuli must, of **necessity**, be about *something*. Whether the assessment is a writing sample, an extended response to a practical problem in **mathematics**, or a portfolio **entry**, the actual exercise will contain words that describe situations encountered in our **culture**. Sound testing **practice**, not to mention simple common **sense**, requires that assessment materials not include **wording**, **references**, or situations that are offensive and insensitive to **test-takers**. "**Sensitivity review**" is a generic term for set procedures for ensuring (1) that stimulus materials used in assessment reflect the diversity in our society and the diversity of contributions to our **culture**, and (2) that the assessment stimuli are free of wording and/or situations that are **sexist**, **ethnically insensitive**, **stereotypic**, or otherwise offensive to subgroups of the **population**. Sensitivity review is becoming a routine part of many large-scale test development **agencies**. Ramsey (1993) provides an excellent review of sensitivity review procedures at **ETS**, a leader in this area.

Selection and training of administrators and scorers. Sound administration and scoring of performance assessments are much more difficult to achieve in performance assessment than in multiple-choice **tests**. In the **latter**, standardization of *instructions* in administering the test is presumed to be a part of the definition of **fairness**, and standardized scoring is assured by a **machine**. In performance **assessment**, however, the selection and training of administrators and scorers are critical features of the overall validity of the **assessment**. The necessity for a diverse group of administrators and scorers has already been **mentioned**. In **addition**, however, close attention must be paid to administrators and **scorers**, and their **biases**.

The objective of assessment should be not so much the standardization of *instructions*, as ensuring that **examinees** have a common understanding of the tasks **involved**. Because

performance assessment **is**, or can **be**, richly **interactive**, it is vitally important that administrators not only understand the constructs being **assessed**, but it is essential that they know how to discern when an **examinee** does not understand what is being asked and what kinds of additional explanation is **needed**. Regarding **scorers**, it is surprising how many people **continue**, even after **training**, to equate and confuse “**writing ability**” with “**understanding**.” Many people find misspellings so annoying that it is impossible for them fairly to ignore this construct-irrelevant ability in essays where the intent is understanding a **discipline**, rather than **spelling**, *per se*. It **is**, **moreover**, often impossible to “**train out**” personal biases some persons hold about the abilities of boys **vs.** girls, or blacks **vs.** whites.

The emphasis in training should not be restricted solely to eliminating sources of construct-irrelevant difficulty for minority group **members**. It may also happen that assessment exercises contain construct-irrelevant elements that specifically *advantage* certain **individuals**. The “**neatness**” and visual attractiveness of a **portfolio**, for **example**, are related to the resources available to a student or **teacher**, but are typically unrelated to the ability being **measured**. **Yet**, despite **training**, these factors **often** figure unconsciously in **scoring**. The bottom line is that often the thought and hard work that go into the specification of the content framework and into the construction of high quality exercises that are tied to that **framework**, can be rendered useless by poorly chosen or inadequately trained **scorers**.

Bias in performance exercises. The sophisticated statistical machinery that has been developed for the identification of biased items in paper-and-pencil tests cannot unfortunately be applied in a straightforward manner to performance **assessment**. For **example**, all of the techniques for identifying differentially **functioning** items by subgroups using item response theory (Holland & Wainer, 1993) require large **numbers** of **examinees** and more items than are typically found in **performance assessment**. The **same** applies to **classically-based** procedures such as the Mantel-Haenszel technique (Holland & Thayer, 1988), and the standardization method (Dorans & Holland, 1988).

A repeated and rather baffling finding in studies that attempt to **identify** patterns of DIF in standardized aptitude and achievement measures is that items identified as functioning differentially do not appear to differ in any discernible way from items not so **identified**. O'Neill and McPeck (1993) have attempted to **summarized** patterns of DIF for the Scholastic Aptitude Test, and although some of their conclusions are **encouraging**, the general rule is that patterns of DIF traceable to substantive characteristics of the items or item-formats are **difficult** to come by. **Moreover**, the connection between DIF (as a technical characteristic of test items defined by reference to internal properties of the **test**), and *test bias* (a much larger issue involving the opportunity to **learn**, among other **things**) is not at **all clear**. Space does not allow a thoroughgoing discussion of this complex **issue**. For an incisive discussion of the relation between *item bias* (a term that predated DIF by some **35 years**), DIF, and test bias broadly conceived, see Camilli Shepard (1994).

Absent any agreed-upon methodology for the systematic **examination** of sources of bias in **performance assessment**, we must rely upon expert judgment and a close adherence to sound assessment **practices**. One of the significant promises of performance assessment is that the sources of any biases in the **assessment**, as well as the sources of any genuine deficiencies of **examinees**, are more likely to be **discovered**. Unlike multiple choice **test**, where the only

permanent record is a series of blackened ovals by the **examinee**, performance assessment allows at least some insight into an **examinee's thinking**.

