New York State Regents Examination in English Language Arts

2023 Technical Report



Prepared for the New York State Education Department by Pearson

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Chapter 1: Introduction

1.1. INTRODUCTION

This technical report for the Regents Examination in English Language Arts (ELA), based on the New York State (NYS) Learning Standards, provides New York State with documentation of the purposes of the Regents Examination, scoring information, evidence of both the reliability and validity of the exams, scaling information, and guidelines for score reporting for the August 2022, January 2023, and June 2023 administrations. As the *Standards for Education and Psychological Testing* discusses in Standard 7, "The objective of the documentation is to provide test users with the information needed to help them assess the nature and quality of the test, the resulting scores, and the interpretations based on the test scores" (AERA et al., 2014, p. 123).¹ **Please note:** A technical report, by design, addresses the technical documentation of a testing program. Other aspects of a testing program (e.g., content standards, scoring guides, the guide to test interpretation) are thoroughly addressed and referenced in supporting documents. All analyses in this report were conducted using the operational items only.

The Regents Examinations are administered each year in August, January, and June to students enrolled in NYS schools. The main administration is in June of each year. As such, the June 2023 results are reported in the main body of this technical report, whereas the August 2022 and January 2023 results are provided in Appendix D and Appendix E, respectively.

1.2. PURPOSES OF THE EXAM

The Regents Examination in ELA measures student achievement against the NYS Learning Standards. The exam is prepared by teacher examination committees and New York State Education Department (NYSED) subject matter and testing specialists. Further, it provides teachers and students with important information about student learning and performance against the established curriculum standards. Results of this exam may be used to identify student strengths and needs to guide classroom teaching and learning. The exam also provides students, parents, counselors, administrators, and college admissions officers with objective and easily understood achievement information that may be used to inform empirically based educational and vocational decisions about students.

As a state-provided objective benchmark, the Regents Examination in ELA is intended for use in satisfying state testing requirements for students who have finished a course in ELA. A passing score on the exam counts toward requirements for a high school diploma, as described in the New York State diploma requirements.² Results of the Regents Examinations may also be used to satisfy various locally established requirements throughout the state.

¹ References to specific *Standards* will be placed in parentheses throughout the technical report to provide further context for each section.

² The New York State diploma requirements are located online at https://www.nysed.gov/sites/default/files/programs/standards-instruction/currentdiplomarequirements.pdf.

1.3. TARGET POPULATION (STANDARD 7.2)

The student population for the Regents Examination in ELA is composed of students who have completed a course in ELA. Table 1.1 presents a demographic breakdown of all students who took the August 2022, January 2023, and June 2023 Regents Examination in ELA. All analyses in this report are based on the population described in this table. Annual Regents Examination results reported in the New York State Report Cards are those reported in the Student Information Repository System (SIRS) as of the reporting deadline. The results reported in SIRS include the exams administered in August 2022, January 2023, and June 2023 (see https://data.nysed.gov/). If a student takes the same exam multiple times in the year, only the highest score is included in these results. Item-level data used for the analyses in this report are reported by districts on a similar timeline but through a different collection system. These data include all student results for each administration. Therefore, the sample sizes (n-counts) in this technical report will differ from publicly reported counts of student test takers.

Table 1.1. Total Student Population in 2022–2023: Regents Examination in ELA

	August 2022		Janua	January 2023		June 2023	
Demographics	N	%	N	%	N	%	
All Students	9,301	100.00	92,758	100.00	148,874	100.00	
Race/Ethnicity							
American Indian/Alaska Native	80	0.86	777	0.84	1,023	0.69	
Asian/Native Hawaiian/Other Pacific Islander	667	7.18	9,008	9.72	15,728	10.57	
Black/African American	2,496	26.87	17,917	19.33	26,613	17.88	
Hispanic/Latino	3,588	38.63	32,484	35.04	44,325	29.78	
Multiracial	183	1.97	1,921	2.07	3,441	2.31	
White	2,274	24.48	30,603	33.01	57,711	38.77	
English Language/Multilingual Learner							
No	7,547	81.14	80,780	87.09	133,633	89.76	
Yes	1,754	18.86	11,978	12.91	15,241	10.24	
Economically Disadvantaged							
No	2,450	26.34	33,855	36.50	69,561	46.72	
Yes	6,851	73.66	58,903	63.50	79,313	53.28	
Gender							
Female	3,887	41.85	44,529	48.03	72,961	49.02	
Male	5,397	58.11	48,127	51.91	75,779	50.91	
Nonbinary	4	0.04	54	0.06	101	0.07	
Student with a Disability							
No	6,469	69.55	74,726	80.56	124,934	83.92	
Yes	2,832	30.45	18,032	19.44	23,940	16.08	

Note: Some students are not reported in the race/ethnicity and gender groups but are reflected in "All Students," including 13 in August 2022, 48 in January 2023, and 33 in June 2023.

Chapter 2: Classical Item Statistics (Standard 4.10)

This chapter provides an overview of the two most familiar item-level statistics obtained from classical item analysis: item difficulty and item discrimination. The following results pertain to the operational Regents Examination in ELA items.

2.1. ITEM DIFFICULTY

At the most general level, an item's difficulty is indicated by its mean score in some specified group (e.g., grade level), calculated as follows:

$$\overline{x} = \frac{1}{n} \cdot \sum_{i=1}^{n} x_i$$

The individual item scores (x_i) are summed and then divided by the total number of students (n). For multiple-choice (MC) items, student scores are represented by 0s and 1s (0 = wrong answer, 1 = correct answer). With 0–1 scoring, the equation above also represents the number of students correctly answering the item divided by the total number of students. Therefore, this is also the proportion correct for the item, or the p-value. In theory, p-values can range from 0.00 to 1.00 on the proportion-correct scale. For example, if an MC item has a p-value of 0.89, it means 89% of the students answered the item correctly. This value might also suggest that the item was relatively easy and/or that the students who attempted the item were relatively high achievers. For constructed-response (CR) items, mean scores can range from the minimum possible score (usually 0) to the maximum possible score (e.g., 6 points for some items). To facilitate average score comparability across MC and CR items, mean item performance for CR items is divided by the maximum score possible so that the p-values for all items are reported as a ratio from 0.0 to 1.0.

Although the *p*-value statistic does not consider individual student ability in its computation, it provides a useful view of overall item difficulty and can provide an early and simple indication of items that are too difficult for the population of students taking the exam. Items with very high or very low *p*-values receive added scrutiny during all follow-up analyses, including item response theory (IRT) analyses that factor student ability into estimates of item difficulty. Such items may be removed from the item pool during the test development process, as field testing typically reveals that they add very little measurement information.

Items for the June 2023 Regents Examination in ELA show a range of *p*-values consistent with the targeted exam difficulty. Item *p*-values, presented in Table 2.1 and Table 2.2 for MC and CR items, respectively, range from 0.55 to 0.82, with a mean of 0.67. The tables also show a standard deviation (SD) of item score and item mean (Table 2.2 only).

2.2. ITEM DISCRIMINATION

At the most general level, estimates of item discrimination indicate an item's ability to differentiate between high and low performance on an exam. It is expected that students who perform well on the exam would be more likely to answer any given item correctly, while low-performing students (i.e., those who perform poorly on the exam overall) would be more likely to answer the same item incorrectly. Pearson's product-moment correlation coefficient (also commonly referred to as a point-biserial correlation) between item scores and test scores is used to indicate discrimination (Pearson, 1896). The correlation coefficient can range from -1.0 to +1.0. If high-scoring students

tend to get the item correct while low-scoring students do not, the correlation between the item score and the total test score will be both positive and noticeably large in its magnitude (i.e., above 0), indicating that the item is likely discriminating well between high- and low-performing students. Point-biserial correlations are computed for each answer option, including correct and incorrect options (commonly referred to as distractors). Point-biserial values for each distractor are typically negative and an important part of the analysis. Positive values can indicate that higher-performing students are selecting an incorrect answer or that the item key for the correct answer should be checked.

Table 2.1 and Table 2.2 present the point-biserial values on the correct response and three distractors (Table 2.1 only) for the June 2023 Regents Examination in ELA. The values for correct answers are 0.22 or higher for all items, indicating that the items are generally discriminating well between high- and low-performing students. Point-biserial values for all distractors are negative, indicating that students are generally responding to the items as expected during item and rubric development.

Table 2.1. MC Item Analysis Summary: Regents Examination in ELA—June 2023

Item	#Students	<i>p</i> -Value	SD	Point- Biserial	Point- Biserial Distractor 1	Point- Biserial Distractor 2	Point- Biserial Distractor 3
1	148,874	0.55	0.50	0.32	-0.10	-0.23	-0.18
2	148,874	0.57	0.49	0.27	-0.18	-0.24	-0.01
3	148,874	0.64	0.48	0.41	-0.21	-0.24	-0.17
4	148,874	0.68	0.47	0.35	-0.23	-0.16	-0.16
5	148,874	0.77	0.42	0.40	-0.18	-0.25	-0.22
6	148,874	0.60	0.49	0.34	-0.22	-0.18	-0.14
7	148,874	0.56	0.50	0.38	-0.10	-0.22	-0.20
8	148,874	0.60	0.49	0.27	-0.30	-0.11	0.00
9	148,874	0.65	0.48	0.43	-0.21	-0.24	-0.18
10	148,874	0.67	0.47	0.44	-0.21	-0.29	-0.22
11	148,874	0.82	0.39	0.22	-0.15	-0.17	-0.08
12	148,874	0.68	0.47	0.48	-0.17	-0.32	-0.23
13	148,874	0.68	0.47	0.34	-0.24	-0.14	-0.15
14	148,874	0.57	0.50	0.40	-0.28	-0.15	-0.22
15	148,874	0.71	0.45	0.38	-0.23	-0.12	-0.27
16	148,874	0.61	0.49	0.44	-0.22	-0.17	-0.25
17	148,874	0.57	0.50	0.44	-0.16	-0.23	-0.25
18	148,874	0.81	0.39	0.43	-0.20	-0.30	-0.19
19	148,874	0.69	0.46	0.46	-0.29	-0.21	-0.22
20	148,874	0.71	0.45	0.47	-0.21	-0.23	-0.28
21	148,874	0.71	0.45	0.41	-0.22	-0.19	-0.22
22	148,874	0.75	0.43	0.53	-0.29	-0.29	-0.24
23	148,874	0.60	0.49	0.30	-0.17	-0.13	-0.21
24	148,874	0.79	0.40	0.53	-0.21	-0.38	-0.24

Table 2.2. CR Item Analysis Summary: Regents Examination in ELA—June 2023

Item	#Students	Min. Score	Max. Score	Mean	SD	p-Value	Point- Biserial
25	148,874	0	6	4.01	1.32	0.67	0.90
26	148,874	0	4	2.65	1.11	0.66	0.78

2.3. DISCRIMINATION ON DIFFICULTY SCATTERPLOT

Figure 2.1 presents a scatterplot of the item discrimination values (*y*-axis) and item difficulty values (*x*-axis). Table 2.3 presents the descriptive statistics of *p*-value and point-biserial values, including mean, minimum, Q1, median, Q3, and maximum.

Figure 2.1. Scatterplot: Regents Examination in ELA—June 2023

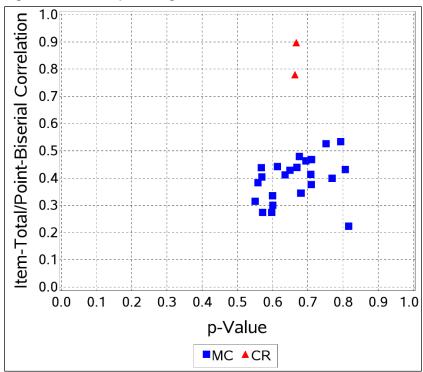


Table 2.3. Descriptive Statistics in *p-V*alue and Point-Biserial Correlation: Regents Examination in ELA—June 2023

Statistic	N	Mean	Min.	Q1	Median	Q3	Max.
<i>p</i> -Value	26	0.67	0.55	0.60	0.67	0.71	0.82
Point-Biserial	26	0.43	0.22	0.34	0.41	0.46	0.90

2.4. OBSERVATIONS AND INTERPRETATIONS

The p-values for the MC items ranged from 0.55 to 0.82, while the p-values for the CR items ranged from 0.66 to 0.67. The difficulty distributions illustrated in the plot indicate that a wide range of item difficulties appeared on each exam, which was one test development goal.

Chapter 3: IRT Calibrations, Equating, and Scaling (Standards 2 and 4.10)

The item response theory (IRT) model used for the Regents Examinations is based on the work of Georg Rasch (Rasch, 1960). The Rasch model has a long-standing presence in applied testing programs. IRT has several advantages over classical test theory and has become the standard procedure for analyzing item response data in large-scale assessments. According to van der Linden and Hambleton (1997), "The central feature of IRT is the specification of a mathematical function relating the probability of an examinee's response on a test item to an underlying ability." Ability, in this sense, can be thought of as performance on the test and is defined as "the expected value of observed performance on the test of interest" (Hambleton et al., 1991). This performance value is often referred to as θ . Performance and θ will be used interchangeably throughout the remainder of this report.

A fundamental advantage of IRT is that it links student performance and item difficulty estimates and places them on the same scale, allowing for an evaluation of student performance that considers the difficulty of the test. This is particularly valuable for final test construction and test form equating, as it facilitates a fundamental attention to fairness for all students across items and test forms.

This chapter outlines the procedures used for calibrating the operational items, including an introduction to the Rasch model, the results from evaluations of the adequacy of the Rasch assumptions, and the Rasch item statistics. Generally, item calibration is the process of assigning a difficulty, or item "location," estimate to each item on an assessment so that all items are placed onto a common scale.

3.1. DESCRIPTION OF THE RASCH MODEL

The Rasch model (Rasch, 1960) was used to calibrate the MC items, and the partial credit model (PCM; Wright & Masters, 1982) was used to calibrate the CR items. The PCM extends the Rasch model for dichotomous (0, 1) items so that it accommodates the polytomous CR item data. Under the PCM model, for a given item i with m_i score categories, the probability of person n scoring x ($x = 0, 1, 2, ... m_i$) is given as follows:

$$P_{ni}(X = x) = \frac{\exp \sum_{j=0}^{x} (\theta_n - D_{ij})}{\sum_{k=0}^{m_i} \exp \sum_{j=0}^{k} (\theta_n - D_{ij})}$$

where θ_n represents student ability, and D_{ij} is the step difficulty of the jth step on item i. D_{ij} can be expressed as $D_{ij} = D_i - F_{ij}$, where D_i is the difficulty for item i, and F_{ij} is a step deviation value for the jth step. For dichotomous MC items, the PCM reduces to the standard Rasch model, and the single step difficulty is referred to as the item's difficulty. The Rasch model predicts the probability of person n getting item i correct, as follows:

$$P_{ni}(X=1) = \frac{\exp(\theta_n - D_i)}{1 + \exp(\theta_n - D_i)}$$

The Rasch model places both performance and item difficulty (estimated in terms of log-odds or logits) on the same continuum. When the model assumptions are met, the Rasch model provides estimates of student performance and item difficulty that are theoretically invariant across random samples of the same student population.

