Growth Model for Educator Evaluation 2018/19 Technical Report

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Introduction

New York Education Law §3012-d(5) requires Annual Professional Performance Reviews (APPRs), resulting in a single overall rating that incorporates multiple measures of effectiveness, including measures of student growth. The overall rating is determined using a matrix table that combines a Student Performance category rating and an Observation/School Visit category rating. In the 2011/12 through 2014/15 school years, State-provided growth scores were required to be used as at least one of the student performance measures for NYS teachers of Grades 4-8 English Language Arts (ELA) and mathematics, building principals covering those grade levels, and high school principals.

In December 2015, the Board of Regents established a 4-year transition period for APPRs, during which time State-provided growth scores were provided to educators for advisory purposes only. During the 2015/16 through 2018/19 school years, educators who had previously received State-provided growth scores (i.e., Grades 4-8 ELA and mathematics teachers and principals of schools that includes Grades 4-8 or all Grades 9-12) received two sets of scores and ratings: original scores and ratings and transition scores and ratings. The State-provided growth score was excluded from the scores and ratings used to calculate the transition score and rating. Only the transition score and rating were permitted to be used for purposes of employment decisions, including tenure determinations and for purposes of proceedings under Education Law §3020-a and §3020-b and teacher and principal improvement plans and the individual's employment record.

On April 12, 2019, Education Law §3012-d was amended and eliminated the original requirement to use State-provided growth scores for NYS teachers of grades 4-8 ELA/Math, building principals covering those grade levels, and high school principals. The amended law now requires that all teachers and principals have a student learning objective (SLO) with a minimum growth target representing at least one year of expected student growth. For more information about New York State educator evaluations, please see the Department's website.

This document describes the models used to measure student growth in New York State for the 2018/19 school year. In 2018/19, growth models were implemented in Grades 4-8 ELA and mathematics and for principals serving students in all of Grades 9-12. All models are based on assessing each student's change in performance between 2017/18 and prior years and 2018/19 on State assessments compared with students having similar characteristics.

The Regents Task Force on Teacher and Principal Effectiveness, made up of representatives from key stakeholder groups, including educators, educator unions, educator professional organizations, and other interested parties, gave input into the development of APPR regulations and the design of the current State-provided growth scores. In addition, a technical

advisory committee reviewed the technical accuracy and utility of the statistical methodology used to calculate scores. A list of the technical advisory committee members is provided in Appendix B. Revisions to the State-provided growth model will be considered during the 2019/20 school year.

Content and Organization of this Report

The results presented in this report are based on 2018/19 and prior school years' data, with some comparisons to prior-year results. For the technical report describing models and full results for 2017/18 and 2016/17, visit the Archived Resources of NYSED Growth Measures Toolkits website. For the technical reports describing models and full results from 2015/16 and earlier, visit the Engage NY website.

This technical report contains four main sections:

- **Data.** Description of what student growth measures, the data used to implement the student growth model, including data processing rules and relevant issues that arose during processing
- **Model.** Description of the statistical model
- **Reporting.** Description of reporting metrics and computation of effectiveness scores
- **Results.** Overview of key model results aimed at providing information on model quality and characteristics

Data

To measure student growth and attribute that growth to educators, at least two sources of data are required: student test scores that can be observed across time and information describing how students are linked to schools, teachers, and courses (i.e., identifying which teachers teach which students for which tested subjects and which school[s] those students attended). In addition, New York State models also use other information about students and schools, such as student demographics.

The following sections describe the data used for model estimation in New York in more detail, including some of the issues and challenges that arose and how they were handled.

Test Scores

New York's student growth models drew on test score data from statewide testing programs in Grades 3-8 in ELA and mathematics for the growth models for teachers, schools, and principals of students in Grades 4-8 and on Regents Exam scores for schools and principals of students in Grades 9-12. In Grades 4-8, models are estimated separately by grade and subject using scores from each grade (e.g., Grade 5 mathematics) as the outcome, with predictors as described in the following section. Scores from the State's test of English language proficiency (New York State English as a Second Language Achievement Test [NYSESLAT]) also are used as predictors in the growth models. These data are described further in the section that follows on English language learner (ELL) variables.

State Tests in ELA and Mathematics (Grades 3–8)

The New York State tests at the elementary and middle school grade levels measure a range of knowledge and skills in mathematics and ELA. State tests in ELA and mathematics for Grades 3-8 are given in the spring. The 2018/19 school year was the seventh school year that the State tests were designed to measure the Common Core State Standards. In 2017/18, the Department conducted a standards review process because the Grades 3-8 ELA and mathematics assessments were administered over the course of two days rather than over the course of three days as in previous years. Due to the State's new two-session test design and performance standards, the 2018 and 2019 Grades 3-8 ELA and mathematics results are not directly comparable to with prior-year results. While test scores cannot be compared to prior year scores, growth results may still be computed.

In 2014/15, across all grades and subjects, approximately 200,000 fewer students participated in State tests in mathematics and ELA than in previous years. This lower test taking rate continued in 2015/16, 2016/17, 2017/18, and 2018/19. For more information about how lower participation relates to growth model results, view the [2014/15 growth model technical report on the EngageNY website](#).

The New York Grades 4-8 growth model uses prior test scores in mathematics and ELA to predict test scores in those subjects. For example, we predicted the mathematics scores based on students' mathematics scores in up to the prior three years (depending on the grade) and the prior year ELA score. In addition, the other subject's scores are used because they reflect the general achievement of students prior to the outcome year (e.g., ELA scores are used in mathematics models and vice versa).

Specifically, New York's Grades 4-8 growth model includes three prior test scores in the same subject area and one prior test score in the other subject. If the immediate prior-year test score in the same subject was missing from the immediate prior grade, the student was not included in the growth measure for that subject. Two examples of how students would not have growth scores computed for them are:

1. Students without a prior-year test score (e.g., a 6th grade student with a valid 6th grade ELA test score in 2018/19 did not have a valid ELA test score in 2017/18); or
2. Students with a prior-year test score for the same grade as the current year test score (e.g., a 6th grade student with a valid 6th grade ELA test score in 2018/19 had a 6th grade ELA test score in 2017/18).