External Sources of Bias in Performance Assessment

In many applied uses of **tests**, the relationship between test performance and some external criterion is **crucial**, and the extent to which test scores of various social groups are *differentially* related to the external criterion becomes a potential source of **bias**.

The behaviors elicited by traditional group-administered paper-and-pencil tests are often different from the behaviors the test is intended to **predict**. The differences are quite apparent in such applications as certification and **licensure** and in employment **contexts**. In the performance of their professional legal and accounting **duties**, for **example**, lawyers and accountants are virtually never required to perform the kinds of tasks typically found on group administered **multiple-choice**. Nor are the circumstances under which the tests are administered reflective of the circumstances under which their actual duties are **performed**. (Most lawyers, for **example**, would never decide on a specific legal strategy without extensive prior **research**).

In the context of student **assessment**, much of the criticism of traditional paper-and-pencil measures of academic achievement derives from the **impoverished, decontextualized** nature of the stimulus **questions**. Their predictive validity for future academic performance **notwithstanding**, verbal analogies are not what teachers of English Language Arts should be **about**. It is entirely likely that excellent instruction in English Arts instruction coupled with performance assessment (**which** properly viewed are **complementary**) would result in students who do well on verbal **analogies**, but that is a derivative side **effect**, rather than an object of **instruction**. One point appears **certain**: to the extent that instruction and assessment become merged into one coherent **process**, performance assessment has the potential to remove one major criticism of current testing **practice**, to **wit**, that biases in the prediction of external criteria are more likely when the relationship between **teaching, testing**, and later performance is only **implicit**.

Fairness and Equity

An unbiased **assessment**—one that is equally valid for all relevant **subgroups**—is, of course, a minimal requirement for fairness. Here we move beyond questions of bias and validity to raise questions about the fairness of the educational system in which the assessment is **used**. Many of the state and national assessment-based reform efforts reflect high expectations about the extent to which assessments can facilitate excellence and equity in **education**. These expectations must be **carefully** and continually evaluated as assessments are designed and **implemented**. Here, we will focus on differences between groups varying in social **background**—differences in **performance** on the **assessments**, differences in opportunity to learn the material assessed as well as other valued goals of **education**, and difference in the consequences of assessment **use**. In **particular**, we focus on differences between **underserved** minority groups and the majority group to which they are **compared**. By **underserved** minority groups, we refer to students from poor communities and from **black, Hispanic, and American Indian** racial/ethnic **backgrounds**. We characterize these groups as **underserved** because of the large body of evidence (e.g., Oakes, Gamoran and Page, 1992) that documents inequalities in access to high quality education—an issue to which we will return in the section on opportunity to **learn**.

Differential Performance

A large body of evidence shows **that**, on the **average**, students from undersexed minority groups-those from poor communities and from **black**, **Hispanic**, and American Indian racial/ethnic backgrounds-have scored lower than students from the relevant majority comparison group on multiple choice tests across a wide range of assessment purposes involving students in **elementary**, **secondary**, post-secondary and vocational education (**National Commission on Testing and Public Policy**, 1991). This difference in performance between socially defined groups is often referred to as the "**achievement gap**." Some educators have theorized that performance assessments will narrow the achievement gap because they **sample** a broader range of **capabilities**, provide **examinees** with the opportunity to explain their **responses**, **permit** them more latitude in choosing tasks that best represent their **capabilities**, **and**, in the case of hands on **assessment**, rely less heavily on language **skills** (see Office of Technology Assessment, 1991; Hambleton and Murphy, 1992). Others raise concerns that these unfamiliar assessment **formats** may disadvantage students when compared to the more **familiar** multiple choice **format** (see Badger, in press; Office of Technology Assessment, 1991). These expectations must be **carefully evaluated**.

Relevant evidence is scanty and **mixed**. (Baker and O'Neill, in press; Bond, 1992, Office of Technology Assessment, 1991; Darling-Hammond, 1994; Linn, 1993; Linn, Baker, and Dunbar, 1991; Madaus and Kelleghan, 1992; Madaus, 1994; in press; National Commission on Testing and Public Policy, 1990; Nettles, in press; Winfield and Woodard, 1994.) Typically, differences in performance between groups are compared for traditional and alternative assessments in **terms** of effect size or by comparing the proportions passing (**or** scoring above or below some cut **score**). Effect sizes are computed by subtracting the mean for the minority group of concern from the mean for the majority comparison group and dividing by the standard deviation of the majority comparison **group**. A narrowing of the "**achievement gap**" would be indicated by a smaller effect size or greater equivalence across groups in proportions passing when **performance** assessments are compared to more traditional **assessments**.