3.2. SOFTWARE AND ESTIMATION ALGORITHM

Item calibration was implemented via the WINSTEPS 3.60 computer program (Linacre, 2005), which employs unconditional (UCON) joint maximum likelihood estimation (JMLE).

3.3. ITEM DIFFICULTY-STUDENT PERFORMANCE MAP

The distributions of the Rasch item logits (item difficulty estimates) and student performance are shown on the item difficulty–student performance map presented in Figure 3.1. This graphic illustrates the location of student performance and item difficulty on the same scale, along with their respective distributions and cut scores (indicated by the horizontal dotted lines). The figure shows more difficult items and higher student performance at the top and lower performance and easier items at the bottom.

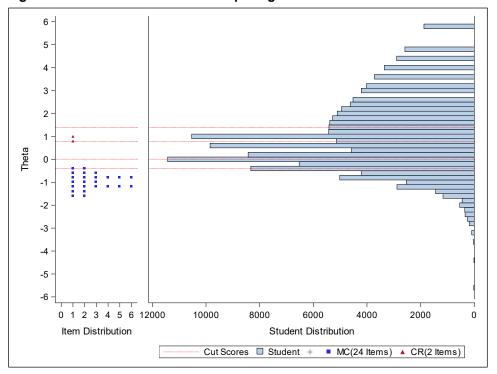


Figure 3.1. Student Performance Map: Regents Examination in ELA—June 2023

3.4. CHECKING RASCH ASSUMPTIONS

Because the Rasch model was the basis of all calibration, scoring, and scaling analyses associated with the Regents Examinations, the validity of the inferences from these results depends on the degree to which the assumptions of the model were met and how well the model fits the test data. Therefore, it is important to check these assumptions. This section evaluates the dimensionality of the data, local item independence, and item fit. Only operational items were analyzed as they are the basis of student scores.

3.4.1. Unidimensionality

Rasch models assume that one dominant dimension determines the differences in students' performances. Principal components analysis (PCA) can be used to assess the unidimensionality assumption. The purpose of the analysis is to verify whether any other dominant components exist among the items. If any other dimensions are found, the unidimensionality assumption would be violated.

A parallel analysis (Horn, 1965) was conducted to help distinguish components that are real from components that are random. Parallel analysis is a technique used to determine how many factors exist in principal components. For the parallel analysis, 100 random datasets of sizes equal to the original data were created. For each random dataset, a PCA was performed and the resulting eigenvalues were stored. For each component, the upper 95th percentile value of the distribution of the 100 eigenvalues from the random data sets was then plotted. Given the size of the data generated for the parallel analysis, the reference line is essentially equivalent to plotting a reference line for an eigenvalue of 1.

Figure 3.2 presents the PCA and parallel analysis results for the exam. The results include the eigenvalues and the percentage of variance explained for the first five components, as well as the scree plots. The scree plots show the eigenvalues plotted by component number and the results of a parallel analysis. Although the total number of components in the PCA is the same as the total number of items in a test, Figure 3.2 shows only the first 10 components. This view is sufficient for interpretation because components are listed in descending eigenvalue order. The fact that the eigenvalues for components 2–10 are much lower than the first component demonstrates that there is one dominant component, showing evidence of unidimensionality.

Reckase (1979) proposed that the variance explained by the primary dimension should be greater than 20% to indicate unidimensionality. However, as this rule is not absolute, it is helpful to consider three additional characteristics of the PCA and parallel analysis results: (a) whether the ratio of the first to the second eigenvalue is greater than 3, (b) whether the second value is not much larger than the third value, and (c) whether the second value is not significantly different from those from the parallel analysis.

As shown in Figure 3.2, the primary dimension explained 22.85% of the total variance for the Regents Examination in ELA. The eigenvalue of the second dimension is less than one-third of the first at 1.09, and the second value is not significantly different from the parallel analysis. Overall, the PCA suggests that the test is reasonably unidimensional.

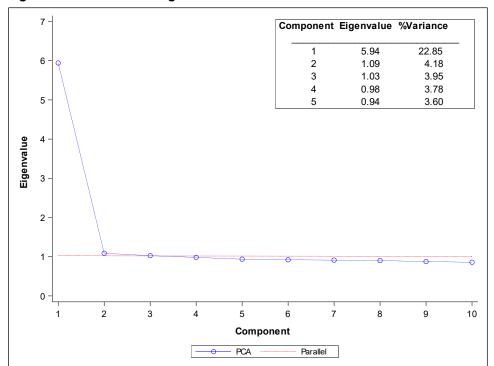


Figure 3.2. Scree Plot: Regents Examination in ELA—June 2023

3.4.2. Local Independence

Local independence is a fundamental assumption of IRT. This means that, for statistical purposes, a student's response to any one item should not depend on the student's response to any other item on the test. In formal statistical terms, Test X, which comprises items $X_1, X_2, ...X_n$, is locally independent with respect to the latent variable θ if, for all $x = (x_1, x_2, ...x_n)$ and θ :

$$P(\mathbf{X} = \mathbf{x} \mid \theta) = \prod_{i=1}^{I} P(X_i = x_i \mid \theta)$$

This formula essentially states that the probability of any pattern of responses across all items (x), after conditioning on the student's true score (θ) as measured by the test, should be equal to the product of the conditional probabilities across each item (i.e., the multiplication rule for independent events where the joint probabilities are equal to the product of the associated marginal probabilities).

The equation above shows the condition after satisfying the strong form of local independence. A weak form of local independence (WLI) is proposed by McDonald (1979). The distinction is important because many indicators of local dependency are framed by WLI. For WLI, the conditional covariances of all pairs of item responses, conditioned on the abilities, are assumed to be equal to zero. When this assumption is met, the joint probability of responses to an item pair, conditioned on abilities, is the product of the probabilities of responses to these two items, as shown below. Based on the WLI, the following expression can be derived:

$$P(X_i = x_i, X_j = x_j \mid \theta) = P(X_i = x_i \mid \theta)P(X_j = x_j \mid \theta)$$

Marais and Andrich (2008) point out that local item dependence in the Rasch model can occur in two ways that may be difficult to distinguish. The first way occurs when the assumption of unidimensionality is violated. Here, other nuisance dimensions besides a dominant dimension determine student performance (this can be called "trait dependence"). The second way occurs when responses to an item depend on responses to another item. This is a violation of statistical independence and can be called "response dependence." By distinguishing the two sources of local dependence, one can see that, while local independence can be related to unidimensionality, the two are different assumptions and therefore require different tests.

Residual item correlations, provided in WINSTEPS for each item pair, were used to assess the local dependence between the Regents Examination items. In general, these residuals are computed as follows: First, expected item performance based on the Rasch model is determined using θ and item parameter estimates. Next, deviations (residuals) between the students' expected and observed performance are determined for each item. Finally, for each item pair, a correlation between the respective deviations is computed.

Three types of residual correlations are available in WINSTEPS: raw, standardized, and logit. The raw score residual correlation essentially corresponds to Yen's Q3 index, a popular statistic used to assess local independence. The expected value for the Q3 statistic is approximately -1/(k-1) when no local dependence exists, where k is test length (Yen, 1993). Thus, the expected Q3 values should be approximately -0.04 for the items on the exam. Index values greater than 0.20 indicate a degree of local dependence that should be examined by test developers (Chen & Thissen, 1997).

The default "standardized residual correlation" in WINSTEPS was used for these analyses because the three residual correlations are very similar. Table 3.1 presents the summary statistics (i.e., the mean, standard deviation, minimum, maximum, and several percentiles [P_{10} , P_{25} , P_{50} , P_{75} , P_{90}]) for all the residual correlations. The table also presents the total number of item pairs (N) and the number of pairs with residual correlations greater than 0.20. There is one item pair with residual correlations greater than 0.20. The mean residual correlations are slightly negative, and the values are close to -0.03. Most of the correlations are very small, suggesting that local item independence generally holds for the Regents Examination in ELA.

Table 3.1. Summary of Item Residual Correlations: Regents Examination in ELA—June 2023

Statistic	Value
N	325
Mean	-0.03
SD	0.04
Minimum	-0.15
P ₁₀	-0.09
P ₂₅	-0.05
P ₅₀	-0.03
P ₇₅	-0.01
P ₉₀	0.01
Maximum	0.21
> 0.20	1.00

3.4.3. Item Fit

An important assumption of the Rasch model is that the data for each item fit the model. WINSTEPS provides two item fit statistics (INFIT and OUTFIT) for evaluating the degree to which the Rasch model predicts the observed item responses for a given set of test items. Each fit statistic can be expressed as a mean square (MnSq) statistic or on a standardized metric (Zstd with mean = 0 and variance = 1). MnSq values are more oriented toward practical significance, while Zstd values are more oriented toward statistical significance. INFIT MnSq values are the average of standardized residual variance (the difference between the observed score and the Rasch-estimated score divided by the square root of the Rasch-model variance). The INFIT statistic is weighted by the θ relative to item difficulty and tends to be affected more by unexpected responses close to the person, item, or rating scale category measure (i.e., informative, on-target responses).

The expected MnSq value is 1.0 and can range from 0.0 to infinity. Deviation in excess of the expected value can be interpreted as noise or lack of fit between the items and the model. Values lower than the expected value can be interpreted as item redundancy or overfitting items (too predictable, too much redundancy), and values greater than the expected value indicate underfitting items (too unpredictable, too much noise). Rules regarding "practically significant" MnSq values vary.

Table 3.2 presents the summary statistics of INFIT mean square statistics for the Regents Examination in ELA, including the mean, standard deviation, and minimum and maximum values. The table also presents the number of items within a targeted range of [0.7, 1.3]. The mean INFIT value is 1.00, with all 26 items falling in a targeted range of [0.7, 1.3]. As the range of [0.7, 1.3] is used as a guide for ideal fit, fit values outside of the range are considered individually. Overall, these results indicate that, for most items, the Rasch model fits the Regents Examination in ELA item data well.

Table 3.2. Summary of INFIT Mean Square Statistics: Regents Examination in ELA—June 2023

N	Mean	SD	Min.	Max.	[0.7, 1.3]
26	1.00	0.10	0.81	1.17	[26/26]

3.5. SCALING OF OPERATIONAL TEST FORMS

Operational test items were selected based on content coverage, content accuracy, and statistical quality. The sets of items on each operational test conformed to the coverage determined by content experts working from the learning standards established by NYSED and explicated in the test blueprint. Each item's classical and Rasch statistics were used to assess item quality. Items were selected to vary in difficulty to accurately measure students' abilities across the ability continuum. Appendix A contains the operational test maps for the 2022–2023 administrations. The statistics presented in the test maps were generated based on the field test data.

All Regents Examinations are pre-equated, meaning that the parameters used to derive the relationship between the raw and scale scores are estimated prior to the construction and administration of the operational form. These field tests are administered to as small a sample of students as possible to minimize the effect on student instructional time throughout the state. The small n-counts associated with such administrations are sufficient for reasonably accurate estimation of most items' parameters.

The Regents Examination in ELA has four cut scores, which are set at the scale scores of 55, 65, 79 (floating), and 85. The third cut point at 79 was determined at the most recent ELA standard setting session, during which a raw score associated with assessment performance at this performance level was determined. The scale score associated with this raw score was then determined and used for the performance level cut moving forward.

A primary consideration during test construction was to select items that would minimize changes in the raw scores corresponding to these scale scores. Maintaining a consistent mean Rasch difficulty level from administration to administration facilitates this. For this assessment, the target value for the mean Rasch difficulty was set at 0.052. The raw scores corresponding to the scale score cut scores may still fluctuate, even if the mean Rasch difficulty level is maintained at the target value. This is due to differences in the distributions of the Rasch difficulty values between the items from administration to administration.

The relationship between raw and scale scores is explicated in the scoring table for the 2022–2023 administrations, presented in Appendix B. These tables are the product of the following scaling procedure.

All Regents Examinations are equated back to a base scale held constant from year to year. Specifically, they are equated to the base scale using a calibrated item pool. The Rasch difficulties from the items' initial administration in a previous year's field test are used to equate the scale for the current administration to the base administration. For ELA, the base administration was the June 2014 administration. Scale scores from the June 2023 administration are on the same scale and can be directly compared to scale scores on all previous administrations back to the June 2014 administration.

When the base administration was concluded, the initial raw-score-to-scale-score relationship was established. Three raw scores were fixed at specific scale scores. Scale scores of 0 and 100 were fixed to correspond to the minimum and maximum possible raw scores. A standard setting had been held to determine the "passing" and "passing with distinction" cut scores in the raw score metric. The scale score points of 55, 65, and 85 were set to correspond to those raw score cuts. A fourth-degree polynomial is required to fit a line exactly to five arbitrary points (e.g., the raw scores corresponding to the five critical scale scores of 0, 55, 65, 85, and 100). The general form of this best-fitting line is as follows:

$$SS = m4 * RS^4 + m3 * RS^3 + m2 * RS^2 + m1 * RS^1 + m0$$

where *SS* is the scale score, *RS* is the raw score, and *m0* through *m4* are the transformation constants that convert the raw score into the scale score (*m0* will always be equal to 0 in this application, as a raw score of 0 corresponds to a scale score of 0). A subscript for a person on both dependent and independent variables is not present for simplicity. The above relationship and the values of *m1* to *m4* specific to this subject were then used to determine the scale scores corresponding to the remainder of the raw scores on the exam. This initial relationship between the raw and scale scores became the base scale.

The Rasch difficulty parameters for the items on the base form were then used to derive a raw score to Rasch student ability (theta score) relationship. This allowed the relationship between the Rasch theta score and the scale score to be known, mediated through their common relationship with the raw scores. In succeeding years, each test form was selected from the pool of items that had been tested in previous years' field tests, each of which had known Rasch item difficulty parameter(s). These known parameters were then used to construct the relationship between the raw and Rasch theta scores for that particular form. Because the Rasch difficulty parameters are all on a common scale, the Rasch theta scores were also on a common scale with previously administered forms. The remaining step in the scaling process was to find the scale score equivalent for the Rasch theta score corresponding to each raw score point on the new form using the theta-to-scale score relationship established in the base year. This was done via linear interpolation.

This process results in a relationship between the raw scores on the form and the overall scale scores. The scale scores corresponding to each raw score are then rounded to the nearest integer for reporting on the conversion chart (posted at the close of each administration). The only exceptions are for the minimum and maximum raw scores and the raw scores that correspond to the scaled cut scores of 55, 65, 79, and 85. The minimum (0) and maximum possible raw scores are assigned scale scores of 0 and 100, respectively. If there are raw scores less than the maximum with scale scores that round to 100, their scale scores are set equal to 99. A similar process is followed with the minimum score; if any raw scores other than 0 have scale scores that round to 0, their scale scores are instead set equal to 1.

When two or more scale scores round to 55, 65, or 85, the lowest raw score's scale score is set equal to 55, 65, or 85, and the scale scores corresponding to the higher raw scores are set to 56, 66, or 86 as appropriate. This rule does not apply for the third floating cut score. If no scale score rounds to these critical cuts, the raw score with the largest scale score that is less than the cut is set equal to the cut. The overarching principle, when two raw scores both round to either scale score cut, is that the lower of the raw scores is always assigned to be equal to the cut so that students are never penalized for this ambiguity.