For the other prior scores, missing data indicators were used. These missing indicator variables allow the model to include students who do not have the maximum possible test history and mean that the model results measure outcomes for students with and without the maximum possible assessment history. This approach was taken to include as many students as possible. For the 2018/19 analyses, data from 2018/19 were used as outcomes, with prior achievement predictors coming from as many as three years prior (going back to 2015/16). The specific tests used as predictors vary by grade and subject and are as follows:

- Grade 4 ELA and mathematics models used scores from Grade 3 in ELA and mathematics. Students were **NOT** included if they lacked Grade 3 scores from the immediate prior year in the same subject.
- Grade 5 ELA and mathematics models used scores from Grades 3 and 4 in ELA and mathematics. Students were **NOT** included if they lacked Grade 4 scores from the immediate prior year in the same subject.
- Grades 6-8 ELA and mathematics models used scores from Grades 3-7 in ELA and mathematics. Students were **NOT** included if they lacked the immediate prior-year score in the same subject (e.g., Grade 6 students must have had a Grade 5 score in the same subject from 2017/18).

In addition to test scores, the New York Grades 4-8 growth model also used the conditional standard errors of measurement of those test scores. All assessments contain some amount of

measurement error, and the New York Grades 4-8 growth model accounts for this error (as described in more detail in the Model section of this report). Conditional standard errors were obtained from published technical reports for the assessments' prior-year test scores, and the State's test vendor provided a similar table for the 2018/19 test scores.

Regents Exams

One growth measure for Grades 9-12 schools and principals is the calculation of a mean growth percentile (MGP) based on student growth on the Algebra 1 or the ELA Regents Exam, compared with those of similar students. Regents Exams in these subjects are the most commonly taken examinations in high school.

Because Regents Exams are offered multiple times each year and students take Regents Exams at different points in their schooling, in 2018/19, the Grades 9-12 New York MGP model included students and test scores using the following rules:

- Students who take the Algebra or ELA Regents Exam prior to high school are **NOT** included in the MGP of a school or principal of Grades 9-12.
- Regents Exam scores from the following administrations were counted: August of the prior year (except for Grade 9 students) and January and June of the current year.
- Student scores were used until the students passed. (Scores are not included after students pass because we do not want to incentivize additional, unnecessary test taking.)
- If a student took a Regents Exam more than once during the school year, the higher test score was used.
- Students were included for up to 8 years after first entering Grade 9, to give credit to schools and principals that keep students beyond 4 years in high school to complete graduation requirements.

Another growth measure used for Grades 9-12 schools and principals is the Comparative Growth in Regents Exams Passed model (GRE model). Because a major graduation requirement is for students to pass four required Regents Exams and one additional Regents Exam or an approved alternative (more for advanced Regents diplomas), this measure compares how much progress a school's students are making from one year to the next toward passing up to eight Regents Exams (the four required Regents Exams, plus a second social studies examination and up to three more examinations).[1] A school's or principal's score on this measure reflects whether or not students exceeded the average number of Regents Exams passed each year by

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[1] Prior to the 2014/15 school year, students were required to pass five specific Regents Exams. Section 100.5(d) of the Commissioner's regulations was amended to allow multiple pathways to graduation.

similar students statewide. Two reasons for not including students in a Grades 9-12 school's GRE measure include a student lacking Grade 7 or 8 State test scores or a student having already passed the maximum number of Regents Exams used in this measure.

As noted, Regents Exams are offered multiple times each year, and students take Regents Exams at different points in their schooling. In 2018/19, the GRE model included students and test scores using the following rules:

- Regents Exam scores from the following administrations were counted: August of prior year (2018) and January and June of current year (2019).
- Only count the first time a student passes a specific Regents Examination.
- Four required Regents Exams, plus a second social studies examination, and no more than three additional examinations, were counted. The scores for students who passed more than eight Regents Exams were NOT included in a school or principal's results.
- Students must have had a valid prior score from Grade 7 or 8 ELA or mathematics.
- The State's modified passing score rules for students with disabilities were used to determine passing for these students.[2]
- All students who met the minimum enrollment requirement (i.e., students who were enrolled on BEDS day and at the beginning of the June Regents administration) were included in determining a school's or principal's score whether or not they took a Regents Exam during the year.
- Students were included for up to 8 years after first entering Grade 9, to acknowledge schools and principals that keep students beyond four years in high school to complete graduation requirements.
- Students who dropped out were counted in the school from which they dropped out until they would have reached their fourth year since entering Grade 9 or enrolled at another school, starting with those who dropped out in the 2015/16 school year.

Demographics

The results of growth models are used to measure the effects of educators on student learning gains, taking into account a student's prior achievement; however, some factors beyond an educator's control may impact student learning gains. For example, different learning

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[2] Modified passing score rules for students with disabilities state that students may pass the four required Regents exams with a score of 55-64. Students may also use a score of 65 or higher on one Regents Exam to compensate for a score of 45-54 on a Regents Exam other than ELA and math, unless a score of 65 or higher is to compensate for a score of 45-54 on a second math Regents exam. Students with disabilities may also graduate without passing Regents Exams based on their Superintendent's determination that they have met the academic requirements necessary to earn a Local Diploma. For the GRE model, these students do not pass any Regents exams.

trajectories often are statistically related to students living in poverty, beyond what would be expected based only on the student's prior achievement.

For all growth measures used in New York State for educator evaluation, a student's academic history and other defined characteristics are controlled for to compare similar students in the state[3] – that is, in computing student-level growth, New York's growth models always assess a student's progress relative to similar students. The rules of the Board of Regents provide that three specific types of characteristics (ELL status, students with disabilities status, and poverty status) be included in the growth models that produce scores used for educator evaluation.

Both student and course or school-level characteristics are included in growth measures used for educator evaluation for 2018/19. For instance, we account for whether a student is an English language learner, and we also account for the percentage of ELL students in a class or course (in Grades 4-8) or school (in Grades 9-12). This type of class- or school-level factor is intended to take peer effects into account, acknowledging that a student may have a different growth trajectory in a classroom, course, or school with many ELL students compared with a classroom, course, or school with few ELL students. Table 1 provides a complete list of the factors included in 2018/19. Additional descriptions of these variables follow Table 1.

Factors are the same for growth measures for teachers, schools, and principals of students in Grades 4-8 as for schools and principals of Grades 9-12, with a few additions or changes for the high school context (e.g., Grades 9-12 models also account for the number of Regents Exams a student had already passed). The New York State Education Department (NYSED or "the Department") reports unadjusted growth scores that include only prior achievement as predictor variables and adjusted growth scores including the list of approved predictor variables shown in Table 1. Unadjusted scores are reported for informational purposes to educators and are used for school accountability in Grades 4-8. In this report, results are shown for the adjusted model and the terms SGP (student growth percentile) and MGP (mean growth percentile) refer to adjusted versions of the measures (those that include all predictor variables) unless specifically identified as unadjusted.

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[3] This comparison is done through a regression modeling approach; see the Model section of this report for more details.