Some studies have found no difference in the achievement gap between **performance** assessments and more traditional **assessments**. For instance, Linn, Baker and Dunbar (1991), using results from the 1988 NAEP, found that differences in achievement between black and white students in grades **4**, **8** and **12** were essentially the same for the open-ended essay items in the writing assessment and the primarily multiple choice items in the reading **assessment**. Other studies have found a narrowing of the achievement **gap**. For instance, Badger (in press), using results on open-ended and multiple choice items from the Massachusetts testing **program** found smaller performance gaps for open-ended items when comparing students in **low-** and **high-SES** **schools**, and when comparing black and Hispanic students to white and Asian **students**. Her results were consistent across grades and subject **areas**. A comparison of the magnitude of differences between black and white military personnel on paper and pencil (**Armed Services Vocational Aptitude Battery (ASVAB)** and other job knowledge **tests**) versus hands on tests resulted in smaller differences for the hands on tests (**Wigdor & Green**, 1991). Others have found an increase in the achievement **gap**. Elliott (1993), comparing differences in performance for white and Asian students to those for black and Hispanic students on multiple choice and constructed response items for the 1992 NAEP in mathematics found a larger gap for constructed response **items**. LeMahieu (1992) reported larger differences between black and white students when portfolio **scores**, reflecting self-selected pieces of **writing**, were compared to

an independent measure of writing. LeMahieu attributes the differences to **self-selection**: black students did not choose material from their writing folders that best represented their writing.

Given the existing **evidence**, there is no reason to believe that differences observed with traditional assessments between **underserved** minority groups and the majority comparison group will necessarily diminish with performance-based **assessment**. To offer any other generalization based on the existing evidence would be **inappropriate**. The studies differ widely in **terms** of the purposes of **assessment**, the **constructs assessed**, the **format** in which performances are presented and **evaluated**, the antecedent instructional **conditions**, and the social and academic characteristics of the students **assessed**. The only appropriate advice is to highlight the crucial importance of investigating and then attempting to understand the reason for group differences for *each* context of **assessment**.

Careful attention must be given to the way in which group differences are reported and **interpreted**. Reporting simple differences in performance between schools or **districts**, between racial/ethnic **groups**, between students from poor and wealthier families, without additional information to assist in *explaining* the **differences**, may result in serious **misinterpretations**. For **instance**, changes in assessment scores from year to year may simply reflect changes in the student population (**dropouts** or **transfers**, for **instance**) rather than changes in the capabilities of **students**. Differences in assessment scores across ethnic groups may reflect differences in socioeconomic status of the **communities** in which they **live**. Differences in assessment scores from school to school may reflect differences in resources and activities such as the qualification of teachers or the **number** of advanced course **offerings**. Of most serious concern here is the potential misinterpretation of differences between racial/ethnic **groups**. Lee cautions (**personal communication**, April 14, 1994) that providing racial breakdowns may "**overestimate** the importance of racial differences in academic achievement . . . because they are confounded with uncontrolled social class **differences**"; this risks misinforming the nation by "**allowing** people to conclude that it is only race (and not **poverty**) which is driving these **differences**". Haertel (1989) similarly notes that controls for **socio-economic** status reduce apparent disparities based on **race/ethnicity**. **Moreover**, differences in achievement between **underserved minority** and majority groups (**whether** defined by **race/ethnicity** or **socio-economic status**) can be substantially explained by differences in access to high quality education (Darling-Hammond, 1994; Barr and Dreeben, 1983; Lee and Bryk, 1988, 1989; Oakes, 1985; Oakes, Gamoran, and Page, 1992). As Darling-Hammond and Ascher (1991) note, "**comparisons** of test scores that ignore these factors hold little promise of directing policy **makers'** attention to the real sources of the **problem**, so that it can be **rectified**" (p.16). This is the issue to which we turn **next**.

Opportunity to Learn and Other Antecedent Conditions

All of the policy research we reviewed appropriately places a high **premium** on providing a "**level playing field**" for students and **schools**. In the context of **assessment**, there are at least four reasons why it is important to provide **information** about opportunity to **learn**. **First**, when assessments are used for high stakes **decisions**, such as promotion for **graduation**, it is a legal and ethical responsibility to ensure that students have had the opportunity to acquire the capabilities for which they are held **accountable**. **Second**, regardless of whether the stakes of assessment are high or **low**, such information is essential both for understanding the results and for directing attention to needed reforms in policy and **practice**. **Third**, it must be recognized that assessments not only document the success of learning **opportunities**, they constrain and enable **future**

learning opportunities (an issue to which we will return in the section on differential consequences). And so, when assessments are introduced into the system, it is essential to study both the antecedent and consequent conditions of their use. Finally, the issue of “opportunity to learn” is no longer solely a measurement or even an ethical one, but a matter of legal precedent. The most directly relevant case law is, of course, *Debra P. vs. Burlington* (730 F.2d 140 11th Cir.1984), in which the court held that a state’s minimum level of performance on a standardized competency test as a prerequisite for a high school diploma is proper only if the assessment tested what was actually taught in the schools. The Due Process Clause of the Fifth Amendment, Title VI of the Civil Rights Act of 1964, Title IX of the Education Amendments of 1972, and Section 504 of the Rehabilitation Act of 1973 are implied if student assessments have an adverse impact on students because of their race, national origin, gender, or disability. If, for example, a test results in substantially different rates of eligibility for educational benefits or services on the basis of a student’s group membership, then the use of that test may be in violation of civil rights laws and statutes.