Chapter 4: Reliability (Standard 2)

Test reliability is a measure of the internal consistency of a test (Cronbach, 1951), or a measure of the extent to which the items on a test provide consistent information about student mastery of a domain. Reliability should ultimately demonstrate that student score estimates maximize consistency and therefore minimize error or, theoretically speaking, that students who take a test multiple times would get the same score each time.

According to the *Standards for Educational and Psychological Testing*, "A number of factors can have significant effects on reliability/precision, and in some cases, these factors can lead to misinterpretations of test scores, if not taken into account" (AERA et al., 2014, p. 38). First, test length and the variability of observed scores can influence reliability estimates. Tests with fewer items or with a lack of heterogeneity in scores tend to produce lower reliability estimates. Second, reliability is concerned with random sources of error. Accordingly, the degree of inconsistency due to random error sources is what determines reliability: less consistency is associated with lower reliability, and more consistency is associated with higher reliability. Systematic error sources also exist.

This chapter discusses reliability results for the Regents Examination in ELA and three additional statistical measures used to address the multiple factors affecting an interpretation of the exam's reliability: standard errors of measurement (SEMs), decision consistency, and group means.

4.1. RELIABILITY INDICES (STANDARD 2.20)

Classical test theory describes reliability as a measure of the internal consistency of test scores. The reliability (ρ_X^2) is defined as the ratio of true score variance (σ_T^2) to the observed score variance (σ_X^2) , as presented in the equation below. The total variance contains two components: (a) the variance in true scores and (b) the variance due to the imperfections in the measurement process (σ_E^2) . In other words, total variance equals true score variance plus error variance.³

$$\rho_X^2 = \frac{\sigma_T^2}{\sigma_X^2} = \frac{\sigma_T^2}{\sigma_T^2 + \sigma_E^2}$$

Reliability coefficients indicate the degree to which differences in test scores reflect true differences, rather than random fluctuations, in the attribute being tested. Total test score variance (i.e., individual differences) is partly due to real differences in the construct (true variance) and partly due to random error in the measurement process (error variance).

Reliability coefficients range from 0.0 to 1.0. The index is 0.0 if none of the test score variances are true. Such scores would be pure random noise (i.e., all measurement error). If all test score variances were true, the index would equal 1.0. If the index achieved a value of 1.0, scores would be perfectly consistent (i.e., contain no measurement error). Although values of 1.0 are never achieved in practice, larger coefficients are more desirable because they indicate that the test scores are less influenced by random error.

³ A covariance term is not required, as true scores and errors are assumed to be uncorrelated in classical test theory.

Reliability is most often estimated using the formula for coefficient alpha, which provides a practical internal consistency index. Coefficient alpha can be conceptualized as the extent to which an exchangeable set of items from the same domain would result in a similar rank ordering of students. Relative error is reflected in this index. Excessive variation in student performance from one sample of items to the next should be of particular concern for any achievement test user. A general computational formula for coefficient alpha is as follows:

$$\alpha = \frac{N}{N-1} \left(1 - \frac{\sum_{i=1}^{N} \sigma_{Yi}^2}{\sigma_X^2} \right)$$

where N is the number of parts (items), σ_X^2 is the variance of the observed total test scores, and σ_{Yi}^2 is the variance of part i.

4.2. STANDARD ERROR OF MEASUREMENT (STANDARDS 2.13, 2.14, 2.15)

Reliability coefficients best reflect the extent to which measurement inconsistencies may be present or absent. The standard error of measurement (SEM) is another indicator of test score precision that is better suited for determining the effect of measurement inconsistencies for the scores obtained by individual students. This is particularly so for conditional SEMs (CSEMs).

4.2.1. Traditional Standard Error of Measurement

The SEM is defined as the standard deviation of the distribution of observed scores for students with identical true scores, as shown below. The SEM is an index of the random variability in test scores within test score units and therefore represents important information for test score users.

$$SEM = SD\sqrt{1-\alpha}$$

This formula indicates that the value of the SEM depends on both the reliability coefficient (the coefficient alpha) and the standard deviation of test scores. If the reliability were equal to 0.00 (the lowest possible value), the SEM would be equal to the standard deviation of the test scores. If test reliability were equal to 1.00 (the highest possible value), the SEM would be 0.0. Therefore, a perfectly reliable test has no measurement error (Harvill, 1991). The value of the SEM also takes the group variation (i.e., score standard deviation) into account. For example, an SEM of 3 on a 10-point test would be very different from an SEM of 3 on a 100-point test.

4.2.2. Traditional Standard Error of Measurement Confidence Intervals

The SEM is an index of the random variability in test scores reported in actual score units, which is why it has such great utility for test score users. SEMs allow statements regarding the precision of individual test scores and help place "reasonable limits" (Gulliksen, 1950) around observed scores through construction of an approximate score band. Often referred to as confidence intervals, these bands are constructed by taking the observed scores, X, and adding and subtracting a multiplicative factor of the SEM. For example, students with a given true score will have observed scores that fall between ±1 SEM about two-thirds of the time.⁴ For ±2 SEM confidence intervals, this increases to about 95%.

⁴ Some prefer the following interpretation: If a student were tested an infinite number of times, the ±1 SEM confidence intervals constructed for each score would capture the student's true score 68% of the time.

Table 4.1 presents the coefficient alpha and associated SEM for the Regents Examination in ELA.

Table 4.1. Coefficient Alpha and SEM: Regents Examination in ELA—June 2023

Coefficient Alpha	SEM
0.85	4.22

Assuming normally distributed scores, one would expect about two-thirds of the observations to be within one standard deviation of the mean. An estimate of the standard deviation of the true scores can be computed as follows:

$$\hat{\sigma}_{T} = \sqrt{\hat{\sigma}_{X}^{2} - \hat{\sigma}_{X}^{2}(1 - \hat{\rho}_{XX})}$$

4.2.3. Conditional Standard Error of Measurement

Every time an assessment is administered, a student's score contains some error. If the same exam were administered an infinite number of times to the same student, the mean of the distribution of the student's raw scores would be equal to the student's true score (θ) (i.e., the score obtained with no error), and the standard deviation of the distribution of the student's raw scores would be the conditional standard error. There is a one-to-one correspondence between the raw score and θ in the Rasch model, so this concept can be applied more generally to all students who obtained a particular raw score. The probability of obtaining each possible raw score can be calculated given the students' estimated θ . The standard deviation of this conditional distribution is defined as the CSEM. The computer program POLYCSEM (Kolen, 2004) was used to carry out the mechanics of this computation.

The relationship between θ and the scale score is not expressible in a simple mathematical form because it is a blend of the fourth-degree polynomial relationship between the raw and scale scores and the nonlinear relationship between the expected raw and θ scores. In addition, as the exam is equated from year to year, the relationship between the raw and scale scores moves away from the original fourth-degree polynomial relationship to one that is also no longer expressible in simple mathematical form. In the absence of a simple mathematical relationship between θ and the scale scores, the CSEMs available for each θ score via Rasch IRT cannot be converted directly to the scale score metric.

The use of Rasch IRT to scale and equate the Regents Examinations does, however, make it possible to calculate CSEMs by using the procedures described by Kolen et al. (1996) for dichotomously scored items and extended by Wang et al. (2000) to polytomously scored items. For tests such as the Regents Examinations that do not have a one-to-one relationship between raw (θ) and scale scores, the CSEM for each achievable scale score can be calculated by using the compound multinomial distribution to represent the conditional distribution of raw scores for each level of θ .

Consider a student with a certain performance level. If it were possible to measure this student's performance perfectly, without any error, this measure could be called the student's true score, which is equal to the expected raw score. However, whenever a student takes a test, the observed test score always includes some level of measurement error. Sometimes this error is positive, and the student achieves a higher score than would be expected given the student's level of θ . Other times it is negative, and the student achieves a lower-than-expected score. If a student could be given the same test multiple times and their observed test scores recorded, the resulting distribution would be the conditional distribution of raw scores for that student's level of θ with a mean value equal to the student's expected raw (true) score. The CSEM for that level of θ in the raw score metric is the square root of the variance of this conditional distribution.

The conditional distribution of raw scores for any level of θ is the compound multinomial distribution (Wang et al., 2000). An algorithm to compute this can be found in Hanson (1994) and in Thissen et al. (1995) and is also implemented in the computer program POLYCSEM (Kolen, 2004). The compound multinomial distribution yields the probabilities that a student with a given level of θ has of attaining each achievable raw (and accompanying scale) score. The point values associated with each achievable raw or scale score point can be used to calculate the mean and variance of this distribution in the raw or scale score metric, respectively. The square root of the variance is the CSEM of the raw or scale score point associated with the current level of θ .

4.2.4. Conditional Standard Error of Measurement Confidence Intervals

CSEMs allow statements regarding the precision of individual test scores. Like SEMs, they help place reasonable limits around observed scale scores through the construction of an approximate score band. The confidence intervals are constructed by adding and subtracting a multiplicative factor of the CSEM.

4.2.5. Conditional Standard Error of Measurement Characteristics

The relationship between the scale score CSEM and θ depends on the nature of the raw-to-scale score transformation (Kolen & Brennan, 2005; Kolen & Lee, 2011) and on whether the CSEM is derived from the raw scores or from θ (Lord, 1980). The pattern of CSEMs for raw scores and linear transformations of the raw score tend to have a characteristic "inverted-U" shape, with smaller CSEMs at the ends of the score continuum and larger CSEMs toward the middle of the distribution. Achievable raw score points for these distributions are spaced equally across the score range. Kolen and Brennan state, "When, relative to raw scores, the transformation compresses the scale in the middle and stretches it at the ends, the pattern of the conditional standard errors of measurement will be concave up (U-shaped), even though the pattern for the raw scores was concave down (inverted-U shape)" (2005, p. 357).

4.2.6. Results and Observations

The relationship between raw and scale scores for the Regents Examinations tends to be roughly linear from scale scores of 0 to 65 and then concave down from about 65 to 100. In other words, the scale scores track linearly with the raw scores for the lower two-thirds of the scale score range and are then compressed relative to the raw scores for the remaining one-third of the range, though there are variations. The CSEMs for the Regents Examinations can be expected to have inverted-U shaped patterns, with some variations.

Figure 4.1 shows this type of CSEM variation for the Regents Examination in ELA, where the compression of raw score to scale scores around the cut score of 85 noticeably changes the shape of the curve. This type of expansion and compression can be seen in the figure by looking

at the changing density of raw score points along the scale score range on the horizontal axis. Specifically, at the lower end of the scale, scale scores 0–16 span raw scores 0–13. Over the range from scale scores 17–60, the raw score range is 14–29 (44 scale score points to 15 raw score points). Finally, scale scores over the range of 61–100 span raw scores of 30–56 (39 scale score points for 27 raw score points).

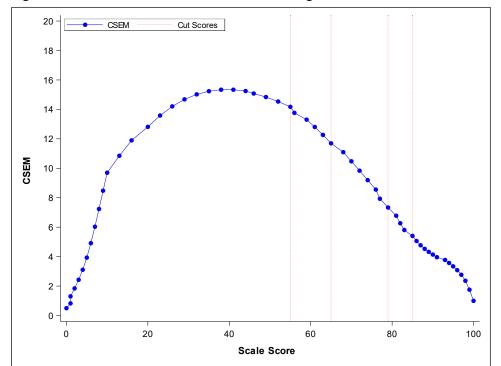


Figure 4.1. Conditional Standard Error Plot: Regents Examination in ELA—June 2023

4.3. DECISION CONSISTENCY AND ACCURACY (STANDARD 2.16)

In a standards-based testing program, there is interest in knowing how accurately students are classified into performance categories. In contrast to the coefficient alpha that is concerned with the relative rank-ordering of students, it is the absolute values of student scores that are important in decision consistency and accuracy.

Classification consistency refers to the degree to which the performance level for each student can be replicated upon retesting using an equivalent form (Huynh, 1976). Decision consistency answers the question, what is the agreement in classifications between the two nonoverlapping, equally difficult forms of the test? If two parallel forms of the test were given to the same students, the consistency of the measure would be reflected by the extent to which the classification decisions based on the first set of test scores matched the decisions based on the second set of test scores.

Consider Table 4.2 and Table 4.3. If a student is classified as being in one category based on the Test 1 score, how probable would it be that the student would be reclassified as being in the same category if the student took Test 2 (a nonoverlapping, equally difficult form of the test)? This proportion is a measure of decision consistency. The proportions of correct decisions, φ , for two and four categories are computed by the following formulas, respectively:

$$\varphi = \varphi_{11} + \varphi_{22}$$
$$\varphi = \varphi_{11} + \varphi_{22} + \varphi_{33} + \varphi_{44}$$

The sum of the diagonal entries (i.e., the proportion of students classified by the two forms into the same performance level) signifies the overall consistency.

Table 4.2. Pseudo-Decision Table for Two Hypothetical Categories

			Test 1	
		Level 1	Level 2	Marginal
	Level 1	φ11	φ12	φ1∙
Test 2	Level 2	φ21	φ22	φ2∙
	Marginal	φ●1	φ•2	1

Table 4.3. Pseudo-Decision Table for Four Hypothetical Categories

				T 14		
				Test 1		
		Level 1	Level 2	Level 3	Level 4	Marginal
	Level 1	φ11	φ12	φ13	φ14	φ1∙
	Level 2	φ21	φ22	φ23	φ24	φ2∙
Test 2	Level 3	φ31	φ32	φ33	φ34	φ3∙
	Level 4	φ41	φ42	φ43	φ44	φ4●
	Marginal	φ∙1	φ∙2	φ•3	φ∙4	1

Classification accuracy refers to the agreement of the observed classifications of students with the classifications made based on their true scores. An observed score contains measurement error, while a true score is theoretically free of measurement error. A student's observed score can be formulated by the sum of the true score plus measurement error (Observed = True + Error). Decision accuracy is an index to determine the extent to which measurement error causes a classification different than the one expected from the true score. Because true scores are unobserved and decision consistency is computed based on a single administration of the Regents Examination in English Language Arts, a statistical model using data only from the available administration is used to estimate the true scores and to project the consistency and accuracy of classifications (Hambleton & Novick, 1973). Although several procedures are available, a well-known method developed by Livingston and Lewis (1995) that implements a specific true score model is used.

Several factors might affect decision consistency and accuracy. One important factor is the reliability of the scores. All other things being equal, more reliable test scores tend to result in more similar reclassifications and less measurement error. Another factor is the location of the cut score in the score distribution. More consistent and accurate classifications are observed when the cut scores are located away from the mass of the score distribution. The number of performance levels is also a consideration. Consistency and accuracy indices based on four performance levels should be lower than those based on two performance levels. This is not

surprising, as classification and accuracy based on four performance levels would allow more opportunity to change performance levels. Hence, there would be more classification errors and less accuracy with four performance levels, resulting in lower consistency indices.