Table 1. Variables Included in the Adjusted Model[4]

Variable[a]	Grades 4-8		Grades 9-12		
	ELA	Mathematics	Regents ELA	Regents Algebra 1	Comparative Growth in Regents Exams Passed
Academic History Variables					
Prior-year ELA scale score (student level)[b]	✓	✓	✓	✓	✓
Two-year-prior ELA scale score if available (student-level)[b]	✓		✓	✓	✓
Three-year-prior ELA scale score if available (student level)[b]	✓				
Prior-year mathematics scale score (student level)[b]	✓	✓	✓	✓	✓
Two-year-prior mathematics scale score if available (student level)[b]		✓	✓	✓	✓
Three-year-prior mathematics scale score if available (student level)[b]		✓			
Retained in grade (student level)	✓	✓			
Mean prior score (aggregate level)[b, c]	✓	✓	✓	✓	✓
Range around mean prior score (aggregate level)[b]	✓	✓			
New to school in nonarticulation year (student level)[d]	✓	✓	✓	✓	✓
Number of years since entering ninth grade (student level)[e]			✓	✓	See note e
Count of prior required Regents Exams passed (student level)			✓	✓	✓
Students with Disabilities (SWD) Variables					
SWD status (student level)	✓	✓	✓	✓	✓
SWD in the general education classroom less than 40% of the time (student level)	✓	✓	✓	✓	✓
Percentage of SWD (aggregate level)[c]	✓	✓	✓	✓	✓
English Language Learner (ELL) Variables					

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[4] Additional detail on the variables included in the adjusted model are in Appendix A.

Variable[a]	Grades 4-8		Grades 9-12		
	ELA	Mathematics	Regents ELA	Regents Algebra 1	Comparative Growth in Regents Exams Passed
ELL status (student level)	✓	✓	✓	✓	✓
Percentage of ELLs (aggregate level)[c]	✓	✓	✓	✓	✓
NYSESLAT scores (student level)[f]	✓	✓	✓	✓	✓
Economically Disadvantaged Variables (ED)					
ED status (student level)	✓	✓	✓	✓	✓
Percentage of ED (aggregate level)[c]	✓	✓	✓	✓	✓

a: Table 1 does not display missing variable indicators. Also, for Grades 9—12 models, prior scores are measured relative to the start of high school (Grade 9). Therefore, "prior-year" means Grade 8 and "two-year-prior" means Grade 7.

b: For Grades 9-12 models, separate predictor variables are included for Common Core-aligned Grades 3–8 State assessments (2017/18), previous Common Core-aligned Grades 3-8 assessments (2012/13 through 2016/17), and previous versions of State assessments (2011/12 and earlier).

c: Aggregate-level variables are computed at the class/course level for Grades 4-8 and at the school level for Grades 9-12.

d: For Grades 9-12 models, the articulation year is Grade 9. Students entering a school that serves Grades 9-12 in a year other than Grade 9 are considered "new to school."

e: GRE models are estimated separately by cohort (based on number of years since entering Grade 9) for five cohorts (1, 2, 3, 4, and 5+ years after Grade 9 entry).

f: Only scores from the Grade 7/8 form of the NYSESLAT are used in the Grades 9-12 models. Separate predictor variables are included where possible for NYSESLAT scale scores from different years because the scales have changed across time (in 2011/12 and earlier, two separate scale scores for Listening/Speaking and Reading/Writing were used, and different scales were used in subsequent years also).

Note: See Appendix I for a complete list of predictor variables by grade and subject (including missing variable indicators) with model coefficients.

Attribution Data and Weighting

Student-level growth scores are attributed to educators based on records of educational links between educators and students. Several different data sources and procedures are used to link students to teachers and principals of Grades 4-8 and 9-12 and determine the weighting of each student's score for teachers, as described in the sections that follow.

Attributing Students to Teachers of Grades 4-8

A critical element of growth analyses is the accurate identification of the courses students are taking in which they learn the content and skills covered on the tests used to measure their learning. Another critical element is identifying who is teaching those courses.

A first step is to identify which courses are considered "relevant"—that is, courses in which instruction is provided that is aligned to the test being used to measure student growth. New York has developed a common set of course codes across the State, and these were used to identify courses as relevant for analysis. Appendix D provides a list of the item descriptions (grade and subject of relevant courses) used in analysis.

Students enrolled in relevant courses were attributed to the teacher(s) who was identified as a teacher of record for that course. Teachers' scores may reflect multiple classrooms of students in the same content area. For example, a Grade 7 mathematics teacher might provide instruction for several sections of Grade 7 mathematics.

Students who were enrolled for less than 60% of a course's duration $\left(\frac{Enrollment\ Duration}{Course\ Duration} < 0.60\right)$ were not included in a teacher's MGP. Students with course enrollment of 60% or more were included in a teacher's MGP, and their SGPs were weighted based on the percentage of time the students were enrolled in and attended the course $\left(\frac{Enrollment\ Duration}{Course\ Duration} \times \frac{Attendance\ Duration}{Enrollment\ Duration}\right)$.[5] SGPs for students who were in a teacher's course for longer periods of time and who attended the class/course more regularly counted more heavily in a teacher's MGP than those who were enrolled and attended for less time.

A teacher received a single HEDI (Highly Effective, Effective, Developing, Ineffective) rating for each district in which they had a sufficient number of student scores (i.e., teachers who may work across schools within a district received one rating). For this purpose, New York City is treated as a single district.

Table 2 shows the attribution of students with at least two years of valid same-subject test results. Attribution means that a student is linked to that teacher and included in the calculation of that teacher's MGP. Note that students can have test scores in both ELA and mathematics, so the count of students with valid test data does not represent unique students, but rather student test scores. Note also that the attribution rate is not expected to be 100% because students may move within and across schools and teacher assignments also may change. Appendix C provides an overview of data processing for Grades 4-8 models, and Appendix G provides an overview of processing for Grades 9-12 models.

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[5] Education Analytics calculated the course duration variable directly from the teacher-student data linkage file.

Table 2. Grades 4-8 Teacher-Student Attribution Rates

Grade	Valid Student Records	Valid Student Records Attributed to at Least One Teacher	Attribution Rate
4	318,410	289,033	91%
5	314,756	284,662	90%
6	302,243	270,771	90%
7	286,657	257,070	90%
8	230,718	197,447	86%
Total	1,452,784	1,298,983	89%

Note: Student records are considered valid for the purposes of growth modeling when there are at least two consecutive years of valid assessment scores. Students can have as many as two valid records per year, one for ELA and one for mathematics.