Unquestionably, differential learning opportunities play a major role in determining student achievement (Barr and Dreeben, 1983; Gamoran, 1987; Lee and Bryk, 1988, 1989; Oakes, 1985; Oakes, Gamoran, and Page, 1992). Moreover, there is evidence of social stratification-differences attributable to race/ethnicity and socioeconomic status-in the learning opportunities available to students and consequently in educational outcomes. Differences in access to high quality education occur within schools, in course taking patterns, for instance, as well as between schools in the resources they have available. Oakes, Gamoran, and Page (1992), summarizing literature on curriculum differentiation, report that “disproportionate percentages of poor and minority students (principally black and Hispanic) are found in curricula designed for low ability or non-college bound studentsFurther, minority students are consistently underrepresented in programs for the gifted and talented . . . Although it seems clear that consistently lower performance on tests by non-Asian minorities under girds this pattern, the result is that curriculum differentiation leads to considerable race and class separation and race- and class-linked differences in opportunities to learn” (p.58x, 590). Darling-Hammond (1994) summarizing evidence on the social distribution of qualified teachers notes that “because teacher salaries and working conditions are inadequate to ensure a steady supply of qualified teachers in poor districts, low-income and minority students are routinely taught by the least experienced and least prepared teachers” (Darling-Hammond, 1994, p.16).

No consequential decision about individual students or educational programs should be made on the basis of a single assessment. With respect to individuals, additional information about the student’s achievement and past experience should be considered. As Haertel (1989) notes in his paper for the National Commission on Testing and Public Policy, “simplistic policies, where action is triggered by scores above or below a cutting point on a single test. . . are contrary to the consensus of professional practice in testing.” (p.32). At the program level, assessment scores should be part of a comprehensive system of indicators, that includes information about student characteristics and about school activities and resources over which policymakers have some control.

While it is beyond the scope of this chapter to provide specific advice about how to design an informative and equitable indicator system, we highlight some general guidelines from policy researchers and direct readers to the original articles for more specific advice about how to operationalize their suggestions. We note, as does Oakes (1989), that there are many obstacles to

overcome in developing standards and indicators of opportunity to **learn**, including insufficient measurement technology and **feasibility**. Porter (1993) raises the question of whether school delivery standards should be intended to inspire and to provide a vision for **educators, students, parents**, and the public of what good practice might **be**, or whether delivery standards should provide prescriptions of required practice that can be used to police the action of **teachers, school administrators**, and public **officials**. He suggests distinguishing between school process **standards**, which would provide the **vision**, and school performance **standards**, which would support school **monitoring**, thus meeting accountability requirements for ethical and legal **purposes**. Our goal here is simply to suggest some possible frameworks for studying these issues.

Some of the advice on documenting opportunity to learn focuses on whether students have had the opportunity to **learn** the material *on which they are tested*. Mehrens and Popham (1992), focusing on legal **defensibility**, suggest gathering the following evidence regarding students opportunity to learn the material covered by a particular test or **assessment**:

(1) Self report from teachers regarding the extent to which instruction for tested content was **supplied**....(2) Self reports from students regarding whether the students have been taught 'how to answer the illustrated kinds of test **questions**....(3) Analysis of required textbooks to determine if test content is **addressed**....(4) Analyses of curricular documents such as course syllabi to establish whether tested content is supposed to be addressed by **teachers**....(5) Analyses of teacher's classroom tests to discern if the content contained in the high-stakes examination is also addressed in **teachers'** routine classroom **tests**. (p. 275-276).

They advise against use of classroom observation because the expense and the difficulty of obtaining an adequately representative sample to permit inferences about all the tested **material**. While this advice addresses the opportunity to learn the tested **material**, it does not address the quality of education more broadly **nor**, with the possible exception of examining teachers **tests**, does it address the way in which the material is taught-an issue crucial to performance assessments that encourage students to interpret and evaluate material for **themselves**. (See Pullin, 1994, for a more extensive review of the legal issues involved in current national educational reform **proposals**.)

