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ABSTRACT

Reliability is one of the chief characteristics researchers consider when judging the quality of data used in their studies. Within the positivist paradigm, data are typically quantified, and thus it is relatively easy to derive estimates of reliability. Within the interpretivist paradigm, however, the idea of data reliability is a looser science. This paper makes the case that the positivist and interpretivist paradigms are not as disparate as many suppose in terms of conceptualizations of reliability. A variety of methods for assessing the reliability, or trustworthiness, of qualitative data are reviewed, including the important process of triangulation. Terminology appropriate to specific data features that affect reliability is compared across the positivist and interpretivist paradigms. (Contains 1 table and 31 references.) (Author/SLD)

Running Head: RELIABILITY AND QUALITATIVE DATA

Reliability and Qualitative Data: Are Psychometric Concepts Relevant Within an
Interpretivist Research Paradigm?

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Abstract

Reliability is one of the chief characteristics researchers consider when judging the quality of data utilized in their studies. Within the positivist paradigm, data are typically quantified, and, thus, it is relatively easy to derive estimates of reliability. Within the interpretivist paradigm, however, the idea of data reliability is a looser science. In the present paper, we argue that the positivist and interpretivist paradigms are not as disparate as many suppose in terms of conceptualizations of reliability. A variety of methods for assessing the reliability, or trustworthiness, of qualitative data are reviewed, including the important process of triangulation. Terminology appropriate to specific data features that affect reliability are compared across the positivist and interpretivist paradigms.

Reliability and Qualitative Data: Are Psychometric Concepts Relevant Within an Interpretivist Research Paradigm?

Reliability is one of the chief characteristics researchers consider when judging the quality of data utilized in their studies. Within the positivist paradigm, data are typically quantified, and, thus, it is relatively easy to derive estimates of reliability based on various statistical indices developed for this purpose (Pedhazur & Schmelkin, 1991). In qualitative research, however, the idea of data reliability is a looser science, considering that the researcher serves as the instrument and that the researcher's understandings and interpretations serve as the data gathered with the "instrument." Consequently, some have argued that reliability of qualitative findings cannot (and should not) be estimated or assessed at all. In fact, many who advocate for the importance of an interpretivist research paradigm (e.g., Smith, 1984) refrain from using the term "reliability," fearing that the positivist framework of reliability will be considered as the standard against which all data integrity issues are conceptualized and assessed.

In the present paper, however, we argue that the positivist and interpretivist paradigms are not as disparate as many suppose in terms of conceptualizations of issues surrounding reliability. Logical connections between the two paradigms as regards reliability issues are discussed, and a list of terminology is presented to illustrate how 12 specific data features relative to reliability are addressed within the two paradigms.

Reliability Within the Interpretivist Paradigm

In qualitative research, information gleaned from observations, interviews, and the like must be “trustworthy” (Eisenhart & Howe, 1992; Lincoln & Guba, 1985); otherwise any themes that emerge from these data will not be credible. An important component of trustworthiness is “dependability” (Lincoln & Guba, 1985). Interestingly, dependability is analogous to reliability (Eisenhart & Howe, 1992; Onwuegbuzie, in press), and, the term was perhaps used originally by Cronbach, Gleser, Nanda, and Rajaratnam (1972) to refer to a rather classical/postivist view of reliability vis-a-vis generalizability theory. Onwuegbuzie (in press) identified 24 methods for assessing the trustworthiness of qualitative data. Many of these techniques can be utilized to assess the dependability or reliability of qualitative data extracted. Techniques for evaluating this dimension of trustworthiness include triangulation, which involves the use of multiple and different methods, investigators, sources, and theories to obtain corroborating evidence (Ely, Anzul, Friedman, Garner, & Steinmetz, 1991; Glesne & Peshkin, 1992; Lincoln & Guba, 1985; Merriam, 1988; Miles & Huberman, 1984, 1994; Onwuegbuzie, in press; Patton, 1990).