Overall, in 2018/19, 89% of the 1,452,784 valid student records were linked to at least one teacher. In 2017/18, 89% of the 1,405,900 valid student records were attributed to teachers.[6]

Attributing Students to Schools of Grades 4-8

Students were attributed to schools and districts based on a continuous enrollment indicator found in the assessment score files. This variable describes whether a student was enrolled at the start and end of the year in a school or district (on BEDS day and at the beginning of the State test administration in the spring). Students who met this criterion were included in school-level MGPs. The same continuous enrollment indicator is used for institutional accountability purposes. Note that student results were not weighted by attendance in determining a school MGP and growth score. The policy rationale for not using attendance weighting for schools (although it is used for teachers) is that school leaders may have more influence on student attendance, and on the integrity of attendance data, than do teachers.

Because of the difference in data sources and indicators used to attribute students to teachers and schools, students can be linked to a school but not a teacher and, in rare cases, vice versa. Table 3 shows attribution rates for schools.

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[6] Details can be found in the *2017/18 Growth Model for Educator Evaluation Technical Report*, which is available on the NYSED growth measures toolkits page.

Table 3. Grades 4-8 School-Student Attribution Rates

Grade	Valid Student Records	Valid Student Records Attributed to at Least One School	Attribution Rate
4	318,410	309,887	97%
5	314,756	306,914	98%
6	302,243	294,865	98%
7	286,657	280,325	98%
8	230,718	225,555	98%
Total	1,452,784	1,417,546	98%

Note: Student records are considered valid for the purposes of growth modeling when there are at least two consecutive years of valid assessment scores. Students can have as many as two valid records per year, one for ELA and one for mathematics.

The attribution rate at the school level in 2018/19 (98%) was slightly higher than the attribution rate in 2017/18 (97%). More student records overall were attributed to schools in 2018/19 than in 2017/18.[7]

Attributing Students to Principals of Grades 4-8

New York's growth models make use of district-reported staff assignment data in growth model reporting. The use of this staff assignment data allows results to be reported for individual principals for the grade levels to which they are assigned or across multiple schools for which a principal was responsible. Students were attributed to principals based on the school-level continuous enrollment indicator found in the assessment score files (see previous section for more information on this variable). Students at each grade level in a school who met the continuous enrollment requirement were attributed to a principal if that principal was assigned to that grade level in the staff assignment file. As with schools, note that student results were not weighted by attendance in determining a principal MGP.

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[7] Details can be found in the *2017/18 Growth Model for Educator Evaluation Technical Report*, which is available on the NYSED growth measures toolkits page.

Table 4. Grades 4-8 Principal-Student Attribution Rates

Grade	Valid Student Records	Valid Student Records Attributed to at Least One Principal	Attribution Rate
4	318,410	303,645	95%
5	314,756	300,252	95%
6	302,243	287,433	95%
7	286,657	273,153	95%
8	230,718	219,054	95%
Total	1,452,784	1,383,537	95%

Note: Student records are considered valid for the purposes of growth modeling when there are at least two consecutive years of valid assessment scores. Students can have as many as two valid records per year, one for ELA and one for mathematics.

The attribution rate at the principal level in 2018/19 (95%) was one percentage point lower than in 2017/18 (96%). As with teachers and schools, more student records overall were attributed to principals 2018/19 than in 2017/18.[8]

Attributing Students to Schools and Principals of Grades 9-12

Students in Grades 9-12 were linked to schools and principals based on a continuous enrollment indicator created from a school enrollment file. Using school entry and exit dates, the indicator describes whether or not a student was enrolled at the start and end of the year in a school or district (on BEDS day and at the beginning of June Regents Exam administration). Students who were enrolled at these two points in time in a given school were attributed to that school and to any principals assigned to all of Grades 9-12 at that school (based on the staff assignment file). These rules are similar to those used for schools and principals of Grades 4-8, although the sources of data used to implement the rule are somewhat different.[9] Note also that scores are reported only for schools serving all of Grades 9-12.

Table 5 shows school attribution rates for the MGP model, and Table 6 shows attribution rates for principals for both the MGP and the GRE models. For the MGP models (based on ELA and Algebra Regents Exams), students are included in the model if they had a current year score, had at least one valid Grade 7 or 8 assessment in the same subject (mathematics for Algebra and ELA for ELA), and had not passed that Regents Exam in a prior year. For the GRE model, students are included in the model when they had at least one valid Grade 7 or 8 assessment in

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[8] Details can be found in the *2017/18 Growth Model for Educator Evaluation Technical Report*, which is available on the NYSED Growth Measures Toolkits page.

[9] For Grades 4-8, NYSED provided an indicator (the school_in flag) of student enrollment/attribution for schools. For Grades 9-12, Education Analytics calculated a similar variable directly from the enrollment file.

either subject, were enrolled in Grades 9-12 for 1-8 years, had not passed eight Regents Exams as of the end of the prior year, and were attributed to at least one school.[10]

Table 5. Grades 9-12 School-Student Attribution Rates

Model	Students Included in Analysis	Students Included in Analysis Attributed to Schools	Attribution Rate
ELA	169,560	155,209	92%
Algebra 1	139,692	127,157	91%
GRE	519,094	519,094	100%

Table 6. Grades 9-12 Principal-Student Attribution Rates

Model	Students Included in Analysis	Students Included in Analysis Attributed to Principals	Attribution Rate
ELA	169,560	150,165	89%
Algebra 1	139,692	122,193	87%
GRE	519,094	501,818	97%

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[10] Schools need to meet the following criteria: not be a transfer or portfolio or non-public school.

Model

Two different types of models were used to produce growth measures in New York State. The first is the MGP model, which was implemented for Grades 4-8 using State assessments in ELA and mathematics and for Grades 9-12 using Regents Exams in ELA and Algebra. To produce scores describing how well students are progressing toward passing Regents Exams, a second model was implemented for Grades 9-12. This model is referred to as the GRE model. These two models are described in detail in the sections that follow.

Mean Growth Percentile Model

This section describes the statistical model used to measure student growth in New York between two points in time on a single subject of a State assessment. The section begins with a description of the statistical model used to form the comparison point against which students are measured—based on similar students—and then describes how SGPs are derived from the comparison point. In addition, this section describes how MGPs and all variance estimates are produced.

At the core of the New York State growth model is the production of an SGP. This statistic characterizes the student's current year score relative to other students with similar measured characteristics and prior test score histories. For example, an SGP equal to 75 denotes that the student's current year score is the same as or better than 75% of the students in the State with prior test score histories and other measured characteristics that are similar. It does **NOT** mean that the student's growth is better than that of 75% of all other students in the population.

One common approach to estimating SGPs is to use a quantile regression model (Betebenner, 2009). This approach models the current year score as a function of prior test scores and finds the SGP by comparing the current year score to the predicted values at various quantiles of the conditional distribution.