Moving beyond the content of tested **material**, the recently passed goals 2000 legislation offers the following guidelines for elements to be included in voluntary national opportunity to learn standards

- the quality and availability to **all** students of **curricula**, instructional **materials**, and **technologies**, including distance **learning**;
- the capability of teachers to provide high-quality instruction to meet diverse learning needs in each content area to **all students**;
- the extent to which **teachers, principals**, and administrators have ready and continuing access to professional **development**, including the best knowledge about **teaching, learning**, and school **improvement**;
- the extent to which **curriculum**, instructional **practices**, and assessments are aligned to voluntary national content **standards**;

- the extent to which school facilities provide a safe and secure environment for learning and instruction and have the requisite **libraries, laboratories**, and other resources necessary to provide an **opportunity-to-learn**; and
- the extent to which schools utilize **policies, curricula**, and instructional practices which ensure nondiscrimination on the basis of **gender**. (Goals 2000 Legislation, Sec. 213(c)(2)(A-F)).

Factors that effect opportunity to learn occur at a variety of **levels**. At the school **level**, factors such as **financial resources**, qualification of **teachers**, curriculum **offerings**, **teachers' work loads**, materials resources such as **libraries, laboratories**, and **computers**, and so **on**, should be **investigated**. **However**, equality of opportunity cannot be adequately addressed by simply looking at school level resources and **assuming** the students within schools have equal access to those **resources**. Substantial variation in access to resources exists within schools and even within classes that school level indicators can't observe (**e.g.**, Barr and **Dreeben**, 1983; Lee and **Bryk**, 1989; **Oakes**, 1989; **Porter**, 1993).

Oakes (1989) notes that school resources are mediated by organization policies and structures as well as by the school culture that is reflected in norms and **relationships**.

Even though we do not **fully** understand how schools produce the results we **want**, context **information** may provide clues to **policymakers** about why we get the outcomes we **do**. Measures of what goes on in schools can add important **information** to the political discussion about how to **improve** them (**Oakes**, 1989, p.182).

Similarly, **Porter** (1993) notes that evidence of opportunity to learn must include evidence of effective **pedagogy**. With **performance** assessment, this covers instructional strategies such as "a need for instruction to emphasize active student **learning**, where students take responsibility for constructing knowledge through **writing, discussion, lab work, manipulatives**, and computer simulations" (p.26). **Moreover**, knowledge of school **resources**, such as libraries and labs must be accompanied by evidence of course specific resources indicating access for **all** students in **all** subjects.

Oakes (1989) suggests gathering **information** about the following sets of "**context indicators**" to monitor schooling resources and **processes**. *Access to knowledge* refers to the "**extent** to which schools provide students with opportunities to learn various domains of knowledge and **skills**" (p.186). Indicators might **include**: teacher **qualifications**, instructional **time**, course **offerings**, class grouping **practices**, **materials/laboratories/equipment**, academic support **programs**, enrichment **activities**, parent **involvement**, staff **development**, and faculty **beliefs**. *Press for achievement* refers to the "**pressure** which the school exerts to get students to work hard and **achieve**" (p.186). Possible indicators **include**: focus on **academics**, graduation **requirements**, graduation **rates**, enrollment in rigorous **programs**, recognition of academic **accomplishments**, academic expectations for **students**, uninterrupted class **instruction**, administrative involvement in **academics**, quantity and type of **homework**, and teacher evaluation emphasizing **learning**. *Professional teaching conditions* are the "**conditions** that can empower or constrain teachers and administrators as they attempt to create and implement instruction **programs**" p.186). **Here**, potential indicators include teacher **salaries**, pupil load/class **size**, teacher time for planning **collegial work**, teacher involvement in decision **making**, teacher **certainty**, teacher **autonomy/flexibility**, administrative support for **innovation**, and clerical **support**. Clearly, to

address the equity **question**, these indicators must be **examined** to see whether they are equally distributed across all groups of **concern**.

Factors that exist outside of school influence students readiness to take advantage of the opportunities school affords **them**. These include factors such as **health, nutrition**, living conditions, family support for **education**, educational resources in the **home**, and so on (Lee and Croninger, in press; Madaus, 1994, in press; Koretz, et al., 1992). Also of concern are differences between the cultural context of the home and the **school**, which are too **often** viewed as deficits rather than differences (Garcia and Pearson, 1994; O'Connor, 1989; LaCelle-Peterson and Rivera, 1994). Darling-Hammond (1994), speaking in the context of **assessment, notes**, "the choice of **items**, responses deemed **appropriate**, and content deemed important are the product of culturally and contextually determined **judgments**, as **well** as the privileging of certain ways of knowing and modes of **performance** over **others**. (p. 17; See also, Garcia & Pearson, 1994; Gardener, 1983; O'Connor, 1989; Steinberg, 1985)." (Darling-Hammond, 1994, p. 17) Gordon, while **affirming** his **commitment** to standards of competence that are the same for all **populations**, highlights the importance of allowing differential indicators of progress toward those **standards**: "The task is to find assessment probes which measure the same criterion from different contexts and perspectives which reflect the life space and values of the learner ... Thus options and choice become critical features in any assessment system created to be responsive to **equity**" (Gordon, 1992, pp. 5-6). These issues concerning potential disjunctions among cultural contexts are equally relevant to any school-based learning or assessment **activity**.

