

**NC EOG Grades 3 through 8 and NC Math 1**  
**with**  
**The Quantile Framework<sup>®</sup> for Mathematics**

**October 2019**  
**Updated March 2020**  
**Redacted**

*Prepared by MetaMetrics for:*

**North Carolina Department of Public Instruction**  
301 N Wilmington St  
Raleigh, NC 27601



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**Linking the  
NC EOG Grades 3 through 8 and NC Math 1  
with  
The Quantile<sup>®</sup> Framework for Mathematics:**

**Linking Study Report  
Redacted**

Prepared by MetaMetrics for North Carolina Department of Public Instruction  
(Request for Quote #: 40-RQ21164619, dated July 20, 2018).

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October 2019 — Updated March 2020

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## Introduction

Often it is desirable and necessary to convey more information about test performance than can be incorporated into a single primary score scale. Two examples arise in large-scale assessment. In one situation, a test is administered less frequently but can provide a unique type of information (such as national or international comparisons). Alternatively, another test is administered more frequently, but does not provide the same breadth of information. An auxiliary score scale for a test can be established to provide this additional information through assessment scale linkages. Once a linkage is established between the two assessments, then the results of the more-frequently-administered assessment can be translated in terms of the scale for the less-frequently-administered assessment.

In yet another situation, the linkage between two score scales can be used to provide a context for understanding the results of the assessment. For example, sometimes it is hard to explain what mathematical skills and concepts a student actually understands based on the results of a mathematics test. Parents typically ask the question, “Based on my child’s test results, what math problems can he or she understand and how well?” Once a linkage is established with an assessment that is reported in relation to specific concepts and skills, the results of the assessment can then be explained and interpreted in the context of the specific concepts and skills that a student will likely understand.

Auxiliary score scales can be used to “convey additional normative information, test-content information, and information that is jointly normative and content based. For many test uses, an auxiliary scale conveys information that is as meaningful as the information conveyed by the primary score scale. In such instances, the auxiliary score is the one that is focused on, and the primary scale can be viewed more as a vehicle for maintaining interpretability over time” (Petersen, Kolen, and Hoover, 1989, p. 222). One such auxiliary scale is The Quantile Framework<sup>®</sup> for Mathematics, which was developed to appropriately match students with materials at a level where the student has the background knowledge necessary to be ready for instruction on new mathematical skills and concepts.

The Quantile Framework for Mathematics takes the guesswork out of mathematics instruction. It serves as a hands-on tool for demonstrating which mathematics skills and concepts a learner has likely learned, and which require further instruction. Because the Quantile Framework uses a common developmental scale to measure both student mathematical achievement and mathematical task difficulty, teachers can use the Quantile Framework to determine a student’s readiness to learn more advanced skills and concepts. The Quantile Framework targets instruction, forecasts understanding, and helps improve mathematics instruction and achievement. It places the mathematics curriculum, the materials to teach mathematics, and the students themselves on the same scale.

The Quantile Framework for Mathematics can be used to:

- monitor student mathematics progress,
- forecast student performance on end-of-year assessments,

- match students with appropriate materials at their level,
- determine if a student is ready for a new mathematics skill or concept,
- link big mathematical concepts with state curriculum objectives,
- identify student strengths and weaknesses,
- understand the prerequisite skills needed to learn more advanced concepts in mathematics, and
- adapt instructional methods in the classroom to ensure a greater level of understanding and application.

The Quantile Framework for Mathematics is a unique resource for accurately estimating a student’s ability to think mathematically and matching him/her with appropriate mathematical content. With this valuable information in the hands of educators, instruction can be more accurately tailored to the mathematical achievement of individual students. The structure of the Quantile Framework is organized around two principles: (1) mathematics and mathematical achievement are developmental in nature and (2) mathematics is a specific domain of knowledge and skills.

Linking assessment results with the Quantile Framework provides a mechanism for matching each student with materials on a common scale. It serves as an anchor to which resources, concepts, skills, and assessments can be connected, allowing parents, teachers, and administrators to speak the same language. Because the Quantile scale can be a common, supplemental metric to the scales of many assessments, linking the North Carolina End-of-Grade (EOG) Grades 3 through 8 Mathematics and the End-of-Course NC Math 1 assessments provides a way to evaluate the progress of students across years. By using the Quantile Framework, the same metric is applied to the materials students use, the tests they take, and the results that are reported.

Parents often ask questions such as the following:

- How much has my student grown in mathematics ability?
- How can I help my child become better at mathematics?
- How do I challenge my child to think mathematically?

Questions like these can be challenging for parents and educators. By linking the North Carolina EOG Grades 3 through 8 Mathematics and the NC Math 1 assessments with the Quantile Framework, educators and parents will be able to answer these questions, and will be better able to use the results from the tests to improve instruction and to develop each student’s level of mathematics understanding.

In 2009, the North Carolina End-of-Grade Grades 3 through 8 Mathematics and the End-of-Course assessments in Algebra I, Geometry, and Algebra II were linked with the Quantile Framework (MetaMetrics, 2010). With the revision of the assessment system in 2012, the NC READY EOG Mathematics and Algebra I/Integrated Math I were relinked with the Quantile Framework (MetaMetrics, 2014). This current research study was designed to implement a mechanism to provide mathematics achievement levels that can be matched with mathematical skills and concepts based on scale scores from the North Carolina EOG Mathematics Grades 3

through 8 and NC Math 1 assessments. The study was conducted by MetaMetrics with the North Carolina Department of Public Instruction (contract dated July 27, 2018). The primary purposes of this study were to:

- provide the North Carolina Department of Public Instruction with Quantile measures on the North Carolina EOG Mathematics and NC Math 1 assessments;
- provide tools (e.g., Quantile Math@Home, Quantile Teacher Assistant, and Quantile Math Skills Database) and information that can be used to answer questions related to standards, student-level accountability, test score interpretation, and test validation;
- develop tables for converting North Carolina EOG Mathematics and NC Math 1 scale scores to Quantile measures; and
- produce a report that describes the linking analysis procedures.



# The Quantile Framework for Mathematics

The Quantile Framework is a scale that describes a student’s mathematical achievement. Similar to how degrees on a thermometer measure temperature, the Quantile Framework uses a common metric—the Quantile—to scientifically measure a student’s ability to reason mathematically, monitor a student’s readiness for mathematics instruction, and locate a student on its taxonomy of mathematical skills, concepts, and applications.

The Quantile Framework uses this common metric to measure many different aspects of education in mathematics. The same metric can be applied to measure the materials used in instruction, to calibrate the assessments used to monitor instruction, and to interpret the results that are derived from the assessments. The result is an anchor to which resources, concepts, skills, and assessments can be connected.

There are dozens of mathematics tests that measure a common construct and report results in proprietary, nonexchangeable metrics. Not only are all of the tests using different units of measurement, but all use different scales on which to make measurements. Consequently, it is difficult to connect the test results with materials used in the classroom. The alignment of materials and linking of assessments with the Quantile Framework provides educators, parents, and students a common vocabulary to communicate and improve mathematics learning. The benefits of having a common metric include being able to:

- Develop individual multiyear growth trajectories that denote a developmental continuum from the early elementary level to Algebra II and Precalculus. The Quantile scale is vertically constructed, so the meaning of a Quantile measure is the same regardless of grade level.
- Monitor and report student growth that meets the needs of state accountability systems.
- Help classroom teachers make day-to-day instructional decisions that foster acceleration and growth toward algebra readiness and through the next several years of secondary mathematics.
- Build links between mathematics curricula and major mathematics tests.
- Develop classroom/interim assessments that can link to the major mathematics tests and forecast how likely the student is to meet the state performance standards.

In developing the Quantile Framework, the following tasks were undertaken:

- The development of a structure of mathematics that spans the developmental continuum from first-grade content through Algebra I, Geometry, and Algebra II or Math 1 through Math 3 content.
- The production of a bank of items that have been field tested.
- The development of the Quantile scale (multiplier and anchor point) based on the calibrations of the field-test items.

- The validation of the measurement of mathematics ability as defined by the Quantile Framework.

Each of these tasks is described in the subsequent sections.

## Structure of the Quantile Framework for Mathematics

The structure of the Quantile Framework is organized around two principles—(1) mathematics and mathematical ability are developmental in nature and (2) mathematics is a specific domain of knowledge and skills.

The Common Core State Standards for Mathematics describe one of the key shifts in mathematics – the call for greater rigor in mathematics instruction. Rigor is defined as the pursuit of “conceptual understanding, procedural skills and fluency, and application with equal intensity” (National Governor’s Association and Council of Chief State School Officers, 2014).

- *Conceptual understanding.* The standards call for conceptual understanding of key concepts, such as place value and ratios. Students must be able to access concepts from a number of perspectives in order to see mathematics as more than a set of mnemonics or discrete procedures.
- *Procedural skills and fluency.* The standards call for speed and accuracy in calculation. Students must practice core functions, such as single-digit multiplication, in order to have access to more complex concepts and procedures. Fluency must be addressed in the classroom or through supporting materials.
- *Application.* The standards call for students to use mathematics in situations that require mathematical knowledge. Correctly applying mathematical knowledge depends on students having a solid conceptual understanding and procedural fluency.

When developing the Quantile Framework, MetaMetrics recognized that in order to adequately address the scope and complexity of mathematics, multiple proficiencies and competencies must be assessed. The Quantile Framework is an effort to recognize and define a developmental context of mathematics instruction. This notion is consistent with the National Council of Teachers of Mathematics’ (NCTM) conclusions about the importance of school mathematics for college and career readiness presented in the *Administrator’s Guide: Interpreting the Common Core State Standards to Improve Mathematics Education* and published in 2011.

**Mathematical strands.** A strand is a major subdivision of mathematical content. Strands describe what students should know and be able to do. The National Council of Teachers of Mathematics’ (NCTM) publication *Principles and Standards for School Mathematics* (2000, hereafter NCTM Standards) outlined ten standards—five content standards and five process standards. These content standards are Number and Operations, Algebra, Geometry, Measurement, and Data Analysis and Probability. The process standards are Communications, Connections, Problem Solving, Reasoning, and Representation.

The Common Core State Standards for Mathematics (CCSSM) have been adopted or adapted by a majority of states. The CCSSM identify critical areas of mathematics that students are expected to learn each year from kindergarten through high school (National Governors Association Center for Best Practices [NGA Center] & the Council of Chief State School Officers [CCSSO], 2010a, 2010b). The critical areas in kindergarten through Grade 8 are divided into domains which differ at each grade level and include Counting and Cardinality, Operations and Algebraic Thinking, Number and Operations in Base Ten, Number and Operations-Fractions, Ratios and Proportional Relationships, The Number System, Expressions and Equations, Functions, Measurement and Data, Statistics and Probability, and Geometry. The CCSSM for Grades 9–12 are organized by six conceptual categories: Number and Quantity, Algebra, Functions, Modeling, Geometry, and Statistics and Probability.

The six strands of the Quantile Framework bridge the Content Standards of the NCTM Standards and the domains specified in the CCSSM.

- *Algebra and Algebraic Thinking.* The use of symbols and variables to describe the relationships between different quantities is covered by algebra. By representing unknowns and understanding the meaning of equality, students develop the ability to use algebraic thinking to make generalizations. Algebraic representations can also allow the modeling of an evolving relationship between two or more variables.
- *Number Sense.* Students with number sense are able to understand a number as a specific amount, a product of factors, and the sum of place values in expanded form. These students have an in-depth understanding of the base-ten system and understand the different representations of numbers.
- *Numerical Operations.* Students perform operations using strategies and standard algorithms on different types of numbers but can also use estimation to simplify computation and to determine how reasonable their results are. This strand also encompasses computational fluency.
- *Measurement.* The description of the characteristics of an object using numerical attributes is covered by measurement. The strand includes using the concept of a unit to determine length, area and volume in the various systems of measurement, and the relationship between units of measurement within and between these systems.
- *Geometry.* The characteristics, properties, and comparison of shapes and structures are covered by geometry, including the composition and decomposition of shapes. Not only does geometry cover abstract shapes and concepts, but it provides a structure that can be used to observe the world.
- *Data Analysis, Statistics, and Probability.* The gathering of data and interpretation of data are included in data analysis, probability, and statistics. The ability to

apply knowledge gathered using mathematical methods to draw logical conclusions is an essential skill addressed in this strand.

**The Quantile Skill and Concept.** Within the Quantile Framework, a Quantile Skill and Concept, or QSC, describes a specific mathematical skill or concept a student can learn. These QSCs are arranged in an orderly progression to create a taxonomy called the Quantile scale. Examples of QSCs include:

- Know and use addition and subtraction facts to 10 and understand the meaning of equality.
- Use addition and subtraction to find unknown measures of non-overlapping angles.
- Determine the effects of changes in slope and/or intercepts on graphs and equations of lines.

During Spring 2003, the QSCs used within the Quantile Framework were developed for Grades 1 through 8, Grade 9 (Algebra I) and Grade 10 (Geometry). The framework was extended to Algebra II and revised during Summer/Fall 2003. The content was finally extended to include material typically taught in Kindergarten and Grade 12 (Precalculus) during the Summer/Fall of 2007.

The first step in developing a content taxonomy was to review the curricular frameworks from a variety of sources (e.g., National Council of Teachers of Mathematics [NCTM], National Assessment of Educational Progress: 2005 Pre-Publication Edition, North Carolina, California, Florida, Illinois, and Texas). The review of the content frameworks resulted in the development of a list of QSCs spanning the content typically taught in kindergarten through Algebra I, Geometry, and Algebra II. Each QSC consists of a description of the content, a unique identification number, the grade at which it typically first appears, and the strand with which it is associated.

The Quantile Framework for Mathematics Map (Appendix A) presents a visual representation of the construct of mathematics ability. The map is organized by the six strands and describes the development of mathematics from basic skills to sophisticated problem solving. Exemplar QSCs and problems are used to annotate the Quantile scale and the strands. QSCs are located on the Quantile scale at the point corresponding to the mean of the ensemble of items addressing that QSC from two large, national studies (Quantile Framework field study and *PASeries* Mathematics field study described later in this document) and from additional field studies as new QSCs are proposed and investigated.

## Quantile Scale Development

The second step in the process of developing The Quantile Framework for Mathematics was to develop and field test a bank of items that could be used in future linking studies. Item bank development for the Quantile Framework went through several stages—content specification, item writing and review, field-testing and analyses, and final evaluation.

**Item specification and development.** Each QSC developed during the design of the Quantile Framework was aligned to a strand and identified as typically being taught at a particular grade level. The curricular frameworks from Florida, North Carolina, Texas, and California were synthesized to identify the QSCs instructed and/or assessed at each grade level. If a QSC was included in any state framework it was included in the list of QSCs for which items were to be developed for use with the Quantile Framework field study.

During the summer and fall of 2003, over 1,400 items were developed to assess the QSCs associated with content in Grades 1 through Algebra II. The items were written and reviewed by mathematics educators trained to develop multiple-choice items (Haladyna, 1994). Each item was associated with a strand and a QSC. In the development of the Quantile Framework item bank, the reading demand of the items was kept as low as possible to ensure that the items were testing mathematics achievement and not reading.

**Item writing and review.** Item writers were experienced teachers and item-development specialists who had experience with the everyday mathematical ability of students at various levels. The use of individuals with these types of experiences helped to ensure that the items were valid measures of mathematics. Item writers were provided with training materials concerning the development of multiple-choice items and the Quantile Framework. The item writing materials also contained incorrect and ineffective items that illustrated the criteria used to evaluate items and make corrections based on those criteria. The final phase of item writer training was a short practice session with three items.

Item writers were also given additional training related to sensitivity issues. Part of the item writing materials address these issues and identify areas to avoid when developing items. The following areas are covered: violence and crime, sources of common phobias, negative emotions such as death and family issues, offensive language, drugs/alcohol/tobacco, sex/attraction, race/ethnicity, class, gender, religion, supernatural/magic, parent/family, politics, animal cruelty and hunting, environmental issues, brand names, and junk food. These materials were developed based on material published by McGraw-Hill (Guidelines for Bias-Free Publishing, 1983) on universal design and fair access—equal treatment of the sexes, fair representation of minority groups, and the fair representation of disabled individuals.

Items were reviewed and edited by a group of specialists that represented various perspectives—test developers, editors, and curriculum specialists. These individuals examined each item for sensitivity issues and for the quality of the response options. During the second stage of the item review process, items were approved, approved with edits, or deleted.

**Linking and field-test design.** The next stage in the development of the Quantile item bank was the field-testing of all of the items. First, individual test items were compiled into leveled assessments and distributed to groups of students. The data gathered from these assessments were then analyzed using a variety of statistical methods. The final result was a bank of test items appropriately placed within the Quantile scale, suitable for determining the mathematical achievement of students on this scale. Assessment forms were developed for 10 levels for the purposes of field-testing. Levels 2 through 8 were aligned with the typical content taught in Grades 2 through 8, Level 9 was aligned with the typical content taught in Algebra I, Level 10

was aligned with the typical content taught in Geometry, and Level 11 was aligned with the typical content taught in Algebra II. For each level, three forms were developed with each form containing 30 items.

The final field tests were composed of 685 unique items. Besides the 660 items mentioned above, two sets of 12 linking items were developed to serve as below-level items for Grade 2 and above-level items for Algebra II. Two additional Algebra II items were developed to ensure coverage of all the QSCs at that level.

Linking the test levels vertically (across grades) employed a common-item test design (design in which items are used on multiple forms). In this design, multiple tests are given to nonrandom groups, and a set of common items is included in the test administration to allow some statistical adjustments for possible sample-selection bias. This design is most advantageous where the number of items to be tested (treatments) is large and the consideration of cost (in terms of time) forces the experiment to be smaller than is desired (Cochran and Cox, 1957).

**Quantile Framework field study and analysis.** The Quantile Framework field study was conducted in February 2004. Thirty-seven schools from 14 districts across six states (California, Indiana, Massachusetts, North Carolina, Utah, and Wisconsin) agreed to participate in the study. Data were received from 34 of the schools (two elementary and one middle-school did not return data). A total of 9,847 students in Grades 2 through 12 were tested. The number of students per school ranged from 74 to 920. The schools were diverse in terms of geographic location, size, and type of community (e.g., urban; suburban; and small town, city, or rural communities). See *Table 1* for information about the sample at each grade level and the total sample. See *Table 2* for test administration forms by level.

*Table 1. Field-study participation by grade and gender.*

| <b>Grade Level</b> | <b>N</b>     | <b>Percent Female (N)</b> | <b>Percent Male (N)</b> |
|--------------------|--------------|---------------------------|-------------------------|
| 2                  | 1,283        | 48.1 (562)                | 51.9 (606)              |
| 3                  | 1,354        | 51.9 (667)                | 48.1 (617)              |
| 4                  | 1,454        | 47.7 (644)                | 52.3 (705)              |
| 5                  | 1,344        | 48.9 (622)                | 51.1 (650)              |
| 6                  | 976          | 47.7 (423)                | 52.3 (463)              |
| 7                  | 1,250        | 49.8 (618)                | 50.2 (622)              |
| 8                  | 1,015        | 51.9 (518)                | 48.1 (481)              |
| 9                  | 489          | 52.0 (252)                | 48.0 (233)              |
| 10                 | 259          | 48.6 (125)                | 51.4 (132)              |
| 11                 | 206          | 49.3 (101)                | 50.7 (104)              |
| 12                 | 143          | 51.7 (74)                 | 48.3 (69)               |
| Missing            | 74           | 39.1 (9)                  | 60.9 (14)               |
| <b>Total</b>       | <b>9,847</b> | <b>49.6 (4,615)</b>       | <b>50.4 (4,696)</b>     |

Table 2. Test-form administration by level.

| Test Level | N     | Missing | Form 1 | Form 2 | Form 3 |
|------------|-------|---------|--------|--------|--------|
| 2          | 1,283 | 4       | 453    | 430    | 397    |
| 3          | 1,354 | 7       | 561    | 387    | 399    |
| 4          | 1,454 | 17      | 616    | 419    | 402    |
| 5          | 1,344 | 3       | 470    | 448    | 423    |
| 6          | 917   | 13      | 322    | 293    | 289    |
| 7          | 1,309 | 6       | 463    | 429    | 411    |
| 8          | 1,181 | 16      | 387    | 391    | 387    |
| 9          | 415   | 4       | 141    | 136    | 134    |
| 10         | 226   | 5       | 73     | 77     | 71     |
| 11         | 313   | 10      | 102    | 101    | 100    |
| Missing    | 51    | 31      | 9      | 8      | 3      |
| Total      | 9,847 | 116     | 3,596  | 3,119  | 3,016  |

Students administered Levels 2 through 11 were provided with rulers and students administered Levels 3 through 11 were provided with protractors. For students administered Levels 5 through 8 and 10 and 11, formulas were provided on the back of the test booklet. Administration time was approximately 45 minutes at each level. Students administered Level 2 could have the test read aloud and mark in the test booklet if that was typical of instruction.

**Field-test analyses.** At the conclusion of the field test, complete data was available for 9,678 students. Data were deleted if test level or test form was not indicated or the answer sheet was blank. The field-test data were analyzed using both the classical measurement model and the Rasch (one-parameter logistic item response theory) model. Item statistics and descriptive information (item number, field test form and item number, QSC, and answer key) were printed for each item and attached to the item record. The item record contained the statistical, descriptive, and historical information for an item; a copy of the item itself as it was field-tested; any comments by reviewers; and the psychometric notations. Each item had a separate item record.

**Field-test analyses—classical measurement.** For each item, the  $p$ -value (percent correct) and the point-biserial correlation between the item score (correct response) and the total test score were computed. Point-biserial correlations were also computed between each of the incorrect responses and the total score. In addition, frequency distributions of the response choices (including omits) were tabulated (both actual counts and percents). Items with point-biserial correlations less than 0.10 were removed from the item bank. Table 3 displays the summary item statistics.

Table 3. Summary item statistics from the Quantile Framework field study (February 2004).

| Level | Number of Items Tested | <i>p</i> -value Mean(Range) | Correct Response Point-Biserial Correlation Mean (Range) | Incorrect Responses Point-Biserial Correlation Mean (Range) |
|-------|------------------------|-----------------------------|--|---|
| 2     | 90                     | 0.58 (0.12 – 0.95)          | 0.32 (-0.15 – 0.56)                                      | -0.21 (-0.43 – 0.12)  |
| 3     | 90                     | 0.53 (0.11 – 0.93)          | 0.26 (-0.08 – 0.52)                                      | -0.22 (-0.54 – 0.02)  |
| 4     | 90                     | 0.55 (0.12 – 0.92)          | 0.24 (-0.21 – 0.50)                                      | -0.22 (-0.48 – 0.12)  |
| 5     | 90                     | 0.54 (0.12 – 0.95)          | 0.28 (-0.05 – 0.50)                                      | -0.23 (-0.45 – 0.05)  |
| 6     | 90                     | 0.52 (0.04 – 0.86)          | 0.24 (-0.08 – 0.45)                                      | -0.22 (-0.46 – 0.09)  |
| 7     | 90                     | 0.44 (0.10 – 0.77)          | 0.29 (-0.12 – 0.56)                                      | -0.21 (-0.46 – 0.25)  |
| 8     | 90                     | 0.43 (0.10 – 0.81)          | 0.26 (-0.15 – 0.50)                                      | -0.20 (-0.45 – 0.13)  |
| 9     | 90                     | 0.40 (0.10 – 0.79)          | 0.21 (-0.19 – 0.52)                                      | -0.19 (-0.53 – 0.22)  |
| 10    | 88                     | 0.51 (0.01 – 0.97)          | 0.19 (-0.26 – 0.53)                                      | -0.21 (-0.55 – 0.18)  |
| 11    | 90                     | 0.53 (0.09 – 0.98)          | 0.26 (-0.09 – 0.51)                                      | -0.22 (-0.52 – 0.07)  |

**Field-test analyses—bias.** Differential item functioning (DIF) examines the relationship between the score on an item and group membership while controlling for ability. The Mantel-Haenszel procedure has become “the most widely used methodology [to examine differential item functioning] and is recognized as the testing industry standard” (Roussos, Schnipke, and Pashley, 1999, p. 293). The Mantel-Haenszel procedure examines DIF by examining  $j \times 2 \times 2$  contingency tables, where  $j$  is the number of different levels of ability actually achieved by the examinees (actual total scores received on the test). The focal group is the group of interest and the reference group serves as a basis for comparison for the focal group (Dorans and Holland, 1993; Camilli and Shepherd, 1994).

The Mantel-Haenszel chi-square statistic tests the alternative hypothesis that there is a linear association between the row variable (score on the item) and the column variable (group membership). The  $\chi^2$  distribution has 1 degree of freedom and is determined as:

$$Q_{MH} = (n - 1)r^2 \quad \text{Equation (1)}$$

where  $r$  is the Pearson correlation between the row variable and the column variable (SAS Institute, 1985).