The methods described here do not rely on the quantile regression method for two reasons. First, the typical implementation of the quantile regression makes no correction for measurement variance in the predictor variables or the outcome variable. Ignoring the measurement variance in the predictor variables yields bias in the model coefficients (e.g., Wei and Carroll, 2009). Further complicating the issue, the measurement variance in the outcome variable also adds to the bias in a quantile regression (Hausman, 2001), an issue that does not occur with linear regression.

The model implemented for New York State is a linear regression model designed to account for measurement variance in the predictor variables, as well as the outcome variable, to yield unbiased estimates of the model coefficients. Subsequently, these model coefficients are used

to form a predicted score, which is ultimately the basis for the SGP. Because the prediction is based on the observed score, it is necessary to account for measurement variance in the prediction as well. Hence, the model accounts for measurement variance in two steps: first in the model estimation and second in forming the prediction. The next section describes this model in detail.

Covariate Adjustment Model

The statistical model implemented as the MGP model is typically referred to as a *covariate adjustment model* (McCaffrey, Lockwood, Koretz, and Hamilton, 2004), as the current year observed score is conditioned on prior levels of student achievement as well as other possible covariates.

In its most general form, the model can be represented as follows:

$$y_{ti} = X_i\beta + \sum_{r=1}^{L} Y_{t-r,i}\gamma_{t-r} + e_i$$

where y_{ti} is the observed score at time t for student i, X_i is the model matrix for the student- and school-level demographic variables, β is a vector of coefficients capturing the effect of any demographics included in the model, $Y_{t-r,i}$ is the observed lag score at time $t - r(r \in \{1,2,\dots,L\})$, and γ is the coefficient vector capturing the effects of lagged scores.

Accounting for Measurement Variance in the Predictor Variables

All test scores are measured with variance, and the magnitude of the variance varies across the range of test scores. The standard errors (variances) of measurement are referred to as *conditional standard errors of measurement* (CSEMs) because the variance of a score is heteroscedastic and depends on the score itself. Figure 1 shows a sample from the Grade 8 ELA test in New York.

Figure 1. Conditional Standard Error of Measurement Plot (Grade 8 ELA, 2018/19)
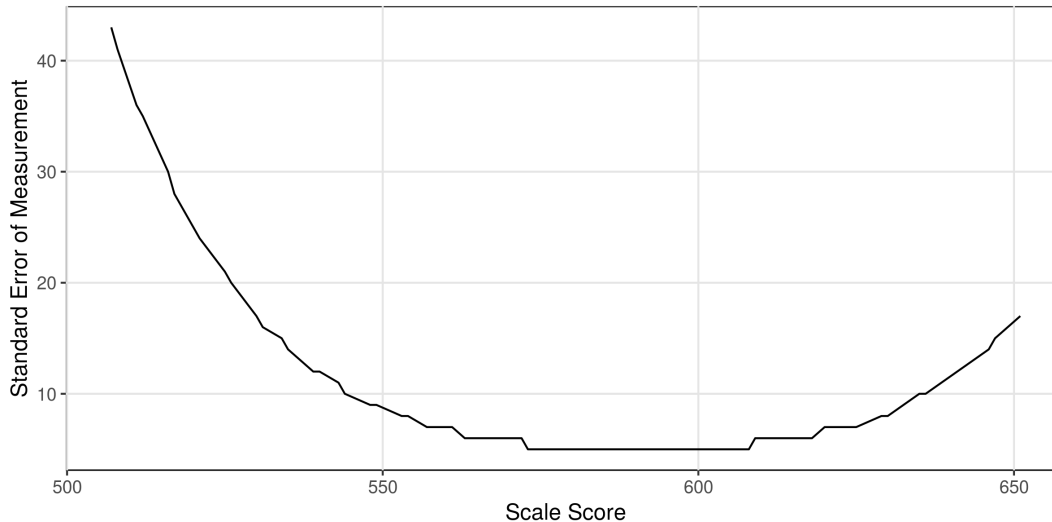
Treating the observed scores as if they were the true scores introduces a bias in the regression, and this bias cannot be ignored within the context of a high-stakes accountability system (Greene, 2003). In test theory, the observed score is described as the sum of a true score plus an independent variance component, $X = X^* + U$, where U is a matrix of unobserved disturbances with the same dimensions as X.

Our estimator accounting for the error in the predictor variables is derived in a manner similar to that of Goldstein (1995). The estimator and a complete theoretical derivation are provided in Appendix E.

Specification for MGP Model for Grades 4-8 and Grades 9-12

The preceding section provides details on the general modeling approach and specifically how measurement variance is accounted for in the model. The exact specification for the New York Grades 4—8 model in 2018/19 is described as follows:

$$y_{gi} = \mu + \sum_{l=1}^{K} \beta_l y_{g-r,i} + \sum_{s=1}^{M} \tau_s m_{si} + \sum_{q=1}^{J} \gamma_q x_{qi} + \varepsilon_i$$

where y_{gi} is the current year test scale score for student i in grade g, μ is the intercept, β_l is the set of coefficients associated with the three prior test scores, τ_s is the set of coefficients associated with the missing variable indicators, γ_q is the set of coefficients associated with the student-level measured characteristics (which are described in Appendix A), and ε_i is the student residual. For the MGP model used for Grades 9-12, scale scores from assessments taken before Grade 9 were used as predictors (not prior Regents Exam scores themselves, although the number of Regents Exams passed prior to the outcome year was used as a

predictor). The form of the model is the same as shown previously, where y_{gi} is the Regents Exam scale score for student i in subject s, μ is the intercept, β_l is the set of coefficients associated with the Grades 7 and 8 test scores and is estimated with an error-in-variables (EiV) approach[11], τ_s is the set of coefficients associated with the missing variable indicators, γ_q is the set of coefficients associated with the student-level measured characteristics (which are described in Appendix A), and ε_i is the student residual.

MGP models were implemented separately for each grade and subject. Two models were estimated. The "adjusted" model is the model as described previously. The "unadjusted" model is a special case of the adjusted model that does not contain any variables (such as the ELL status) except prior test scores and missing indicators for the two- and three-year-prior scores. In all models, special procedures are used to adjust standard errors of measurement. These procedures are described in Appendix F.

Student Growth Percentiles

The previously described regression models yield unbiased estimates of the coefficients by accounting for the measurement error in observed scores. The resulting estimates are then used to form a student-level student growth percentile (SGP) statistic. For purposes of the growth model, a predicted value and its variance for each student are required to compute the SGPs as follows:

$$SGP_i = \phi\left(\frac{y_i - \hat{y}_i}{\sqrt{\sigma_{yf,i}^2}}\right)$$

where y_i is the observed value of the outcome variable and $\hat{y}_i = w'\hat{\delta}$ where w' is the ith row of the model matrix W, and the notation $\sigma_{yf,i}^2$ is used to mean the variance of the predicted value of y for the ith student.