Frameworks for more comprehensive indicator systems that tie together background **information, information** about school context and **processes**, and **information** about valued learning outcomes have been suggested by Darling-Hammond and Ascher (1991), Porter (1991, 1993) and Bryk and Hermanson (1993). Porter, drawing on Shavelson (1987, in Porter, 1991), suggests a comprehensive model organized around **inputs, processes**, and outputs of **education**. Inputs include fiscal and other **resources**, teacher **quality**, student **background**, and parental/community **norms**. Processes are divided into organizational and instructional characteristics of **schooling**. Organization characteristics of schooling include **national, state, district**, and school **quality**; instructional characteristics include general **curriculum/content** and teaching/pedagogy **quality**, student nonacademic **activities**, course specific teacher **quality**, and course specific **resources**. Outputs include **achievement, participation**, and attitudes and **aspirations**.

Bryk and Hermanson (1993), drawing on the work of the Special Study Panel on Education Indicators (NCES, 1991, in Bryk and Hermanson, 1993), caution that inputs-processes-outputs models risk conveying a false sense of control based upon external **manipulation**. They suggest that a comprehensive indicator system should focus *on the* following "**enduring concerns**" in **education**, the purpose of which is to inform public discourse about the means and ends of **education**:

Clearly the system must focus on *student learning* and the *quality of its educational institutions*. The system **should**, as **well**, report on the larger social context that also **educates**. This implied a concern for *children's readiness to learn as they enter the formal educational system* **and**, in more general **terms**, *societal support for learning*. It must also **inform** us about the broader aspirations we hold for education in a modern democratic

society. This meant a dual focus on *educational equity* and the contributions of education to *economic productivity*. (Bryk and Hermanson, 1993, p. 468).

They go on to suggest that the system should use a number of different reporting and data collection levels.

As a trend in a particular indicator engages our **attention**, it should naturally lead to more detailed statistical **information**, including in-depth studies and case analyses that might **illuminate** some of the forces at work The system should have a strong conceptual **organization**, capturing both established means-ends generalizations from social science and the best clinical **expertise**. (p. 468).

As a comparison of these more comprehensive indicators systems **suggest**, one of the major issues underlying assessment-based reform efforts is the way in which information is expected (and intended) to influence the educational **system**. That is the issue to which we turn **next**.

Differential Consequences²

It is widely hoped that performance assessments will not only permit valid inferences about the quality of education but serve as instruments of reform by raising standards for all **students**, thus promoting excellence and **equity**. Given our past experience with high stakes assessment and the current state of our knowledge about the construct validity of performance **assessments**, this anticipation is optimistic at best—the **assumptions** about the quality of information and the consequences of the assessment must be **carefully evaluated**. In understanding and evaluating **consequences**, it is important to look not just at the format of the assessment (e.g., multiple choice versus performance **based**) but also at the way in which the assessment is used to promote reform.

A growing body of evidence indicates that when assessments are visible and have consequences for individuals or **programs**, they alter educational **practice**, sending an unequivocal message to teachers and students about what is important to teach and learn (e.g., Johnston, Weiss, and Afflerbach, 1990; National Commission on Testing and Public Policy, 1989; Resnick and Resnick, 1992; Smith, 1991). In a review of literature on the impact of classroom evaluation on **students**, Crooks (1988) concluded assessment not only affects students' judgments about what is important to **learn**, but also their **motivation**, perceptions of competence, approaches to personal **study**, and development of enduring learning **strategies**. Similar conclusions have been drawn about the impact of district and state mandated assessment on the **judgment**, **perceptions**, and instructional strategies of **teachers**. The salience of this influence seems to be directly related to the importance of the consequences of testing to students and teachers and to the administrative and supervisory practices of a school or **district**. In a paper prepared for the National **Commission** on Testing and Public Policy, Resnick and Resnick concluded that "when the stakes are high—when **schools'** ratings and budgets or **teachers'** salaries depend on test scores—efforts to improve **performance** on a particular assessment seem to drive out most other educational **concerns**" and "to progressively restrict curricular attention to the objectives that are tested and even the particular item forms that will appear on the test." (1992, p. 58).