The Mantel-Haenszel (MH) Log Odds Ratio statistic is used to determine the direction of differential item functioning (SAS Institute Inc., 1985). This measure is obtained by combining the odds ratios,  $\alpha_j$ , across levels with the formula for weighted averages (Camilli and Shepherd, 1994, p. 110):

$$\alpha_j = \frac{p_{Rj} / q_{Rj}}{p_{Fj} / q_{Fj}} = \frac{\Omega_{Rj}}{\Omega_{Fj}} \quad \text{Equation (2)}$$

For this statistic, the null hypothesis of no relationship between score and group membership, or that the odds of getting the item correct are equal for the two groups, is not rejected when the odds ratio equals 1. For odds ratios greater than 1, the interpretation is that an individual at score level  $j$  of the Reference Group has a greater chance of answering the item correctly than an individual at score level  $j$  of the Focal Group. Conversely, for odds ratios less than 1, the interpretation is that an individual at score level  $j$  of the Focal Group has a greater chance of answering the item correctly than an individual at score level  $j$  of the Reference Group. The Breslow-Day Test is used to test whether the odds ratios from the  $j$  levels of the score are all equal. When the null hypothesis is true, the statistic is distributed approximately as a  $\chi^2$  with  $j-1$  degrees of freedom (Camilli and Shepherd, 1994; SAS Institute, 1985).

For the gender analyses, males (approximately 50.4% of the population) were defined as the reference group and females (approximately 49.6% of the population) were defined as the focal group.

The results from the Quantile Framework field study were reviewed for inclusion on later linking studies. The following statistics were reviewed for each item:  $p$ -value, point-biserial correlation, and DIF estimates. Items that exhibited extreme statistics were removed from the item bank (47 out of 685).

From the studies conducted with the Quantile Framework item bank (Palm Beach County [FL] linking study, Mississippi linking study, DoDEA/TerraNova linking study, and Wyoming linking study), approximately 6.9% of the items in any one study were flagged as exhibiting DIF using the Mantel-Haenszel statistic and the  $t$ -statistic from Winsteps. For each linking study the following steps were used to review the items: (1) flag items exhibiting DIF, (2) review items to determine if the content of the item is something that all students should know and be able to do, and (3) make decision to retain or delete the item.

**Field-test analyses—Rasch item response theory.** Classical test theory has two basic shortcomings: (1) the use of item indices whose values depend on the particular group of examinees from which they were obtained, and (2) the use of examinee ability estimates that depend on the particular choice of items selected for a test. The basic premises of item response theory (IRT) overcome these shortcomings by predicting the performance of an examinee on a test item based on a set of underlying abilities (Hambleton and Swaminathan, 1985). The relationship between an examinee's item performance and the set of traits underlying item performance can be described by a monotonically increasing function called an item characteristic curve (ICC). This function specifies that as the level of the trait increases, the probability of a correct response to an item increases.

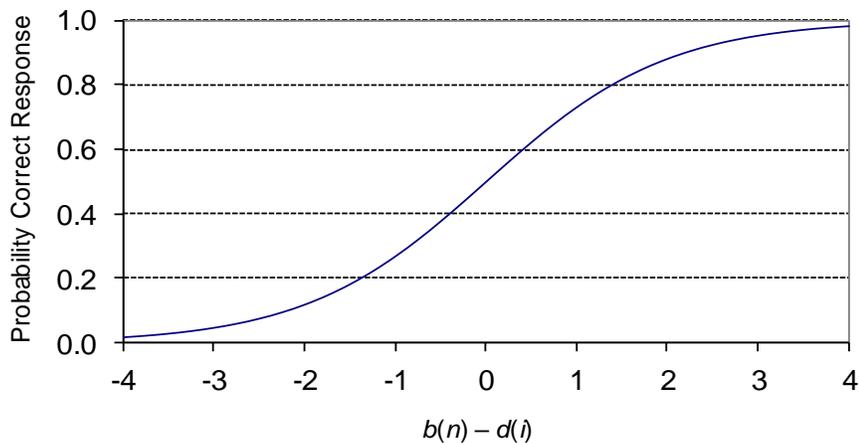
The conversion of observations into measures can be accomplished using the Rasch (1980) model, which states a requirement for the way that item calibrations and observations (count of correct items) interact in a probability model to produce measures. The Rasch IRT model expresses the probability that a person ( $n$ ) answers a certain item ( $i$ ) correctly by the following relationship:

$$P_{ni} = \frac{e^{b_n - d_i}}{1 + e^{b_n - d_i}} \quad \text{Equation (3)}$$

where  $d_i$  is the difficulty of item  $i$  ( $i = 1, 2, \dots$ , number of items);  
 $b_n$  is the ability of person  $n$  ( $n = 1, 2, \dots$ , number of persons);  
 $b_n - d_i$  is the difference between the ability of person  $n$  and the difficulty of item  $i$ ; and  
 $P_{ni}$  is the probability that examinee  $n$  responds correctly to item  $i$   
(Hambleton and Swaminathan, 1985; Wright and Linacre, 1994).

This measurement model assumes that item difficulty is the only item characteristic that influences the examinee’s performance such that all items are equally discriminating in their ability to identify low-achieving persons and high-achieving persons (Bond and Fox, 2001; and Hambleton, Swaminathan, and Rogers, 1991). In addition, the lower asymptote is zero, which specifies that examinees of very low ability have zero probability of correctly answering the item. The Rasch model has the following assumptions: (1) unidimensionality—only one ability is assessed by the set of items; and (2) local independence—when abilities influencing test performance are held constant, an examinee’s responses to any pair of items are statistically independent (conditional independence, i.e., the only reason an examinee scores similarly on several items is because of his or her ability, not because the items are correlated). The Rasch model is based on fairly restrictive assumptions, but it is appropriate for criterion-referenced assessments. *Figure 1* graphically shows the probability that a person will respond correctly to an item as a function of the difference between a person’s ability and an item’s difficulty.

*Figure 1. The Rasch Model—the probability person  $n$  responds correctly to item  $i$ .*



An assumption of the Rasch model is that the probability of a response to an item is governed by the difference between the item calibration ( $d_i$ ) and the person’s measure ( $b_n$ ). From an examination of the graph in *Figure 1*, when the ability of the person matches the difficulty of the item ( $b_n - d_i = 0$ ), then the person has a 50% probability of responding to the item correctly.

The number of correct responses for a person is the probability of a correct response summed over the number of items. When the measure of a person greatly exceeds the calibration (difficulties) of the items ( $b_n - d_i > 0$ ), then the expected probabilities will be high and the sum of these probabilities will yield an expectation of a high “number correct.” Conversely, when the item calibrations generally exceed the person measure ( $b_n - d_i < 0$ ), the modeled probabilities of a correct response will be low and the expectation will be a low “number correct.”

Thus, Equation 3 can be rewritten in terms of the number of correct responses of a person on a test:

$$O_p = \sum_{i=1}^L \frac{e^{b_n - d_i}}{1 + e^{b_n - d_i}} \quad \text{Equation (4)}$$

where  $O_p$  is the number of correct responses of person  $p$  and  $L$  is the number of items on the test.

When the sum of the correct responses and the item calibrations ( $d_i$ ) is known, an iterative procedure can be used to find the person measure ( $b_n$ ) that will make the sum of the modeled probabilities most similar to the number of correct responses. One of the key features of the Rasch IRT model is its ability to place both persons and items on the same scale. It is possible to predict the odds of two individuals being successful on an item based on knowledge of the relationship between the abilities of the two individuals. If one person has an ability measure that is twice as high as that of another person (as measured by  $b$ —the ability scale), then he or she has twice the odds of successfully answering the item.

Equation 4 possesses several distinguishing characteristics:

- The key terms from the definition of measurement are placed in a precise relationship to one another.
- The individual responses of a person to each item on an instrument are absent from the equation. The only information that appears is the “count correct” ( $O_p$ ), thus confirming that the raw score (i.e., number of correct responses) is “sufficient” for estimating the measure.
- For any set of items the possible raw scores are known. When it is possible to know the item calibrations (either theoretically or empirically from field studies), the only parameter that must be estimated in Equation 4 is the person measure that corresponds to each observable count correct. Thus, when the calibrations ( $d_i$ ) are known, a correspondence table linking observation and measure can be constructed without reference to data on other individuals.

All students and items were submitted to a Winsteps analysis using a logit convergence criterion of 0.0001 and a residual convergence criterion of 0.001. Items that a student skipped were treated as missing, rather than being treated as incorrect. Only students who responded to at least 20 items were included in the analyses (22 students were omitted, 0.22%). The Quantile measure comes from multiplying the logit value by 180 and is anchored at 656Q. The multiplier and the anchor point will be discussed in a later section. *Table 4* shows the mean and median Quantile

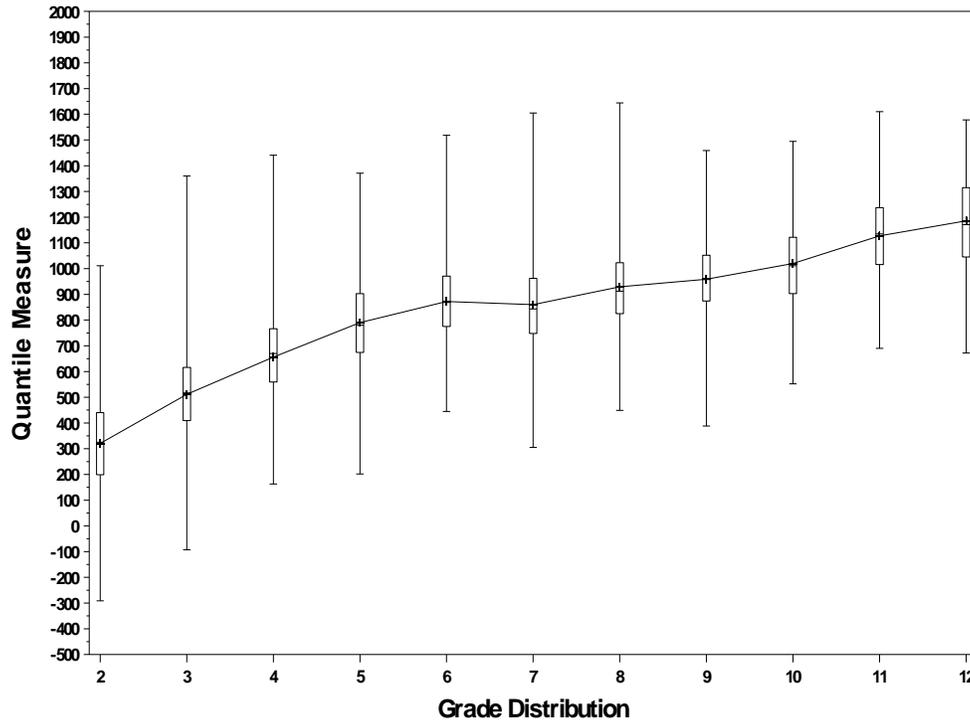
measures for all students with complete data at each grade level. While there is not a monotonically increasing trend in the mean and median Quantile measures in Grades 6 and 7, the measures are not significantly different. Results from other studies (e.g., *PASeries Mathematics* described beginning on page 26) exhibit a monotonically increasing function.

*Table 4. Mean and median Quantile measures for students with complete data (N = 9,656).*

| <b>Grade Level</b> | <b>N</b> | <b>Quantile measure<br/>Mean (SD)</b> | <b>Quantile measure<br/>Median</b> |
|--------------------|----------|---------------------------------------|------------------------------------|
| 2                  | 1,275    | 321 (189.1)                           | 323                                |
| 3                  | 1,339    | 511 (157.7)                           | 516                                |
| 4                  | 1,427    | 655 (157.5)                           | 667                                |
| 5                  | 1,337    | 790 (167.7)                           | 771                                |
| 6                  | 959      | 872 (153.0)                           | 865                                |
| 7                  | 1,244    | 861 (174.2)                           | 841                                |
| 8                  | 1,004    | 929 (157.6)                           | 910                                |
| 9                  | 482      | 959 (152.8)                           | 953                                |
| 10                 | 251      | 1020 (162.9)                          | 1005                               |
| 11                 | 200      | 1127 (178.6)                          | 1131                               |
| 12                 | 138      | 1186 (189.2)                          | 1164                               |

*Figure 2* shows the relationship between grade level and Quantile measure. The following box and whisker plots (*Figures 2, 3, and 4*) show the progression of the y-axis scores from grade to grade (the x-axis). For each grade, the box refers to the inter-quartile range. The line within the box indicates the median and the + indicates the mean. The end of each whisker shows the minimum and maximum values of the y-axis which is the Quantile measure. Across all students, the correlation between grade and Quantile measure was 0.76.

Figure 2. Box-and-whisker plot of the Rasch ability estimates of all students with complete data ( $N = 9,656$ ).



All students with outfit mean square statistics greater than or equal to 1.8 were removed from further analyses. A total of 480 students (4.97%) were removed from further analyses. The number of students removed ranged from 8.47% (108) in Grade 2 to 2.29% (22) in Grade 6 with a mean percent decrease of 4.45% per grade.

All remaining students (9,176) and all items were submitted to a Winsteps analysis using a logit convergence criterion of 0.0001 and a residual convergence criterion of 0.001. Items that a student skipped were treated as missing, rather than being treated as incorrect. Only students who responded to at least 20 items were included in the analyses. *Table 5* shows the mean and median Quantile measures for the final set of students at each grade level. *Figure 3* shows the results from the final set of students. The correlation between grade level and Quantile measure was 0.78.

Table 5. Mean and median Quantile measures for the final set of students (N = 9,176).

| Grade Level | N     | Logit Value Median | Quantile measure Mean (Median) |
|-------------|-------|--------------------|--------------------------------|
| 2           | 1,167 | -2.800             | 289 (292)                      |
| 3           | 1,260 | -1.650             | 502 (499)                      |
| 4           | 1,352 | -0.780             | 653 (656)                      |
| 5           | 1,289 | 0.000              | 795 (796)                      |
| 6           | 937   | 0.430              | 881 (874)                      |
| 7           | 1,181 | 0.370              | 878 (863)                      |
| 8           | 955   | 0.810              | 951 (942)                      |
| 9           | 466   | 1.020              | 983 (980)                      |
| 10          | 244   | 1.400              | 1044 (1048)                    |
| 11          | 191   | 2.070              | 1160 (1169)                    |
| 12          | 134   | 2.295              | 1220 (1210)                    |

Figure 3. Box-and-whisker plot of the Rasch ability estimates for the final sample of students with outfit statistics less than 1.8 (N = 9,176).

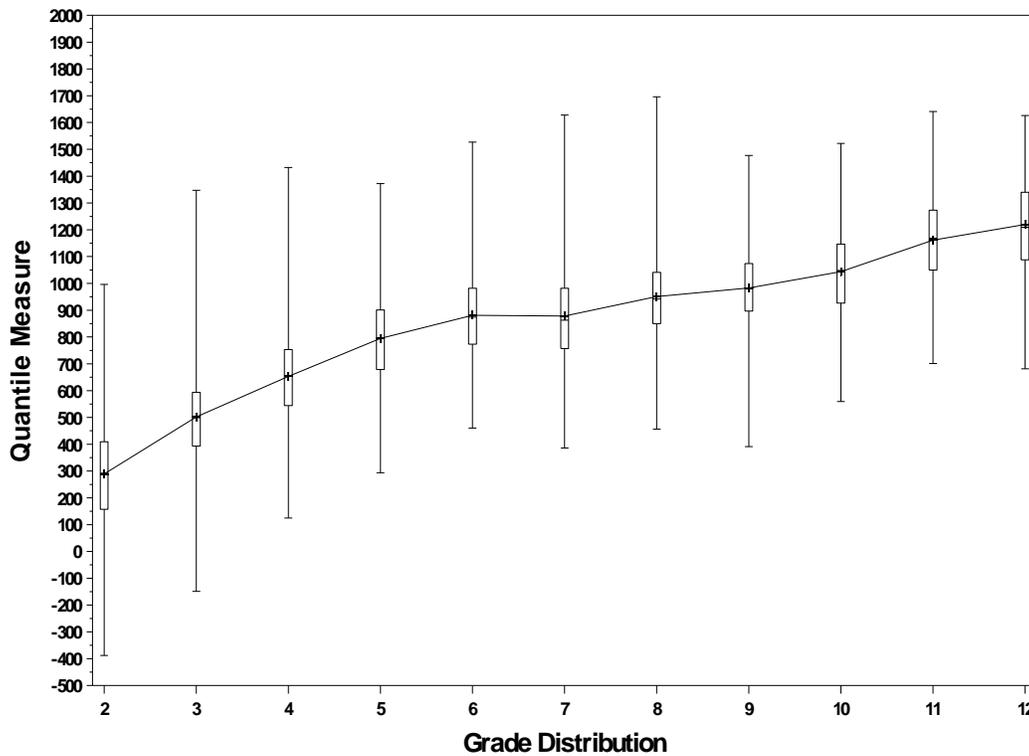
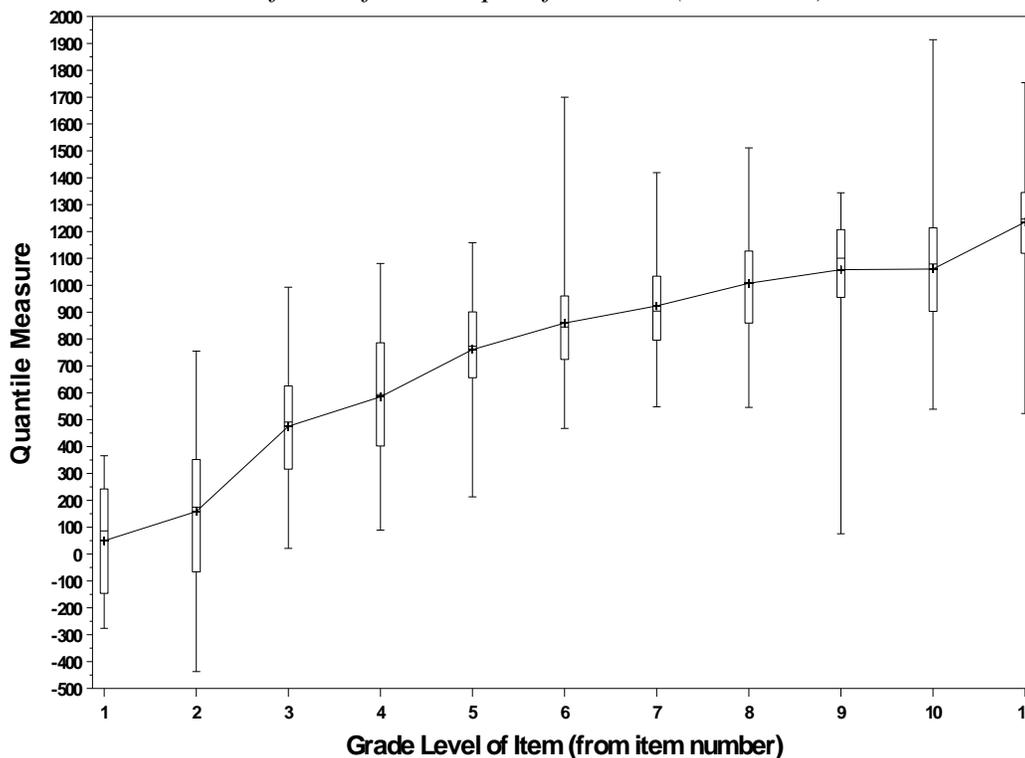


Figure 4 shows the distribution of item difficulties based on the final sample of students. For this analysis, missing data were treated as “skipped” items and not counted as wrong. There is a gradual increase in difficulty when items are sorted by level of test for which the items were written. This distribution appears to be non-linear, which is consistent with other studies. The

correlation between the grade level for which the item was written and the Quantile measure of the item was 0.80.

Figure 4. Box-and-whisker plot of the Rasch difficulty estimates of the 685 Quantile Framework items for the final sample of students ( $N = 9,176$ ).



The field testing of the items written for the Quantile Framework indicates a strong correlation between the grade level of the item and the item difficulty.

## The Specification of the Quantile Scale

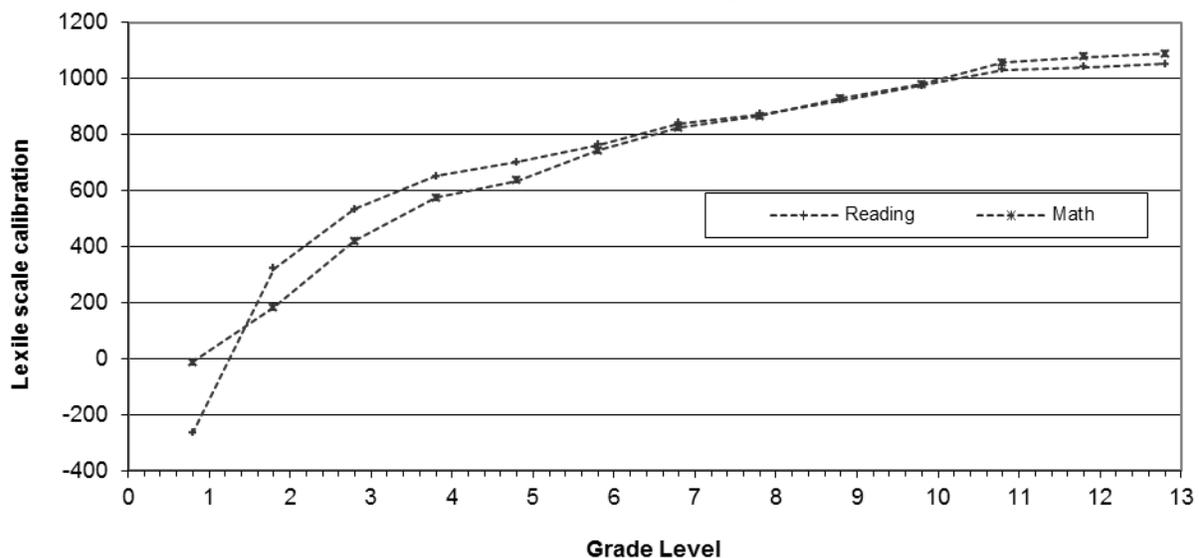
In developing the Quantile scale, two features of the scale were needed: (1) scale multiplier (conversion factor) and (2) anchor point.

As described in the previous section, the Rasch item response theory model (Wright and Stone, 1979) was used to estimate the difficulties of items and the abilities of persons on the logit scale. The calibrations of the items from the Rasch model are objective in the sense that the relative difficulties of the items will remain the same across different samples of persons (specific objectivity). When two items are administered to the same person it can be determined which item is harder and which item is easier. This ordering should hold when the same two items are administered to a second person. If two different items are administered to the second person, there is no way to know which set of items is harder and which set is easier.

The problem is that the location of the scale is not known. General objectivity requires that scores obtained from different test administrations be tied to a common zero—absolute location must be sample independent (Stenner, 1990). To achieve general objectivity, the theoretical logit difficulties must be transformed to a scale where the ambiguity regarding the location of zero is resolved.

The first step in developing the Quantile scale was to determine the conversion factor used to go from logits to Quantile measures. Based on prior research with reading and the Lexile scale, the decision was made to examine the relationship between reading and mathematics scales used with other assessments. The median scale score for each grade level on a norm-referenced assessment linked with the Lexile scale is plotted in *Figure 5* using the same conversion equation for both reading and mathematics.

*Figure 5. Relationship between reading and mathematics scale scores on a norm-referenced assessment linked to the Lexile scale for reading.*

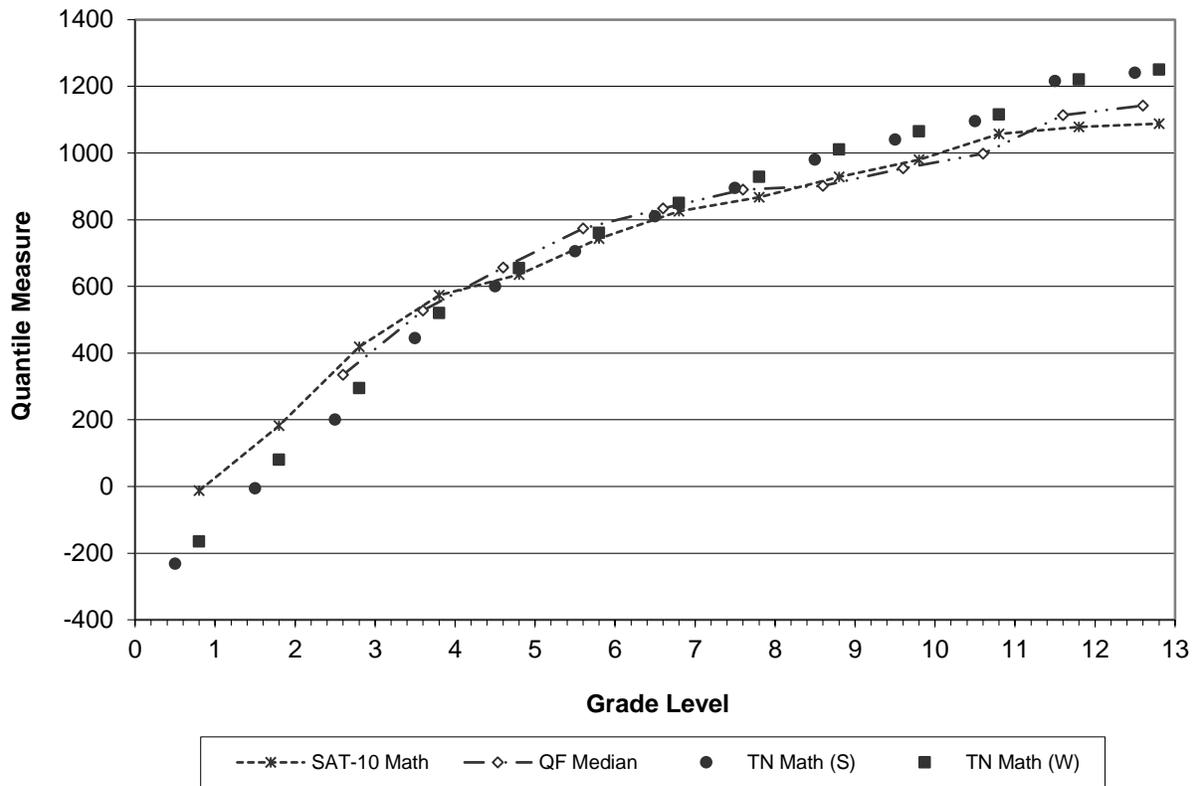


Based on an examination of *Figure 5*, it was concluded that the same conversion factor of 180 that is used with the Lexile scale could be used with the Quantile scale. Both sets of data exhibited a similar pattern across grades.