Table 4.4 presents the results for the dichotomies created by the four corresponding cut scores. For example, the statistics under "2/3" indicate the decision consistency and accuracy when the performance levels are divided into two categories: one for the second and lower performance level, and the other for the third and higher performance levels. The tabled values are derived with the program *BB-Class* (Brennan, 2004) using the Livingston and Lewis method (1995). The decision consistency ranged from 0.83 to 0.90, and the decision accuracy ranged from 0.88 to 0.93. Both decision consistency and accuracy values are high, indicating very good consistency and accuracy of student classifications.

Table 4.4. Decision Consistency and Accuracy Results: Regents Examination in ELA—June 2023

Statistic	1/2	2/3	3/4	4/5
Consistency	0.90	0.87	0.83	0.83
Accuracy	0.93	0.91	0.88	0.88

4.4. GROUP MEANS (STANDARD 2.17)

The student population for the Regents Examination in ELA is composed of students who have completed a course in ELA. Table 4.5 presents the overall mean scale score that was computed based on all students who took the exam.

Table 4.5. Group Means: Regents Examination in ELA—June 2023

		Mean Scale	SD Scale	
Demographics	#Students	Score	Score	
All Students	148,874	73.89	20.70	
Race/Ethnicity				
American Indian/Alaska Native	1,023	68.06	20.91	
Asian/Native Hawaiian/Other Pacific Islander	15,728	80.28	18.70	
Black/African American	26,613	66.39	20.42	
Hispanic/Latino	44,325	66.14	22.87	
Multiracial	3,441	77.74	18.53	
White	57,711	81.45	15.67	
English Language/Multilingual Learner				
No	133,633	77.21	17.60	
Yes	15,241	44.80	22.94	
Economically Disadvantaged				
No	69,561	81.18	16.96	
Yes	79,313	67.50	21.55	
Gender				
Female	72,961	76.36	19.49	
Male	75,779	71.51	21.54	
Nonbinary	101	81.28	15.30	
Student with a Disability				
No	124,934	76.70	19.39	
Yes	23,940	59.23	21.13	

Note: Thirty-three students are not reported in the race/ethnicity and gender groups but are reflected in "All Students."

4.5. STATE PERCENTILE RANKINGS

Table 4.6 presents the state percentile rankings based on scale score distributions of all students taking the Regents Examination in ELA for the June 2023 administration. The scale scores range from 0 to 100, and some scale scores may not be obtainable depending on the raw score-to-scale score relationship for a specific administration. The percentile ranks are computed in the following manner:

- A student's assigned "state percentile rank" will be the cumulative percentage of students scoring at the immediate lower score plus half of the percentage of students obtaining the given score.
- Students who obtain the highest possible score will receive a percentile rank of 99.

Table 4.6. State Percentile Ranking for Scale Score: Regents Examination in ELA—June 2023

Scale Score	Percentile Rank	Scale Score	Percentile Rank	Scale Score	Percentile Rank
0	1	38	7	81	51
1	1	41	9	82	55
2	1	44	10	83	58
3	1	46	11	85	62
4	1	49	12	86	66
5	1	52	14	87	69
6	1	55	16	88	73
7	1	56	18	89	76
8	1	59	20	90	79
9	1	61	22	91	82
10	2	63	24	93	85
13	2	65	27	94	88
16	3	68	29	95	91
20	3	70	32	96	93
23	4	72	35	97	95
26	4	74	38	98	97
29	5	76	41	99	99
32	6	77	44	100	99
35	6	79	48	_	_

Chapter 5: Validity (Standard 1)

To restate the purposes and uses of the Regents Examination in ELA, this exam measures student achievement against the NYS Learning Standards and was prepared by teacher examination committees and NYSED subject matter and testing specialists. Further, it provides teachers and students with important information about student learning and performance against the established curriculum standards. Results of this exam may be used to identify student strengths and needs, which may then be used to guide classroom teaching and learning. The exam also provides students, parents, counselors, administrators, and college admissions officers with objective and easily understood achievement information that may be used to inform empirically based educational and vocational decisions about students. As a state-provided objective benchmark, the Regents Examination in ELA is intended for use in satisfying state testing requirements for students who have finished a course in ELA. A passing score on the exam counts toward requirements for a high school diploma, as described in the New York State diploma requirements (https://www.nysed.gov/sites/default/files/programs/standards-instruction/currentdiplomarequirements.pdf). Results of the exam may also be used to satisfy various locally established requirements throughout the state.

The validity of score interpretations is supported by the following sources of evidence, which are important to gather and document to support validity claims for an assessment as provided by the *Standards for Educational Psychological Testing* (AERA et al., 2014): test content, response processes, internal test structure, relation to other variables, and consequences of testing. While these categories are not mutually exclusive, as one source of validity evidence often falls into more than one category, they provide a useful framework within the *Standards* for the discussion and documentation of validity evidence. The process of gathering evidence of the validity of score interpretations is best characterized as ongoing throughout test development, administration, scoring, reporting, and beyond.

5.1. EVIDENCE BASED ON TEST CONTENT

The validity of test content is fundamental to arguments that test scores are valid for their intended purpose. It demands that a test developer provide evidence that test content is well-aligned within the framework and standards used in curriculum and instruction. Accordingly, detailed attention was given to this correspondence between standards and test content during test design and construction. The Regents Examination in ELA measures student achievement on the NYS Learning Standards for ELA.

5.1.1. Content Validity

Content validity is concerned with the proper definition of the construct and evidence that the test provides an accurate measure of student performance within the defined construct. The test blueprint is essentially the design document for test construction that provides an explicit definition of the construct domain that is to be represented on the exam. The test development process is in place to ensure, to the extent possible, that the blueprint is met in all operational forms of the exam. Table 5.1 presents a summary of the ELA blueprint.

Table 5.1. Test Blueprint: Regents Examination in ELA

Test Part	Suggested Time	Standards Addressed	Text Description	Student Task
Part 1 Reading Comprehension	60 minutes	RL.1-6, 10 RI.1-6, 8-10 L.3-5	3 texts Up to approximately 2,600 words total Each test will contain one literature text, one poem, and one informational text.	Students will perform a close reading of the texts and answer 24 multiplechoice questions.
Part 2 Writing from Sources: Argument	90 minutes	RI.1-10 W.1, 4,9 L.1-6	 4 texts Up to approximately 2,600 words total Each test will contain 4 informational texts. Texts may include graphically presented information. 	Students will perform a close reading of the texts and write a source-based argument, as directed by the task.
Part 3 Text Analysis: Exposition	30 minutes	RL.1-6, 10 RI.1-6, 8-10 W.2,4,9 L.1-6	1 text Up to approximately 1,000 words Each test will contain one literature or one informational text.	Students will perform a close reading of the text and write a two- to three-paragraph expository response that identifies a central idea in the text and analyzes how the author's use of one writing strategy (literary element or literary technique or rhetorical device) develops this central idea.

Overall, the test requires that students closely read eight texts of up to approximately 6,200 words total and that they answer 24 MC items, write one source-based argument, and one text-based response that identifies a central idea in the text and analyzes how the author's use of one writing strategy develops this central idea. The test assesses the Learning Standards in Reading, Writing, and Language for Grades 11–12. However, due to the integrative and cumulative nature of the standards, items may also assess standards in Grades 9–10. Exact standard coverage will vary from test to test based on the texts and writing tasks used.

5.1.2. Item Development Process

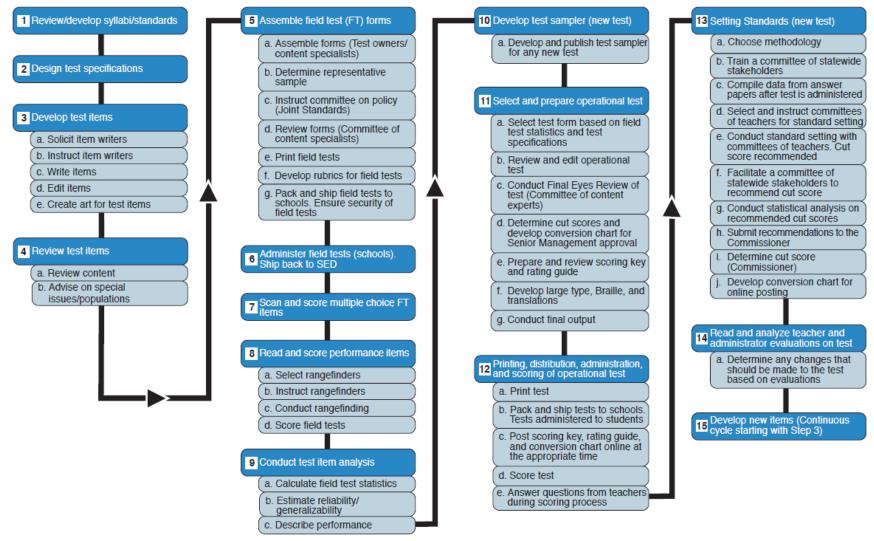
Test development is a detailed, step-by-step process of development and review cycles. An important element of this process is that all test items are developed by NYS educators in a process facilitated by state subject matter and testing experts. Bringing experienced classroom teachers into this central item development role serves to draw a strong connection between classroom and test content.

Only NYS-certified educators may participate in this process. NYSED asks for nominations from districts, and all recruiting is done with diversity of participants in mind, including diversity in gender, race/ethnicity, geographic region, and teaching experience. Educators with item writing skills from throughout the state are retained to write all items for the Regents Examinations under strict guidelines that leverage best practices (see Appendix C). State educators also conduct all item quality and bias reviews to ensure that item content is appropriate to the construct being measured and fair for all students. Finally, educators use the defined standards, test blueprint targets, and statistical information generated during field testing to select the highest-quality items for use in the operational test.

Initial item development is conducted under the criteria and guidance provided by multiple documents, including the blueprint, item writing criteria, and a content verification checklist. Both MC and CR items are included on the exam to ensure appropriate coverage of the construct domain. The item writing guidelines in Appendix C provide detailed information about how the items are developed.

Figure 5.1 summarizes the full test development process, with Steps 3 and 4 addressing initial item development and review. This figure also demonstrates the ongoing nature of ensuring the content validity of items through field test trials and final item selection for operational testing.

Figure 5.1. NYSED Test Development Process



5.1.3. Item Review Process

The item review process assists in the consistent application of rigorous item reviews intended to assess the quality of the items developed and identify items that require edits or removal from the pool of items to be field tested. The following criteria help to ensure that high-quality items are continually developed in a manner that is consistent with the test blueprint.

All reviewers participate in rigorous training designed to assist in a consistent interpretation of the standards throughout the item review process. This is a critical step in item development because consistency between the standards and what the items are asking students is a fundamental form of evidence of the validity of the intended score interpretations. Another integral component of this item review process is to review the scoring rules, or rubrics, for their clarity and consistency in what the student is being asked to demonstrate by responding to each item. Each of these elements is in place to target fairness for all students by targeting consistency in student scores and providing evidence of the validity of their interpretations. Following these reviews, only items that are approved by an assigned educator panel move forward for field testing.

Specifically, the item review process articulates four major item characteristics that NYSED looks for when developing quality items:

- 1. Language and graphical appropriateness
- 2. Sensitivity/bias
- 3. Alignment of measurement to standards
- 4. Conformity to the expectations for the specific item types and formats

Each section of the criteria includes pertinent questions that help reviewers determine if an item is of sufficient quality. Within the first two categories, criteria for language appropriateness are used to help ensure that students understand what is asked in each item and that the language in the item does not adversely affect a student's ability to perform the required task. Similarly, the sensitivity/bias criteria are used to evaluate whether items are unbiased, inoffensive, and not disadvantageous to any given subgroup(s). The art criteria assess the appropriateness and clarity of any graphics that are used within items.

The third category of the item review addresses how each item measures a given standard. This criterion asks the reviewer to comment on key aspects of how the item addresses and calls for the skills demanded by the standards. These criteria also prompt reviewers to comment on how more than one standard is addressed by a given item. The fourth category addresses the demands for different item types and formats. Reviewers evaluate each item to ensure that it conforms to the given requirements. For example, MC items must have one unambiguously correct answer and several plausible, but incorrect, answer choices.

Ongoing attention is also given to the relevance of the standards used to guide curriculum and assessment. Consistent with a desire to assess this relevance, NYSED is committed to ongoing standards review over time and periodically solicits thoughtful, specific responses from stakeholders about individual standards within the NYS Learning Standards.

5.2. EVIDENCE BASED ON RESPONSE PROCESSES

The second source of validity evidence is based on student response processes. It requires evidence that students are responding in the manner intended by the test items and rubrics and that raters are scoring those responses in a manner that is consistent with the rubrics. It is important to control and monitor whether construct-irrelevant variance in response patterns has been introduced at any point in the test development, administration, or scoring processes.

The controls and monitoring in place for the Regents Examinations include the item development process, with attention paid to mitigating the introduction of construct-irrelevant variance. The test development process details the methods used and attention given to reducing the potential for construct irrelevance in response processes. This is accomplished by attending to the quality and alignment of test content to the test blueprint and to the item development guidelines (Appendix C). Further evidence is documented in the test administration and scoring procedures and in the results of statistical analyses.

5.2.1. Administration and Scoring

Adherence to standardized administration procedures is fundamental to the validity of test scores and their interpretation, as such procedures allow for adequate and consistently applied conditions for scoring the work of every student who takes the examination. For this reason, guidelines, as contained in the *School Administrator's Manual* (https://www.nysed.gov/state-assessment/test-manuals), have been developed and implemented. All secondary-level Regents Examinations are administered under these standard conditions to support valid inferences for all students. These procedures also cover testing students with disabilities who are provided testing accommodations consistent with their Individualized Education Programs (IEPs) or Section 504 Accommodation Plans (504 Plans). Full test administration procedures are available at https://www.nysed.gov/state-assessment/high-school-regents-examinations.

The implementation of rigorous scoring procedures supports the validity of the scores. Regents test scoring practices therefore focus on producing high-quality scores. MC items are scored via local scanning at testing centers, and trained educators score CR items. Many studies focus on various elements of producing valid and reliable scores for CR items, but generally, attention to the following all contribute to valid and reliable scores for CR items:

- 1. Quality training (Hoyt & Kerns, 1999; Lumley & McNamara, 1995; Wang et al., 2010; Gorman & Rentsch, 2009; Schleicher et al., 2002; Woehr & Huffcutt, 1994; Johnson et al., 2008; Weigle, 1998)
- 2. Detection and correction of rating bias (McQueen & Congdon, 1997; Congdon & McQueen, 2000; Myford & Wolfe, 2009; Barkaoui, 2011; Patz et al., 2002)
- Consistency or reliability of ratings (Congdon & McQueen, 2000; Harik et al., 2009; McQueen & Congdon, 1997; Myford & Wolfe, 2009; Mero & Motowidlo, 1995; Weinrott & Jones, 1984)
- 4. Rubric designs that facilitate consistency of ratings (Pecheone & Chung, 2007; Wolfe & Gitomer, 2000; Cronbach et al., 1995; Cook & Beckman, 2009; Penny et al., 2000; Smith, 1993; Leacock et al., 2014).