Triangulation reduces the possibility of chance associations, as well as of systematic biases prevailing due to a specific method being utilized, thereby allowing greater confidence in any interpretations made (Fielding & Fielding, 1986; Maxwell, 1992). Hence, Lancy (1993, p. 20) noted, “The qualitative researcher’s most effective defense against the charge of being subjective is to buttress what she has observed

with material that reinforces these observations from other semi-independent sources.” Likewise, Eisner (1998) proposed “structural corroboration” as a synonym for triangulation, noting, “Structural corroboration is the term I use to describe the confluence of multiple sources of evidence or the recurrence of instances to support a conclusion” (p. 55).

According to Denzin (1978), three outcomes arise from triangulation: convergence, inconsistency, and contradiction. Each of these outcomes clearly represents issues pertaining to reliability. Nevertheless, many interpretivists refrain from using the term “reliability” when pertaining to qualitative data, probably because of an attempt to distance qualitative analytical techniques from statistical method (Madill et al., 2000). However, this line of thinking is counterproductive. Indeed, as noted by Conostas (1992, p. 255), unless methods for examining rival hypotheses in qualitative research are developed, “the research community will be entitled to question the analytical rigor of qualitative research”--where rigor is defined as the attempt to make data and categorical schemes as public and as replicable as possible (Denzin, 1978).

Analyzing and Comparing Reliability Issues Across Paradigms

As previously noted, we maintain that issues relative to reliability of social science data do not vary appreciably across the positivist and interpretivist paradigms, with specific data features that affect reliability being constant across the paradigms. The major differences revolve around the nature of the data and the philosophical assumptions of the paradigms. Hence, terminology has developed that is distinctive to

each paradigm. As presented in Table 1, we have identified 12 key features of data that affect reliability. Each of these features is discussed in light of its applicability to the two paradigms.

Consistency of Evidence

Within the positivist paradigm, response variance on variables of interest is the focus of most all data analyses. In a conceptual sense, reliability coefficients are an estimate of the percent of the total variance in the scores on the measurement of interest that is attributable to true score variance (Cronbach, 1951). When this estimate is high, the researcher has enough evidence to place confidence in the scores and in the scores' use in additional descriptive, parametric, or non-parametric analyses.

Within the interpretivist paradigm, consistency of evidence is defined more loosely as the degree to which the data are "trustworthy." Although the term "trustworthiness" is defined in varying ways, it seems generally to cover at least some of the issues addressed by "research validity," "measurement validity" and "measurement reliability" within the positivist paradigm. For example, Lincoln and Guba (1985) posed four standards that should be used when judging qualitative, or naturalistic, studies: credibility, transferability, and confirmability, and dependability (or consistency). Credibility and transferability would apply to both research validity and measurement validity, whereas dependability (consistency) would be a standard for judging something roughly equivalent to reliability. Confirmability (i.e., objectivity) would be applicable across all areas. Wolcott (1990) noted that consistency is the degree to

which a study is free of inner contradictions, cautioning, however, against researchers assuming a totally contradiction-free approach as this would “set us to wondering how they could be [accurately] describing human behavior” (p. 134). Trustworthiness, therefore would involve looking for a high degree of consistency in the findings and presenting an explanation for factors to which any inconsistent findings might be attributed.

Data Integrity

Within the postivist paradigm (and more particularly within the circles of classical measurement theorists), it is commonplace for researchers to speak of the “psychometric integrity of the data,” a term that normally implies some set of assumptions about validity, reliability, and other related measurement characteristics (Crocker & Algina, 1986). One would normally expect to see at least one estimate of reliability among whatever other data might accompany the reference to data integrity. For the interpretivist, data integrity would be essentially equivalent to the “consistency of evidence” and would refer to consistency or dependability of the data. Dependability is often addressed in terms of data triangulation, with a variety of qualitative data collection and analysis strategies used simultaneously and, in some cases, supplemented with quantitative methods in a mixed methods approach.