The second step in developing the Quantile scale with a fixed zero was to identify an anchor point for the scale. Given the number of students at each grade level in the field study, it was concluded that the scale should be anchored at Grade 4 or 5 (middle of grade span typically tested by state assessment programs). Median performance at the end of Grade 3 on the Lexile scale is 590L. The Quantile Framework field study was conducted in February and this point would correspond to six months (0.6) through the school year. Median performance at the end of Grade 4 on the Quantile scale is 700Q. To determine the location of the scale, 66Q were added to the median performance at the end of Grade 3 to reflect the growth of students in Grade 4 prior to the field study ( $700 - 590 = 110$ ;  $110 \times 0.6 = 66$ ).

Therefore, the value of 656Q was used for the location of Grade 4 median performance. The anchor point was validated with other assessment data and collateral data from the Quantile Framework field study (see *Figure 6*).

*Figure 6. Relationship between grade level and mathematics performance on the Quantile Framework field study and other mathematics assessments.*



Finally, a linear equation of the form:

$$[(\text{Logit} - \text{Anchor Logit}) \times \text{CF}] + 656 = \text{Quantile measure} \quad \text{Equation (5)}$$

was developed to convert logit difficulties to Quantile calibrations where the anchor logit is the median for Grade 4 in the Quantile Framework field study.

## Quantile Skill and Concept (QSC) Measures

The next step in the development process was to use the Quantile Framework to estimate the Quantile measure of each QSC. Having a measure for each QSC on the Quantile scale will allow the difficulty of skills and concepts and the complexity of other resources to be evaluated. The Quantile measure of a QSC estimates the solvability, or a prediction of how difficult the skill or concept will be for a learner.

The QSCs are assembled into Knowledge Clusters along a content continuum. Recall that the Quantile Framework is a content taxonomy of mathematical skills and topics. Knowledge Clusters are a family of skills, like building blocks, that depend one upon the other to connect and demonstrate how comprehension of a mathematical topic is founded, supported, and extended along the continuum. The Knowledge Clusters illustrate the interconnectivity of the Quantile Framework and the natural progression of mathematical skills (content trajectory) needed to solve increasingly complex problems (Hudnutt, 2012).

The Quantile measures and Knowledge Clusters for QSCs were determined by a group of three to five subject-matter experts (SMEs). Each SME had classroom experience at multiple developmental levels, had completed graduate-level courses in mathematics education, and understood basic psychometric concepts and assessment issues.

For the development of Knowledge Clusters, certain terminology was developed to describe the relationships between QSCs.

- A **focus QSC** is the skill and concept that is the focus of instruction.
- A **prerequisite QSC** is a QSC that describes a skill or concept that provides a building block necessary for another QSC. For example, adding single-digit numbers is a prerequisite for adding two-digit numbers.
- A **supporting QSC** is a QSC that describes associated skills or knowledge that assists and enriches the understanding of another QSC. For example, two supporting QSCs are multiplying two fractions and determining the probability of compound events.
- An **impending QSC** describes a skill or concept that will further augment understanding, building on another QSC. An impending QSC for using division facts is simplifying equivalent fractions.

Each focus QSC was classified with prerequisite QSCs and supporting QSCs, or was identified as a foundational QSC. As a part of the taxonomy, QSCs are either a single link in a chain of skills that lead to the understanding of larger mathematical concepts, or they are the first step toward such an understanding. A QSC that is classified as foundational requires only general readiness to learn.

The SMEs examined each QSC to determine where the specific QSC comes in the content continuum based on their classroom experience, instructional resources (e.g., textbooks), and other curricular frameworks (e.g., NCTM Standards). The process called for each SME to independently review the QSC and develop a draft Knowledge Cluster. The second step

consisted of the 3-5 SMEs meeting and reviewing the draft clusters. Through discussion and consensus, the SMEs developed the final Knowledge Cluster.

Once the Knowledge Cluster for a QSC was established, the information was used when determining the Quantile measure of a QSC, as described below. If necessary, Knowledge Clusters were reviewed and refined if the Quantile measures of the QSCs in the cluster were not monotonically increasing (steadily increasing) or there was not an instructional explanation for the pattern.

The Quantile Framework is a theory-referenced measurement system of mathematical understanding. As such, a QSC Quantile measure represents the “typical” difficulty of all items that could be written to represent the QSC and the collection of items can be thought of as an *ensemble* of the all of the items that could be developed for a specific skill or concept. During 2002, Stenner, Burdick, Sanford, and Burdick (2006) conducted a study to explore the “ensemble” concept to explain differences across reading items with The Lexile Framework for Reading. The theoretical Lexile reading measure of a piece of text is the mean theoretical difficulty of all items associated with the text. Stenner and his colleagues state that the “Lexile Theory replaces statements about individual items with statements about ensembles. The ensemble interpretation enables the elimination of irrelevant details. The extra-theoretical details are taken into account jointly, not individually, and, via averaging, are removed from the data text explained by the theory” (p. 314). The result is that when making text-dependent generalizations, text readability can be measured with high accuracy and the uncertainty in expected comprehension is largely due to the unreliability in reader measures.

To determine the Quantile measure of a QSC, actual performance by examinees is used. While expert judgment alone could be used to scale the QSCs, empirical scaling is more replicable. Items and resulting data from two national field studies were used in the process:

- Quantile Framework field study (685 items,  $N = 9,647$ , Grades 2 through Algebra II) which is described earlier in this section; and
- *PASeries* Mathematics field study (7,080 items,  $N = 27,329$ , Grades 2 through 9/Algebra I) which is described in the *PASeries* Mathematics Technical Manual (MetaMetrics, 2005).

The items initially associated with each QSC were reviewed by SMEs and accepted for inclusion in the set of items, moved to another QSC, or not included in the set. The following criteria were used:

- Psychometric (responded to by at least 50 examinees, administered at the target grade level, point-biserial correlation greater than or equal to 0.16);
- Matched grade level of introduction of concept/skill from national review of curricular frameworks; and
- Appropriate for instruction of concept (e.g., first night’s homework; from the A and B sections of the lesson problems in textbooks) based on consensus of the SMEs.

Once the set of items meeting the inclusion criteria was identified, the set of items was reviewed to ensure that the curricular breadth of the QSC was covered. If the group of SMEs considered the set of items to be acceptable, then the Quantile measure of the QSC was calculated. The Quantile measure of a QSC was defined as the mean Quantile measure of items that met the criteria.

The final step in the process was to review the Quantile measure of the QSC in relationship to the Quantile measures of the QSCs identified as pre-requisite and supporting to the QSC. If the group of SMEs did not consider the set of items to be acceptable, then the Quantile measure of the QSC was estimated and assigned a Quantile zone (e.g., 200Q-290Q, 800Q-890Q).

In 2007, with the extension of the Quantile Framework to include Kindergarten and Precalculus, the Quantile measures of the QSCs were reviewed. Where additional items had been tested and the data was available, estimated QSC Quantile measures were calculated. In 2014, a large data set was analyzed to examine the relationship between the original QSC Quantile measures and empirical QSC means from the items administered. The overall correlation between QSC Quantile measures and empirically estimated Quantile measures was 0.98 ( $N = 7,993$  students). Based on the analyses, 12 QSCs were identified with larger-than-expected deviations given the “ensemble” interpretation of a QSC Quantile measure. Each QSC was reviewed in terms of the items that generated the data, linking studies where the QSC was employed, and data from other assessments developed using the Quantile Framework. Of the 12 QSCs identified, it was concluded that the Quantile measure of nine of the QSCs should be recalculated. Five of the QSCs are targeted for Kindergarten and Grade 1 and the current data set provided data to calculate a Quantile measure (the Quantile measure for the QSC had been previously estimated). The other four QSC Quantile measures were revised because the type of “typical” item and the technology used to assess the skill or concept had shifted from the time that the QSC Quantile measure was established in 2004 (QSCs: 79, 654, 180, and 217). Three of the QSC Quantile measures were not changed (QSCs: 134, 604, 408) because (1) some of the items did not reflect the intent of the QSC, or (2) not enough items were tested to indicate that the Quantile measure should be recalculated.

## Validity Evidence for The Quantile Framework for Mathematics

Validity is the extent to which a test measures what its authors or users claim it measures. Specifically, test validity concerns the appropriateness of inferences “that can be made on the basis of observations or test results” (Salvia and Ysseldyke, 1998, p. 166). The 2014 *Standards for Educational and Psychological Testing* (American Educational Research Association, American Psychological Association, and National Council on Measurement in Education) state that “validity refers to the degree to which evidence and theory support the interpretations of test scores for proposed uses of tests” (p. 11). In other words, a valid test measures what it is supposed to measure.

In applying this definition to the Quantile Framework, the question that should be asked is “What evidence supports the use of the Quantile Framework to describe mathematics skill and concept complexity and student ability?” Stenner, Smith, and Burdick state that “[t]he process of ascribing meaning to scores produced by a measurement procedure is generally recognized as the most important task in developing an educational or psychological measure, be it an achievement test, interest inventory, or personality scale” (1983). For the Quantile Framework, which measures student understanding of mathematical skills and concepts, the most important aspect of validity that should be examined is construct-identification validity. This global form of validity encompassing content-description and criterion-prediction validity may be evaluated for The Quantile Framework for Mathematics by examining how well Quantile measures relate to other measures of mathematical achievement.

### **Relationship of Quantile measures to Other Measures of Mathematical Understanding.**

Scores from tests purporting to measure the same construct, for example “mathematical achievement,” should be moderately correlated (Anastasi, 1982). The Quantile Framework for Mathematics has been linked with numerous standardized tests of mathematics achievement. When assessment scales are linked, a common frame of reference can be used to interpret the test results. This frame of reference can be “used to convey additional normative information, test-content information, and information that is jointly normative and content-based. For many test uses ... [this frame of reference] conveys information that is more crucial than the information conveyed by the primary score scale” (Petersen, Kolen, and Hoover, 1989, p. 222).

*Table 6* presents the results from linking studies conducted with the Quantile Framework. For each of the tests listed, student mathematics scores were reported using the test’s scale, as well as by Quantile measures. This dual reporting provides a rich, criterion-related frame of reference for interpreting the standardized test scores. Each student who takes one of the standardized tests can receive, in addition to norm- or criterion-referenced test results, information related to the specific QSCs on which he or she is ready to be instructed. *Table 6* also shows that measures derived from the Quantile Framework are more than moderately correlated to other measures of mathematical understanding.

Table 6. Results from linking studies conducted with the Quantile Framework.

| Standardized Test   | Grades in Study       | N            | Correlation Between Test Score and Quantile measure |
|---|-----------------------|--------------|---|
| Mississippi Curriculum Test, Mathematics (MCT)                        | 2 – 8                 | 7,039        | 0.89  |
| TerraNova (CTB/McGraw-Hill)   | 3, 5, 7, 9            | 4,253        | 0.92  |
| Proficiency Assessments for Wyoming Students (PAWS)                   | 3, 5, 8<br>11         | 2,616<br>537 | 0.87<br>0.91  |
| Progress Towards Standards (PTS3)                                     | 3-8 and 10            | 8,544        | 0.86 to 0.90*                                       |
| Comprehensive Testing Progressing (CPT 4 – ERB)                       | 3, 5, and 7           | 802          | 0.90  |
| Kentucky Core Content Tests (KCCT)                                    | 3 - 8 and 11          | 12,660       | 0.80 to 0.83*                                       |
| Oklahoma Core Competency Tests (OCCT)                                 | 3 – 8                 | 5,649        | 0.81 to 0.85*                                       |
| Iowa Assessments  | 2, 4, 6, 8, and 10    | 7,365        | 0.92  |
| Virginia Standards of Learning (SOL)                                  | 3-8, A1, G, and A2    | 9,519        | 0.86 to 0.89*                                       |
| Kentucky Performance Rating for Educational Progress (K-PREP)         | 3 – 8                 | 6,859        | 0.81 to 0.85*                                       |
| North Carolina ACT  | 11                    | 2,707        | 0.90  |
| North Carolina READY End-of-Grade/End-of-Course Tests (NC EOG/NC EOC) | 3, 4, 6, 8, and A1/I1 | 8,720        | 0.87 to 0.90*                                       |
| aimsweb – Math Concepts and Applications (Pearson)                    | 2 – 8                 | 2,547        | 0.87  |
| ACT Aspire Math   | 4, 6, 8, and EHS      | 1,269        | 0.81  |
| South Carolina READY Mathematics                                      | 3 – 8                 | 11,104       | 0.88  |
| ISIP Early Math   | K, 1                  | 1,155        | 0.57, 0.67  |
| ISIP Math   | 2 - 8                 | 4,332        | 0.63 – 0.76*  |
| State of Texas Assessments of Academic Readiness (STAAR)              | 3 – 8,<br>Alg. I      | 6,350<br>909 | 0.86<br>0.84  |

Notes: \* Tests were not vertically scaled; separate linking equations were derived for each grade/course.

**Multidimensionality of Quantile Framework Items.** Test dimensionality is defined as the minimum number of abilities or constructs measured by a set of test items. A construct is a theoretical representation of an underlying trait, concept, attribute, process, and/or structure that a test purports to measure (Messick, 1993). A test can be considered to measure one latent trait, construct, or ability (in which case it is called unidimensional); or a combination of abilities (in which case it is referred to as multidimensional). The dimensional structure of a test is intricately tied to the purpose and definition of the construct to be measured. It is also an important factor in many of the model(s) used in data analyses. Though many of the models assume unidimensionality, this assumption cannot be strictly met because there are always other cognitive, personality, and test-taking factors that have some level of impact on test performance (Hambleton and Swaminathan, 1985).

The complex nature of mathematics and the curriculum standards most states have adopted also contribute to unintended dimensionality. Application and process skills, the reading demand of items, and the use of calculators could possibly add features to an assessment beyond what the developers intended. Strands, or sub-domains of mathematics, are useful in organizing mathematics instruction in the classroom. These standards could represent different constructs and thereby introduce more sources of dimensionality to tests designed to assess these standards. The following studies were conducted to examine the dimensionality of the Quantile scale.

***Study 1 – Comparison of Mathematics with Reading.*** The multidimensionality of the Quantile scale was examined using the Principal Components Analysis of Residuals in Winsteps (PRCOMP=S) (MetaMetrics, 2014). A three-step process was undertaken in order to examine the results and provide a context for interpreting the results.

The first step in the process was to run the Principal Components Analysis on all Quantile Framework field study items ( $N = 898$ ). Next, the residual matrix was factor analyzed. The variance that is unexplained by the first factor (the Rasch measurement model) is 0.2% of the residual variance or 2.5 items of information. Based upon this set of data, it cannot be concluded that mathematics achievement as measured by the Quantile scale is multidimensional. The results supported the use of a unidimensional item response model on the items.

Next, the items were ordered by factor loading. Based on an examination of the item names with strand listed first, there did not appear to be any effect of strand. As a sub-analysis, items from the Geometry and Algebra and Algebraic Thinking strands were analyzed. It was hypothesized, that if multidimensionality were to be evidenced in the data, this would be the most likely contrast. The Rasch model explained 54.1% of the variance in the Geometry and Algebra and Algebraic Thinking items. The results from the study are consistent with the interpretation of a single construct for each of the analyses (mathematics).

The third step was to examine the results of reading (considered a unidimensional construct) with the mathematics results. The Rasch model explained 60.6% of the variance in the reading comprehension items. Along with the results presented in the first two steps of the process, these data are consistent with the use of a unidimensional item response theory model for each of the analyses (reading and mathematics).

**Study 2 – Burg (2007).** A study conducted by Burg (2007) analyzed the dimensional structure of mathematical achievement tests aligned to the NCTM content standards. Since there is not a consensus within the measurement community on a single method to determine dimensionality, Burg employed four different methods for assessing dimensionality:

- exploring the conditional covariances (DETECT),
- assessment of essential unidimensionality (DIMTEST),
- item factor analysis (NOHARM), and
- principal component analysis (WINSTEPS).

All four approaches have been shown to be effective indices of dimensional structure. Burg analyzed Grades 3 through 8 data from the Quantile Framework field study previously described.

Each set of on-grade items for a test form from Grades 3 through 8 were analyzed for possible sources of dimensionality related to the five mathematical content strands. The analyses were also used to compare test structures across grades. The results indicated that although mathematical achievement tests for Grades 3 through 8 are complex and exhibit some multidimensionality, the sources of dimensionality are not related to the content strands. The complexity of the data structure, along with the known overlap of mathematical skills, suggests that mathematical achievement tests could represent a fundamentally unidimensional construct. While these sub-domains of mathematics are useful for organizing instruction, developing curricular materials such as textbooks, and describing the organization of items on assessments, they do not describe a significant psychometric property of the test or impact the interpretation of the test results. Mathematics, as measured by the Quantile Framework, can be described as one construct with various sub-domains.

These findings support the NCTM Connections Standard, which states that all students (prekindergarten through Grade 12) should be able to make and use connections among mathematical ideas and see how the mathematical ideas interconnect. Mathematics can be best described as an interconnection of overlapping skills with a high degree of correlation across the mathematical topics, skills, and strands.

Furthermore, these findings support the goals of the Common Core State Standards for Mathematics by providing the foundations of a growth model by which a single measure can inform progress toward college and career readiness.

**Study 3 – Hennings and Simpson (2012).** Results from Hennings and Simpson (2012) also suggest that the mathematics assessments used in MetaMetrics’ linking studies are functionally unidimensional. Data from a Quantile Framework linking study involving the end-of-grade tests from a southeastern state were examined. Scored student responses to items on the combined Quantile Linking Test and the state end-of-grade test were used. The end-of-grade tests had three polytomous items worth two points each on the forms for Grades 3 through 8, and one polytomous item worth four points on the forms for Grades 4 through 8. The remaining items on both tests were dichotomous and scored 0/1. *Table 7* shows the number of students and the number of items, combined and by test, for each grade.

Table 7. Number of items included in analyses (Hennings and Simpson, 2012).

| Grade | N of Students | Quantile Linking Test | End-of-Grade Test | Total |
|-------|---------------|-----------------------|-------------------|-------|
| 3     | 897           | 40                    | 47                | 87    |
| 4     | 1,161         | 42                    | 48                | 90    |
| 5     | 1,029         | 46                    | 48                | 94    |
| 6     | 1,327         | 44                    | 48                | 92    |
| 7     | 1,475         | 43                    | 48                | 91    |
| 8     | 933           | 47                    | 48                | 95    |

The polychoric item correlation matrix was analyzed for each test and grade. Because the principal components method of factor extraction in SAS does not require a positive-definite correlation matrix as input, principal component analyses were conducted instead of factor analyses.

The results support treating the data as unidimensional. The first component was dominant in all analyses. The first eigenvalue accounted for greater than 20% of the total variance in the analyses. Ratios of first-to-second eigenvalues ranged from approximately 6 to slightly over 9 (Gorsuch, 1983; Reckase, 1979). Secondary dimensions, i.e., the second and third components, accounted for approximately 5 - 6.5% of the total variance for each grade. *Table 8* lists the eigenvalues for the first five principal components by grade, *Table 9* shows the ratios of first-to-second eigenvalues, and *Table 10* shows the proportion of variance accounted for by the first five principal components for each grade.

Table 8. Eigenvalues for the first five principal components, by grade.

| Grade | Principal Components |       |       |       |       |
|-------|----------------------|-------|-------|-------|-------|
|       | 1                    | 2     | 3     | 4     | 5     |
| 3     | 24.152               | 3.463 | 2.411 | 2.253 | 2.011 |
| 4     | 23.252               | 3.637 | 2.257 | 1.894 | 1.829 |
| 5     | 22.770               | 3.222 | 2.407 | 2.239 | 1.935 |
| 6     | 21.400               | 3.058 | 2.297 | 2.185 | 1.866 |
| 7     | 23.919               | 3.922 | 2.442 | 1.744 | 1.648 |
| 8     | 24.572               | 2.654 | 2.152 | 2.076 | 1.914 |

Table 9. Ratio of the first-to-second eigenvalues, by grade.

| Grade | Ratio |
|-------|-------|
| 3     | 6.975 |
| 4     | 6.394 |
| 5     | 7.066 |
| 6     | 6.997 |
| 7     | 6.099 |
| 8     | 9.257 |

Table 10. Proportion of variance explained for the first five principal components, by grade.

| Grade | Principal Components |       |       |       |       |
|-------|----------------------|-------|-------|-------|-------|
|       | 1                    | 2     | 3     | 4     | 5     |
| 3     | 0.278                | 0.040 | 0.028 | 0.026 | 0.023 |
| 4     | 0.258                | 0.040 | 0.025 | 0.021 | 0.020 |
| 5     | 0.242                | 0.034 | 0.026 | 0.024 | 0.021 |
| 6     | 0.233                | 0.033 | 0.025 | 0.024 | 0.020 |
| 7     | 0.263                | 0.043 | 0.027 | 0.019 | 0.018 |
| 8     | 0.259                | 0.028 | 0.023 | 0.022 | 0.020 |

# The North Carolina End-of-Grade Mathematics and NC Math 1– Quantile Framework Linking Process

## Description of the Assessments

**North Carolina End-of-Grade and End-of-Course Mathematics Assessments.** The North Carolina End-of-Grade (NC EOG) and North Carolina End-of-Course (NC EOC) Assessments are administered annually and provide a measure of student understanding of the skills outlined by the North Carolina Standard Course of Study for Mathematics (NCSCOS; NCDPI, 2018a). The results of the annual assessments provide accountability for programs and student achievement across North Carolina (NCDPI, n.d.). NC EOG assessments are administered to students enrolled in English Language Arts/Reading and Mathematics (Grades 3-8) and in Science (Grades 5 and 8). NC EOC assessments are administered to students enrolled in Biology, English II, NC Math 1, and NC Math 3.

The NC EOG and NC EOC assessments in mathematics are constructed to assess an understanding of the skills outlined within the NCSCOS. The NCSCOS defines the key concepts and skills that students need to be successful in their future academic and post-academic careers. The NCSCOS for Mathematics is organized by grade level and domains. *Table 11* shows the number and percentage of NC EOG and NC EOC items by grade and domain for Grades 3-8 and NC Math 1 (NCDPI, 2017).

*Table 11. Number and percentage of NC EOG and NC Math 1 items, by grade and domain.*

| Domain                                       | Grade 3 | Grade 4 | Grade 5 | Grade 6 | Grade 7 | Grade 8 | NC Math 1 |
|--|---------|---------|---------|---------|---------|---------|-----------|
| Operations and Algebraic Thinking            | 32-36%  | 14-18%  | 9-13%   |         |         |         |           |
| Number and Operations in Base Ten            | 9-13%   | 25-29%  | 25-29%  |         |         |         |           |
| Number and Operations - Fractions            | 28-32%  | 30-34%  | 39-43%  |         |         |         |           |
| Measurement and Data, Geometry               | 23-27%  | 23-23%  | 19-23%  |         |         |         |           |
| Ratios and Proportional Relationships        |         |         |         | 24-28%  | 24-28%  |         |           |
| The Number System                            |         |         |         | 20-24%  | 8-12%   |         |           |
| Expressions and Equations                    |         |         |         | 22-26%  | 20-24%  |         |           |
| The Number System, Expressions and Equations |         |         |         |         |         | 24-28%  |           |
| Functions                                    |         |         |         |         |         | 28-32%  | 34-38%    |
| Geometry                                     |         |         |         | 12-16%  | 16-20%  | 24-28%  | 10-10%    |
| Statistics and Probability                   |         |         |         | 12-16%  | 22-26%  | 16-20%  | 16-18%    |
| Number and Quantity and Algebra              |         |         |         |         |         |         | 36-38%    |

The NC EOG Mathematics assessment contains 40 operational items in Grades 3 through 5 and 45 operational items in Grades 6 through 8; the NC Math 1 assessment contains 50 operational

items (NCDPI, 2018b). The online NC Math 1 assessment contains multiple-choice, numeric entry, and technology-enhanced items. The assessment at all grades includes both a calculator inactive and a calculator active section. In Grades 3 and 4, all items are multiple-choice, while in Grades 5 through 8, both calculator inactive and active sections include multiple-choice and gridded response/numeric entry item types.

The NC EOG Mathematics and NC Math 1 results are reported on separate, horizontal scales. Each grade level is individually scaled with specific ranges. The NC EOG Mathematics and NC Math 1 scales range from 515 to 575. The three-parameter logistic (3PL) and the two-parameter logistic (2PL) item response theory models were used to scale the multiple-choice items and the gridded response items, respectively. Because scores are not reported on a vertical scale, scale scores across grades should not be directly compared.