The distinct steps for operational test scoring include close attention to each of these elements and begin before the operational test is selected. After the field test process, a set of anchor papers representing student responses across the range of possible responses for CR items is selected. The objective of these range-finding efforts is to create a training set for scorer training and execution. The scores calculated for this training set are used to generate important statistical information about the item. Training scorers to produce reliable and valid scores is the basis for creating rating guides and scoring ancillaries to be used during operational scoring.

To review and select these anchor papers, NYS educators serve as table leaders during the range-finding session. In the range-finding process, committees of educators receive a set of student papers for each field tested item. Committee members familiarize themselves with each item type and score several responses that are representative of each score point. After the independent scoring is completed, the committee reviews and discusses their results and determines consensus scores for the student responses. Atypical responses are important to identify and annotate for use in training and live scoring. The range-finding results are then used to build training materials for the vendor's scorers, who then score the rest of the field test responses to CR items. The Rating Guides for the August 2022, January 2023, and June 2023 administrations of the Regents Examination in ELA are located at https://www.nysedregents.org/hsela/.

During the range-finding and field test scoring processes, it is important to control for sources of variation in scoring. One possible source of variation in CR scores is unintended rater bias associated with items and student responses. The rater is often unaware of such bias, so this type of variation may be the most challenging source of variation in scoring to control and measure. Rater biases can appear as severity or leniency in applying the scoring rubric. Bias also includes phenomena such as the halo effect, which occurs when good or poor performance on one element of the rubric encourages inaccurate scoring of other elements. These types of rater bias can be controlled by training practices with a strict focus on rubric requirements.

The training process for operational scoring by NYS educators begins with a review and discussion of actual student work on CR items. This helps raters understand the range and characteristics typical of student responses and the kinds of mistakes students commonly make. This information is used to train raters on how to consistently apply key elements of the scoring rubric across the domain of student responses.

Raters then receive training consistent with the guidelines and ancillaries produced after field testing and are allowed to practice scoring prior to the start of live scoring. Throughout the scoring process, there are important procedures for correcting inconsistent scoring or the misapplication of scoring rubrics for CR items. When monitoring and correction do not occur during scoring, construct-irrelevant variation may be introduced. Accordingly, a scoring lead may be assigned to review the consistency of scoring for the lead's assigned staff against model responses and to be available for consultation throughout the scoring process.

Attention to the rubric design also contributes to the validity of student response processes. The rubric specifies what the student needs to provide as evidence of learning based on the question asked. The more explicit the rubric (and the item), the clearer the response expectations are for students. To facilitate the development of CR scoring rubrics, NYSED training for writing items includes specific attention to rubric development, as follows:

- The rubric should clearly specify the criteria for awarding each credit.
- The rubric should be aligned to what is asked for in the item and correspond to the knowledge or skill being assessed.
- Whenever possible, the rubric should be written to allow for alternate approaches and other legitimate methods.

In support of the goal of valid score interpretations for each student, such scoring training procedures are implemented for the Regents Examinations. Operational raters are selected based on expertise in the exam subject and are assigned a specific set of items to score. No more than one-third of the items on the test are assigned to any one rater. This increases the consistency of scoring across student responses by allowing each rater to focus on a subset of items. It also ensures that no one rater is allowed to score the entire test for any one student. This practice reduces the effect of any potential bias of a single rater on individual students. Raters are also not allowed to score the responses of their own students.

5.2.2. Statistical Analysis

A useful statistic for evaluating the response processes for MC items is an item's point-biserial correlation on the distractors. A high point-biserial on a distractor may indicate that students are not able to identify the correct response for a reason other than the difficulty of the item. A finding of poor model fit for an item may also support a finding that students are not responding in the way in which the item developer intended. As shown in Table 2.1, the point-biserial correlations for distractors in the MC items all appear to be negative or close to zero, indicating that, for the most part, students are not being drawn to an unintended construct.

5.3. EVIDENCE BASED ON INTERNAL STRUCTURE

The third source of validity evidence comes from the internal structure of the test and requires that test developers evaluate the test structure to ensure that it is functioning as intended. Such an evaluation may include attention to item interactions, tests of dimensionality, or indications of test bias for or against one or more student subgroups detected by differential item functioning (DIF) analysis. Evaluation of internal test structure also includes a review of the results of classical item analyses, test reliability, and the IRT scaling and equating. The following analyses were conducted for the exam:

- Item difficulty
- Item discrimination
- Differential item functioning (DIF)
- IRT model fit
- Test reliability
- Classification consistency
- Test dimensionality

5.3.1. Item Difficulty

Multiple analyses allow for an evaluation of item difficulty. For this exam, *p*-values and Rasch difficulty (item location) estimates were computed for MC and CR items. Items for the June 2023 Regents Examination in ELA show a range of *p*-values consistent with the targeted exam difficulty. Item *p*-values range from 0.55 to 0.82, with a mean of 0.67. The difficulty distribution illustrated in Figure 2.1 shows a wide range of item difficulties on the exam. This is consistent with general test development practice, which seeks to measure student ability along a full range of difficulty. Refer to Chapter 2 of this report for additional details.

5.3.2. Item Discrimination

How well the items on a test discriminate between high- and low-performing students is an important measure of the test structure. Items that do not discriminate well generally provide less reliable information about student performance. Table 2.1 and Table 2.2 provide point-biserial values on the correct responses; Table 2.1 also provides point-biserial values on the three distractors. The values for correct answers are 0.22 or higher for all items, indicating that most items are discriminating well between high- and low-performing students. Point-biserial values for all distractors are negative, indicating that students are responding to the items as expected during item development. Refer to Chapter 2 of this report for additional details.

5.3.3. Differential Item Functioning

Differential item functioning (DIF) was conducted for gender, race/ethnicity, needs/resource capacity (NRC) categories, and ELL status based on the data for the June 2023 administration. DIF data are only available after the administration as all Regents exams are pre-equated, meaning that the parameters used to derive the relationship between the raw and scale scores are estimated prior to the construction and administration of the operational form.

The Mantel-Haenszel (Mantel & Haenszel, 1959) and standardized mean difference (Dorans & Schmitt, 1991) methods were used to detect items that may function differently for any subgroup. The Mantel-Haenszel χ^2 is a conditional mean comparison of the ordered response categories for reference and focal groups combined over values of the matching variable score. "Ordered" means that a response earning a score of 1 on an item is better than a response earning a score of 0, a 2 is better than 1, and so on. "Conditional," on the other hand, refers to the comparison of members from the two groups who received the same score on the matching variable (the total test score in this analysis). The results of these analyses were examined by NYSED content specialists to identify potential systematic issues that could be addressed in future item writing.

5.3.4. IRT Model Fit

Model fit for the Rasch method used to estimate location (difficulty) parameters for the items on the exam provide important evidence that the internal structure of the test is of high technical quality. The number of items within a targeted range of [0.7, 1.3] is reported in Table 3.2. The mean INFIT value is 1.00, with all items falling in a targeted range of [0.7, 1.3]. As the range of [0.7, 1.3] is used as a guide for ideal fit, fit values outside of this range are considered individually. These results indicate that, for most items, the Rasch model fits the item data well.

5.3.5. Test Reliability

Test reliability is a measure of the internal consistency of a test (Cronbach, 1951). It is a measure of the extent to which the items on a test provide consistent information about student mastery of the domain. Reliability should ultimately demonstrate that student score estimates maximize consistency and therefore minimize error or, theoretically, that students who take a test multiple times would get the same score each time. The reliability estimate for the Regents Examination in ELA is 0.85, showing high reliability of student scores. Refer to Chapter 4 of this report for additional details.

5.3.6. Classification Consistency and Accuracy

Decision consistency measures the agreement between the classifications based on two nonoverlapping, equally difficult forms of the test. If two parallel forms were given to the same students, the consistency of the measure would be reflected by the extent to which the classification decisions based on the first set of test scores matched the decisions based on the second set of scores. Decision accuracy determines the extent to which measurement error causes a classification different from that expected from the true score. High decision consistency and accuracy provide strong evidence that the internal structure of a test is sound.

For the Regents Examination in ELA, both decision consistency and accuracy values are high, indicating very good consistency and accuracy of student classifications. The results for the overall consistency across all five performance levels, as well as for the dichotomies created by the four corresponding cut scores, are presented in Table 4.4. The tabled values are derived with the program BB-Class (Brennan, 2004) using the Livingston and Lewis method. The decision consistency ranged from 0.83 to 0.90, and the decision accuracy ranged from 0.88 to 0.93.

5.3.7. Test Dimensionality

In addition to model fit, a strong assumption of the Rasch model is that the construct measured by a test is unidimensional. Violation of this might suggest that the test is measuring something other than the intended content, indicating that the test structure quality is compromised. The results of a PCA conducted to test the assumption of unidimensionality provide strong evidence that a single dimension in the exam is explaining a large portion of the variance in student response data. This analysis does not characterize or explain the dimension, but a reasonable assumption can be made that the test is largely unidimensional and that the dimension most present is the targeted construct. Refer to Chapter 3 for details of this analysis.

Considering this collection of detailed analyses of the internal structure of the Regents Examination in ELA, strong evidence exists that the exam is functioning as intended and is providing valid and reliable information about student performance.

5.4. EVIDENCE BASED ON RELATIONS TO OTHER VARIABLES

Another source of validity evidence is based on the relation of the test to other variables. This source commonly encompasses two validity categories prevalent in the literature and practice: concurrent and predictive validity. To make claims about the validity of a test that is to be used for high-stakes purposes, such as the Regents Examinations, these claims could be supported by providing evidence that performance on the test correlates well with other tests that measure the same or similar constructs. Although not absolute in its ability to offer evidence that concurrent test score validity exists, such correlations can be helpful for supporting a claim of concurrent validity if the correlation is high. To conduct such studies, matched student score data for other tests measuring the same content as the Regents Examination in ELA is ideal. However, the systematic acquisition of such data is complex and costly.

Importantly, a strong connection between classroom curriculum and test content may be inferred by the fact that NYS educators, deeply familiar with the curriculum standards and their enactment in the classroom, develop all content for the Regents Examination in ELA. In terms of predictive validity, time is a fundamental constraint on gathering evidence. The gold standard for supporting the validity of predictive statements about test scores requires empirical evidence of the relationship between test scores and future performance on a defined characteristic. To the extent that the objective of the standards is to prepare students for meeting graduation requirements, it will be important to gather evidence of this empirical relationship over time.

5.5. EVIDENCE BASED ON TESTING CONSEQUENCES

There are two general approaches in the literature for evaluating consequential validity. Messick (1995) points out that adverse social consequences invalidate test use mainly if they are due to flaws in the test. In this sense, the sources of evidence documented in this report (based on the construct, internal test structure, response processes, and relation to other variables) serve as a consequential validity argument as well. This evidence supports conclusions based on test scores that social consequences are not likely to be traced to characteristics or qualities of the test itself.

Cronbach (1988), on the other hand, argues that negative consequences could invalidate test use. From this perspective, the test user is obligated to make the case for test use and to ensure appropriate and supported uses. Regardless of perspective on the nature of consequential validity, it is important to caution against uses that are not supported by the validity claims documented for this test. For example, use of this test to predict student scores on other tests is not directly supported by either the stated purposes or by the development process and research conducted on student data. A brief survey of websites for NYS universities and colleges finds that, beyond the explicitly defined use as a testing requirement toward graduation for students who have completed a course in ELA, the exam is commonly used to inform admissions and course placement decisions. Such uses can be considered reasonable, assuming the competencies demonstrated in the Regents Examination in ELA are consistent with those required in the courses for which a student is seeking enrollment or placement. Educational institutions using the exam for placement purposes are advised to examine the scoring rules for the Regents Examination in ELA and to assess their appropriateness for the inferences being made about course placement.

5.6. SUMMARY

As stated, the nature of validity arguments is not absolute. Rather it is supported through ongoing processes and studies designed to accumulate support for validity claims. The evidence provided in this report documents the evidence to date that supports the use of the Regents Examination in ELA scores for the purposes described.

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Appendix A: Operational Test Maps

Table A.1. August 2022 Test Map

		Max.				Point-	Rasch	
Position	Item Type	Points	Weight	Cluster	Mean	Biserial	Difficulty	INFIT
1	MC	1	1	RL.4	0.48	0.35	-0.6751	1.15
2	MC	1	1	RL.5	0.45	0.33	-0.5229	1.17
3	MC	1	1	RL.4	0.63	0.31	-1.4489	1.16
4	MC	1	1	L.5	0.61	0.49	-1.3435	0.96
5	MC	1	1	RL.3	0.74	0.54	-2.1391	0.85
6	MC	1	1	RL.3	0.62	0.60	-1.4262	0.83
7	MC	1	1	RL.5	0.49	0.44	-0.7472	1.04
8	MC	1	1	RL.2	0.55	0.47	-1.0644	0.99
9	MC	1	1	RL.2	0.53	0.52	-0.9560	0.94
10	MC	1	1	RL.5	0.56	0.41	-1.1153	1.06
11	MC	1	1	L.5	0.47	0.32	-0.6245	1.18
12	MC	1	1	L.5	0.61	0.55	-1.3660	0.88
13	MC	1	1	RL.2	0.64	0.50	-1.5330	0.92
14	MC	1	1	RL.5	0.50	0.52	-0.7832	0.94
15	MC	1	1	RI.3	0.61	0.61	-1.3510	0.82
16	MC	1	1	RI.4	0.52	0.60	-0.8911	0.84
17	MC	1	1	RI.3	0.52	0.52	-0.9055	0.93
18	MC	1	1	RI.5	0.56	0.60	-1.0862	0.83
19	MC	1	1	RI.3	0.57	0.58	-1.1590	0.86
20	MC	1	1	RI.3	0.38	0.43	-0.1739	1.04
21	MC	1	1	RI.4	0.53	0.64	-0.9632	0.80
22	MC	1	1	RI.2	0.38	0.47	-0.1967	0.98
23	MC	1	1	RI.4	0.53	0.53	-0.9344	0.92
24	MC	1	1	RI.2	0.40	0.49	-0.2720	0.96
25	CR	6	4	RI.1-6 &10, W.1,4&9, L.1-6	2.18	0.70	1.2260	1.18
26	CR	4	2	RI.1-6 &10, W.1,4&9	1.76	0.70	0.8580	0.96