Consistency of Judgments or Interpretations

Some measures of performance-based tasks in education and related disciplines (e.g., writing samples, public speaking, teaching behaviors) require the rater

or scorer to exercise a moderate to appreciable amount of judgment when determining scores for the individual or work sample being rated. Consequently, psychometric specialists have developed various indices of inter-rater agreement and inter-rater reliability. These may be in the form of correlation-type indices or degree-of-difference indices. Calculation of these indices allows for (a) tracking of the consistency (i.e., fairness) of the scoring process across raters and (b) gaining evidence to substantiate possible rater effects that might contaminate the scores (e.g., rater severity, biases, inconsistency in application of scoring criteria).

Fortunately, rater agreement as a concept in qualitative data analysis is increasingly gaining acceptance. In particular, it is no longer unusual for qualitative researchers to report either intrarater (i.e., consistency of a given rater's scores or observations—in essence, a variation of test-retest reliability) or interrater (i.e., consistency of two or more independent raters' scores or observations) reliability estimates (Gay & Airasian, 2000). Evidence of rater agreement can be gleaned from the fact that a leading theory-building qualitative software program called NUD.IST (non-numerical unstructured data indexing searching & theorizing) allows data analysts to determine inter-coder reliability (QSR International Pty Ltd., 2002). Even in the absence of these inter-coder issues, however, it is important for qualitative researchers to realize that all data, regardless of their nature or how they are collected, are subject to the limitations of the specific conditions under which they have been collected (Marshall & Rossman, 1999).

Temporality of the Data

Positivist researchers should be aware of the temporal nature of data and the degree to which temporality can affect reliability and correlation. Quantitative data within the social sciences are subject to conditions of “temporal instability,” namely the tendency for scores on variables of interest to fluctuate over relatively short periods of time. Granted, scores on certain measures would be expected to change over longer time periods due to maturity, effectiveness of interventions, or other natural or imposed changes that take place within an individual over a reasonable period of time. In other cases, scores will tend to vary without a reasonable explanation within a relatively brief period of time, making the data suspect due to temporal instability. Further, as Nunnally (1994) noted, “a measure which has low temporal stability will not be a good predictor of future behavior” (p. 243). Within the interpretivist paradigm, temporality is played out in terms of the relativism, or context specificity, of the data. Bernstein (1983) noted that any reality under study “must be understood as relative to a specific conceptual scheme, theoretical framework, paradigm, form of life, society or culture” (p. 685).

Corroboration of Evidence from Multiple Sources

In traditional measurement integrity studies, coefficients of equivalence are used in cases in which multiple forms of a test have been developed. Participants would be administered both forms of the test, and correlations between the two forms would be computed. Higher coefficients would imply that data from one test are equally meaningful as the other, providing evidence that the construct of interest can be

measured as effectively with one score as with the other. Further, if subjects are tested across a variety of scenarios that simultaneously explore several facets of measurement (e.g., internal consistency, occasion of measurement, and equivalent forms), generalizability theory analysis could be used to better understand the effects of those multiple sources of variation in test scores.

As previously noted, within the interpretivist paradigm, researchers may utilize triangulation procedures (i.e., structural corroboration) to achieve a similar result as obtained when testing for equivalence within the positivist paradigm. Originally a term used in navigation and surveying, triangulation serves as a method for the qualitative analyst to “steer the course” in the direction of a more accurate data interpretation. This gives the researcher an opportunity to account for the strengths and weaknesses of each data collection strategy and to examine the overall data for convergence toward a clear understanding of the phenomena under consideration: “Triangulation assumes that looking at an object from more than one standpoint provides researchers and theorists with more comprehensive knowledge about the object” (Miller, 1997, p. 25).