*The Quantile Framework for Mathematics.* The Quantile Framework was developed to assist teachers, parents, and students in identifying strengths and weaknesses in mathematics and forecast growth in overall mathematical achievement. Items and mathematical content are scaled using the Rasch model. The Quantile Framework spans the developmental continuum from Kindergarten mathematics through the content typically taught in Algebra II, Geometry, Trigonometry and Precalculus. The Quantile scale ranges from below EM400Q to above 1600Q (“EM” -- Emerging Mathematician, below 0Q).

The Quantile Framework was developed to assess how well a student understands the natural language of mathematics, knows how to read mathematical expressions and employ algorithms to solve decontextualized problems, and knows why conceptual and procedural knowledge is important and how and when to apply it.

The Quantile Framework measures mathematical achievement by focusing on mathematics skills and concepts students are expected to learn, denoted as Quantile Skills and Concepts (QSCs). The Quantile Item Bank consists of items aligned with content spanning Kindergarten through Geometry, Algebra II, and Precalculus. Quantile items used in linking studies were developed for administration to students in Grades K-12 and organized into six content strands:

- Algebra and Algebraic Thinking
- Data Analysis, Statistics, and Probability
- Geometry
- Measurement
- Number Sense
- Numerical Operations

The distribution of the items in the Quantile linking item pools reflect the proportions of items in each of the NC EOG Mathematics and NC Math 1 assessment domains. To achieve this alignment, the content of the NC EOG Mathematics and NC Math 1 assessment blueprints were matched to corresponding Quantile QSCs. Quantile linking items were selected to maximize the alignment with the NC blueprints. *Table 12* shows the distribution of items by grade and strand for the Quantile linking item pools, where each grade had 39 items.

Table 12. Number and percent of items in the Quantile linking item pools, by strand and grade.

|           | <b>Algebra and Algebraic Thinking</b> | <b>Data Analysis, Statistics and Probability</b> | <b>Geometry</b> | <b>Measurement</b> | <b>Number Sense</b> | <b>Numerical Operations</b> |
|-----------|---------------------------------------|--|-----------------|--------------------|---------------------|-----------------------------|
| Grade 3   | 7 (18%)                               | 1 (3%)   | 5 (13%)         | 3 (8%)             | 9 (23%)             | 14 (36%)                    |
| Grade 4   | 3 (8%)                                | 1 (3%)   | 4 (10%)         | 6 (15%)            | 10 (26%)            | 15 (38%)                    |
| Grade 5   | 3 (8%)                                | 2 (5%)   | 5 (13%)         | 1 (3%)             | 6 (15%)             | 22 (56%)                    |
| Grade 6   | 11 (28%)                              | 6 (15%)  | 3 (8%)          | 4 (10%)            | 8 (21%)             | 7 (18%)                     |
| Grade 7   | 12 (31%)                              | 8 (21%)  | 2 (5%)          | 6 (15%)            | 1 (3%)              | 10 (26%)                    |
| Grade 8   | 16 (41%)                              | 5 (13%)  | 9 (23%)         | 2 (5%)             | 4 (10%)             | 3 (8%)                      |
| NC Math 1 | 23 (59%)                              | 6 (15%)  | 5 (13%)         | 2 (5%)             | 2 (5%)              | 1 (3%)                      |

All linking items were four-option, multiple-choice items, and had known statistics from previous administrations. Quantile items were delivered as grade level item pools and organized into subsets to be embedded into three base forms of the NC EOG Mathematics and 2 base forms of the NC Math 1 assessments. Each subset contained 4 to 5 items. Common items were included within subsets and across grade levels. The Quantile item pools consisted of calculator neutral, inactive, and active items. Students were permitted to use a calculator on calculator neutral or calculator active items administered in calculator active sections of the NC EOG Mathematics or NC Math 1 assessments.

Each Quantile item had an established difficulty value (Quantile measure) based on data collected from previous test administrations. The mean difficulties of the Quantile linking item pools were as follows: Grade 3, 405Q; Grade 4, 588Q; Grade 5, 673Q; Grade 6, 751Q; Grade 7, 897Q; Grade 8, 977Q; and NC Math 1, 1074Q.

**Evaluation of Quantile Linking Items.** After administration, the Quantile linking items were reviewed for use in the linking analysis. *Table 13* presents the descriptive statistics for the Quantile linking items. A total of 669,252 records were provided to MetaMetrics. During an initial screening process, 82 records were removed from NC Math 1 as duplicated records. Then, 8,823 records across all grades were removed where the data exhibited misfit to the Rasch model, indicated by an infit statistic greater than 1.5 and an outfit statistic greater than 2.0 (Linacre, 2011). A total of 660,347 records remained in the sample used to evaluate the Quantile linking item pool items. Each linking item was evaluated for use in the linking study based on potential alternate answer choices being more attractive than the correct answer choice (i.e., low point-measure correlation). The items with the largest number of responses were common items within grade. In addition to 5 items in the linking item pools that were not administered to students, 8 items were removed from further analysis because they had low point-measure correlations or misfit criteria outside the acceptable range.

Table 13. Descriptive statistics from the administration of the Quantile linking items.

| Test Level | N (Persons)*   | N (Items)** | Percent Correct Mean (Range) | Point-Measure Mean (Range) |
|------------|----------------|-------------|------------------------------|----------------------------|
| 3          | 3,667 - 46,637 | 39          | 81 (65 - 96)                 | 0.41 (0.20 - 0.58)         |
| 4          | 3,901 - 48,139 | 37          | 75 (38 - 93)                 | 0.43 (0.24 - 0.58)         |
| 5          | 4,522 - 56,343 | 37          | 77 (50 - 97)                 | 0.40 (0.25 - 0.56)         |
| 6          | 4,753 - 57,946 | 36          | 68 (44 - 90)                 | 0.40 (0.17 - 0.57)         |
| 7          | 4,637 - 56,762 | 36          | 65 (47 - 82)                 | 0.46 (0.26 - 0.60)         |
| 8          | 2,988 - 36,836 | 38          | 58 (36 - 84)                 | 0.38 (0.12 - 0.58)         |
| NC Math 1  | 3,900 - 48,487 | 37          | 60 (22 - 87)                 | 0.43 (0.32 - 0.56)         |

\* Reflects removal of 8,823 persons due to misfit to the Rasch model.

\*\* Reflects 5 items not administered and 8 items removed for low point-measure or exceeding misfit criteria.

## Study Design

A single-group/non-equivalent anchor test design was chosen for this study (Dorans and Holland, 2000). This design is most useful when (1) administering two sets of items to examinees is operationally possible, and (2) differential order effects are not expected to occur (Kolen and Brennan, 2014, pp. 16–17). For each student, four or five linking items from the Quantile item pools were embedded in the NC EOG Grades 3 through 8 Mathematics and NC Math 1 assessment administrations. The NC EOG Grades 3 through 8 assessments are administered in the final 10 instructional days of the school year; the NC Math 1 assessment is administered in the final 5 days of instruction for a semester course, or final 10 days of instruction for a yearlong course.

## Description of the Sample

The NC EOG Grades 3 through 8 Mathematics and NC Math 1 scale scores and item responses were provided to MetaMetrics by NCDPI. The data file included 669,252 total records. During an initial screening process, 82 duplicate records were identified in the NC Math 1 records and removed from further analysis. The statewide sample included students administered a computer-based test, the primary modality, or a paper-based test. The purpose of the linking study was to link the Quantile scale with the NC EOG Grades 3 through 8 Mathematics and the NC Math 1 scale scores, thus all administration modalities were included in the analysis.

Three samples were used for the linking process. First, an initial sample was established by retaining all records with a valid test scale score. Next, a calibration sample was established to perform two tasks: (1) evaluate the performance of the Quantile linking items, discussed above, and (2) place the NC EOG Grades 3 through 8 Mathematics and the NC Math 1 items on the Quantile scale. Lastly, a linking sample was established to link the NC EOG Grades 3 through 8 Mathematics and NC Math 1 scale scores with the Quantile scale.

During the calibration process, all student records from the initial sample were submitted to a Winsteps analysis. Student records were removed from further analysis if the data did not fit the Rasch model, indicated by an infit statistic greater than 1.5 or outfit statistic greater than 2.0 (Linacre, 2011). A total of 660,347 student records remained in the calibration sample, or 98.7% of the initial sample (see *Table 14*).

*Table 14. Number of records in the initial and calibration samples.*

| <b>Grade</b> | <b>N Initial Sample</b> | <b>N Removed Based on Misfit Person</b> | <b>N Calibration Sample</b> | <b>Calibration Sample Percent of Initial Sample</b> |
|--------------|-------------------------|---|-----------------------------|---|
| 3            | 97,958                  | 2,286                                   | 95,672                      | 97.7  |
| 4            | 100,164                 | 1,766                                   | 98,398                      | 98.2  |
| 5            | 101,740                 | 1,896                                   | 99,844                      | 98.1  |
| 6            | 103,541                 | 948                                     | 102,593                     | 99.1  |
| 7            | 101,508                 | 696                                     | 100,879                     | 99.4  |
| 8            | 66,263                  | 736                                     | 65,527                      | 98.9  |
| Math 1       | 97,996                  | 562                                     | 97,434                      | 99.4  |
| Total        | 669,170                 | 8,890                                   | 660,347                     | 98.7  |

The sample used to link the NC EOG Grades 3 through 8 Mathematics and the NC Math 1 scale scores with the Quantile scale included all records from the initial sample with all lowest observable scale scores (LOSS) and highest observable scale scores (HOSS) removed. The linking sample consisted of 661,766 records, or 98.9% of the initial sample (see *Table 15*).

Table 15. Number of records in the initial and linking samples.

| <b>Grade</b> | <b>N Initial Sample</b> | <b>N Removed HOSS/LOSS</b> | <b>Linking Sample</b> | <b>Linking Sample Percent of Initial Sample</b> |
|--------------|-------------------------|----------------------------|-----------------------|---|
| 3            | 97,958                  | 1,910                      | 96,048                | 98.1  |
| 4            | 100,164                 | 2,169                      | 97,995                | 97.8  |
| 5            | 101,740                 | 1,144                      | 100,596               | 98.9  |
| 6            | 103,541                 | 802                        | 102,739               | 99.2  |
| 7            | 101,508                 | 932                        | 100,576               | 99.1  |
| 8            | 66,263                  | 41                         | 66,222                | 99.9  |
| NC Math 1    | 97,996                  | 406                        | 97,590                | 99.6  |
| Total        | 669,170                 | 7,404                      | 661,766               | 98.9  |

Table 16 presents the demographic characteristics for the initial and linking samples for the NC EOG Grades 3 through 8 Mathematics and the NC Math 1 student records included in this study. Through the trimming process it is important to preserve the demographic characteristics of the original sample to ensure that bias is not introduced. After removals, the linking sample compares well with the initial sample. One observable difference is the NC EOG Grade 8 Mathematics sample, which is approximately 10 percent of the total sample compared to approximately 15 percent of the total sample for all other grade and levels. After consulting with NCDPI, it was determined that a large portion of higher achieving students in Grade 8 were enrolled in the NC Math 1 course and therefore only took the NC Math 1 assessment and not the NC EOG Grade 8 Mathematics assessment.

Table 16. Percentage of students in the NC EOG Grades 3 through 8 Mathematics and NC Math 1 linking study initial and linking samples, for selected demographic characteristics.

| <b>Student Characteristic</b> | <b>Category</b>                         | <b>Initial Sample<br/>N = 669,170</b> | <b>Linking Sample<br/>N = 661,766</b> |
|-------------------------------|---|---------------------------------------|---------------------------------------|
| Test Level                    | 3                                       | 14.6                                  | 14.5                                  |
|                               | 4                                       | 15.0                                  | 14.8                                  |
|                               | 5                                       | 15.2                                  | 15.2                                  |
|                               | 6                                       | 15.5                                  | 15.5                                  |
|                               | 7                                       | 15.2                                  | 15.2                                  |
|                               | 8                                       | 9.9                                   | 10.0                                  |
|                               | NC Math 1                               | 14.6                                  | 14.7                                  |
| Gender                        | F                                       | 50.4                                  | 50.5                                  |
|                               | M                                       | 49.4                                  | 49.4                                  |
|                               | Not Available                           | 0.1                                   | 0.1                                   |
| Race/Ethnicity                | American Indian or Alaska Native        | 1.2                                   | 1.2                                   |
|                               | Asian                                   | 3.4                                   | 3.2                                   |
|                               | Black                                   | 25.5                                  | 25.7                                  |
|                               | Hispanic/Latino                         | 16.9                                  | 17.0                                  |
|                               | Native Hawaii or Other Pacific Islander | 0.1                                   | 0.1                                   |
|                               | Two or more races                       | 4.7                                   | 4.7                                   |
|                               | White                                   | 48.1                                  | 47.9                                  |
|                               | Not Available                           | 0.1                                   | 0.1                                   |
| EDS                           | N                                       | 53.9                                  | 53.5                                  |
|                               | Y                                       | 46.0                                  | 46.4                                  |
|                               | Not Available                           | 0.1                                   | 0.1                                   |
| ELS                           | 1                                       | 0.2                                   | 0.2                                   |
|                               | 2                                       | 0.1                                   | 0.1                                   |
|                               | N                                       | 91.6                                  | 91.5                                  |
|                               | U                                       | 4.5                                   | 4.5                                   |
|                               | Y                                       | 3.6                                   | 3.6                                   |
|                               | Not Available                           | 0.1                                   | 0.1                                   |

## Item Calibration and Scoring

Three steps were performed prior to the linking analysis. First, using the calibration sample shown in *Table 14*, a concurrent calibration at each test level was conducted between the NC EOG Grades 3 through 8 Mathematics and NC Math 1 items and Quantile linking items to evaluate the appropriateness of scaling the Quantile and NC EOG Grades 3 through 8 Mathematics and NC Math 1 items on the same scale, and determine the final set of linking items used to scale NC Math 1 items onto the Quantile scale. Second, a concurrent calibration of the NC EOG Grades 3 through 8 Mathematics and NC Math 1 items with the Quantile linking items anchored to their theoretical Quantile difficulties was conducted to place the NC EOG Grades 3 through 8 Mathematics and NC Math 1 items on the Quantile scale. Finally, a scoring run was conducted using only the NC EOG Grades 3 through 8 Mathematics and NC Math 1 assessment items on the Quantile scale to express student results on the NC EOG Grades 3 through 8 Mathematics and NC Math 1 assessments in the Quantile metric.

*Table 17* provides the descriptive statistics from the NC EOG Grades 3 through 8 Mathematics and NC Math 1 scale scores and their calibrated Quantile measures. The NC EOG Mathematics Grades 3 through 8 and NC Math 1 scales are horizontal scales denoted by the consistent means and standard deviations across each test level. The Quantile scale is a vertical scale, so the mean Quantile measures increase from one grade to the next, with the exception of Grade 8. The decline in the Grade 8 performance is likely due to a number of higher achieving Grade 8 students taking NC Math 1 instead of the NC EOG Grade 8 Mathematics assessment.

*Table 17. Descriptive statistics for the NC EOG Grades 3 through 8 Mathematics and NC Math 1 scale scores and NC EOG Grades 3 through 8 Mathematics and NC Math 1 calibrated Quantile measures for the linking sample, by grade (N = 661,766).*

| Test Level | N       | NC EOG and NC Math 1 Scale Score Mean (SD) | NC EOG and NC Math 1 Calibrated Quantile Measure Mean (SD) | r |
|------------|---------|--|--|---|
| 3          | 96,048  |  |  |   |
| 4          | 97,995  |  |  |   |
| 5          | 100,596 |  |  |   |
| 6          | 102,739 |  |  |   |
| 7          | 100,576 |  |  |   |
| 8          | 66,222  |  |  |   |
| NC Math 1  | 97,590  |  |  |   |

*Figures 7 through 13* show the relationships between NC EOG Grades 3 through 8 Mathematics and NC Math 1 scale scores and their calibrated Quantile measures for the calibration sample. There exists a linear relationship throughout the majority of the distributions and a curvilinear relationship at the lower and upper ends of the distributions at each grade. These departures from linearity occur at or below the 5<sup>th</sup> or 1<sup>st</sup> percentiles and above the 95<sup>th</sup> or 99<sup>th</sup> percentiles of the distributions. Because the vast majority of scores are represented with a linear relationship, linear

methods can be used to represent the relationship between the NC EOG Grades 3 through 8 Mathematics and NC Math 1 scales and the Quantile scale.

Figure 7. Scatter plot of NC EOG Grade 3 Mathematics scale scores and their calibrated Quantile measures, linking sample (N = 96,048)

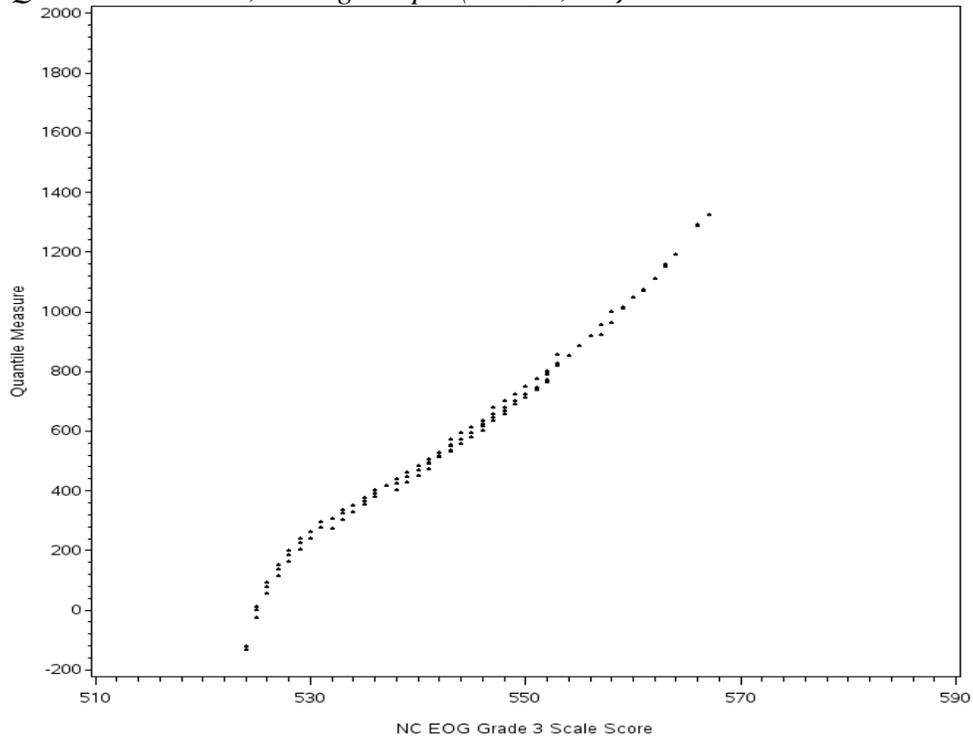


Figure 8. Scatter plot of NC EOG Grade 4 Mathematics scale scores and their calibrated Quantile measures, linking sample (N = 97,995)

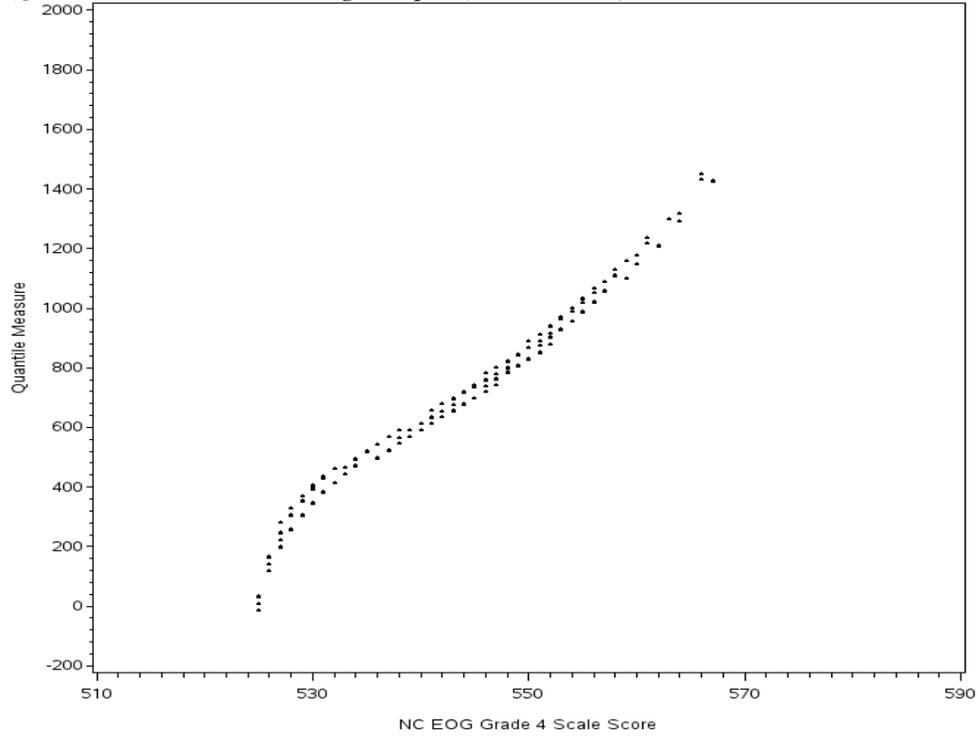


Figure 9. Scatter plot of NC EOG Grade 5 Mathematics scale scores and their calibrated Quantile measures, linking sample (N = 100,596)

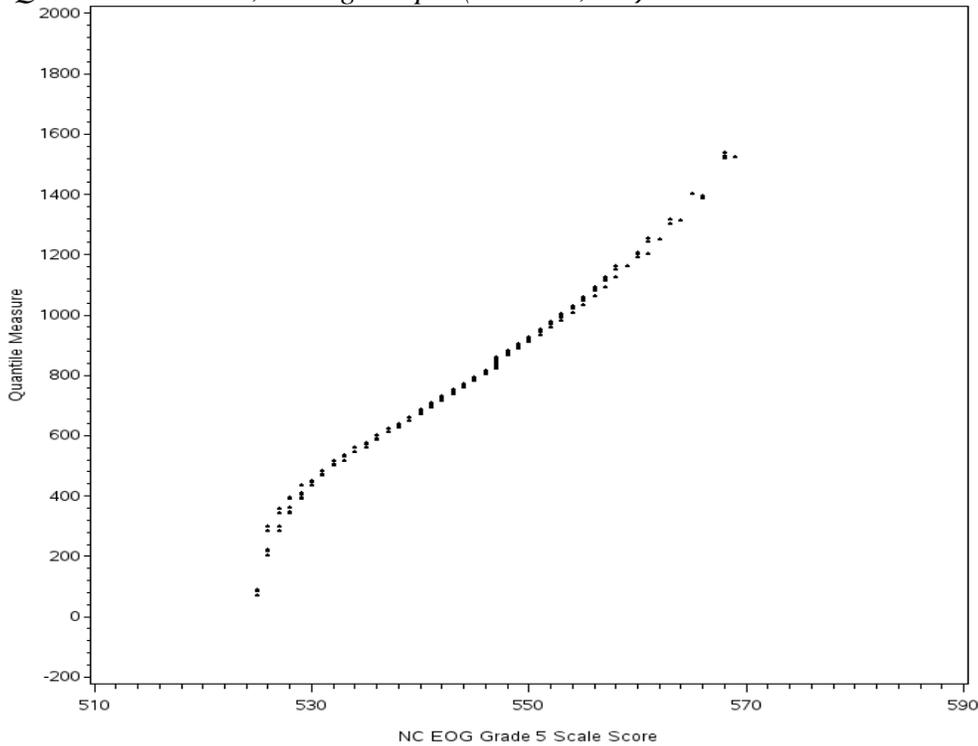


Figure 10. Scatter plot of NC EOG Grade 6 Mathematics scale scores and their calibrated Quantile measures, linking sample (N = 102,739)

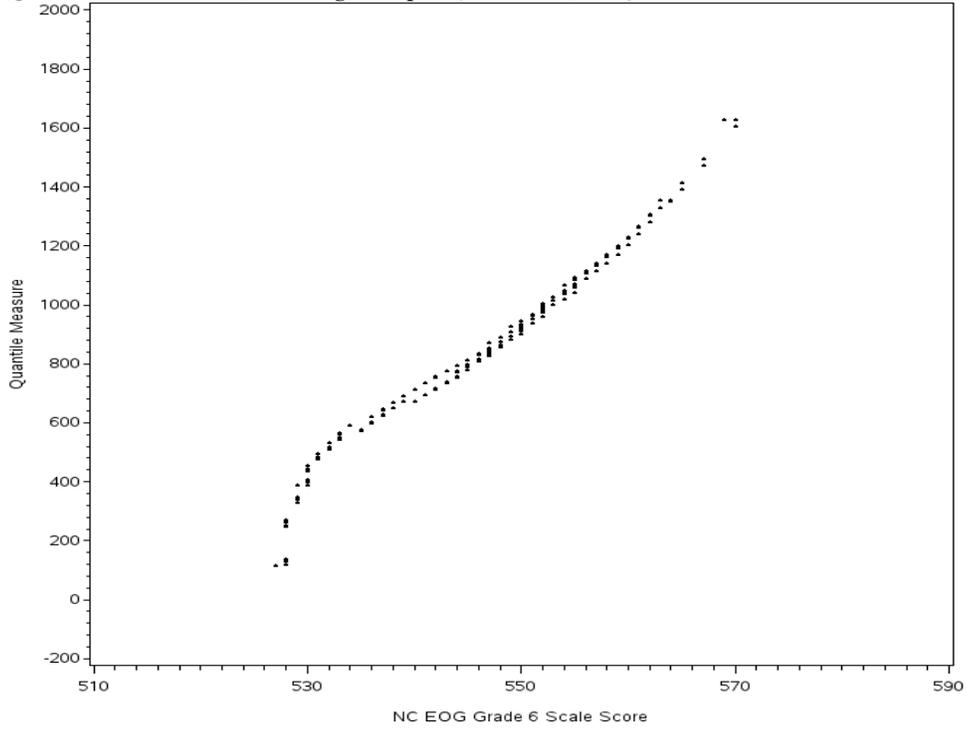


Figure 11. Scatter plot of NC EOG Grade 7 Mathematics scale scores and their calibrated Quantile measures, linking sample (N = 100,576)