Table A.2. January 2023 Test Map

		Max.	•			Point-	Rasch	
Position	Item Type	Points	Weight	Cluster	Mean	Biserial	Difficulty	INFIT
1	MC	1	1	RL.5	0.67	0.34	-1.6341	1.07
2	MC	1	1	RL.5	0.52	0.44	-0.8330	1.00
3	MC	1	1	RL.4	0.61	0.35	-1.2699	1.06
4	MC	1	1	RL.2	0.52	0.40	-0.8469	1.05
5	MC	1	1	L.4	0.68	0.49	-1.6496	0.93
6	MC	1	1	RL.3	0.68	0.47	-1.6419	0.94
7	MC	1	1	RL.3	0.75	0.52	-2.0520	0.87
8	MC	1	1	RL.2	0.61	0.41	-1.3060	1.02
9	MC	1	1	RL.3	0.55	0.50	-0.9864	0.93
10	MC	1	1	RL.3	0.53	0.42	-0.9096	1.02
11	MC	1	1	RL.4	0.65	0.50	-1.4748	0.91
12	MC	1	1	L.4	0.62	0.48	-1.3204	0.94
13	MC	1	1	RL.4	0.55	0.35	-0.9864	1.10
14	MC	1	1	L.5	0.53	0.42	-0.8678	1.02
15	MC	1	1	RI.4	0.58	0.54	-1.1130	0.89
16	MC	1	1	RI.3	0.44	0.42	-0.4490	1.02
17	MC	1	1	L.4	0.32	0.35	0.1935	1.08
18	MC	1	1	RI.2	0.54	0.51	-0.9375	0.92
19	MC	1	1	L.4	0.52	0.61	-0.8191	0.82
20	MC	1	1	RI.3	0.53	0.64	-0.8608	0.78
21	MC	1	1	RI.4	0.55	0.52	-1.0074	0.91
22	MC	1	1	RI.5	0.47	0.43	-0.6033	1.02
23	MC	1	1	RI.6	0.56	0.56	-1.0566	0.87
24	MC	1	1	RI.2	0.55	0.60	-0.9724	0.83
25	CR	6	4	RI.1-6 &10, W.1,4&9, L.1-6	2.43	0.72	1.0238	1.08
26	CR	4	2	RI.1-6 &10, W.1,4&9	1.78	0.70	0.6074	0.98

Table A.3. June 2023 Test Map

		Max.				Point-	Rasch	
Position	Item Type	Points	Weight	Cluster	Mean	Biserial	Difficulty	INFIT
1	MC	1	1	RL.4	0.55	0.36	-1.0408	1.07
2	MC	1	1	RL.3	0.49	0.22	-0.7498	1.21
3	MC	1	1	RL.3	0.57	0.48	-1.1458	0.94
4	MC	1	1	RL.5	0.64	0.38	-1.4916	1.04
5	MC	1	1	L.5	0.67	0.47	-1.6503	0.93
6	MC	1	1	RL.4	0.50	0.36	-0.8051	1.07
7	MC	1	1	RL.3	0.55	0.40	-1.0199	1.02
8	MC	1	1	RL.6	0.46	0.25	-0.5904	1.19
9	MC	1	1	RL.2	0.56	0.46	-1.1037	0.96
10	MC	1	1	RL.2	0.59	0.49	-1.2519	0.92
11	MC	1	1	RL.5	0.67	0.39	-1.5608	1.06
12	MC	1	1	L.5	0.55	0.51	-0.9480	0.94
13	MC	1	1	RL.4	0.64	0.48	-1.4089	0.96
14	MC	1	1	L.5	0.53	0.44	-0.8363	1.03
15	MC	1	1	RI.3	0.58	0.48	-1.0961	0.99
16	MC	1	1	L.4	0.46	0.49	-0.4414	0.96
17	MC	1	1	RI.2	0.48	0.47	-0.5870	0.99
18	MC	1	1	RI.5	0.59	0.64	-1.1460	0.79
19	MC	1	1	RI.3	0.54	0.56	-0.8921	0.89
20	MC	1	1	L.5	0.50	0.57	-0.6907	0.87
21	MC	1	1	RI.3	0.52	0.50	-0.7669	0.96
22	MC	1	1	RI.3	0.54	0.59	-0.8781	0.85
23	MC	1	1	RI.4	0.45	0.48	-0.4206	0.98
24	MC	1	1	RI.2	0.62	0.65	-1.2761	0.78
25	CR	6	4	RI.1-6 &10, W.1,4&9, L.1-6	2.19	0.70	0.9617	1.12
26	CR	4	2	RI.1-6 &10, W.1,4&9	1.76	0.67	0.7567	0.97

Appendix B: Raw-to-Theta-to-Scale Score Conversions

Table B.1. August 2022 Score Table

Raw Score	Ability	Scale Score	Raw Score	Ability	Scale Score	Raw Score	Ability	Scale Score
0	-5.5691	0.000	19	-0.9670	34.279	39	0.9519	81.051
1	-4.3424	0.588	20	-0.8899	37.162	40	1.1178	82.586
2	-3.6181	1.357	21	-0.8139	39.951	41	1.2994	84.017
3	-3.1829	2.024	22	-0.7384	42.686	42	1.4992	85.354
4	-2.8671	2.857	23	-0.6631	45.353	43	1.7199	86.647
5	-2.6172	3.739	24	-0.5874	47.972	44	1.9640	87.944
6	-2.4094	4.679	25	-0.5110	50.547	45	2.2330	89.300
7	-2.2309	5.677	26	-0.4333	53.058	46	2.5264	90.738
8	-2.0741	6.715	27	-0.3540	55.516	47	2.8415	92.220
9	-1.9339	7.786	28	-0.2725	57.970	48	3.1740	93.680
10	-1.8067	8.872	29	-0.1884	60.358	49	3.5208	95.029
11	-1.6901	10.360	30	-0.1010	62.705	50	3.8816	96.241
12	-1.5821	13.151	31	-0.0098	65.029	51	4.2608	97.167
13	-1.4811	16.130	32	0.0859	67.282	52	4.6693	98.091
14	-1.3860	19.165	33	0.1868	69.486	53	5.1302	98.861
15	-1.2956	22.224	34	0.2937	71.642	54	5.6966	99.807
16	-1.2092	25.309	35	0.4073	73.706	55	6.5384	99.999
17	-1.1261	28.354	36	0.5287	75.705	56	7.8396	100.000
18	-1.0455	31.334	38	0.7997	79.386	38	0.7997	79.386

Table B.2. January 2023 Score Table

Raw		Scale	Raw		Scale	Raw		Sca
Score	Ability	Score	Score	Ability	Score	Score	Ability	Scor
0	-5.6469	0.000	19	-1.0121	32.587	38	0.6659	77.7
1	-4.4196	0.523	20	-0.9360	35.437	39	0.8069	79.46
2	-3.6937	1.243	21	-0.8615	38.207	40	0.9589	81.12
3	-3.2565	1.903	22	-0.7877	40.904	41	1.1229	82.62
4	-2.9381	2.670	23	-0.7145	43.548	42	1.3002	84.02
5	-2.6853	3.482	24	-0.6413	46.116	43	1.4919	85.30
6	-2.4745	4.360	25	-0.5678	48.645	44	1.6994	86.53
7	-2.2930	5.306	26	-0.4934	51.124	45	1.9242	87.73
8	-2.1331	6.305	27	-0.4178	53.541	46	2.1680	88.97
9	-1.9899	7.342	28	-0.3404	55.929	47	2.4329	90.28
10	-1.8601	8.408	29	-0.2607	58.306	48	2.7212	91.66
11	-1.7411	9.680	30	-0.1782	60.631	49	3.0359	93.10
12	-1.6310	11.856	31	-0.0924	62.924	50	3.3808	94.53
13	-1.5282	14.725	32	-0.0023	65.205	51	3.7612	95.87
14	-1.4318	17.688	33	0.0924	67.424	52	4.1856	96.98
15	-1.3405	20.685	34	0.1926	69.606	53	4.6724	98.09
16	-1.2536	23.715	35	0.2993	71.746	54	5.2691	99.09
17	-1.1703	26.732	36	0.4130	73.806	55	6.1410	99.99
18	-1.0900	29.686	37	0.5349	75.803	56	7.4615	100.00

Table B.3. June 2023 Score Table

Table B	.o. oune z	2023 30016						
Raw	Ability	Scale	Raw	Ability	Scale		Raw	The second secon
Score	Ability	Score	Score	Ability	Score		Score	-
0	-5.5421	0.000	19	-0.9479	34.994		38	38 0.7756
1	-4.3190	0.608	20	-0.8694	37.916		39	39 0.9139
2	-3.5993	1.386	21	-0.7922	40.743		40	40 1.0609
3	-3.1680	2.063	22	-0.7157	43.506		41	41 1.2172
4	-2.8554	2.888	23	-0.6396	46.179		42	42 1.3830
5	-2.6080	3.774	24	-0.5635	48.793		43	43 1.5592
6	-2.4021	4.715	25	-0.4869	51.340		44	44 1.7465
7	-2.2249	5.713	26	-0.4095	53.801		45	45 1.9463
8	-2.0687	6.752	27	-0.3308	56.220		46	46 2.1603
9	-1.9286	7.829	28	-0.2503	58.601		47	47 2.3907
10	-1.8011	8.921	29	-0.1676	60.915		48	48 2.6404
11	-1.6837	10.475	30	-0.0820	63.187		49	49 2.9126
12	-1.5745	13.372	31	0.0068	65.421		50	50 3.2112
13	-1.4722	16.413	32	0.0996	67.584		51	51 3.5417
14	-1.3754	19.520	33	0.1970	69.696		52	52 3.9136
15	-1.2833	22.658	34	0.2997	71.754		53	53 4.3472
16	-1.1952	25.822	35	0.4082	73.721		54	54 4.8929
17	-1.1103	28.937	36	0.5233	75.620		55	55 5.7188
18	-1.0280	31.992	37	0.6455	77.427	_	56	56 7.0106

Appendix C: Item Writing Guidelines

GENERAL RULES FOR WRITING MULTIPLE-CHOICE ITEMS

1. Use either a direct question or an incomplete statement as the item stem, whichever seems more appropriate to effective presentation of the item.

Some item ideas can be expressed more simply and clearly in the incomplete statement style of question. On the other hand, some items seem to require direct question stems for the most effective expression. Teachers should use the item style that seems most appropriate.

2. Items should be written in clear and simple language, with vocabulary kept as simple as possible.

Like any other item, the multiple-choice item should be perfectly clear. Difficult and technical vocabulary should be avoided unless essential for the purpose of the question. The important elements should generally appear early in the statement of the item, with qualifications and explanations following.

3. Each item should have one and only one correct answer.

While this requirement is obvious, it is not always fulfilled. Sometimes writers produce items involving issues so controversial and debatable that even experts are unable to agree on one correct answer. More often the trouble is failure to consider the full implications of each response.

4. Base each item on a single central problem.

A multiple-choice item functions most effectively when the student is required to directly compare the relative merits of a number of specific responses to a definite problem. An item consisting merely of a series of unrelated true-false statements, all of which happen to begin with the same phrase, is unacceptable.

5. State the central problem of the item clearly and completely in the stem.

The stem should be meaningful by itself. It should be clear and should convey the central problem of the item. It should not be necessary for the student to read and reread all the responses before being able to understand the basis upon which the student is to make a choice.

6. In general, include in the stem any words that must otherwise be repeated in each response.

The stem should contain everything the answers have in common or as much as possible of their common content. This practice serves to make the item shorter so that it can be read and grasped more quickly.

7. Avoid negative statements.

Negative statements in multiple-choice items lead to unnecessary difficulties and confusion. Special care must be exercised to avoid the double negative.

8. Avoid excessive "window dressing."

The item should contain only material relevant to its solution unless selecting what is relevant is part of the problem.

9. Make the responses grammatically consistent with the stem and parallel with one another in form.

10. Make all responses plausible and attractive to students who lack the information or ability tested by the item.

The incorrect responses should be plausible answers. So far as possible, each response should be designed specifically to attract students who have certain misconceptions or who tend to make certain common errors.

11. Arrange the responses in logical order, if one exists.

Where the responses consist of numbers or letters, they should ordinarily be arranged in ascending order. Events should be listed in the order in which they occurred, from earliest to most recent, except when this order would clue the answer. This practice helps ensure the student will mark the answer correctly.

12. Make the responses independent and mutually exclusive.

Responses should not be interrelated in meaning. Responses that are not mutually exclusive aid the student in eliminating wrong answers and reduce the reliability of the item by decreasing the number of effective, functioning responses.

13. Avoid extraneous clues.

Since the student is required to associate one of several alternative responses with the stem, any aspect of the question that provides an extraneous basis for correctly associating the right answer or for eliminating a wrong response constitutes an undesirable clue.

- 14. Avoid using "all of the above" and "none of the above" as alternatives.
- 15. Avoid using the phrase "of the following" in the stem.

CHECKLIST OF TEST CONSTRUCTION PRINCIPLES

(Multiple-Choice Items)

		YES	NO
1.	Is the item significant?		
2.	Does the item have curricular validity?		
3.	Is the item presented in clear and simple language, with vocabulary kept as simple as possible?		
4.	Does the item have one and only one correct answer?		
5.	Does the item state one single central problem completely in the stem? (See Helpful Hint below.)		
6.	Does the stem include any extraneous material ("window dressing")?		
7.	Are all responses grammatically consistent with the stem and parallel with one another in form?		
8.	Are all responses plausible (attractive to students who lack the information tested by the item)?		
9.	Are all responses independent and mutually exclusive?		
10.	Are there any extraneous clues due to grammatical inconsistencies, verbal associations, length of response, etc.?		
11.	Were the principles of Universal Design used in constructing the item?		

HELPFUL HINT

To determine if the stem is complete (meaningful all by itself):

- Cover up the responses and read just the stem.
- Try to turn the stem into a short-answer question by drawing a line after the last word. If it is not a good, short-answer item, then there may be a problem with the stem.
- The stem must consist of a statement that contains a verb.

STANDARDS INTERPRETATIONS FOR ELA MULTIPLE-CHOICE ITEMS

Grade 11- RL.2

Determine two or more themes or central ideas of a text and analyze their development over the course of the text, including how they interact and build on one another to produce a complex account; provide an objective summary of the text.

Determine central ideas or themes of a text and analyze their development; summarize the key supporting details and ideas. **Central aspect:** Items measure students' ability to analyze the development of a central idea across the text, and to summarize the text's ideas.

Item Measuring CCLS R	Yes/No	If "No," Explain or Describe
Measures central aspect: The item requires students to identify or summarize how a story's central themes/ideas develop .		
Possible stems may include:		
Stem: The author develops the central idea in the text by		
<u>Stem</u> : Which direct quote from the passage best reflects the moment when the theme changes?		
<u>Stem</u> : Which alternative title best expresses the central idea of the poem/story?		
Measures supporting aspect: The item requires students to identify or summarize a story's central themes/ideas.		
Possible stems may include:		
Stem : Which detail would be most important to include in a summary of the text?		
<u>Stem</u> : Which direct quote from the passage best summarizes the theme of the text?		
Stem: Which statement/claim best states the central idea of the text?		
The item stem does not * reveal: - the central themes/ideas - how the central themes/ideas develop *unless the theme is presented in the stem in order to ask the student to identify or explain a contrast		
The analysis in the item is supported by the text (i.e., there is development of central themes/ideas).		
The item requires students to comprehend the majority of the passage to answer the item correctly.		
THE ITEM MEASURES THIS STANDARD.	_	

Analyze the impact of the author's choices regarding how to develop and relate elements of a story or drama (e.g., where a story is set, how the action is ordered, how the characters are introduced and developed).

Analyze how and why individuals, events, or ideas develop and interact over the course of a text.

Central aspect: Items measure students' ability to analyze how different elements of a story interact with one another. Elements may include how a plot is developed in episodes, how the setting shapes the plot, how characters respond to changes in the story, or how specific dialogue and action propel the plot.