Here, the regression is of the form

$$y = W\delta + \epsilon$$

where

$$\epsilon \sim N(0, \sigma_\epsilon^2)$$

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[11] EiV regression is a method to estimate consistent coefficients when variables are measured with error, such as assessment scores. EiV regression allows us to acknowledge and account for that error when estimating value-added for teachers.

For this case, the classic variance of a predictor is

$$\sigma_{yf,i}^2 = [1 + w_i'(W'W)^{-1}w_i]\hat{\sigma}_e^2$$

where $\hat{\sigma}_e^2$ is the variance of the predictor. However, in this case, we make two refinements to acknowledge the effect of measurement error on the residual variance. The first is to use the actual variance on y_i, called σ_{yi}^2, rather than the population variance on y_i, called $\bar{\sigma}_{yi}^2$, which is already included in $\hat{\sigma}_e^2$. This is done by subtracting the population variance and adding back the individual variance. Thus, the variance on the predictor becomes

$$\sigma_{yf,i}^2 = [1 + w_i'(W'W)^{-1}w_i][\sigma_e^2 - \bar{\sigma}_{yi}^2] + \sigma_{yi}^2$$

The second refinement is to replace the population variance in w_i, called $\bar{\Sigma}$, with the individual variance in w_i, called Σ_i. This replacement is done in the same way as with the variance in y_i, so the variance estimate is now

$$\sigma_{yf,i}^2 = [1 + w_i'(W'W)^{-1}w_i][\sigma_e^2 - \bar{\sigma}_{yi}^2 - \delta'\bar{\Sigma}\delta] + \sigma_{yi}^2 + \delta'\Sigma_i\delta$$

A predicted value for each student is used to compute the SGP. However, that prediction is based on the estimates of the fixed effects that were corrected for measurement variance but based on the observed score in the vector w.

Figure 2 illustrates how the SGPs are found from the previously described approach. The illustration considers only a single predictor variable, although the concept can be generalized to multiple predictor variables, as presented earlier. For each student, we find a predicted value conditional on his or her observed prior scores and the model coefficients. To illustrate the concept, assume we find the prediction and its variance but do not account for the measurement variance in the observed scores used to form that prediction. We would form a conditional distribution around the predicted value and find the portion of the normal distribution that falls below the student's observed score. This is equivalent to

$$SGP_i = \int_{-\infty}^{yi} f(x)dx$$

with $f(x) \sim N(\hat{y}_i, \sigma_{yfi}^2)$, although this is readily accomplished using the cumulative normal distribution function, $\phi(\cdot)$.

Figure 2. Sample Growth Percentile from Model
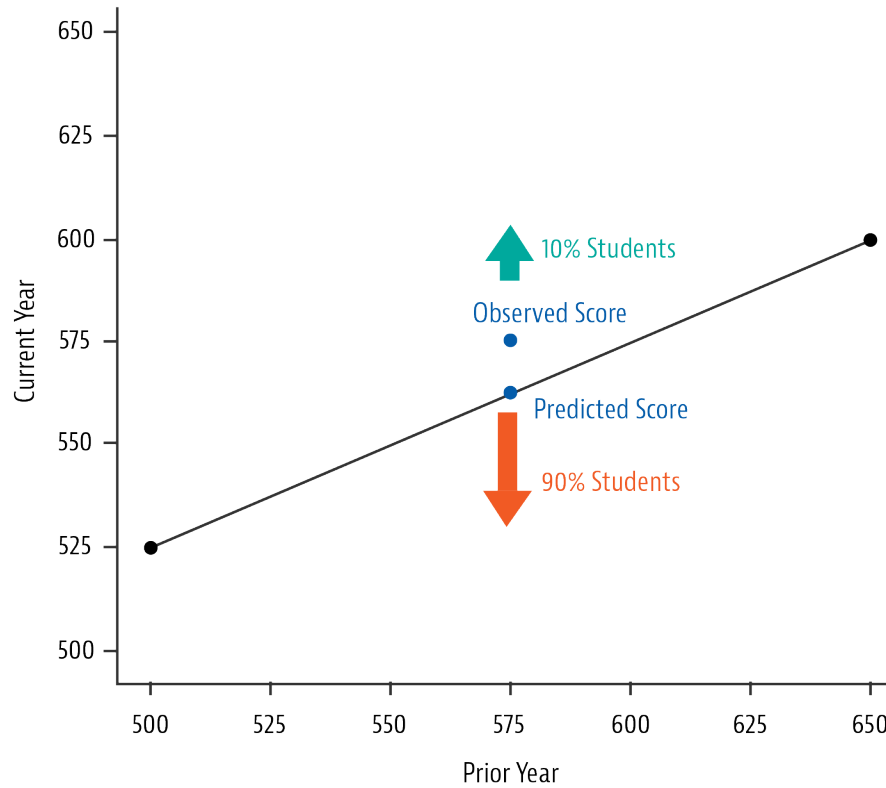
Figure 3 illustrates the same hypothetical student shown in Figure 2. Note that the observed score and predicted value are exactly the same. However, the prediction variance is larger than in Figure 2. As a result, when we integrate over the normal from $-\infty$ to y_i, the SGP is 60, not 90 as in the previous example. This difference occurs because the conditional density curve has become more spread out, reflecting less precision in the prediction.

Figure 3. Sample Growth Percentile from Model
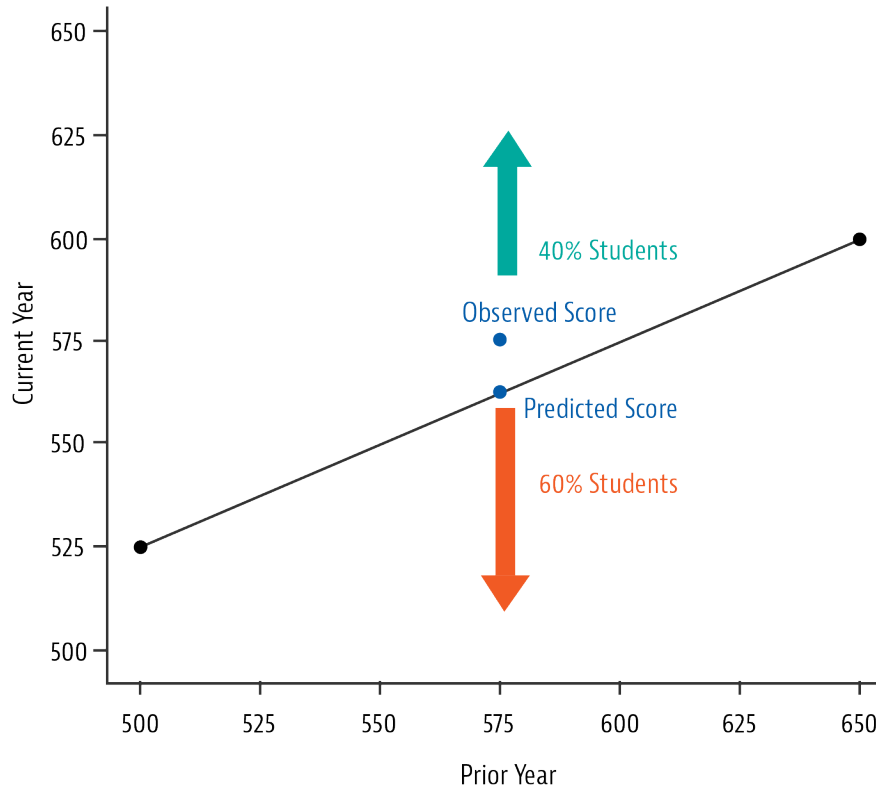
Mean Growth Percentiles