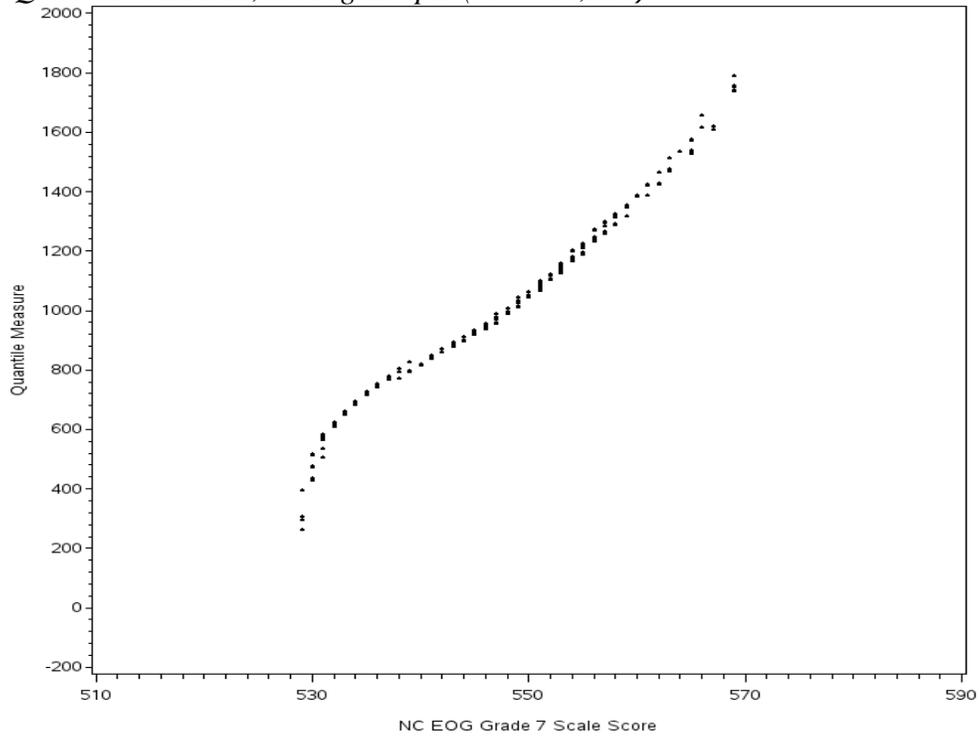


Figure 12. Scatter plot of NC EOG Grade 8 Mathematics scale scores and their calibrated Quantile measures, linking sample (N = 66,222)

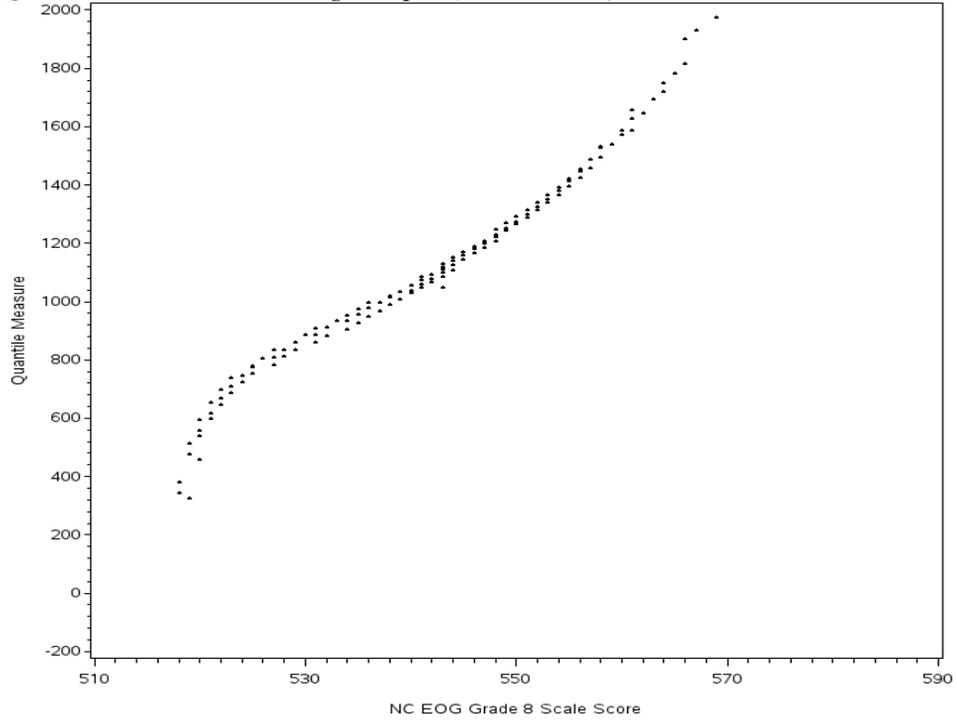
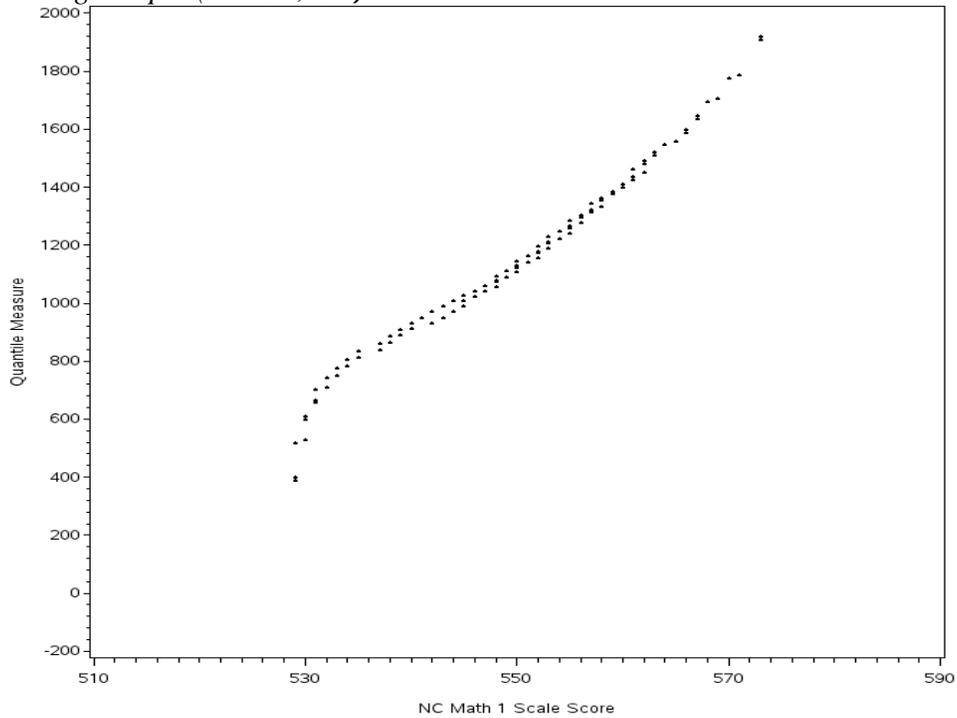


Figure 13. Scatter plot of NC Math 1 scale scores and their calibrated Quantile measures, linking sample (N = 97,590)



## Linking the North Carolina EOG Grades 3 through 8 Mathematics and NC Math 1 Scales with the Quantile Scale

Linking in general means “putting the scores from two or more tests on the same scale” (National Research Council, 1999, p.15). MetaMetrics and the North Carolina Department of Public Instruction conducted this linking study to provide information that could be used to match students’ mathematical achievement with instructional resources—to identify the materials, concepts, and skills a student should be matched with for successful mathematical instruction, given their performance on the NC EOG Grades 3 through 8 Mathematics or the NC Math 1 assessments.

**Linking Analyses.** In scale alignment, which often uses the same methods as linear equating (Dorans, Moses, and Eignor, 2010), the equating relationship requires that the transformations between two scales be symmetric (Lord, 1980). This requirement means that the function used to transform Form  $X$  to Form  $Y$  can be inversely applied. Two score scales (e.g., the NC EOG Grades 3 through 8 Mathematics and NC Math 1 scales and the Quantile scale) can then be linked using linear equating methods when simplicity in developing conversion tables or equations, in conducting analyses, and in describing procedures are desired (Kolen and Brennan, 2014).

In linear linking, a transformation is chosen such that two sets of scores are considered to be linked if they correspond to the same number of standard deviations above (or below) the mean in some group of examinees (Angoff, 1984, cited in Petersen, Kolen, and Hoover, 1989; Kolen and Brennan, 2014). Given scores  $x$  and  $y$  on tests  $X$  and  $Y$ , the linear relationship is:

$$\frac{(x - \mu_x)}{\sigma_x} = \frac{(y - \mu_y)}{\sigma_y} \quad \text{Equation (2)}$$

and the linear transformation  $l_x$  (called the SD line in this report) used to transform scores on test  $Y$  to scores on test  $X$  is:

$$x = l_x(y) = \left( \frac{\sigma_x}{\sigma_y} \right) y + \left( \mu_x - \frac{\mu_y \sigma_x}{\sigma_y} \right) \quad \text{Equation (3)}$$

Linear transformation by definition has the same mean and standard deviation for the linking equation because the means and standard deviations are the same for the tests being linked. Linear linking using an SD-line approach is preferable to linear regression because the tests are not perfectly correlated. With less than perfectly reliable tests, linear regression is dependent on which way the regression is conducted: predicting scores on test  $X$  from scores on test  $Y$  or predicting scores on test  $Y$  from scores on test  $X$ . The SD line provides the symmetric linking function that is desired.

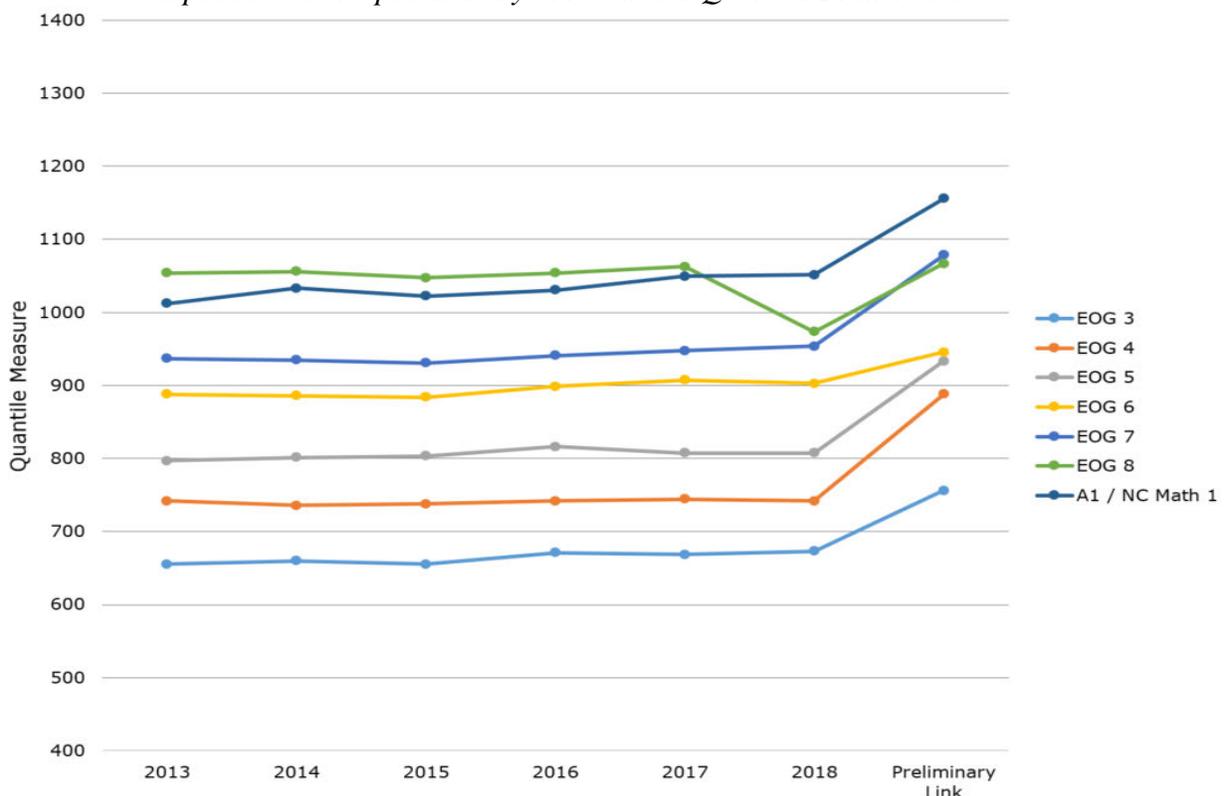
The linking equation between NC EOG Grades 3 through 8 Mathematics and NC Math 1 scale scores and Quantile measures can be written as:

$$\text{Quantile measure} = \text{Slope}_g(\text{NC EOG/NC Math 1 scale score}) + \text{intercept}_g \quad \text{Equation (4)}$$

where the slope is the ratio of the standard deviations of the NC EOG Grades 3 through 8 Mathematics and the NC Math 1 scale scores and the Quantile measures and  $g$  represents the test level.

**Preliminary Analyses.** Preliminary analyses were conducted to develop linking functions from the results in *Table 17* using Equation 4. First, historical trends were reviewed comparing Quantile measures calculated from the previous linking study (MetaMetrics, 2014) and the published state scale scores in the “Greenbook” (NCDPI, 2018c). Next, the historical Quantile measures were compared with the preliminary linking results from the current study. *Figure 14* provides the cross-sectional historical trends of Quantile measures compared with the preliminary link. From 2013 to 2018, the distributions of Quantile measures for all grades were stable from year-to-year with the exception of Grade 8 from 2017 to 2018, where a decrease was observed. This decrease was likely due, at least in part, to Grade 8 students enrolled in NC Math 1 or higher no longer being administered the Grade 8 assessment and only participating in their enrolled end-of-course assessment. Thus, a population change occurred for the NC EOG Grade 8 assessment. An increase was observed in Quantile measures using the preliminary link. While some variability in scores was expected, the observed systematic increase urged further investigation.

*Figure 14. Historical trends of North Carolina EOG “Greenbook” scores as Quantile measures compared with the preliminary link with the Quantile Framework.*



The increase in Quantile measures from the historical data to the preliminary link prompted a closer look at the item level statistics for each grade. *Figures 15 through 21* compare the  $p$ -value distributions for the Quantile linking item pools (the upper box) used in each grade with the  $p$ -value distributions of the NC EOG Grades 3 through 8 Mathematics and NC Math 1 item pools (the lower box). The  $x$ -axis shows the  $p$ -value intervals and the  $y$ -axis shows the proportion of items in each interval. The figures help to explain the extent to which the Quantile item difficulties were properly targeted to the NC EOG Grades 3 through 8 Mathematics and NC Math 1 item difficulties. In *Figures 15 through 17* for Grades 3 through 5, the Quantile item pool mean  $p$ -values all center between .75 and .81 with large proportions of items at or above .90. In comparison, the NC EOG item pool mean  $p$ -values for Grades 3 through 5 center between .62 and .65 and item difficulties are distributed fairly evenly across the  $p$ -value scale.

The large proportion of Quantile item  $p$ -values at the upper end of the distribution for Grades 3 through 5 (*Figures 15 through 17*) caused the model to inflate the Quantile measures of NC EOG Grades 3 through 5 Mathematics items, and therefore the Quantile measures for students as observed in *Figure 14*. The  $p$ -value distributions for Grades 6 through NC Math 1 show considerable overlap between the distributions of difficulties of Quantile linking items and NC EOG Grades 6 through 8 Mathematics and NC Math 1 items, which resulted in more accurate Quantile calibrations of the NC EOG Grades 6 through 8 Mathematics and NC Math 1 items and, therefore, more accurate Quantile student measures.

Figure 15. Comparing p-values of the Quantile linking and NC EOG Grade 3 Mathematics item pools.

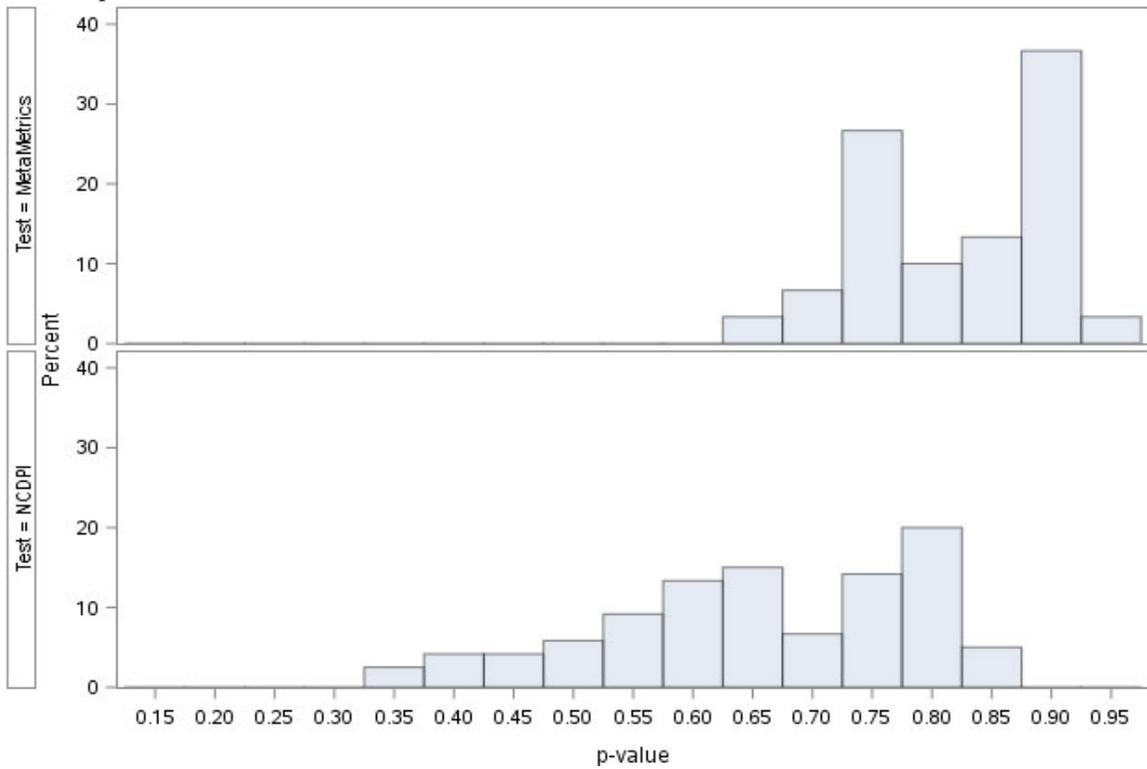


Figure 16. Comparing p-values of the Quantile linking and NC EOG Grade 4 Mathematics item pools.

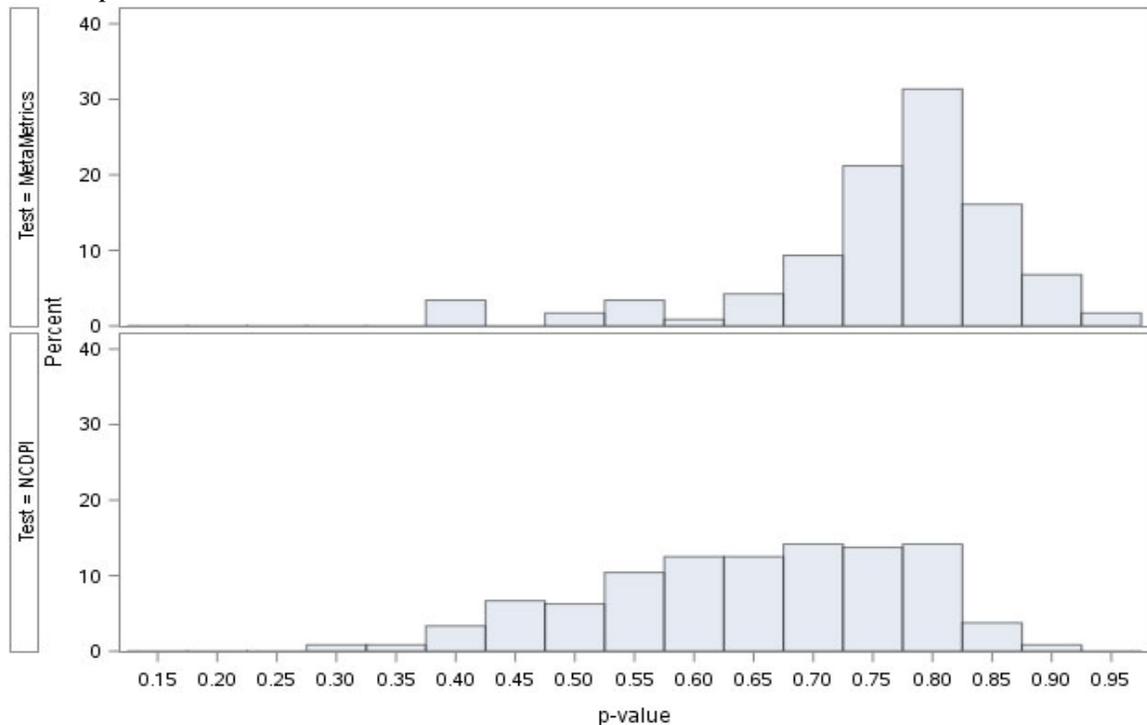


Figure 17. Comparing p-values of the Quantile linking and NC EOG Grade 5 Mathematics item pools.

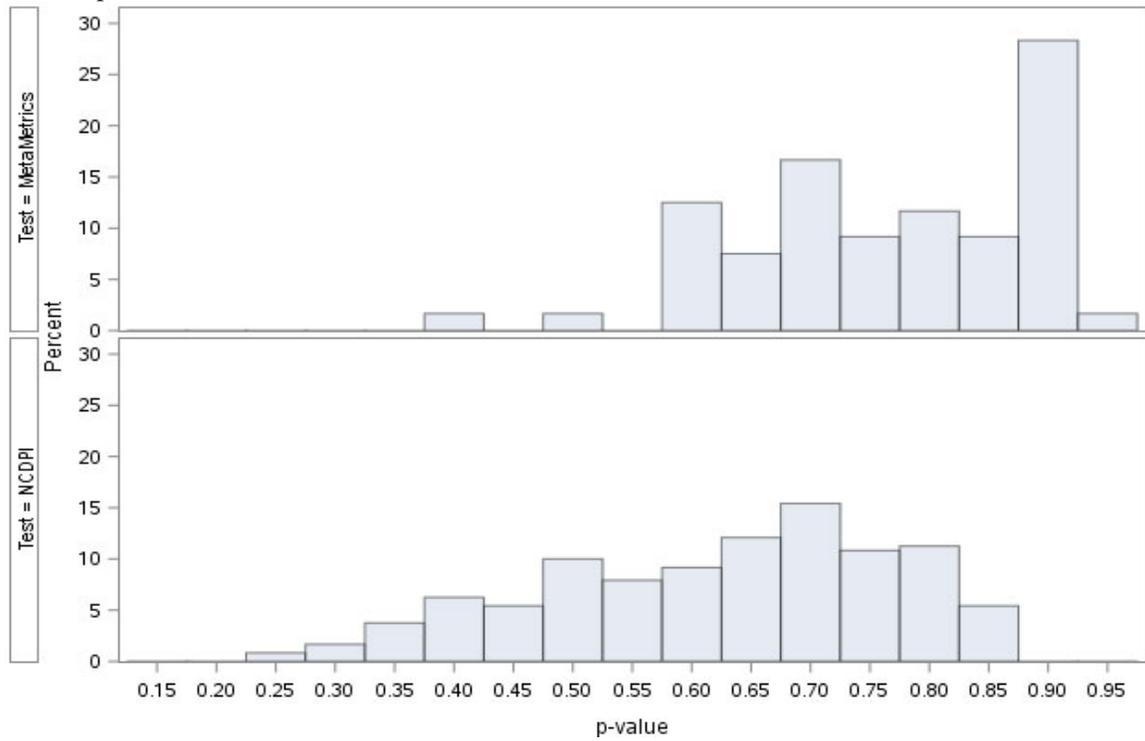


Figure 18. Comparing p-values of the Quantile linking and NC EOG Grade 6 Mathematics item pools.