Yes/No	If "No," Explain or Describe
	Yes/No

Determine the meaning of words and phrases as they are used in the text, including figurative and connotative meanings; analyze the impact of specific word choices on meaning and tone, including words with multiple meanings or language that is particularly fresh, engaging, or beautiful.

Interpret words and phrases as they are used in a text, including determining technical, connotative, and figurative meanings, and analyze how specific word choices shape meaning or tone.

Central aspects: Items measure the students' ability to determine the meaning of multiple-meaning words as they are used in text OR analyze the impact of specific word choices.

Item Measuring CCLS R.4	Yes/No	If "No," Explain or Describe
Measures central aspect: Item requires that the student glean meaning of multiple-meaning vocabulary word from context.		
Possible stems may include:		
<u>Stem</u> : Which word or phrase from sentence X helps the reader understand the meaning of word Y in the passage?		
Stem: Which line best helps define the meaning of word X?		
$\underline{\textbf{Stem}}\text{: Define X as used in lines $\#-$\# (figurative/connotative/analogous word use)}$		
OR		
Measures central aspect: Item requires analysis of the impact of word choice and literary devices on the meaning, tone, or analysis developed in the text.		
Possible stems may include:		
<u>Stem</u> : What does the word/phrase X in the story suggest about the narrator's actions/decisions?		
<u>Stem</u> : What affect does X word in line # have on the tone/plot/description of character?		
The item stem does not reveal: - the connotation or impact of the literary device - the meaning of the word or phrase		
The analysis in the item is supported by the text.		
THE ITEM MEASURES THIS STANDARD.		

Analyze how an author's choices concerning how to structure specific parts of a text contribute to its overall structure and meaning as well as its aesthetic impact.

Analyze the structure of texts, including how specific sentences, paragraphs, and larger portions of the text (e.g., a section, chapter, scene, or stanza) relate to each other and the whole.

Central aspect: Items measure students' ability to analyze how specific phrases, paragraphs, and sections of a text develop the plot and themes of a story. In contrast to RL.3, RL.5 focuses on students' ability to identify how structure shapes meaning in a story.

Item Measuring CCLS RL.5	Yes/No	If "No," Explain or Describe
Measures central aspect: The item requires analysis of how a particular sentence/paragraph develops/contributes to the overall structure, theme, setting, or plot of the passage.		
Possible stems may include:		
Stem : What effect does paragraph X have on the story/poem?		
<u>Stem</u> : Lines # – # develop a key theme in the story by		
<u>Stem</u> : Which statement accurately describes the effect of the story's structure?		
The item stem does not reveal: - key analysis - key differences/similarities		
The analysis in the item is supported by the text.		
The item requires students to comprehend the majority of the passage to answer the item correctly.		
THE ITEM MEASURES THIS STANDARD.		

Analyze a case in which grasping point of view requires distinguishing what is directly stated in a text from what is really meant. Assess how point of view or purpose shapes the content and style of a text.

Central aspects: Items measure students' ability to assess how point of view or purpose shapes the content and style of a text **OR** to analyze how an author builds a point of view through the narrator, different characters, and/or the audience.

Item Measuring CCLS RL.6	Yes/No	If "No," Explain or Describe
Measures central aspect: The item requires students to identify how an author uses specific techniques to distinguish the narrator's point of view from those of the other characters and the audience.		
Possible stems may include:		
<u>Stem</u> : How does the narrator's point of view contrast with the point of view of another character in the story?		
<u>Stem</u> : How does the author help the reader better understand the points of view of X and Y?		
Stem: What is an effect of telling the story from X's point of view?		
Stem: How does the author develop X's character in lines # – #		
Measures supporting aspect: The item requires students to identify author, narrator, or characters' point of view.		
Possible stems may include:		
Stem: What do lines # – # suggest about author's point of view?		
Stem : In which sentence does the narrator reveal the irony/suspense/etc. of his situation?		
The item stem does not reveal: - all necessary points of view - how the point of view is developed using a specific strategy - clues that would limit reliance on text		
The analysis in the item is supported by the text.		
THE ITEM MEASURES THIS STANDARD.		

Determine two or more central ideas of a text and analyze their development over the course of the text, including how they interact and build on one another to provide a complex analysis; provide an objective summary of the text.

Determine central ideas or themes of a text and analyze their development; summarize the key supporting details and ideas. **Central aspect:** Items measure students' ability to analyze the development of a central idea across the text, and to summarize the text's ideas.

Manager control concept. The item requires students to identify or	
Measures central aspect: The item requires students to identify or summarize how a story's central ideas develop or interact .	
Possible stems may include:	
Stem: The author develops a central idea/claim in the text by	
Stem: Lines # – # develop a central idea/claim by	
<u>Stem</u> : Which statement best describes how lines # – # develop the author's claim/central idea?	
Measures supporting aspect: The item requires students to identify or summarize a text's central themes or an author's central claim. Possible stems may include:	
<u>Stem</u> : Which detail would be most important to include in a summary of the text?	
<u>Stem</u> : Which direct quote from the passage best summarizes a central idea of the text/central claim of the author?	
Stem: Which statement best states the central idea of the text/ central claim of the author?	
The item stem does not * reveal: - the central ideas/claims - how the central ideas/claims develop *unless the idea/claim is presented in the stem in order to ask the student to identify or explain a contrast	
The analysis in the item is supported by the text (i.e., there is development of central themes/ideas).	
The item requires students to comprehend the majority of the passage to answer the item correctly.	
THE ITEM ALIGNS TO THIS STANDARD.	

Analyze a complex set of ideas or sequence of events and explain how specific individuals, ideas, or events interact and develop over the course of the text.

Analyze how and why individuals, events, or ideas develop and interact over the course of a text.

Central aspect: Items measure students' ability to analyze how different elements of a text interact with one another. Elements may include how ideas, events or claims connect and overlap over the course of the text to support a central claim/idea.

Item Measuring CCLS RI.3	Yes/No	If "No," Explain or Describe
Measures central aspect: The item requires identification or analysis of the interaction between developing text elements (events, ideas, and supporting claims).		
Possible stems may include:		
<u>Stem</u> : How do lines # – # develop an idea/support a claim?		
Stem: What does line # reveal about an event/idea?		
Stem: The examples provided in lines # – # serve mostly to illustrate		
Measures supporting aspect: The item requires identification or analysis of the interaction between text elements (events, ideas, and points/claims).		
Possible stems may include:		
Stem: How is idea introduced into the passage?		
<u>Stem</u> : What statement best describes the relationship between events/individuals/ideas X and Y?		
The item stem does not reveal: - the interaction of elements* - the key change/development * unless the interaction is identified in the stem to set up a question regarding the contrast		
The analysis in the item is supported by the text (i.e., there are development of text elements).		
The item requires students to comprehend the majority of the passage to answer the item correctly.		
THE ITEM MEASURES THIS STANDARD.		

Determine the meaning of words and phrases as they are used in a text, including figurative, connotative, and technical meanings; analyze how an author uses and refines the meaning of a key term or terms over the course of a text (e.g., how Madison defines faction in Federalist No. 10).

Interpret words and phrases as they are used in a text, including determining technical, connotative, and figurative meanings, and analyze how specific word choices shape meaning or tone.

Central aspects: Items measure the students' ability to determine the meaning of multiple-meaning words as they are used in text **OR** analyze the impact of specific word choices.

Item Measuring CCLS RI.4	Yes/No	If "No," Explain or Describe
Measures central aspect: Item requires that the student glean meaning of multiple-meaning vocabulary word from context.		
Possible stems may include:		
Stem : Which word or phrase from sentence X helps the reader understand the meaning of word Y in the passage?		
<u>Stem</u> : Define X as used in lines # – # (figurative/connotative/analogous word use)		
OR		
Measures central aspect: Item requires analysis of the impact of word choice and literary devices on the meaning, tone, or analysis developed in the text.		
Possible stems may include:		
Stem : What does the word/phrase X in the story suggest about the author's actions/decisions?		
<u>Stem</u> : What affect does X word in line # have on the tone/description of the text?		
The item stem does not reveal: - the interaction of elements * - the key change/development * unless the interaction is identified in the stem to set up a question regarding the contrast		
The analysis in the item is supported by the text.		
THE ITEM MEASURES THIS STANDARD.		

Analyze and evaluate the effectiveness of the structure an author uses in his or her exposition or argument, including whether the structure makes points clear, convincing, and engaging.

Analyze the structure of texts, including how specific sentences, paragraphs, and larger portions of the text (e.g., a section, chapter, scene, or stanza) relate to each other and the whole.

Central aspect: Items measure students' ability to analyze how specific phrases, paragraphs, and sections of a text develop an analysis or argument. In contrast to RI.3, RI.5 focuses on students' ability to identify how structure shapes meaning and efficacy in a text.

Item Measuring CCLS RI.5	Yes/No	If "No," Explain or Describe
Measures central aspect: The item requires analysis of how a particular sentence/paragraph develops/contributes to the overall structure, theme, meaning, or efficacy of the passage.		
Possible stems may include:		
Stem: What effect does paragraph X have on the text?		
Stem: Lines # – # develop a key concept of the article by		
Measures supporting aspect: The item requires identification of how a particular sentence/paragraph develops/contributes to the overall structure, theme, meaning, efficacy of the passage. Possible stems may include: Stem: The purpose of lines # – # in the passage is to (support a claim/give necessary information/offer a counterclaim).		
The item stem does not reveal: - key analysis - key differences/similarities		
The analysis in the item is supported by the text.		
The item requires students to comprehend the majority of the passage to answer the item correctly.		
THE ITEM MEASURES THIS STANDARD.		

Determine an author's point of view or purpose in a text in which the rhetoric is particularly effective, analyzing how style and content contribute to the power, persuasiveness, or beauty of the text.

Assess how point of view or purpose shapes the content and style of a text.

Central aspects: Items measure students' ability to assess how point of view or purpose shapes the content and style of a text **OR** analyze how an author builds a point of view/purpose through the use of rhetoric.

Item Measuring CCLS RI.6	Yes/No	If "No," Explain or Describe
Measures central aspect: The item requires students to assess how the author's point of view shapes the content or style of a text.		
Possible stems may include:		
<u>Stem</u> : What do lines # – # suggest about author's point of view/purpose		
<u>Stem</u> : What does the author's choice of words in lines #-# suggest about his or her point of view/purpose?		
Stem: The function of lines # – # in the passage is to		
Measures central aspect: The item requires students to analyze how an author uses rhetoric to build a point of view or purpose.		
Possible stems may include:		
Stem: Paragraph X supports the author's central claim by		
Stem: Lines # – # and lines # – # reinforce the author's argument by		
Stem: The choice of words in lines # serve the purpose of		
The analysis in the item is supported by the text (i.e., the text contains a distinct point of view/purpose that shapes the content and style of the text and/or the text contains effective rhetoric).		
The item requires students to comprehend the majority of the passage to answer the item correctly.		
THE ITEM MEASURES THIS STANDARD.		

Determine or clarify the meaning of unknown and multiple-meaning words and phrases based on grades 11–12 reading and content.

Determine or clarify the meaning of unknown and multiple-meaning words and phrases by using context clues, analyzing meaningful word parts, and consulting general and specialized reference materials, as appropriate.

Central aspect: Items measuring this standard test students' ability to determine the meaning of words and phrases in context.

Central aspect: Items measuring this standard test students ability to determine the meaning or words and phrases in con-					
Item Measuring CCLS L.4	Yes/No	If "No," Explain or Describe			
Aligned: Item requires that the student glean meaning of an above-					
grade-level or otherwise unfamiliar vocabulary word from context.					
Possible stems may include:					
<u>Stem</u> : Which word or phrase from sentence X helps the reader understand the meaning of word Y in the passage?					
Stem: Which line best helps define the meaning of word X?					
<u>Stem</u> : What is the definition of X as used in lines # – # (figurative/connotative/analogous word use)					
The item stem does not reveal: - the meaning of the word or phrase					
The text provides the required context.					
THE ITEM MEASURES THIS STANDARD.					

Demonstrate understanding of figurative language, word relationships, and nuances in word meanings.

Central aspect: Items measuring this standard test students' ability to demonstrate understanding of figurative language, word relationships, and nuances in word meanings.

relationships, and huances in word meanings.						
Item Measuring CCLS L.5	Yes/No	If "No," Explain or Describe				
Aligned: The item requires students to demonstrate understanding of figurative language, word relationships, and nuances in word meanings.						
Possible stems may include:						
Stem: The statement in line # suggests that						
<u>Stem</u> : The author's use of figurative language in lines # – # of the text serves to illustrate/emphasize						
Stem: In the context of the entire text, phrase X as used in lines # – # most likely means						
The item stem does not reveal: - the meaning of the figurative language/word relationships - the nuances in word meanings						
The analysis in the item is supported by the text.						
THE ITEM MEASURES THIS STANDARD.						