Once SGPs are estimated for each student, group-level (e.g., teacher-level) statistics can be formed that characterize the typical performance of students within a group. New York's growth model Technical Advisory Committee recommended using a mean SGP for educator scores. Hence, group-level statistics are expressed as the mean SGP within a group. This statistic is referred to as the *MGP*.

For each aggregate unit j ($j \in \{1, 2, \ldots, J\}$), such as a class or course, the statistic of interest is a summary measure of growth for students within this group. Within group j, there are $\{SGP_{j(1)}, SGP_{j(2)}, \ldots, SGP_{j(N)}\}$. That is, there is an observed SGP for each student within group j.

Then the MGP for unit j is produced as the simple mean

$$\theta_j = mean\left(SGP_{j(i)}\right)$$

for Grades 4-8 and Grades 9-12 schools and principals, and as the weighted mean

$$\theta_j = \frac{1}{\sum w_{j(i)}} \sum w_{j(i)} SGP_{j(i)}$$

for Grades 4—8 teachers, where $w_{j(i)}$ is a weight for student i in teacher j's class or course based on the student's enrollment and attendance.

As with all statistics, the MGP is an estimate, and it has a variance term. The following measures of variance are produced for the MGP.

The analytic standard error of the unweighted MGP (schools and principals) is computed within unit j as

$$se(\theta_j) = \frac{sd(SGP_{ij})}{\sqrt{N_j}}$$

and in the weighted case (teachers) as

$$se(\theta_j) = \frac{sd(SGP_{ij})}{\sqrt{\dfrac{(\sum w_s)^2}{(\sum w_s^2)}}}$$

where $sd(SGP_{ij})$ is the sample standard deviation of the SGPs in group j, and N_j is the number of students in group j.

Combining Student Growth Percentiles across Grades and Subjects
Many teachers, schools, and principals serve students from different grades and with results from different tested subjects. For evaluation purposes, there is a need to aggregate these SGPs and form summary measures.

Because the SGPs are expressed as percentiles, they are free from scale-specific inferences and can be combined. For any aggregate-level statistics to be provided (in this case, MGPs), all SGPs of relevant students are pooled and the average of the pooled SGPs is found. In the case of Grades 4–8 teachers, the average is a weighted average, as described earlier. Variances of these MGPs are found using the same methods described previously. More detail on reported scores can be found in the Reporting section.

Comparative Growth in Regents Exams Passed
For the GRE model, the outcome of interest is the number of Regents Exams that a student passes for the first time in the outcome or current year (in this case, 2018/19). Educators whose

students pass more Regents Exams in a year than similar students will have higher scores on this metric than those of other educators. For this model, Regents Exams in the four required subjects, plus a second social studies examination, and up to three additional Regents Exams (for a total possible of eight Regents Exams for each student) were counted as outcomes. Once a student had passed eight Regents Exams, he or she was excluded from the model.

Because the outcome can take on only positive integer values and is bounded by a minimum (a student can never pass fewer than zero Regents Exams in a year) and a maximum (a student can never have more than eight Regents Exams passed in a year), an ordered logit model is implemented. The model is fit separately for each cohort of students (students who entered ninth grade 1 year ago, 2 years ago, and so on) for Years 1, 2, 3, and 4. Students who entered Grade 9 more than 4 years ago are aggregated into a single fifth run.

The linear part of the model is

$$\eta_i = X_i \beta^c$$

where X includes the variables named in the definition of similar students as well as an intercept term, η is the latent variable that dictates the number of Regents Exams a student passes, β is the fitted parameters for the variables in X, the superscript c is used to indicate that the β coefficients depend on the cohort, and the subscript i is used to indicate that η and X are specific to an individual student.

From this, the logistic function and a series of cut points are used to map η to the outcome space, generating an estimated fraction of the time that zero through eight Regents Exams were passed by similar students. The fraction of similar students passing a particular number of Regents Exams is then given by

$$\Pr(\delta_i = k | X_i, \beta^c) = \frac{1}{1 + \exp\left(-\lambda_{k+1} + X_i \beta^g\right)} - \frac{1}{1 + \exp\left(-\lambda_k + X_i \beta^g\right)}$$

where δ is the number of Regents Exams passed this year, and the λ_k are fitted cut points[12] between having passed $k-1$ and k Regents Exams.

This set of nine values is then collapsed into the average number of Regents Exams similar students passed this year using

$$\hat{y}_t = \sum_{k=0}^{8} \Pr\left(\delta_i = k | X_i, \beta^c\right) \cdot min\left(8 - N_{i,yy-1}, k\right)$$

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[12] These sometimes are called intercepts.

where $\hat{y}$ is the estimated number of Regents Exams passed by similar students, and N_{yy-1} is the number of Regents Exams passed at the initiation of this school year. In the previous equation, the first term represents the probability of a similar student having passed k Regents Exams this year, and the second term often multiplies that probability by k. A min function also is included in the second term that imposes a ceiling on the number of Regents Exams passed this year, acknowledging that the total number passed this year plus the number that had been passed at the beginning of this year (N_{yy-1}) cannot exceed eight.

Finally, values of $\hat{y}$ that are larger than two are set to two because to meet a projection larger than two Regents Exams per year, students would have to complete the eight Regents Exams counted in this model on a schedule faster than eight Regents Exams in 4 years. Because NYSED did not wish to encourage unnecessary Regents Exam taking, this cap on projected Regents Exams was applied.