Cohesiveness of Evidence

Because experimentation is prohibitive in many practical measurement situations (e.g., it is difficult to do the test-retest studies needed to assess for score stability within a regular first-grade classroom), researchers often limit themselves to reliability studies that feature internal consistency measures (e.g., Cronbach alpha [Cronbach, 1951], K-R 20 and K-R 21 estimates [Kuder & Richardson, 1937]). Scores

gathered via a common set of items can yield first line evidence of reliability. Likewise, a series of related pieces of qualitative data that are used to form a narrative argument can be examined collectively for evidence of “coherence.” Eisner (1998) contended that coherence is rooted in “believability” of qualitative findings and the law of good fit, noting, “We scrutinize the argument by looking for inconsistencies, lapses of logic, things that just don’t fit” (p. 53).

Data Inconsistency

The notion of cohesiveness leads logically to an antithetical concept, namely, data inconsistency. Within the positivist paradigm, reliability is diminished when scores contain higher amounts of measurement error, or unexplained/unsystematic variance. Problems with error can be tracked and reported using various estimates of standard error. Obviously, the interpretivist does not normally have tight quantitative data available to make these types of judgments; however, it is possible through triangulation for the interpretivist to explore inconsistencies in the findings and, in larger data sets, to look for negative cases (i.e., cases that stand out as atypical in terms of the relationships among the phenomena of interest in the given qualitative study). These inconsistencies can be useful in generating theories for investigation in future studies (Woods, 1992).

Alternate Explanations

Traditional reliability analyses can sometimes yield totally unexpected and seemingly illogical results. For example, a reliability coefficient can be negative,

indicating, at least *prima facie*, that less than zero percent of the variance is being attributed to the true score! Obviously, such a finding is logically impossible even though it is mathematically accurate. Erroneous values should cue the researcher to examine the data further for evidence of rival hypotheses that account for the data better than the researcher's original hypothesis. In the present example, there may be two competing constructs underlying the data. The researcher might wisely look for evidence to support a rival hypothesis that the items, upon being split into two sets, might yield more reasonable coefficients. Similarly, Denzin (1978), noted that "contradictions" are a likely outcome of triangulation procedures used in qualitative case studies. Similar to inconsistencies, contradictions represent broader patterns within the data in which data from one source do not "line up" with data from other sources. Contradictions may indicate a systematic misunderstanding of the data, a larger concern regarding the reliability of the data, or the need to develop a new theory to support the data if the new theory is determined to be a legitimate representation of the reality being studied.

Surety of Evidence

Estimates of standard error provide the positivist with a means for assessing the surety (or accuracy) of a given data result. Standard errors can be utilized to develop confidence intervals around reliability coefficients or descriptive statistics generated for a variable of interest. If the standard error is low, resulting in a small confidence interval, the researcher can place confidence in the result.

Within the interpretivist paradigm, confidence in (surety of) in the results has a lot more to do with how well that data have been recorded and coded. Data coding allows for the assignment of alpha-numeric symbols to various observations for purposes of tracking the incidence of phenomena of interest. Obviously, these codes would have to be applied uniformly and consistently if any confidence or surety was to be placed in the data generated by the coding processes. Hence, Kelle and Laurie (1995, pp. 24-25) noted:

a coding frame[work] would only be regarded as reliable if in any subsequent re-coding exercise the same codes could be applied to the same incidents, which means that the coding could be repeated by a different coder within an acceptable margin of error. To attain this goal one would be careful to construct coding categories which are mutually exclusive and unambiguous. . . . [I]t is of crucial importance to apply these codes consistently to ensure that the same text segments are assigned the same codes, since otherwise different members of the same research group would draw upon different information when referring to the same topics.

Elusive Goal of Data Collection

Within any area of inquiry is embedded some ultimate goal or reality which the researcher hopes to attain. While these goals are typically elusive, if not utopian, it is important that researchers keep these goals in mind when examining actual results that are obtained from a given study within that area of inquiry. Classical measurement

theorists who conduct reliability analyses are ultimately searching for an accurate understanding of the true score for each individual in a given sample. The true score may be defined as “the average score that would be obtained over repeated testings” (Nunnally, 1994, p. 211). Of course, to actually obtain the true score, the number of “repeated testings” needed would be prohibitive, not to mention that each test score gained via these repetitions would be contaminated by error such that the true score would still remain elusive.