MULTIPLE-CHOICE ITEM REVIEW CHECKLIST FORM

Text	Passage	Set	Item
Reviewer	Da	ate	

	CRITERIA	Accept Revise Reject
1	The item is free of content that might offend, typecast, or lead to offensive or stereotypic inferences regarding individuals or groups of different genders, ethnicities, locations, religions, socio-economic status, political views, family situations, ability, or physical or mental conditions, etc. The item would not be construed as offensive to any constituency—even if taken out of test context.	
2	The item asks a question worth asking—it is not trivial, tricky, unrelated to a close reading of the text, or unnecessary to an understanding or analysis of the text.	
3	The item is purposeful, well-crafted, and succinct—language and word choice is appropriate, clear, direct, graceful, etc.	
4	Item vocabulary is on grade level—unless item assesses vocabulary.	
5	The item only uses acronyms or foreign words whose meaning is clear in the text.	
6	The item is written in the active voice and present tense—unless another tense is appropriate.	
7	Item identification of persons, locations, etc. is identical to the text and consistent throughout entire item set.	
8	Item references to lines in the text are correct.	
9	The item does not reference the author of the text by name.	
10	The item stem presents one problem.	
11	The item stem presents sufficient information and qualifications to enable the student to answer the question.	
12	The item stem does not include words, phrases, number, tense, or grammar that cue particular options.	
13	The item has one and only one correct text-based option which requires understanding and analysis of the text.	
14	Incorrect item options are text based and plausible.	
15	Item options are written such that the correct option is not obvious because of word choice, length, etc.	
16	Item options are consistent with the stem and written using parallel structures—forms, styles, tenses, etc.	
17	Item options are mutually exclusive, but not opposites.	
18	Item options are ordered by their appearance in the text, length, etc. (unless randomized).	
19	The item does not use absolute statements—never, always, impossible, etc.	
20	The item does not use copyrighted or trademarked words or references.	
21	The item is unique in the item set—it does not repeat similar concepts or wording or other items.	
22	The item's correct option is not keyed by other items in the item set.	
	Comments or Suggestions	

Appendix D: Tables and Figures for August 2022 Administration

Table D.1. MC Item Analysis Summary: Regents Examination in ELA—August 2022

Item	#Students	<i>p</i> -Value	SD	Point- Biserial	Point- Biserial Distractor 1	Point- Biserial Distractor 2	Point- Biserial Distractor 3
1	9,301	0.38	0.49	0.23	-0.12	-0.10	-0.04
2	9,301	0.42	0.49	0.24	-0.16	-0.10	-0.04
3	9,301	0.58	0.49	0.32	-0.17	-0.17	-0.12
4	9,301	0.57	0.50	0.34	-0.17	-0.16	-0.14
5	9,301	0.73	0.45	0.41	-0.24	-0.21	-0.18
6	9,301	0.53	0.50	0.39	-0.16	-0.16	-0.22
7	9,301	0.41	0.49	0.28	-0.21	-0.12	-0.01
8	9,301	0.44	0.50	0.32	-0.19	-0.13	-0.08
9	9,301	0.43	0.49	0.38	-0.20	-0.21	-0.08
10	9,301	0.48	0.50	0.21	-0.07	-0.05	-0.16
11	9,301	0.39	0.49	0.25	-0.07	-0.12	-0.12
12	9,301	0.46	0.50	0.33	-0.21	-0.12	-0.10
13	9,301	0.60	0.49	0.33	-0.14	-0.20	-0.18
14	9,301	0.29	0.46	0.32	-0.12	-0.15	-0.08
15	9,301	0.49	0.50	0.41	-0.16	-0.23	-0.14
16	9,301	0.39	0.49	0.35	-0.18	-0.15	-0.12
17	9,301	0.51	0.50	0.33	-0.16	-0.19	-0.08
18	9,301	0.55	0.50	0.32	-0.21	-0.14	-0.12
19	9,301	0.51	0.50	0.36	-0.18	-0.16	-0.13
20	9,301	0.35	0.48	0.30	-0.14	-0.07	-0.13
21	9,301	0.46	0.50	0.38	-0.16	-0.15	-0.18
22	9,301	0.34	0.47	0.26	-0.06	-0.07	-0.16
23	9,301	0.52	0.50	0.36	-0.21	-0.15	-0.12
24	9,301	0.38	0.49	0.30	-0.11	-0.15	-0.09

Table D.2. CR Item Analysis Summary: Regents Examination in ELA—August 2022

Item	#Students	Min. Score	Max. Score	Mean	SD	p-Value	Point- Biserial
25	9,301	0	6	2.87	1.25	0.48	0.86
26	9,301	0	4	1.79	1.15	0.45	0.72

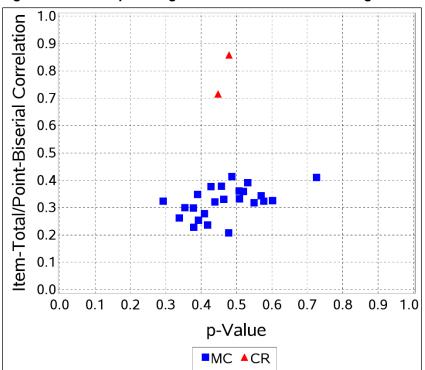


Figure D.1. Scatterplot: Regents Examination in ELA—August 2022

Table D.3. Descriptive Statistics in *p*-Value and Point-Biserial Correlation: Regents Examination in ELA—August 2022

Statistic	N	Mean	Min.	Q1	Median	Q3	Max.
<i>p</i> -Value	26	0.47	0.29	0.39	0.46	0.52	0.73
Point-Biserial	26	0.36	0.21	0.30	0.33	0.38	0.86

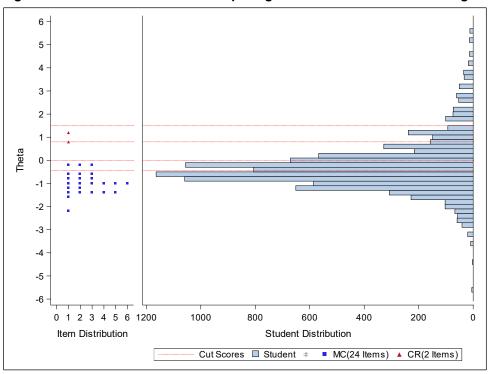


Figure D.2. Student Performance Map: Regents Examination in ELA—August 2022



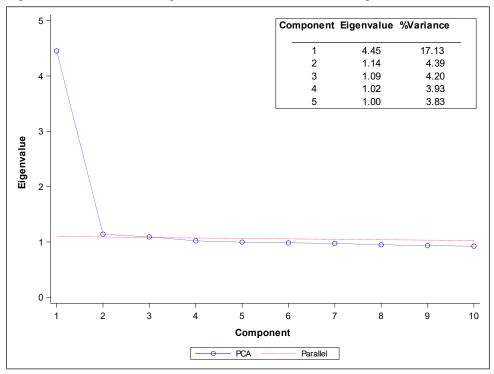


Table D.4. Summary of Item Residual Correlations: Regents Examination in ELA—August 2022

Statistic	Value
N	325
Mean	-0.03
SD	0.04
Minimum	-0.14
P ₁₀	-0.10
P ₂₅	-0.05
P ₅₀	-0.03
P ₇₅	-0.01
P ₉₀	0.01
Maximum	0.24
> 0.20	1.00

Table D.5. Summary of INFIT Mean Square Statistics: Regents Examination in ELA—August 2022

N	Mean	SD	Min.	Max.	[0.7, 1.3]
26	1.00	0.06	0.88	1.11	[26/26]

Table D.6. Reliabilities and Standard Errors of Measurement: Regents Examination in ELA—August 2022

Coefficient Alpha	SEM
0.79	4.44

Figure D.4. Conditional Standard Error Plot: Regents Examination in ELA—August 2022

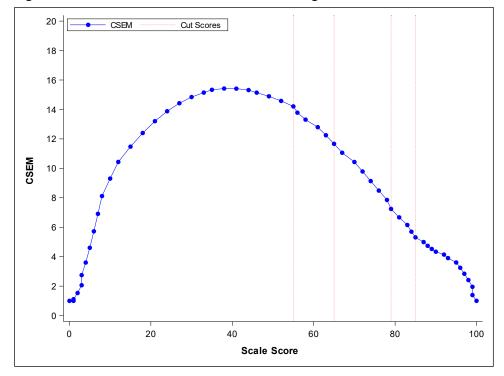


Table D.7. Decision Consistency and Accuracy Results: Regents Examination in ELA—August 2022

Statistic	1/2	2/3	3/4	4/5
Consistency	0.80	0.82	0.90	0.94
Accuracy	0.85	0.87	0.93	0.95

Table D.8. Group Means: Regents Examination in ELA—August 2022

Demographics	#Students	Mean Scale Score	SD Scale Score
All Students	9,301	51.84	22.68
Race/Ethnicity			
American Indian/Alaska Native	80	51.78	22.68
Asian/Native Hawaiian/Other Pacific Islander	667	54.59	25.84
Black/African American	2,496	50.18	21.17
Hispanic/Latino	3,588	48.27	22.40
Multiracial	183	58.42	23.19
White	2,274	58.03	22.21
English Language/Multilingual Learner			
No	7,547	56.10	21.19
Yes	1,754	33.51	19.55
Economically Disadvantaged			
No	2,450	57.52	23.61
Yes	6,851	49.81	21.99
Gender			
Female	3,887	55.02	22.30
Male	5,397	49.57	22.67
Nonbinary	4	64.00	21.09
Student with a Disability			
No	6,469	54.72	23.13
Yes	2,832	45.26	20.13

Note: Thirteen students are not reported in the race/ethnicity and gender groups but are reflected in "All Students."

Appendix E: Tables and Figures for January 2023 Administration

Table E.1. MC Item Analysis Summary: Regents Examination in ELA—January 2023

		•	•	J		•	
ltem	#Students	<i>p</i> -Value	SD	Point- Biserial	Point- Biserial Distractor 1	Point- Biserial Distractor 2	Point- Biserial Distractor 3
1	92,758	0.69	0.46	0.45	-0.14	-0.31	-0.22
2	92,758	0.62	0.49	0.36	-0.24	-0.22	-0.09
3	92,758	0.72	0.45	0.30	-0.07	-0.18	-0.20
4	92,758	0.58	0.49	0.37	-0.11	-0.26	-0.15
5	92,758	0.78	0.41	0.41	-0.24	-0.20	-0.22
6	92,758	0.76	0.43	0.40	-0.25	-0.20	-0.19
7	92,758	0.85	0.36	0.44	-0.19	-0.29	-0.22
8	92,758	0.71	0.45	0.32	-0.18	-0.09	-0.28
9	92,758	0.61	0.49	0.39	-0.25	-0.15	-0.20
10	92,758	0.61	0.49	0.37	-0.20	-0.22	-0.10
11	92,758	0.73	0.45	0.35	-0.14	-0.13	-0.26
12	92,758	0.73	0.45	0.32	-0.25	-0.25	-0.03
13	92,758	0.63	0.48	0.24	-0.30	-0.00	-0.14
14	92,758	0.50	0.50	0.31	-0.06	-0.18	-0.20
15	92,758	0.65	0.48	0.41	-0.22	-0.23	-0.17
16	92,758	0.58	0.49	0.27	-0.24	-0.17	-0.04
17	92,758	0.37	0.48	0.26	-0.08	-0.15	-0.12
18	92,758	0.67	0.47	0.38	-0.18	-0.16	-0.23
19	92,758	0.72	0.45	0.42	-0.19	-0.26	-0.19
20	92,758	0.68	0.47	0.51	-0.20	-0.27	-0.30
21	92,758	0.67	0.47	0.37	-0.21	-0.19	-0.13
22	92,758	0.79	0.40	0.49	-0.25	-0.23	-0.29
23	92,758	0.77	0.42	0.41	-0.25	-0.22	-0.18
24	92,758	0.79	0.41	0.44	-0.22	-0.24	-0.24

Table E.2. CR Item Analysis Summary: Regents Examination in ELA—January 2023

Item	#Students	Min. Score	Max. Score	Mean	SD	p-Value	Point- Biserial
25	92,758	0	6	3.85	1.28	0.64	0.89
26	92,758	0	4	2.32	1.17	0.58	0.78

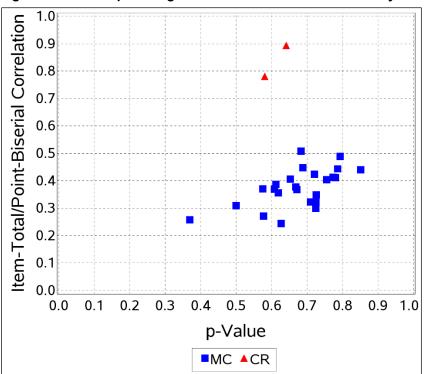


Figure E.1. Scatterplot: Regents Examination in ELA—January 2023

Table E.3. Descriptive Statistics in p-Value and Point-Biserial Correlation: Regents Examination in ELA—January 2023

Statistic	N	Mean	Min.	Q1	Median	Q3	Max.
<i>p</i> -Value	26	0.67	0.37	0.61	0.68	0.73	0.85
Point-Biserial	26	0.41	0.24	0.32	0.39	0.44	0.89

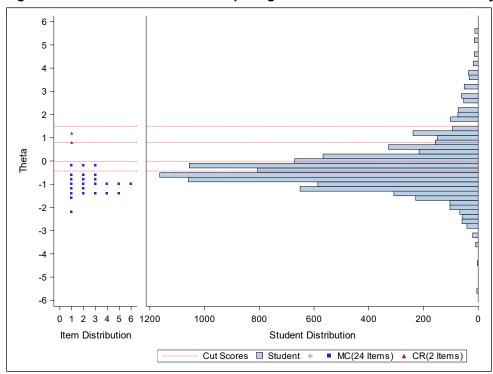


Figure E.2. Student Performance Map: Regents Examination in ELA—January 2023



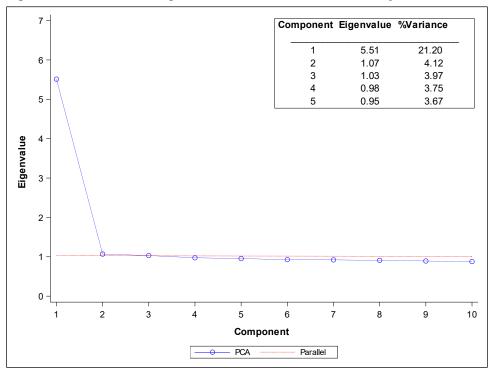


Table E.4. Summary of Item Residual Correlations: Regents Examination in ELA—January 2023

	-
Statistic	Value
N	325
Mean	-0.03
SD	0.04
Minimum	-0.13
P ₁₀	-0.10
P ₂₅	-0.05
P ₅₀	-0.03
P ₇₅	-0.02
P ₉₀	0.00
Maximum	0.19
> 0.20	0.00

Table E.5. Summary of INFIT Mean Square Statistics: Regents Examination in ELA—January 2023

N	Mean	SD	Min.	Max.	[0.7, 1.3]
26	1.00	0.09	0.85	1.18	[26/26]

Table E.6. Reliabilities and Standard Errors of Measurement: Regents Examination in ELA—January 2023

Coefficient Alpha	SEM
0.83	4.25

Figure E.4. Conditional Standard Error Plot: Regents Examination in ELA—January 2023

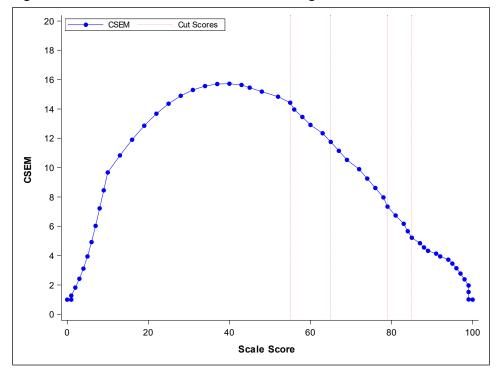


Table E.7. Decision Consistency and Accuracy Results: Regents Examination in ELA—January 2023

Statistic	1/2	2/3	3/4	4/5
Consistency	0.89	0.85	0.82	0.82
Accuracy	0.92	0.89	0.87	0.87

Table E.8. Group Means: Regents Examination in ELA—January 2023

		Mean Scale	SD Scale
Demographics	#Students	Score	Score
All Students	92,758	71.08	21.24
Race/Ethnicity			
American Indian/Alaska Native	777	68.17	19.70
Asian/Native Hawaiian/Other Pacific Islander	9,008	77.90	20.31
Black/African American	17,917	65.73	20.06
Hispanic/Latino	32,484	65.00	22.91
Multiracial	1,921	74.67	18.55
White	30,603	78.54	17.14
English Language/Multilingual Learner			
No	80,780	75.16	17.71
Yes	11,978	43.56	22.54
Economically Disadvantaged			
No	33,855	78.37	18.75
Yes	58,903	66.89	21.45
Gender			
Female	44,529	73.87	19.97
Male	48,127	68.51	22.03
Nonbinary	54	80.09	16.71
Student with a Disability			
No	74,726	73.99	20.45
Yes	18,032	59.02	20.17

Note: Forty-eight students are not reported in the race/ethnicity and gender groups but are reflected in "All Students."