Evidence suggests that the narrowing of the curriculum associated with high stakes standardized assessment may be falling disproportionately on certain groups of students for

whom concerns about equality of education have been most **salient**. Neill and Medina (1989) found that standardized testing was more prevalent in large urban school **systems**. Madaus, West, Harmon, Lomax, and Viator (1992, in Madaus, in press) report national survey results showing that teachers with greater than 60 percent minority students in their classrooms were more likely to teach to standardized **tests**, to spend time in direct test **preparation**, and to spend more of their class time on these activities than were teachers in predominately white **classrooms**. Similar results were found for teachers of students in Chapter 1 **programs**. Herman and colleagues (Herman and Golan, 1993; Dorr-Bremme and Herman, 1986), using **teachers'** and **principals'** **self-reports**, found that in low income **communities**, teachers felt a greater need to spend time preparing students for tests and principals felt that tests counted **far** more in decisions such as planning **curriculum**, making class **assignments**, allocating **funds**, and reporting to district officials and the **community**. To the extent that testing **undergirds** decisions about educational **placement**, studies on the effects of tracking reviewed by Oakes, Gamoran, and Page (1992) also support concerns about the differential impact of testing on **underserved** minority **students**. They report that qualitative differences exist in the educational experiences provided students in different **tracks**, with lower **track** students progressing more slowly through the **curriculum**, having less experience with inquiry **skills**, problem **solving**, and autonomy in their **work**, and losing more educational time to classroom **management**; that the achievement gap between students in higher and lower tracks increases over years of **schooling**; and that track placement can have a long lasting impact on the life chances of students after high **school**. Taken together with our knowledge about the impact of high stakes testing on the **curriculum**, these observations raise substantial concerns about differential access to knowledge for students from **underserved** minority **groups**.

It is anticipated that performance assessment can overcome such influences by providing targets toward which we want teachers to **teach**. **Clearly**, providing a broader range of valued educational outcomes is essential to address concerns about narrowing the **curriculum**, as most advocates of performance assessment have **argued**, as is a **careful** consideration of what outcomes remain **unaddressed**. **However**, past experience with high stakes uses of multiple choice tests suggests the need for **caution**. Although high-stakes testing programs frequently result in improved test **scores**, such improvement does not necessarily imply a rise in the quality of education or a better educated student population (Darling-Hammond and Snyder, 1992; Haertel, 1989; National Commission on Testing and Public Policy, 1990; Shepard, 1992). At **best**, test scores can reflect only a small subset of valued education **goals**. When educators focus their attention on improving test **scores**, they not only narrow the **curriculum**, they undermine the validity of the tests as indicators of a broader range of achievements (Shepard, 1992). Koretz, Linn, Dunbar, and Shepard (1991, in Linn, 1993) showed evidence of test score inflation comparing performance in two districts between the district mandated tests in reading and math and tests constructed to cover the same content **objectives**. Students in the **high-stakes** testing districts scored considerably lower on the alternative tests in reading for both districts and in mathematics for one **district**. When particular tests become targets for instruction they become less valid indicators of the broader capabilities they were intended to **tap**. As Madaus (in press) **notes**, "there is no evidence to support the belief that performance-based measures will not be as corruptible as any multiple choice measure when used in the context of measurement driven instruction" (p. 41).

Further, evidence suggests that test driven reforms may undermine attempts at genuine educational reform by diverting attention from **fundamental** educational **problems**. Ellwein, Glass, and Smith (1989) conducted extended case studies of five competency testing programs at the state and district **level**. They concluded that competency tests and standards served more as symbolic and political gestures rather than as **instrumental** reforms-focusing attention on **the** tests themselves rather than on their **impact, utility, or value**. Similarly, Corbett and Wilson (1992), who studied competency testing programs in two **states**, focusing on six districts per state, found that the pressure to do well on tests did not encourage **fundamental** consideration of the **structures, processes, or purposes of education**, rather it caused "knee-jerk" reactions designed to improve test scores **quickly—actions** which many of the educators involved considered **counter-productive**.

When **performance** is tied to rewards and **sanctions**, including public **disclosure**, the system sometimes results in peculiar and counter-productive incentives to improve the appearance of **progress**, such as retaining students so that their scores **will** be compared with a younger **cohort**, placing students in special education programs so that their scores don't **count**, targeting instruction at those closest to the passing level and ignoring the needs of students unlikely to pass the **exam**, or failing to discourage low-scoring students from dropping **out**. (e.g., Haertel, 1989, Madaus, in press; Slavin and Madden, 1991). Such practices exacerbate inequities in educational **opportunity**.

Some suggest reporting information in ways that will **minimize** the negative effects of high-stakes standardized **assessments**. Haertel (1989) suggests preparing reports that attend to the entire distribution of achievement (e.g., including the **25th, 50th, and 75th percentiles**) so that progress is not just defined in terms of increases in measures of central **tendency**. He also suggests tracking the population of **students—counting** dropouts and **transfers—and**, where **possible**, looking at the progress of individual **students**. Aggregate indices of individual growth are better indicators of program effectiveness than are differences in average scores which may simply reflect differences in the population of **students**. Moreover, they are less likely to confound program effectiveness with incoming capabilities of **students**. Lee (**personal communication, April 14, 1994**) notes that even in the absence of longitudinal achievement **data**, it is possible to control for academic background using proxy measures such as self reports of **students' previous grades** and whether or not students had repeated or skipped a grade **level**. And, as we noted **before**, interpreting educational outcomes in light of school resources and processes is important in controlling misinterpretations leading to counter-productive **policies**. (Darling-Hammond and Ascher, 1991; Darling-Hammond and Snyder, 1992; Haertel, 1989; Oakes, 1989; Porter, 1991, 1993).