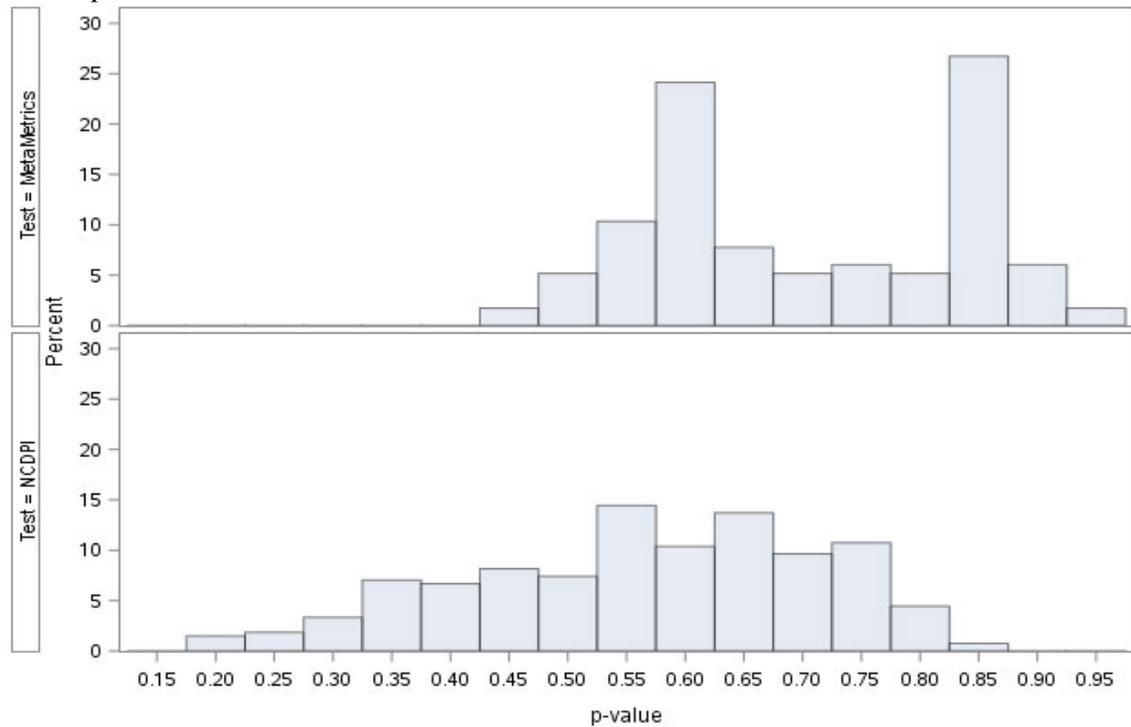


Figure 19. Comparing p-values of the Quantile linking and NC EOG Grade 7 Mathematics item pools.

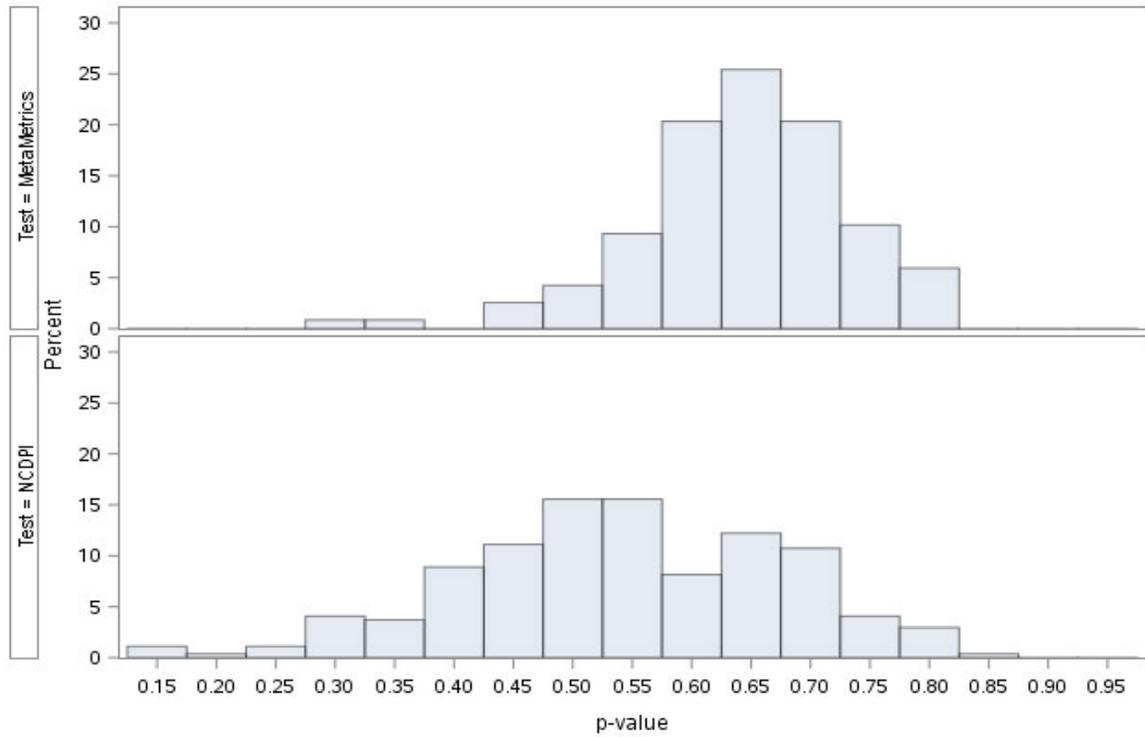


Figure 20. Comparing p-values of the Quantile linking and NC EOG Grade 8 Mathematics item pools.

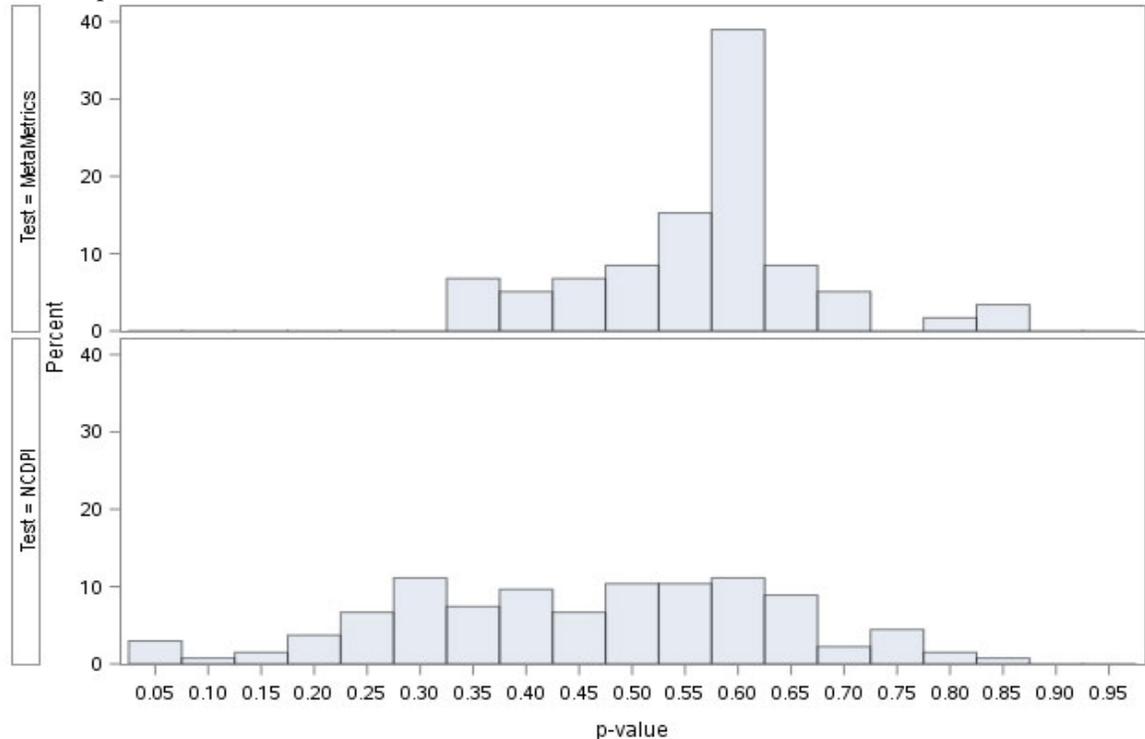
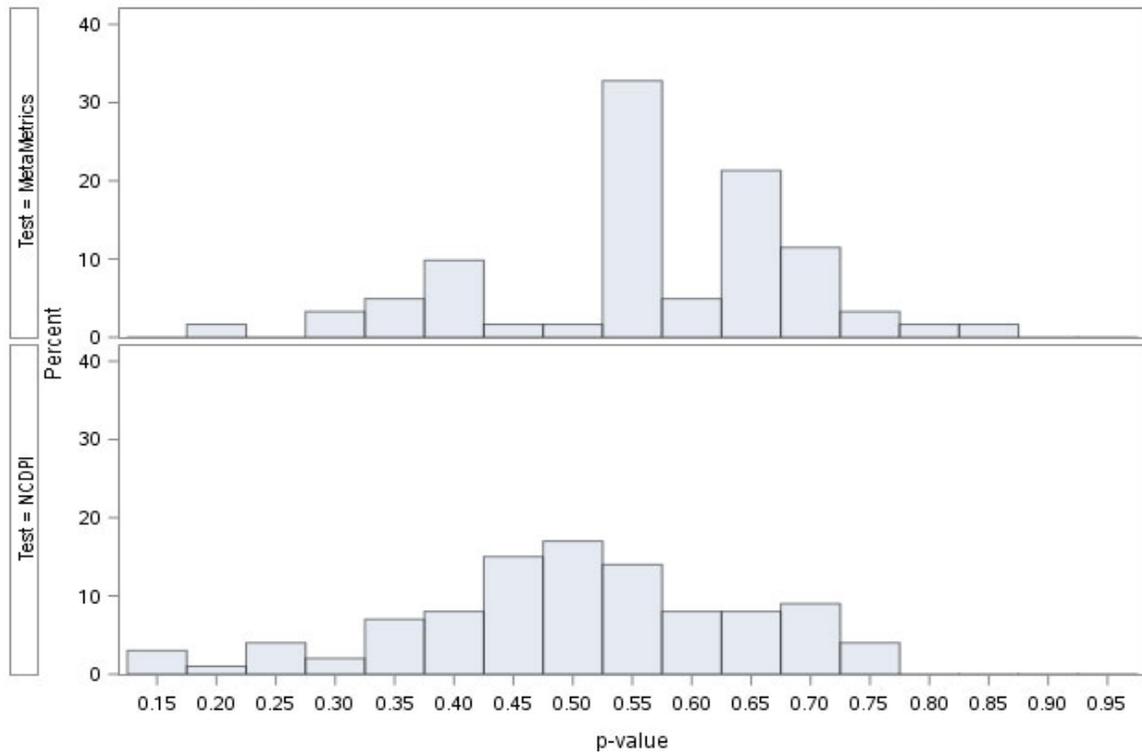


Figure 21. Comparing p-values of the Quantile linking and NC Math 1 item pools.



Additionally, an effect size,  $d$ , on the observed item difficulties from the calibration procedures was calculated between the NC EOG Mathematics and Quantile linking item group means and their pooled standard deviation (Cohen, 1988). An effect size provides an interpretative context for the degree of differences observed in the two distributions. Cohen provided the following suggestions for interpreting an effect size; small,  $d = 0.2$ ; medium,  $d = 0.5$ ; and large  $d = 0.8$ . An additional interpretation provided by Cohen was an effect size of 0.8 indicates a non-overlap of 47.4% in the two distributions. Grades 3, 4 and 5 all had an effect size of 0.87 or higher indicating a large degree of non-overlap between the two item distributions.

Table 18. Effect size difference between the NC EOG Mathematics observed item difficulties and the Quantile linking item difficulties.

| <b>Grade</b> | <b><math>d</math></b> |
|--------------|-----------------------|
| 3            | 1.30                  |
| 4            | 0.87                  |
| 5            | 0.95                  |
| 6            | 0.74                  |
| 7            | 0.73                  |
| 8            | 0.63                  |

MetaMetrics examined how the same items also performed in two other linking studies conducted in Spring 2019 (see *Table 19*). In Grade 4, State A’s 2017 mean NAEP score was six points higher than North Carolina’s while State B’s was one point lower. In Grade 8, the mean 2017 NAEP scale scores for both States A and B were six points higher than the North Carolina mean. Based on 2017 NAEP performance, Quantile item  $p$ -values in States A and B would have been expected to be similar to or higher than item  $p$ -values in North Carolina. The results shown in *Table 19* do not align with expectations for student performance based on the NAEP results. In Grades 3 through 5, the mean  $p$ -value difference is 0.08 or higher, with the largest difference of 0.14 in Grade 4. These results align with the results shown in Figures 15 through 21, where the  $p$ -values for the Quantile linking items in Grades 3 through 5 were the most extreme.

*Table 19. Mean  $p$ -value difference between common items shared by North Carolina and State A or State B.*

| Grade | NC - State A                         | NC - State B                         |
|-------|--------------------------------------|--------------------------------------|
|       | Mean Difference<br>( <i>N</i> items) | Mean Difference<br>( <i>N</i> items) |
| 3     | 0.09 (9)                             | 0.11 (21)                            |
| 4     | 0.14 (4)                             | 0.11 (11)                            |
| 5     | 0.13 (9)                             | 0.08 (15)                            |
| 6     | 0.10 (1)                             | 0.06 (12)                            |
| 7     | 0.06 (7)                             | 0.07 (9)                             |
| 8     | -0.02 (7)                            | 0.01 (16)                            |

Due to the high performance of North Carolina students on the Quantile linking items resulting from possible mistargeting of item difficulties, and a large degree of non-overlap in item difficulties was observed in Grades 3 through 5, it was deemed necessary to make an adjustment in Grades 3 through 5, where the largest differences occurred. The adjustment was implemented in order to provide students with appropriate, rather than out of reach, instructional materials. The North Carolina Standard Course of Study for Grades 3 through 8 was examined to determine whether the mathematical content was similar enough to support maintaining the previous Quantile link. MetaMetrics conducted an alignment on the overall Quantile measures of mathematical content between the previous (2010) and current (2017) curricular frameworks. The Quantile measures for the mathematical content were found to be very similar. The largest difference observed between the mean Quantile measures was 75Q in Grade 5, which is less than one standard deviation on the Quantile scale.

For Grades 3 through 5, the Quantile means and standard deviations from the current study were replaced with the Quantile means and standard deviations from the 2013 link (see MetaMetrics, 2014). All other grades were adequately targeted and supported using the current data collection. *Table 20* provides the means and standard deviations used to establish the linking function for the NC EOG Grades 3 through 8 Mathematics and NC Math 1 scale scores and Quantile measures.

Table 20. Means and standard deviations for the NC EOG Grades 3 through 8 Mathematics and NC Math 1 scale scores and Quantile measures used to establish the linear linking functions, by grade.

| Test Level | NC EOG and NC Math 1<br>Scale Score<br>Mean (SD) | Linking<br>Quantile Measure<br>Mean (SD) |
|------------|--|--|
| 3          |  |  |
| 4          |  |  |
| 5          |  |  |
| 6          |  |  |
| 7          |  |  |
| 8          |  |  |
| NC Math 1  |  |  |

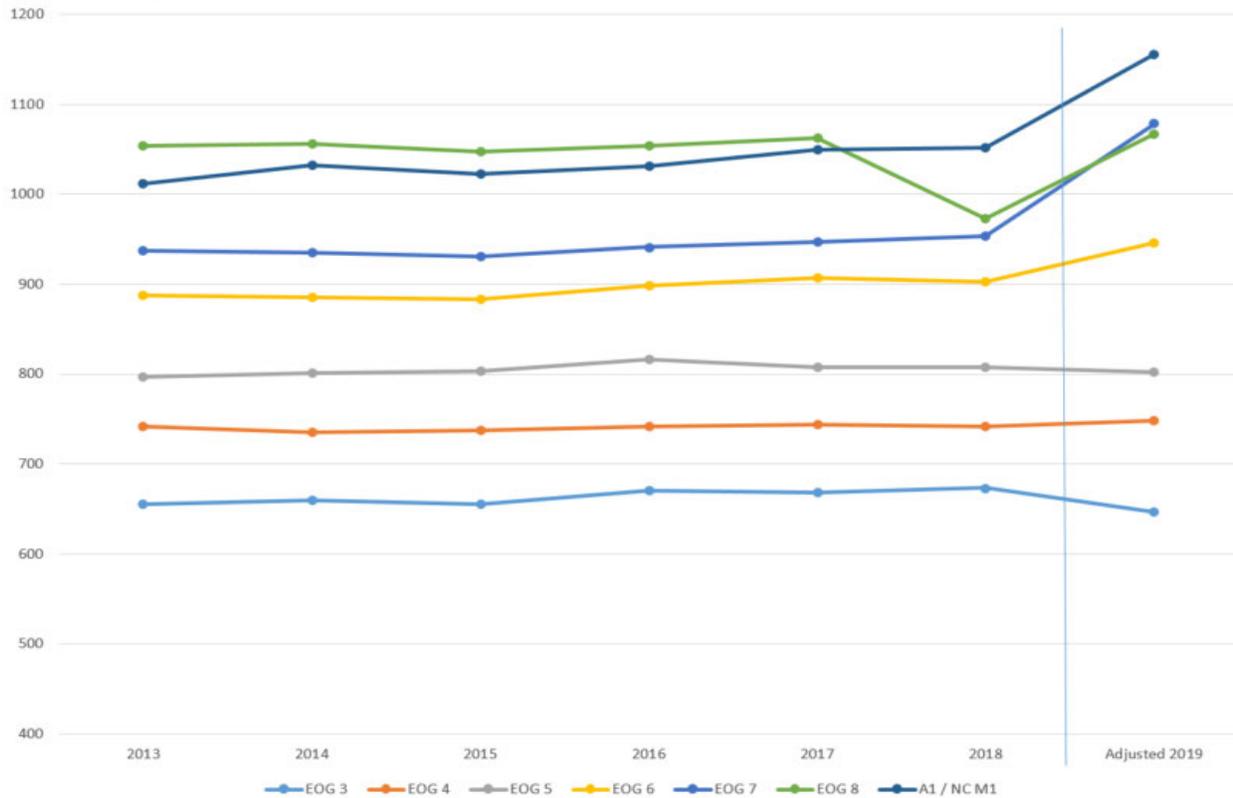
Using Equation 4, Table 21 shows the slope and intercept for each grade-level linking equation. Conversion tables were developed in order to express NC EOG Grades 3 through 8 Mathematics and NC Math 1 scale scores in the Quantile metric and were delivered to NCDPI in electronic format.

Table 21. Linear linking equation coefficients used to predict Quantile measures from NC EOG Grades 3 through 8 Mathematics and NC Math 1 scale scores.

| Grade     | Slope | Intercept |
|-----------|-------|-----------|
| 3         |       |           |
| 4         |       |           |
| 5         |       |           |
| 6         |       |           |
| 7         |       |           |
| 8         |       |           |
| NC Math 1 |       |           |

To affirm the consistency of the adjusted scores from the final link for Grades 3 through 5, historical trends from the “Greenbook” (NCDPI, 2018c) were examined and are presented in Figure 22. The adjustment used to establish the final link provides more stability in the scores across time, with the largest Grades 3 to 5 difference between the 2018 and 2019 mean Quantile measures being -26Q in Grade 3.

Figure 22. Historical NC EOG Mathematics and NC Math 1 performance trends expressed as Quantile measures compared with the final links for 2019.



**Recommendations about reporting Quantile measures.** Quantile measures that are reported for an individual student should reflect the purpose for which they will be used. If the purpose is research (e.g., to measure growth at the student, grade, school, district, or state level), then actual measures should be used at all score points, rounded to the nearest integer. A computed Quantile measure of 772.5Q would be represented as 773Q. If the purpose is instructional, then the Quantile measures should be capped at the upper bound of measurement error (e.g., at or above the 95<sup>th</sup> percentile of the national Quantile user norms) to ensure developmental appropriateness of the instructional material. MetaMetrics expresses these measures used for instructional purposes as “Reported Quantile Measures” and recommends that they be used on individual score reports. The grade level caps used for reporting Quantile measures are shown in *Table 22*.

Table 22. Maximum reported Quantile measures, by Grade.

| Grade     | Quantile Caps |
|-----------|---------------|
| 3         | 975Q          |
| 4         | 1075Q         |
| 5         | 1125Q         |
| 6         | 1280Q         |
| 7         | 1430Q         |
| 8         | 1450Q         |
| NC Math 1 | 1510Q         |

In an instructional environment, all scores below 0Q should be reported as “EMxxxQ”; no student should receive a negative Quantile measure. A Quantile student measure of -150 is reported as EM150Q where “EM” stands for “Emerging Mathematician” and replaces the negative sign in the number. The Quantile scale is like a thermometer, with numbers below zero indicating decreasing mathematical achievement as the number moves away from zero. The smaller the number following the EM code, the more advanced the student is. For example, an EM150Q student is more advanced than an EM200Q student. Above 0Q, measures indicate increasing mathematical achievement as the numbers increase. For example, a 200Q student is more advanced than a 150Q student. The lowest reported value below 0Q is EM400Q.

Some assessments report a Quantile range, which is 50Q above and 50Q below the student’s actual Quantile measure. The Quantile range takes into account measurement error found in the tests and in the Quantile measures of the skills/concepts. If a student attempts material above his or her Quantile range, the level of challenge may be too great for the student to be able to construct an understanding of the skill or concept. Likewise, material below the student’s Quantile range may provide the student with little challenge.

## Validity of the NC EOG Grades 3 through 8 Mathematics and NC Math 1–Quantile Links

**Grade-Level Progressions.** The following box-and-whisker plots (*Figures 23 and 24*) show the progression of NC EOG Grades 3 through 8 Mathematics and NC Math 1 scale scores and Quantile measures (the *y*-axis) from grade to grade (the *x*-axis). For each grade, the box refers to the interquartile range. The line within the box indicates the median. The end of each whisker represents the 5<sup>th</sup> and 95<sup>th</sup> percentile values of the scores (the *y*-axis).

*Figure 23* demonstrates the horizontal nature of the scaling used for the NC EOG Grades 3 through 8 Mathematics and NC Math 1 scores where the NC EOG Grades 3 through 8 Mathematics and NC Math 1 scale scores have a set mean of 550, with the exception of Grade 8 with a mean of 540. *Figure 24* demonstrates the vertical nature of the Quantile scale where, as the grade level increases, the overall Quantile distribution increases, with the exception of Grade 8 where a significant portion of the student population (approximately 27%) participated in NC Math 1 and were not administered the NC EOG Grade 8 Mathematics assessment. This highlights the benefit of having NC EOG Grades 3 through 8 Mathematics and NC Math 1 scores also being reported on a supplemental vertical scale.

Figure 23. Box-and-whisker plot of the NC EOG Grades 3 through 8 Mathematics and NC Math 1 scale scores, linking sample (N = 661,766).