The equivalent elusive goal within the interpretivist paradigm is the capturing of the social understanding or social reality underlying the events, activities, and behaviors being studied. Roman (1992) distinguished between the “practices behaviors, and social meanings arising in the field when a researcher is physically present among the research subjects and when she or he is physically absent” (p. 571). These two social realities are clearly distinct, and even if the researcher argues for the former reality, the elusivity issue still exists considering that reality changes moment by moment, resulting in “the impossibility of knowing the world in its pristine state” (Eisner, 1998, p. 46).

Data Collection Setting

All data collected in any study are subject to the limitations of the scenario in which they were collected. For example if a teacher were to administer a test to a group of students, a host of factors related to the occasion on which the test was given might have an impact on the results (e.g., individual differences in the achievement levels of

the students, degree of fatigue or anxiety a given student is experiencing, inability of a student in understanding the test directions). This entire set of factors that makes the collection of data on any measure somewhat unique is what is referred to in classical measurement circles as the “occasion of measurement.” Within the positivist paradigm, researchers postulate that these “temporal” factors are always present to some degree, and attempts are made to estimate the effects of these factors on the reliability of scores via the computation of coefficients of stability or other similar reliability indices.

As is true in quantitative studies, there is clearly an “occasion of measurement” for any data collected in a qualitative study. However, the nature of the qualitative data precludes the type of statistical analyses used in quantitative approaches when examining the impact of the specific scenario in which data were collected on the researcher’s perception of the results. For example, a qualitative analyst may use a strictly narrative approach to cataloging observational data with no generation of performance scores or other scaled criteria with which the narrative data could be triangulated. In this case, the researcher’s findings would be limited by this single observational setting, and the degree to which the findings would generalize, in absence of additional confirming evidence, to other similar settings would be unknown.

Adequacy of Evidence

A final feature of reliability, adequacy of evidence, has implications for both positivists and interpretivists. To the positivist, reliability evidence is subject to the adequacy of the researcher’s “domain sampling.” In preparing a measurement tool, the

researcher selects a sample of test items (or other data prompts) of some size from a given domain (population, universe) of all possible items. Even though scores on the measurement tool may be shown to be reliable, Nunnally (1994) warned that the amount of evidence one has for making a decision about the meaning is limited to the extent that the selected items are deemed not to adequately reflect the entire domain. This concern for adequacy of evidence is essentially a reminder that reliability does not equal validity.

Similarly, interpretivists must be concerned with the degree to which the narrative descriptions provide an adequate view of the social phenomena of interest. Within this paradigm, the “thickness” of the description will have an impact on the adequacy of the evidence (Marshall & Rossman, 1999). As Eisner (1998, p. 15) noted, “Thick description is an effort aimed at interpretation, at getting below the surface to that most enigmatic aspect of the human condition: the construction of meaning.” If the researcher’s description is overly superficial, the result will be data that are consistent (reliable) to some degree but that will fall short of the trustworthiness criterion expected of good qualitative research.

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Table 1

Comparison of Terminology Relative to Reliability of Measurements/Data Across Two Research Paradigms

DATA FEATURES	PARADIGM	
	Positivist Terminology	Interpretivist Terminology
consistency of evidence	explained variance	trustworthiness, consistency
data integrity	reliability, psychometric integrity	dependability, consistency
consistency of judgments or interpretations	inter-rater agreement	consensus
temporality of the data	temporal stability	relativism
corroboration of evidence from multiple sources	equivalency, generalizability	triangulation, structural corroboration, convergence
cohesiveness of evidence	internal consistency	coherence
data inconsistency	measurement error, unexplained variance	inconsistency, negative case analysis
alternate explanations	rival hypotheses	contradictions
surety of evidence	confidence interval	accuracy of coding
elusive goal of data collection	true score analysis	social understanding, social reality
data collection setting	occasion of measurement	observational setting
adequacy of evidence	domain sampling	thick description



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