However, some researchers argue that simply changing the **format** and reporting requirements of assessments is **insufficient; rather**, what is needed is a change in the way assessments are used to promote reform (Bryk and Hermanson, 1993; Darling-Hammond, 1994; Madaus, 1993, 1994, in press). Darling-Hammond (1994) raises the concern that, "if performance assessments are used in the same **fashion** as current externally developed and mandated tests are **used**, they are likely to highlight differences in **students' learning** even more **keenly**, but they **will** be unlikely to help teachers revamp their teaching or schools rethink their ways of **operating**" (p.19). She contrasts two different approaches to assessment **reform**, which reflect different theories of organizational change and different views of educational **purposes**. One view seeks to induce change through extrinsic rewards and sanctions for both schools and

students The other view seeks to induce change by building knowledge among school practitioners and parents about alternative methods and by stimulating organization rethinking through opportunities to work together on the design of **teaching** and schooling and to experiment with new **approaches**. (Darling-Hammond, in press, p.114).

The recently passed Goals 2000 legislation suggests **purposes** for assessment that appear intended to increase the control of externally imposed assessments on teaching and **learning**. Its authors indicate that **among** the appropriate purposes for state assessments are "**measuring** and motivating individual **students, schools, districts, state**, and the nation to improve educational **performance**" (Sec. 213, (f)(1)(B)(iv)) and that **after** a period of five **years**, such assessments may be used "**to make decisions regarding graduation, grade promotion, or retention of students**" (Sec. 213, (f)(1)(C)(II)). In contrast, Darling-Hammond describes proposals in states like New York, Vermont, Connecticut, and California, where assessments used for policy purposes are distinct from those used for individual **students**.

"**These** states envision carefully targeted state assessments at a few key developmental points that will provide data for **informing** policy makers about program successes and **needs**, areas where assistance and investment are **needed**, and assessment models for local **schools**. **Meanwhile**, locally implemented assessment systems—including **portfolios, projects, performance tasks**, and structured teachers observations of learning—will provide the multiple forms of evidence about student learning needed to make sound judgments about **instruction**." (Darling Hammond, 1994, p. 20)

As research into the consequences of high stakes assessment **suggests**, choices made in designing assessment systems not only impact the nature of teaching and learning (**in both intended and unintended ways**) but also the nature of the discourse about the purposes and processes of **education**. Bryk and Hermanson (1993) offer a useful distinction between two different views of the ways in which indicators enter and influence the discourse and practice of educational **reform**: an "**instrumental use**" model and an "**enlightenment**" model. In "**the instrumental use**" model, the goals **are**: to develop a comprehensive set of outcome **measures**; to examine the relationship between these outcomes and indicators of school resources and **processes**; and, based upon that generalized **knowledge**, to control schools through allocation of **resources**, rewards and **sanctions**, and regulation so as to maximize **performance** on the **outcomes**. As they **note**, the instrumental use model characterizes much of the current rhetoric about the potential of indicators to improve **schools**. In criticizing this **conceptualization**, they argue first that there are many valued outcomes for which available measures do not **exist**. As our past experience **suggests**, any model which attempts to **maximize** measurable outcomes is likely to result in a variety of **unintended**, possibly **undesirable, effects**, including the **undermining** of progress in areas not **addressed**. More **fundamentally**, the instrumental use model, with its reliance on generalizations about the relationship between processes and **outcomes**, under-represents the complexity of **schools**. While "**external** policy-making and administrative action shape **schools' structure and function**" (p. 453), the "**behavior, attitudes, and beliefs of actors inside the school-professional staff, parents, and students-influence its operations**" (p. 453). "**Schools are places where personal meaning and human intentionality matter.**" (p. 457) An "**enlightenment model**" reflects a view of schools where interaction **among** individuals is fundamental and **reform** requires "**changing** the values and tacit understandings that ground these **interactions**". From this **perspective**, the goal of an indicator system is not to manipulate such interactions through external **controls**, but rather to "**enrich** and encourage

sustained conversation about **education**, its **institutions**, and its processes in order ultimately to improve **them**" (p.467).

Notes

¹The authors took responsibility for different aspects of the **paper**. Bond and Carr drafted the section on "**Fairness and Bias**;" Carr provided a discussion of legal issues that informed both sections of the **paper**; Moss drafted the introduction and the section on "**Fairness and Equity**;" and Bond coordinated the **paper**.

²This section draws heavily on Moss (**in press**).

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