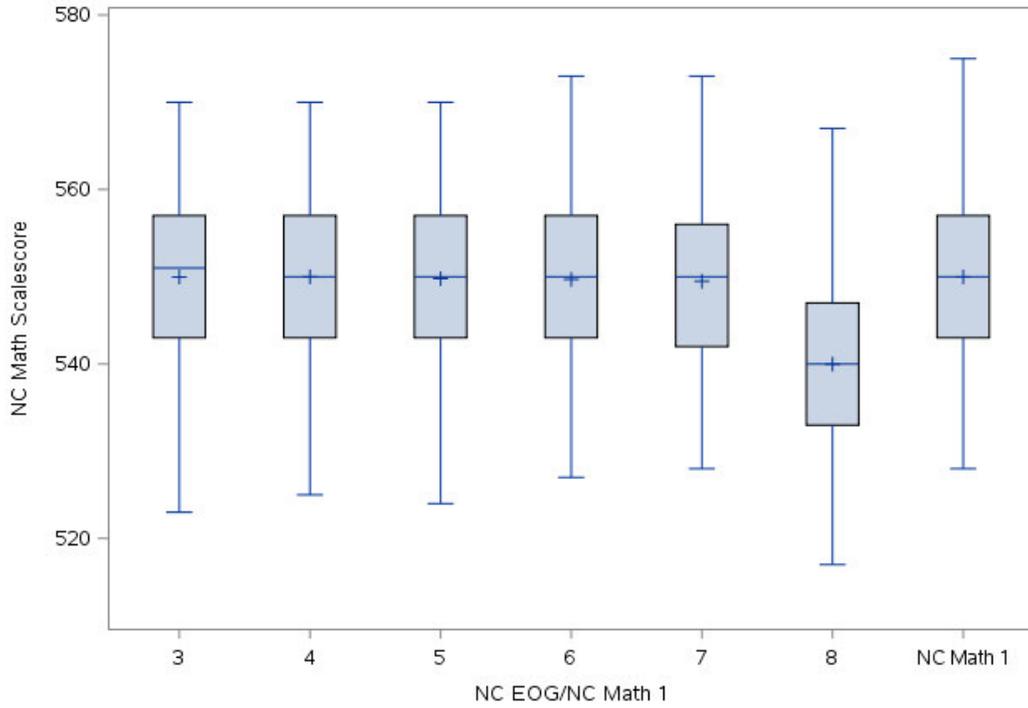
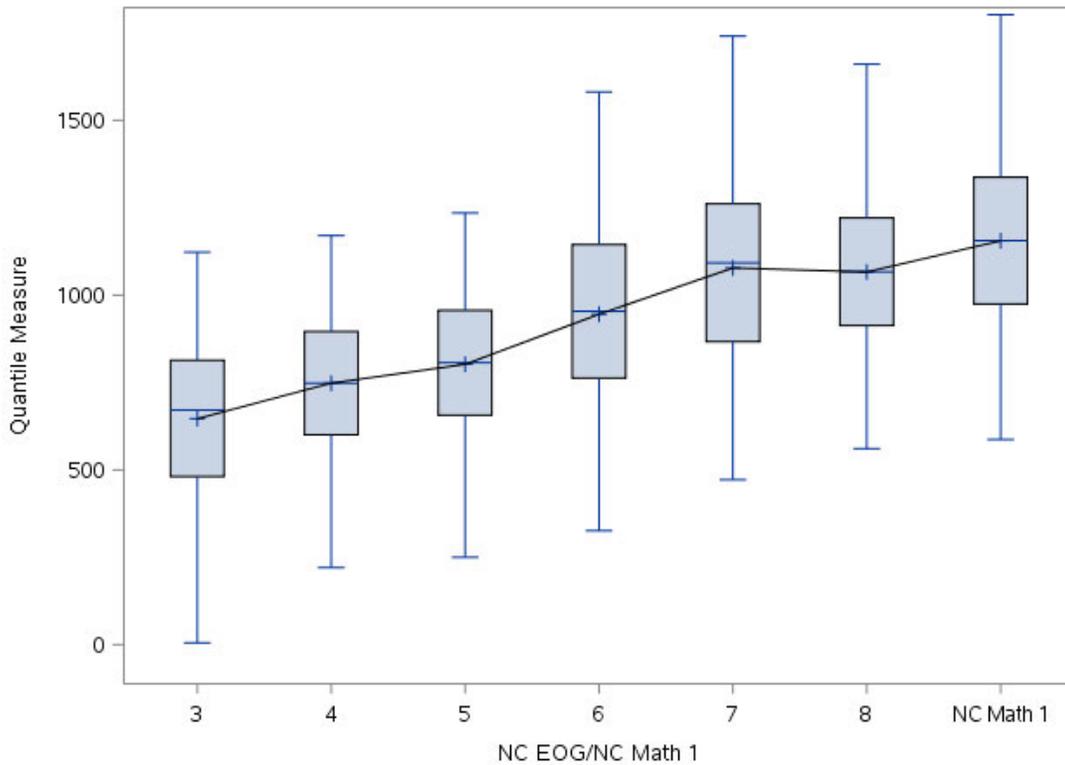


Figure 24. Box-and-whisker plot of the NC EOG Grades 3 through 8 Mathematics and NC Math 1 linking equation Quantile measures, linking sample (N = 661,766).



**NC EOG Grades 3 through 8 Mathematics and NC Math 1 Achievement Levels.** The NC EOG Grades 3 through 8 Mathematics and NC Math 1 assessment scales are divided into four student achievement level categories using three cut points. These four achievement levels and associated cut points are used to describe student results: Not Proficient, Level 3, Level 4, and Level 5. Students who score at or above the “Level 3” cut point are identified as having demonstrated proficiency in grade-level skills and grade-appropriate materials. Student performances that do not reach the cut score for the Level 3 are not considered proficient (North Carolina State Board of Education, 2019).

In *Table 23*, the range of NC EOG Grades 3 through 8 Mathematics and NC Math 1 scale scores and their associated Quantile measures are provided for each achievement level. The achievement levels reported in terms of Quantile measures can provide insight with respect to aligning appropriate instructional materials with student ability.

*Table 23. NC EOG Grades 3 through 8 Mathematics and NC Math 1 scale scores and Quantile measures for achievement levels.*

| Grade     | Not Proficient Scale Score Range | Not Proficient Quantile Measure Range | Level 3 Scale Score Range | Level 3 Quantile Measure Range | Level 4 Scale Score Range | Level 4 Quantile Measure Range | Level 5 Scale Score Range | Level 5 Quantile Measure Range |
|-----------|----------------------------------|---------------------------------------|---------------------------|--------------------------------|---------------------------|--------------------------------|---------------------------|--------------------------------|
| 3         | 520 - 544                        | ██████████                            | 545 - 550                 | ██████████                     | 551 - 559                 | ██████████                     | 560 - 570                 | ██████████                     |
| 4         | 520 - 546                        | ██████████                            | 547 - 551                 | ██████████                     | 552 - 559                 | ██████████                     | 560 - 570                 | ██████████                     |
| 5         | 520 - 545                        | ██████████                            | 546 - 550                 | ██████████                     | 551 - 560                 | ██████████                     | 561 - 570                 | ██████████                     |
| 6         | 525 - 545                        | ██████████                            | 546 - 550                 | ██████████                     | 551 - 560                 | ██████████                     | 561 - 573                 | ██████████                     |
| 7         | 525 - 545                        | ██████████                            | 546 - 549                 | ██████████                     | 550 - 559                 | ██████████                     | 560 - 573                 | ██████████                     |
| 8         | 515 - 542                        | ██████████                            | 543 - 547                 | ██████████                     | 548 - 554                 | ██████████                     | 555 - 570                 | ██████████                     |
| NC Math 1 | 525 - 547                        | ██████████                            | 548 - 554                 | ██████████                     | 555 - 562                 | ██████████                     | 563 - 575                 | ██████████                     |



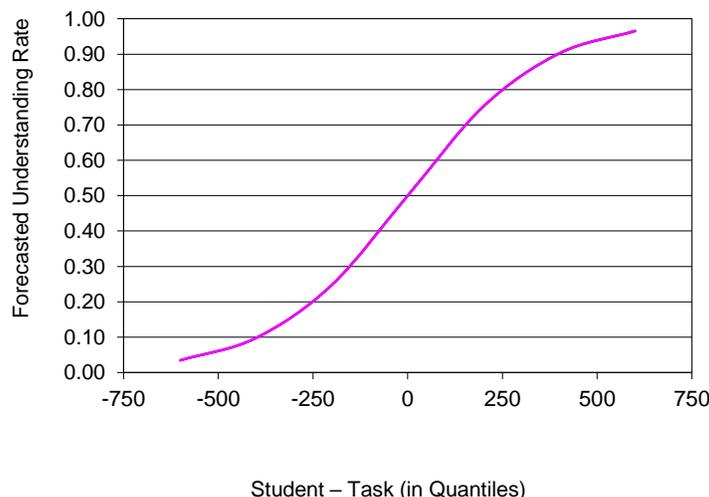
## The Quantile Framework and Instruction

Quantile measures are available from many norm-referenced and criterion-referenced assessments, in addition to state tests and instructional products. Students who take a mathematics achievement test that is linked with the Quantile Framework or one that reports directly in the Quantile metric will receive a Quantile measure. Educators can use these Quantile measures to match students, by *readiness level*, to level-appropriate instructional materials and forecast understanding. For example, a student with a Quantile measure of 500Q should be ready for instruction of mathematics problems at a demand level of 500Q.

**Differentiated Instruction.** A Quantile measure for materials is a number indicating the mathematical demand of the material in terms of the concept/application solvability. The Quantile measure for an individual student is the level at which he or she is ready for instruction (50% competency with the material) and has knowledge of the prerequisite mathematical concepts and skills necessary to succeed. The Quantile scale ranges from below EM400Q to above 1600Q. The Quantile measure does not relate to a specific grade, per se, so the score is developmental as it spans the mathematics continuum from kindergarten mathematics through the content typically taught in Algebra II, Geometry, Trigonometry, and Precalculus. The measure can be used by a teacher to determine what mathematical instruction the student is likely to be ready for next.

*Figure 25* shows the general relationship between the student-task discrepancy and forecasted understanding. When the student measure and the task mathematical demand are the same (difference of 0Q), then the forecasted understanding, or success rate, is modeled as 50% and the student is likely ready for instruction on the particular skill or concept.

*Figure 25. Relationship between student mathematical demand discrepancy and forecasted understanding (success rate).*



An appropriate instructional range for the Quantile measure of a student is 50Q above to 50Q below the Quantile measure of the student (44% - 56% competency). This range identifies the mathematics skills in which a student has the prerequisite knowledge and skills needed to understand the instruction, and will likely have success with tasks related to the skill or concept after this introductory instruction.

Quantile measures provide reliable, actionable results because instruction and assessment are described using the same metric. When instruction is measured at a unique mathematical level of understanding and any form of assessment can be reported using the same scale, equal levels of achievement are observed.

By understanding the interaction between student measures and resource measures (e.g., textbook lessons, instructional materials), any level of understanding can be used as a benchmark. An individual can modulate his or her own likely success rate by lowering the difficulty of the task (i.e., increase to 90% understanding) or increasing the difficulty of the task (i.e., lower to 40% understanding) depending on the situation (refer to *Figure 24*). This flexibility allows the teacher, parent, or student the ultimate control to modulate the fit between person and task.

*Table 24* gives an example of the forecasted understanding (or likely success rates) for specific skills for a specific student. *Table 25* shows forecasted understanding for one specific skill calculated for different student achievement measures.

*Table 24. Success rates for a student with a Quantile measure of 750Q and skills of varying difficulty (demand).*

| <b>Student Mathematics Achievement</b> | <b>Skill Demand</b> | <b>Skill Description</b>  | <b>Forecasted Understanding</b> |
|--|---------------------|---|---------------------------------|
| 750Q                                   | 350Q                | Locate points on a number line.   | 90%                             |
| 750Q                                   | 550Q                | Use order of operations, including parentheses, to simplify numerical expressions.                          | 75%                             |
| 750Q                                   | 750Q                | Translate between models or verbal phrases and algebraic expressions.                                       | 50%                             |
| 750Q                                   | 950Q                | Estimate and calculate areas with scale drawings and maps.  | 25%                             |
| 750Q                                   | 1150Q               | Recognize and apply definitions and theorems of angles formed when a transversal intersects parallel lines. | 10%                             |

Table 25. Success rates for students with different Quantile measures of achievement for a task with a Quantile measure of 850Q.

| Student Mathematics Achievement | Problems Related to "Locate points in all quadrants of the coordinate plane using ordered pairs." | Forecasted Understanding |
|---------------------------------|---|--------------------------|
| 450Q                            | 850Q  | 10%                      |
| 650Q                            | 850Q  | 25%                      |
| 850Q                            | 850Q  | 50%                      |
| 1050Q                           | 850Q  | 75%                      |
| 1250Q                           | 850Q  | 90%                      |

The primary utility of the Quantile Framework is its ability to forecast what will likely happen when students confront resources and instruction on specific mathematical skills and concepts. With every application by teacher, student, or parent there is a test of the Quantile Framework’s accuracy. The Quantile Framework makes a point prediction every time a resource or lesson is chosen for a student. Anecdotal evidence suggests that the Quantile Framework predicts as intended. That is not to say that there is an absence of error in forecasted understanding. There is error in resource measures based on Quantile Skill and Concept (QSC) measures, student measures, and their difference modeled as forecasted understanding. However, the error is sufficiently small that the judgments about students, resources, and understanding rates are useful.

The subjective experience of 25%, 50%, and 75% understanding/success as reported by students varies greatly. A student with a Quantile measure of 1000Q being instructed on materials that measure 1000Q (50% understanding) has a successful instructional experience—he or she has the background knowledge needed to learn and apply the new information. Teachers working with such a student report that the student can engage with the skills and concepts that are the focus of the instruction and, as a result of the instruction, are able to solve problems utilizing those skills. In short, such students appear to understand what they are learning. A student with a Quantile measure of 1000Q being instructed on materials that measure 1200Q (25% understanding) encounters so many unfamiliar skills and difficult concepts that the learning is frequently lost. Such students report frustration and seldom engage in instruction at this level of understanding. Finally, a student with a Quantile measure of 1000Q being instructed on materials that measure 800Q (75% understanding) reports that he or she is able to engage with the skills and concepts with minimal instruction, is able to solve complex problems related to the skills and concepts, is able to connect the skills and concepts with skills and concepts from other strands, and experiences fluency and automaticity of skills.

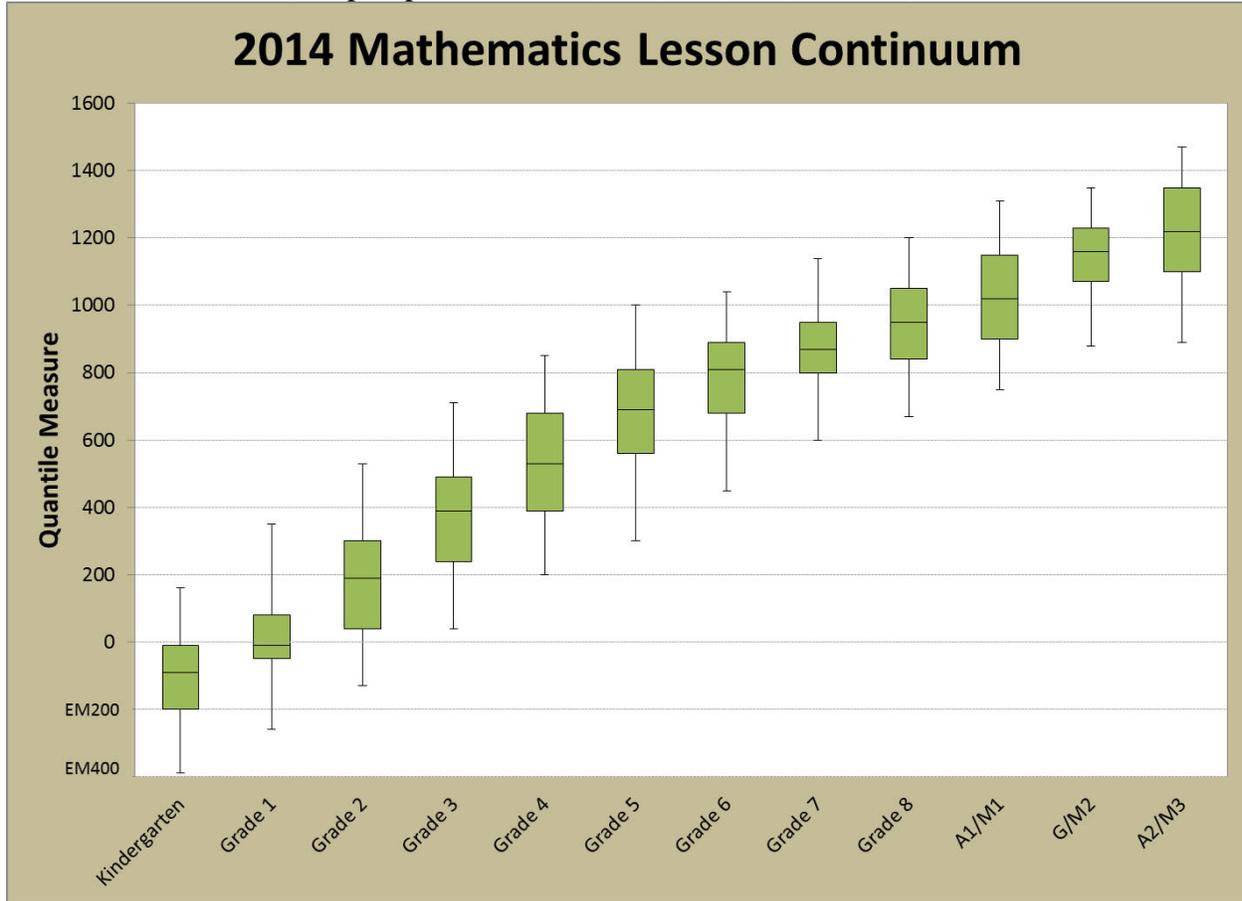
**Quantile Framework and Mathematical Demand in Education and Careers.** There is increasing recognition of the importance of bridging the gap that exists between K-12 and higher education and other postsecondary endeavors. Many state and policy leaders have formed task forces and policy committees such as P-20 councils. Many state curricular frameworks developed over the past 6 to 7 years, along with the Common Core State Standards for Mathematics (CCSSM), were designed to enable all students to become college and career ready by the end of high school. They acknowledge that students are on many different pathways to

this goal: “One of the hallmarks of the Common Core State Standards for Mathematics is the specification of content that all students must study in order to be college and career ready. This ‘college and career ready line’ is a minimum for all students” (NGA Center & CCSSO, 2010b, p. 4). These college- and career-readiness standards for mathematics suggest that “college and career ready” means completing a sequence that covers Algebra 1, Geometry, and Algebra II (or equivalently, Integrated Mathematics 1, 2 and 3) during the middle school and high school years; and leads to a student’s promotion into more advanced mathematics by their senior year. This has led some policy makers to generally equate the successful completion of Algebra II as a working definition of college and career ready. Exactly how and when this content must be covered is left to the states to designate in their implementations throughout K-12.

The *mathematical demand* of a mathematical textbook (in the Quantile metric) quantitatively defines the level of mathematical achievement that a student needs in order to be ready for instruction on the mathematical content of the textbook. Assigning QSCs and Quantile measures to a textbook is done through a calibration process. Textbooks are analyzed at the lesson level and the calibrations are completed by SMEs experienced with the Quantile Framework and with the mathematics taught in mathematics classrooms. The intent of the calibration process is to determine the mathematical demand presented in the materials. Textbooks contain a variety of activities and lessons. In addition, some textbook lessons may include a variety of skills. Only one Quantile measure is calculated per lesson by the Quantile Analyzer and is obtained through analyzing the Quantile measures of the QSCs that have been mapped to the lesson. This Quantile measure represents the composite task demand of the lesson.

MetaMetrics has calibrated more than 80,000 instructional materials (e.g., textbook lessons, instructional resources) across the K-12 mathematics curriculum (Smith & Turner, 2012). *Figure 26* shows the continuum of calibrated textbook lessons from Kindergarten through Algebra II/Math 3 from 27,630 lessons (370 test books) from materials published between 2005 and 2013 (Sanford-Moore, Williamson, Bickel, Koons, Baker, and Price, 2014).

Figure 26. A continuum of mathematical demand for Kindergarten through Precalculus textbooks (box plot percentiles: 5<sup>th</sup>, 25<sup>th</sup>, 50<sup>th</sup>, 75<sup>th</sup>, and 95<sup>th</sup>).

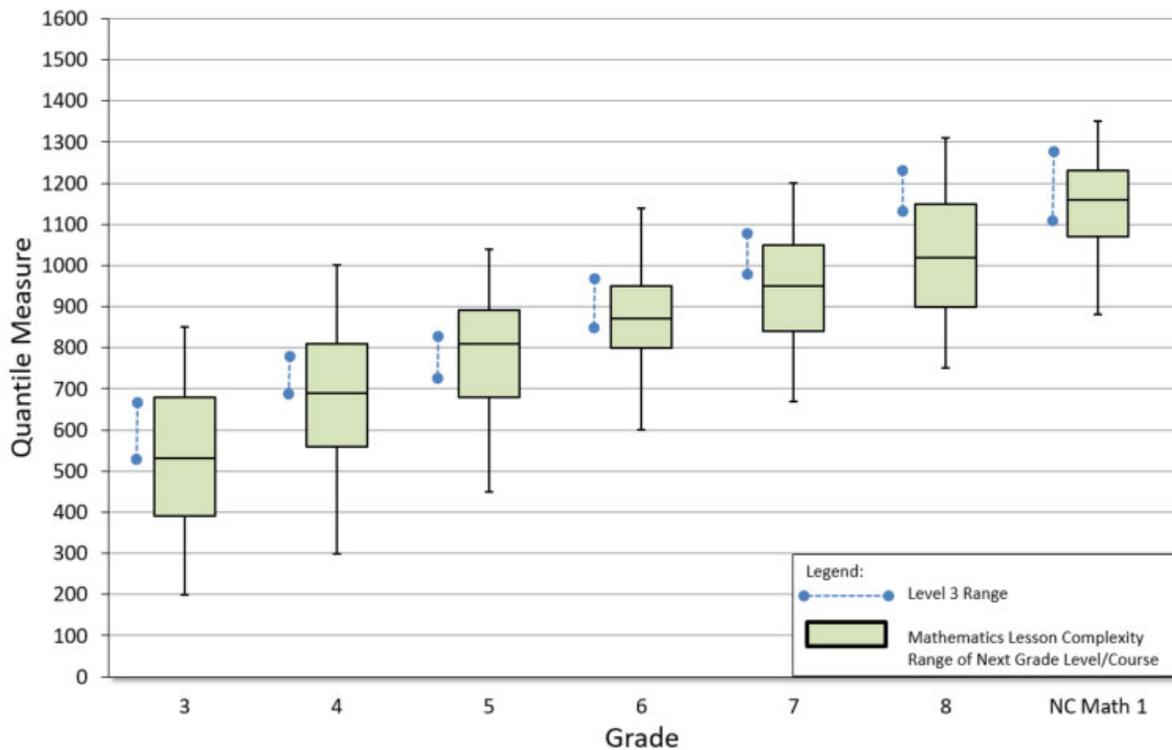


In 2016, Williamson, Sanford-Moore, and Bickel began the examination of the mathematics demands of college and careers to answer the question, “What mathematics will a student likely encounter when entering college or a career?” To address this question, the mathematical concepts and skills that students are likely to encounter as they begin their postsecondary education and/or enter the workplace were examined. For college, being ready for instruction in the types of courses typical of those beyond high school graduation requirements and of first year college were examined (e.g., Precalculus, Trigonometry). For careers, competently performing the mathematics content required for a high school diploma (e.g., Algebra I content, Algebra II content) was examined. In this research, “competently perform” was defined as 75% understanding of the mathematics skills and concepts. The range (interquartile range) of mathematical demands students are likely to encounter as they enter college and careers is 1220Q to 1440Q, with a median of 1350Q.

MetaMetrics’ research on the mathematical demand of college and careers can be used to compare achievement levels from the NC EOG Grades 3 through 8 Mathematics and NC Math 1 assessments with the mathematics skills and concepts a student will likely encounter. *Figure 27* shows the relationship between the “Level 3” achievement level of the NC EOG Grades 3 through 8 Mathematics and NC Math 1 Quantile measures and the mathematics lesson

complexity ranges for the next grade level/course. For each grade/level, the box refers to the interquartile range. The line within the box indicates the median. The end of each whisker represents the 5<sup>th</sup> percentile at the low end of the distribution of mathematical demand distribution and the 95<sup>th</sup> percentile at the high end of the distribution. The Level 3 achievement level is within the mathematics lesson complexity ranges for the next grade level/course across all grades. This supports the interpretation that students at “Level 3” or above will be able to successfully engage with the material at the next grade level.

Figure 27. Comparison of NC EOG Grades 3 through 8 Mathematics and NC Math 1 Quantile measures for the Level 3 achievement level and the mathematical demand at the next grade.

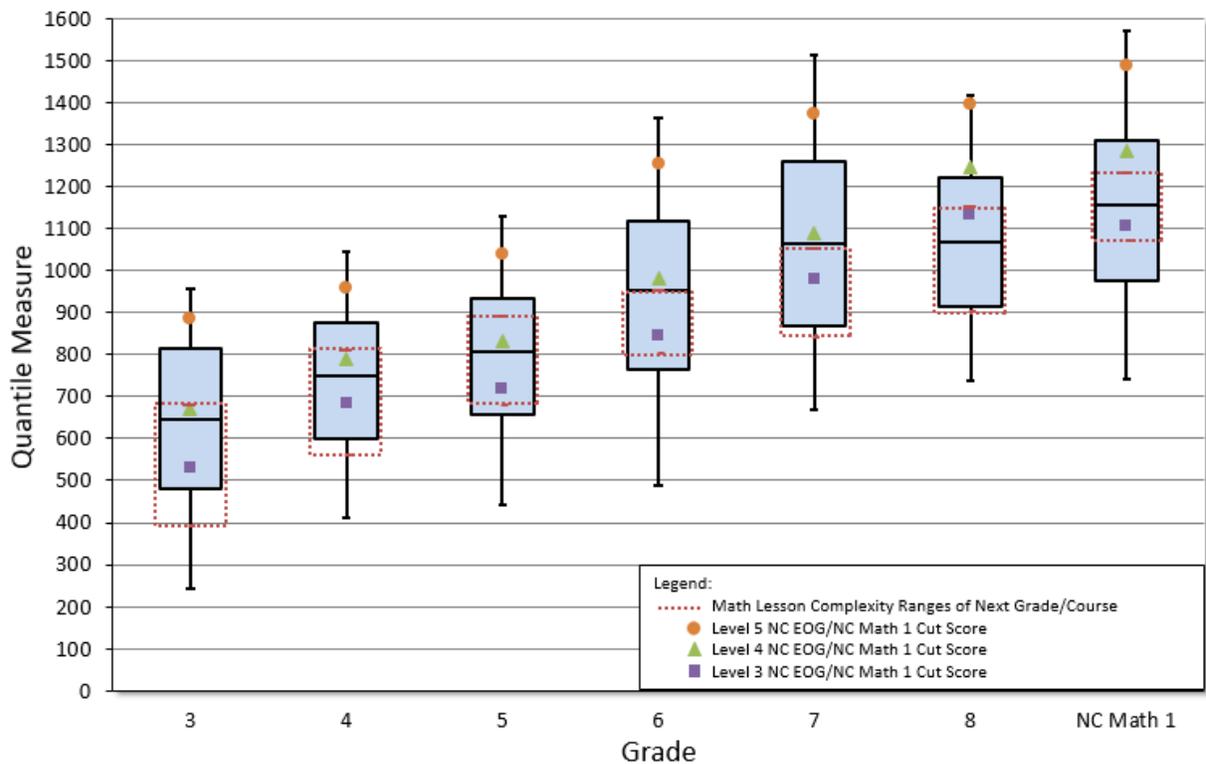


To better understand the results from the current Quantile linking study, student achievement levels from the NC EOG Grades 3 through 8 Mathematics and NC Math 1 assessments can be compared with the distribution of student scores as Quantile measures and the mathematical demands of the instructional materials the students will likely encounter. Figure 28 shows the spring 2019 student results from the NC EOG Grades 3 through 8 Mathematics and NC Math 1 assessments as Quantile measures. For each test level, the box refers to the interquartile range of student results. The line within the box indicates the median. The end of each whisker represents the 5<sup>th</sup> percentile at the low end of the distribution of scores and the 95<sup>th</sup> percentile at the high end of the distribution. The square, triangle, and circle represent the NC EOG Grades 3 through 8 Mathematics and NC Math 1 achievement level-cut scores as Quantile measures for “Level 3”,

“Level 4”, and “Level 5”, respectively. Additionally, the dotted box provides a reference for the complexity of lessons students will encounter at the next grade level in mathematics.

All grades show that the Level 3 cut point is within or above the range of the mathematical demands of the following school year’s mathematics content based on MetaMetrics research by Sanford-Moore et. al. (2014). Ultimately, placing all the information on the same scale allows students to be matched with instructional materials targeted to the skills and concepts students will likely encounter as they enter the next grade level and, ultimately, as they enter college and careers.

Figure 28. NC EOG Grades 3 through 8 Mathematics and NC Math 1 student achievement (Spring 2019) expressed as Quantile measures compared to the NC EOG and NC Math 1 student achievement levels and mathematical lesson demand distributions.





## Conclusions

The purpose of this study was to establish a linkage between the scores on the NC EOG Grades 3 through 8 Mathematics and NC Math 1 assessments and the Quantile scale. A common-person design was employed because it was logically possible to administer two sets of test items to the same group of students and because differential order effects were not expected to occur (Kolen & Brennan, 2014). The linking study was conducted through three major phases: (i) evaluating the relationship between the NC EOG Grades 3 through 8 Mathematics and NC Math 1 assessments and the Quantile Framework, (ii) linking two score scales using linear equating methodology, and (iii) providing validity evidence for the linkage.

The linking procedures included a process to ensure that a similar construct was measured by the NC EOG Grades 3 through 8 Mathematics and NC Math 1 assessments and the Quantile Framework. First, the Quantile linking item pools were constructed to align the QSCs with the NC EOG Grades 3 through 8 Mathematics and NC Math 1 content standards. The Quantile linking items exhibited psychometric quality, including fit to the Quantile Framework and classical item statistics. However, it was observed in Grades 3 through 5 that items were less difficult than have been historically observed in previous linking studies. Additionally, the calibrated student Quantile measures in Grades 3 through 5 were high compared both to the Quantile user norms and previous results from the 2013 North Carolina Quantile linking study (MetaMetrics, 2014). However, a strong correlation between the NC EOG Grades 3 through 8 Mathematics and NC Math 1 scale scores and the calibrated Quantile measures was observed, indicating that both scales yield consistent scores. This information resulted in (1) replacing the NC EOG Grades 3 through 5 Mathematics mean Quantile measures and standard deviations from this study with those of the 2013 linking study to construct the linking equations at those grades and (2) using the current study results to establish the link for the NC EOG Grades 6 through 8 Mathematics and the NC Math 1. The primary purpose of the Quantile Framework is to provide appropriate instructional materials for students given their ability level. The results observed in Grades 3 through 5 necessitated the adjustment to avoid overestimation of Quantile measures and avoid assigning instructional materials that are too challenging for students.

Next, linear linking was employed to establish a link between the NC EOG Grades 3 through 8 Mathematics and NC Math 1 scales and the Quantile scale. A linear link provides a symmetric link with multiple advantages, including ease of interpretation and converting from one scale to the other. Once the link was established, Quantile measures could then be reported alongside the NC EOG Grades 3 through 8 Mathematics and NC Math 1 scale scores.

Finally, validity evidence supporting the link between the NC EOG Grades 3 through 8 Mathematics and NC Math 1 scales and the Quantile scale was provided. NC EOG Grades 3 through 8 Mathematics and NC Math 1 scale scores were placed on a vertical scale, the Quantile scale. Student performance on the NC EOG Grades 3 through 8 Mathematics and NC Math 1 assessments was compared with the Quantile norms showing North Carolina students as a more able group than the Quantile norming population with all selected percentiles being higher than the Quantile user norms. Lastly, when comparing the achievement level standards to a criterion-based outcome, lesson complexity range, the passing and proficient ranges (i.e., Levels 3, 4, and

5) were at or above the lesson complexity students will likely encounter in the next grade or course.

Now that a linkage is established between the NC EOG Grades 3 through 8 Mathematics and NC Math 1 scale and the Quantile scale, educators are able to utilize the assessment results, reported in Quantile measures, to inform classroom instruction. The following sections provide a more detailed description about the caveats associated with the study and recommended uses of the Quantile Framework and associated tools.

## **Caveats**

**Quantile Measures and Grade Levels.** Quantile measures do not translate specifically to grade levels. Within each grade, there will be a range of mathematics ability and a range of instructional materials. In a fourth-grade classroom there will be some students far ahead of others and some students far behind others in terms of their mathematics ability. However, the Quantile Framework can be used to identify students who are ready for instruction of a particular skill, regardless of grade level.

Simply because a student is an excellent mathematician, it should not be assumed that the student will necessarily comprehend a mathematical skill if they have not had the opportunity to learn the prerequisite skills. Without adequate background knowledge and prior instruction, the student may not have had sufficient exposure to the immediate skill being taught. A benefit of the Quantile Framework is that the prerequisite skills for the focal mathematical skill or concept can be identified and taught prior to the lesson as a way to prepare students for success. Moreover, the Quantile Framework provides the connections between skills to facilitate learning and to get students on track with the skills they will likely encounter as they progress throughout school. Likewise, similar features can be used to provide enrichment materials for advanced students by utilizing the impending skills of a Quantile Knowledge Cluster.

**Maintenance of the NC EOG Grades 3 through 8 Mathematics and NC Math 1 scales.** Maintenance of the focal scale (i.e., NC EOG Grades 3 through 8 Mathematics and NC Math 1 scales) is critical to the validity of any link with an auxiliary scale (i.e., Quantile scale). If an update occurs to the focal scale, the integrity of the link needs to be re-examined and additional linking studies may need to occur to incorporate those fundamental changes to the focal scale. Such updates include, but are not limited to, the incorporation of new item types into the scale; or a revision of the assessment program and, therefore, the reported scale scores.

## **Next Steps for Using The Quantile Framework for Mathematics**

To utilize the results from this study, Quantile measures need to be incorporated into the NC EOG Grades 3 through 8 Mathematics and NC Math 1 assessment results processing and interpretation frameworks. Suggested resources need to be developed for ranges of students. Care must be taken to ensure that the resources and materials on the lists are also developmentally appropriate for the students. The Quantile measure is one factor related to

understanding, and is a good starting point in the selection process of materials and resources for a specific student. Other factors such as student developmental level, motivation and interest; amount of background knowledge possessed by the student; and characteristics of the resources and skills also need to be considered when matching resources and instruction with a student.

In this era of student-level accountability and high-stakes assessment, differentiated instruction—the attempt “on the part of classroom teachers to meet students where they are in the learning process and move them along as quickly and as far as possible in the context of a mixed-ability classroom” (Tomlinson, 1999)—is a means for all educators to help students succeed. Differentiated instruction promotes high-level and powerful curriculum for all students, but varies the level of teacher support, task complexity, pacing, and avenues to learning based on student readiness, interest, and learning profile. One strategy for managing a differentiated classroom suggested by Tomlinson is the use of multiple resources and supplementary materials that can be identified with the aid of the Quantile Framework. Equipped with a student’s Quantile measure, teachers can connect a student with textbook lessons, worksheets, games, websites, and trade books that have appropriate Quantile measures (Smith, no date; Smith and Turner, 2012). By incorporating Quantile measures into the planning of mathematics instruction, it becomes possible to forecast with greater probability how successful students are likely to understand the material presented to them. Teachers can provide instruction on QSCs with Quantile measures below the targeted instruction when students are not ready for that instruction by focusing on prerequisite QSCs. On the other hand, teachers can focus enrichment activities on the impending QSCs.

Three resources are available on the Quantile Framework website – the Quantile Math Skills Database, the Quantile Teacher Assistant, and Quantile Math@Home (Smith, 2010; Smith and Turner, 2012, no date). The Math Skills Database ([hub.lexile.com/math-skills-database](http://hub.lexile.com/math-skills-database)) allows teachers and parents to search for Quantile Skills and Concepts (QSCs) by their state standards, by keywords (e.g., adding fractions), and by Quantile measure. The database contains targeted, free resources appropriately matched to students by Quantile measure and math content. In order to support instruction with the many resources connected with the Quantile Framework, the Quantile Teacher Assistant (QTA) was developed to simplify and gather all relevant information. When using the QTA ([hub.lexile.com/quantile-teacher-assistant](http://hub.lexile.com/quantile-teacher-assistant)), teachers can identify a specific state objective or a CCSSM standard and determine the knowledge base. In addition, teachers can differentiate instruction by indicating the range of Quantile measures for their students in their classrooms. Quantile Math@Home ([hub.lexile.com/math-at-home](http://hub.lexile.com/math-at-home)) activities reinforce mathematical skills covered in the previous school year and lay the groundwork for what will be taught when students return to class in the fall. By incorporating fun family games into everyday activities, students can practice mathematical skills year-round and parents can feel more confident about helping their children with mathematics.

MetaMetrics has conducted extensive research to describe the mathematics demands students will likely encounter as they enter college. This research is being extended to describe the mathematics demands of careers student may enter after high school or after additional postsecondary education. Currently, the mathematics demands of 406 careers have been examined and the results are available in the Quantile Career Database ([hub.lexile.com/career-database](http://hub.lexile.com/career-database)).

MetaMetrics, in partnership with The Council of Chief State School Officers, has begun coordinating a national, state-led summer mathematics initiative to bolster student mathematics achievement during summer break. The Summer Math Challenge is designed to raise national awareness of the summer loss epidemic (Cooper, Nye, Charlton, Lindsay, and Greathouse, 1996), share compelling research on the importance of targeted mathematics activities, and provide access to a variety of free resources to support mathematics instruction and the initiative as a whole.

The “Summer Math Challenge” is a six-week, e-mail-based initiative designed to help students on summer vacation fight “summer slide” in mathematics skills. The initiative combats summer math slide by helping students retain mathematics skills acquired during the previous school year. The initiative, started in the summer of 2013, targets Grades 1 through 8 by reinforcing mathematics concepts presented from Kindergarten through Grade 7 aligned with college- and career-readiness standards for mathematics. Participants receive targeted instructional materials for a weekly concept along with personalized e-mail activity suggestions and resources that support each concept. Features include activities grounded in everyday life on “Real World Wednesdays,” and online math fact fluency practice on “Fluency Fridays.” Thirty SEA chiefs requested assistance in launching a 2019 Summer Math initiative in conjunction with the CCSSO Chief’s Summer Reading Challenge. Support materials for states and schools are available on the Quantile web site at <https://www.quantiles.com/parents-students/find-math-resources-to-support-classroom-learning/summer-math-challenge/>. Students from all 50 U.S. states participated in the 2019 Summer Math Challenge.

The following list suggests ways to leverage a student’s Quantile measure in the classroom:

- Start class with warm-up problems and activities related to the prerequisite skills from a Quantile Knowledge Cluster.
- Enhance major themes of mathematics by building a bank of skills at varying levels that not only support a theme but also provide a way for all students to participate in the theme successfully. For example, consider how addition progresses from single numbers to multi-digit numbers, and then moves to decimals and fractions.
- Sequence mathematical skills according to their difficulty as much as possible.
- Develop a mathematics folder that goes home with students and returns weekly for review. The folder can contain examples of practice skills within a student’s range, applications of topics outside the classroom, reports of recent assessments, and a parent form to record the amount of time spent working mathematics problems at home.
- Choose skills lower in a student’s Quantile range when factors make the student view mathematics as more challenging, threatening, or unfamiliar. Select skills at or above a student’s range to stimulate growth, when a topic holds high interest for a student, or when additional support such as background teaching or peer tutoring is provided.
- Develop individualized lists of skills that are tailored to provide appropriately challenging and curriculum suitable for all students.

Below are some suggestions related to leveraging a student’s Quantile measure at home:

- Ensure that each child gets plenty of mathematical practice, concentrating on skills within his or her Quantile range. Parents can ask their child’s teacher to print a list of appropriate skills or search the Quantile Math Skills Database on the Lexile & Quantile Hub ([hub.lexile.com](http://hub.lexile.com)).
- Communicate with the child’s teachers about the child’s mathematical needs and accomplishments. They can use the Quantile scale to describe their assessment of the child’s mathematical achievement.
- When a new topic proves too challenging for a child, use activities or other materials from the website to help. Review the prerequisite QSCs to ensure that gaps or misconceptions are not interfering with the current topic.
- Celebrate a child’s mathematical accomplishments. The Quantile Framework provides an easy way for students to track their own growth. Parents and children can set goals for mathematics—spending a certain amount of time daily working on mathematical problems, discussing situational topics such as statistics from a newspaper or discounts at the store, reading a book about a mathematical topic, trying new kinds of websites and games, or working a certain number of mathematics problems per week. When children reach the goal, make it an occasion!



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## The Quantile<sup>®</sup> Framework for Mathematics Map





Imagine empowering and accelerating students' learning in mathematics by better differentiating instruction and monitoring growth in student ability. With the Quantile Framework, educators can help achieve this goal by identifying level-appropriate mathematical tasks for students and track their progress!

**HOW IT WORKS**

The Quantile Framework for Mathematics is a unique measurement system that uses a common scale and metric to assess a student's mathematical achievement level and the difficulty of specific skills and concepts. The Quantile Framework describes a student's ability to solve mathematical problems and the demand of the skills and concepts typically taught in kindergarten mathematics through Algebra II, Geometry, Trigonometry and Precalculus. The Quantile Map provides educators with a sampling of primary mathematical skills and concepts from over 500 Quantile Skills and Concepts (QSCs) throughout the Quantile scale. This sampling of QSCs ranges from EM (Emerging Mathematician) for early, foundational mathematical skills and concepts to 1500Q for more advanced skills and concepts. As the difficulty, or demand of the skill increases, so does the Quantile measure.

**HOW TO USE IT**

With the Quantile Framework, educators can explore the interconnectedness of mathematical skills and concepts and identify those elements that are critical for progressing student learning. Educators are better able to inform their instruction on how to best teach a skill or concept by pinpointing which skills build upon each other. The skill mapping of mathematical concepts enables educators to build an instructional path that best fits their students'

unique abilities. Both students and QSCs receive a Quantile measure. Numerous tests report Quantile student measures including many state end-of-year assessments, national norm-referenced assessments and math programs. On the QSC side, more than 580 textbooks, 64,000 lessons and 3,100 downloadable resources have received Quantile measures.

Quantile measures provide educators with the information they need to identify gaps in mathematical knowledge, as well as serve as a guide for progressing to more advanced topics. Every QSC is part of a knowledge cluster that shows relationships and connections between mathematical skills and offers their relative difficulty among different skills. Both the *prerequisite* and *impending* skills are elements of knowledge clusters and serve as building blocks that support students' success. Educators can advance student learning by using prerequisite and impending skills to build mathematical knowledge and understanding. Prerequisite skills help educators see the pieces of the puzzle that make up a skill or concept, showing what needs to be understood first. Impending skills are skills and concepts that build upon a focus skill and allow educators to see a trajectory of knowledge across grades and content strands.



James: 1190Q



Sophia: 770Q



Donald: 450Q



Aliyah: EM100Q





## High School Example James

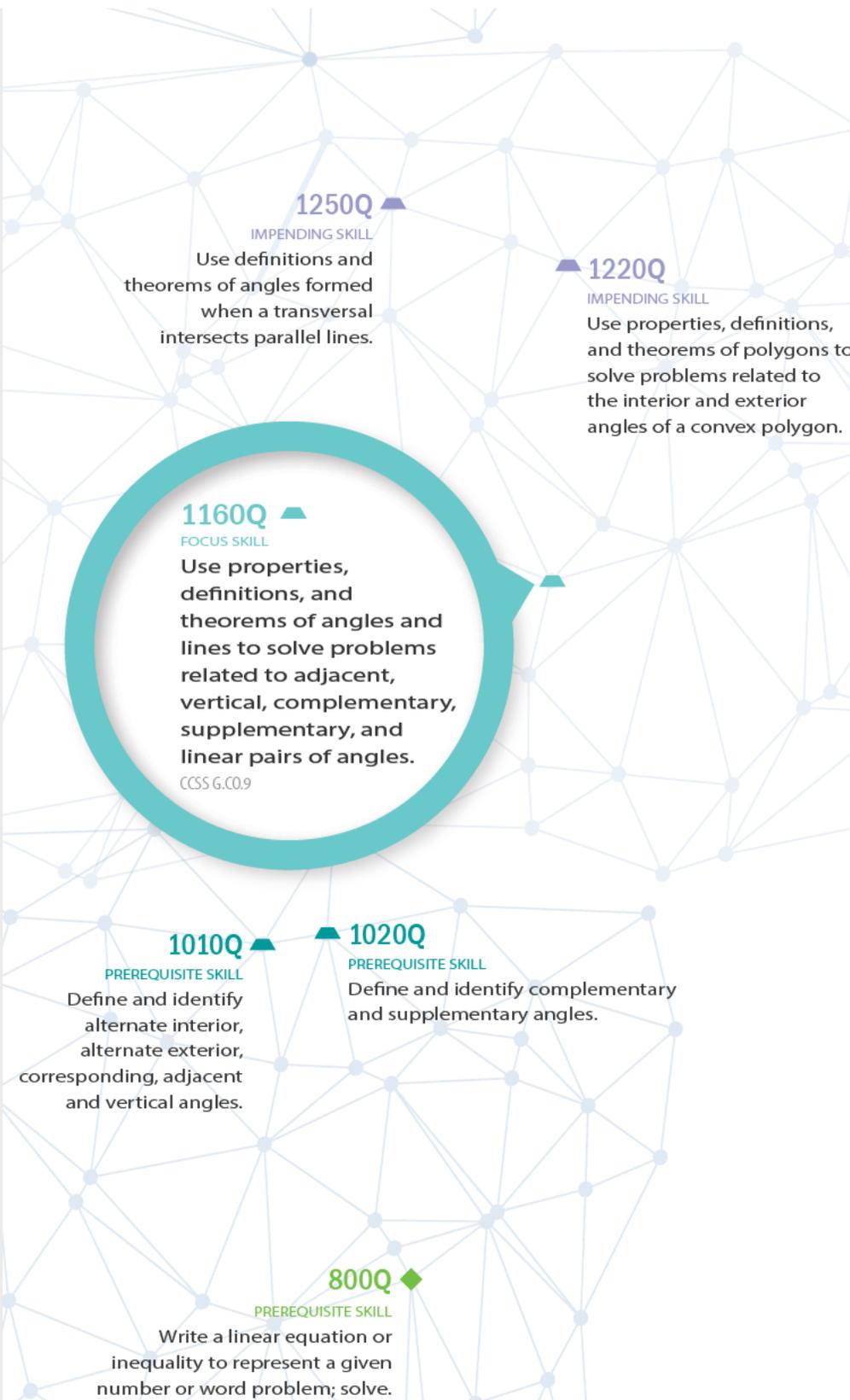
Heritage High School | Geometry Course

Quantile Measure: 1190Q



James is exploring theorems about lines and angles in his Geometry class. In his current learning path, the focus skill being taught is *use properties, definitions, and theorems of angles and lines to solve problems related to adjacent, vertical, complementary, supplementary, and linear pairs of angles*. This focus skill is part of a knowledge cluster that contains prerequisite and impending skills. Working with prerequisite skills can help students struggling to learn and impending skills can help students progress to the next level of learning.

Since James' Quantile measure is within the range of the focus skill being taught (his Quantile measure +/- 50Q), James will be ready for this type of instruction. With his mathematical ability being at the same level as the focus skill, learning will be optimal. Once James is performing well with the focus skill, he will be better prepared to learn the impending skills connected with this focus skill.





## Middle School Example Sophia

Heritage Middle School | Grade 6

Quantile Measure: 770Q



Sophia is using variables to represent mathematical expressions in her math class. In her current learning path, the focus skill being taught is *translate between models or verbal phrases and algebraic expressions*. This focus skill is part of a knowledge cluster that contains prerequisite and impending skills. Working with prerequisite skills can help students struggling to learn and impending skills can help students progress to the next level of learning.

Since Sophia's Quantile measure is within the range of the focus skill being taught (her Quantile measure +/- 50Q), Sophia will be ready for this type of instruction. With her mathematical ability being at the same level as the focus skill, learning will be optimal. Once Sophia is performing well with the focus skill, she will be better prepared to learn the impending skills connected with this focus skill.





## Late Elementary Example Donald

Heritage Elementary School | Grade 4  
Student Quantile Measure: 450Q



Donald is learning about line graphs with very large data values. In his current learning path, the focus skill being taught is *organize, display, and interpret information in graphs containing scales that represent multiple units*. This focus skill is part of a knowledge cluster that contains prerequisite and impending skills. Working with prerequisite skills can help students struggling to learn and impending skills can help students progress to the next level of learning.

Since Donald's Quantile measure is within the range of the focus skill being taught (his Quantile measure +/- 50Q), Donald will be ready for this type of instruction. With his mathematical ability being at the same level as the focus skill, learning will be optimal. Once Donald is performing well with the focus skill, he will be better prepared to learn the impending skills connected with this focus skill.

800Q ▲

IMPENDING SKILL

Identify and use appropriate scales and intervals in graphs and data displays.

480Q ▲

FOCUS SKILL

Organize, display, and interpret information in graphs containing scales that represent multiple units.

CCSS 3.MD.3

480Q ▲

IMPENDING SKILL

Organize, display, and interpret information in bar graphs.

470Q ▲

IMPENDING SKILL

Organize, display, and interpret information in line graphs.

200Q ▲

PREREQUISITE SKILL

Organize, display, and interpret information in line plots and tally charts.

★ 90Q

PREREQUISITE SKILL

Skip count by 3s, 4s, 6s, 7s, 8s, and 9s.

★ 110Q

PREREQUISITE SKILL

Skip count by 2s, 5s and 10s beginning at any number.

EM10Q ▲

PREREQUISITE SKILL

Organize, display, and interpret information in picture graphs and bar graphs using grids.



## Early Elementary Example Aliyah

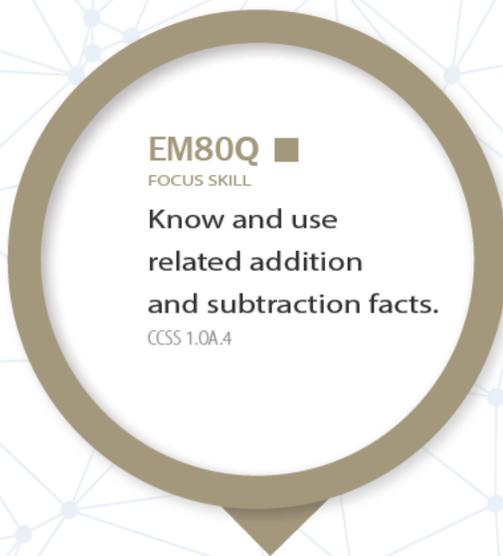
Heritage Elementary School | Kindergarten

Quantile Measure: EM100Q



Aliyah is exploring unknown-addend problems in her class. In her current learning path, the focus skill being taught is *know and use related addition and subtraction facts*. This focus skill is part of a knowledge cluster that contains prerequisite and impending skills. Working with prerequisite skills can help students struggling to learn and impending skills can help students progress to the next level of learning.

Since Aliyah's Quantile measure is within the range of the focus skill being taught (her Quantile measure +/- 50Q), Aliyah will be ready for this type of instruction. With her mathematical ability being at the same level as the focus skill, learning will be optimal. Once Aliyah is performing well with the focus skill, she will be better prepared to learn the impending skills connected with this focus skill.



**EM80Q** ■  
FOCUS SKILL  
**Know and use  
related addition  
and subtraction facts.**  
CCSS 1.OA.4

■ **EM25Q**  
IMPENDING SKILL  
**Model the concept of  
subtraction using numbers  
less than or equal to 10.**

**EM110Q** ◆  
PREREQUISITE SKILL  
**Identify missing addends  
for addition facts.**

■ **EM260Q**  
PREREQUISITE SKILL  
**Model the concept of addition  
for sums to 10.**