Using this approach, each student has an actual number of Regents Exams that he or she passed (y_i), and a number passed by similar students ($\hat{y}_i$); the latter is subtracted from the former to find a student-level GRE:

$$GRE_i = y_i - \hat{y}_i$$

A school or principal's score is then the mean GRE (or MGRE) for students attributed to that school or principal:

$$MGRE = \frac{1}{n}\sum_{i=1}^{n} GRE_i$$

The standard error is found by taking the sample standard deviation of the students GREs. Thus, the variance estimate is

$$Var(MGRE) = \frac{1}{(n-1)n}\sum_{i=1}^{n} [GRE - MGRE]^2$$

and the standard error is the square root of that quantity. Confidence intervals are formed from the variances and point estimates in the same way they were for MGPs.

Reporting

Results of the New York growth models are reported to districts in a series of data files.

Reporting for Teachers, Schools, and Principals of Grades 4-8

The main reporting metrics generated for teachers, schools, and principals of Grades 4-8 were as follows:

- **Number of Student Scores.** The number of SGPs included in an MGP.
- **Unadjusted MGP (School or Principal).** The mean of the SGPs for students attributed to the school or principal based on similar prior achievement scores only, without taking into consideration ELL, disability, economic disadvantage, or other student characteristics.
- **Unadjusted MGP (Teacher).** The weighted mean of the SGPs for students who are linked to a teacher based on similar prior achievement scores only, without taking into consideration ELL, disability, economic disadvantage, or other student characteristics. The weighted mean was calculated based on the amount of time students were enrolled in and attended a course with a teacher.
- **Adjusted MGP (School or Principal).** The mean of the SGPs for students attributed to the school or principal, based on similar prior achievement scores, *including* consideration of ELL, disability, economic disadvantage, and other student characteristics. This MGP is used to determine a school or principal's State-provided growth score and growth rating.
- **Adjusted MGP (Teacher).** Adjusted MGP is the weighted mean of the SGPs for students linked to a teacher, based on similar prior achievement scores, *including* consideration of ELL, disability, economic disadvantage, and other student characteristics. This MGP is used to determine a teacher's State-provided growth score and growth rating.
- **Lower Limit and Upper Limit.** Highest and lowest possible MGP for a 95% confidence range.
- **Growth Rating.** Growth rating describes the educator's HEDI rating on the State-provided growth subcomponent.
- **Growth Score.** A growth score of 0-20 points is assigned to each educator based on his or her overall MGP within each growth rating category using the scoring bands for implementation of Education Law §3012-d.

MGPs disaggregated by grade and subject also are provided. Districts also are provided with student roster files. These files show which students were included in a teacher's MGP along with information about each student, such as whether the student has a disability or is identified as an ELL.

Reporting for Grades 9-12

The main reporting metrics generated for schools and principals of Grades 9-12 are as follows:

- **Number of Student Scores (for MGP Measure) or Students (for GRE Measure).** These numbers refer to the SGPs included in an MGP or the number of students included in the GRE score.
- **Unadjusted Measure.** This measure is based on student growth and accounts for prior achievement scores *only,* without taking into consideration ELL, disability, economic disadvantage, or other student characteristics.
- **Adjusted Measure.** This measure is based on student growth and is adjusted for prior achievement scores and ELL, disability, economic disadvantage, and other characteristics at the student and school levels.
- **Lower Limit and Upper Limit.** Highest and lowest possible measure score for a 95% confidence range.
- **Growth Rating.** Growth rating describes the educator's performance category (HEDI) for each individual measure (MGP or GRE) and overall for Grades 9-12. The overall growth rating is used in a school or principal's evaluation on the State-provided growth subcomponent.
- **Growth Score.** A growth score of 0-20 points is assigned to each school and principal (for each MGP or GRE measure and overall) within each growth rating category using the scoring bands for implementation of Education Law §3012-d. The overall growth score is used in a principal's evaluation on the State-provided growth subcomponent.

As with Grades 4-8 measures, districts also are provided with student-level files that show which students were included growth measures, along with information about each student.

Minimum Sample Sizes for Reporting

Minimum sample size requirements for reporting MGPs and growth ratings were determined to balance statistical reliability and availability of educator growth scores. On one hand, setting no (or a low) minimum sample size will result in the greatest number of educators receiving information; on the other hand, the quality of the information they receive may be reduced. A minimum threshold of 16 student scores or 16 students for the GRE measure was implemented. Educator scores on any measure at any level based on fewer than 16 student scores (or 16 students for the GRE measure) were not reported.

After applying this rule, the fraction of teachers, schools, and principals with reported results is shown for Grades 4-8 in Table 7 and for Grades 9-12 in Table 8. The percentage of teachers, principals, and schools receiving results in 2018/19 compared to 2017/18 are summarized below:

- The percentage of teachers receiving results in 2018/19 (79%) was higher by two percentage points compared to 2017/18 (77%).
- The percentage of principals receiving 4-8 results in 2018/19 remained the same from 2017/18 (98%).
- The percentage of schools receiving 9-12 results in 2018/19 was lower by one percentage point compared to 2017/18 (90% in 2018/19 and 91% in 2017/18).
- The percentage of principals receiving 9-12 results was lower by one percentage point compared to 2017/18 (97% in 2017/18 and 96% in 2018/19).
- The percentage of schools receiving 4-8 results in 2018/19 was unchanged compared to 2017/18 (95%).[13]

Table 7. Grades 4-8 Reporting Rates

	Number With At Least One Student Attributed	Number Meeting the Minimum Sample Size Requirement	Percentage Meeting the Minimum Sample Size Requirement
Teachers	47,025	36,921	79%
Principals	3,802	3,716	98%
Schools	3,805	3,596	95%

Table 8. Grades 9-12 Reporting Rates

	Number With At Least One Student Attributed	Number Meeting the Minimum Sample Size Requirement	Percentage Meeting the Minimum Sample Size Requirement
Principals	1,381	1,326	96%
Schools	1,483	1,335	90%

Performance Categories

To determine an educator's growth rating (HEDI category) and growth points (0-20), NYSED has developed a set of general rules that describe how similar or different a score on each measure is from the State average. The general rules used to obtain growth ratings are shown in Figure 4. Specific values used to determine growth ratings are shown in Appendix H.

Within each growth rating category, points are then assigned so that educators are approximately uniformly distributed at each HEDI point value (with higher MGPs or GRE results earning more points than lower MGPs or GRE results in that category). Growth scores are

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[13] Details can be found in the *2017/18 Growth Model for Educator Evaluation Technical Report*, which is available on the NYSED Growth Measures Toolkits page.

assigned using the scoring bands for implementation of Education Law §3012-d. Additional detail about the assignment of HEDI point values also can be found in Appendix H.

Figure 4. Determining Growth Ratings
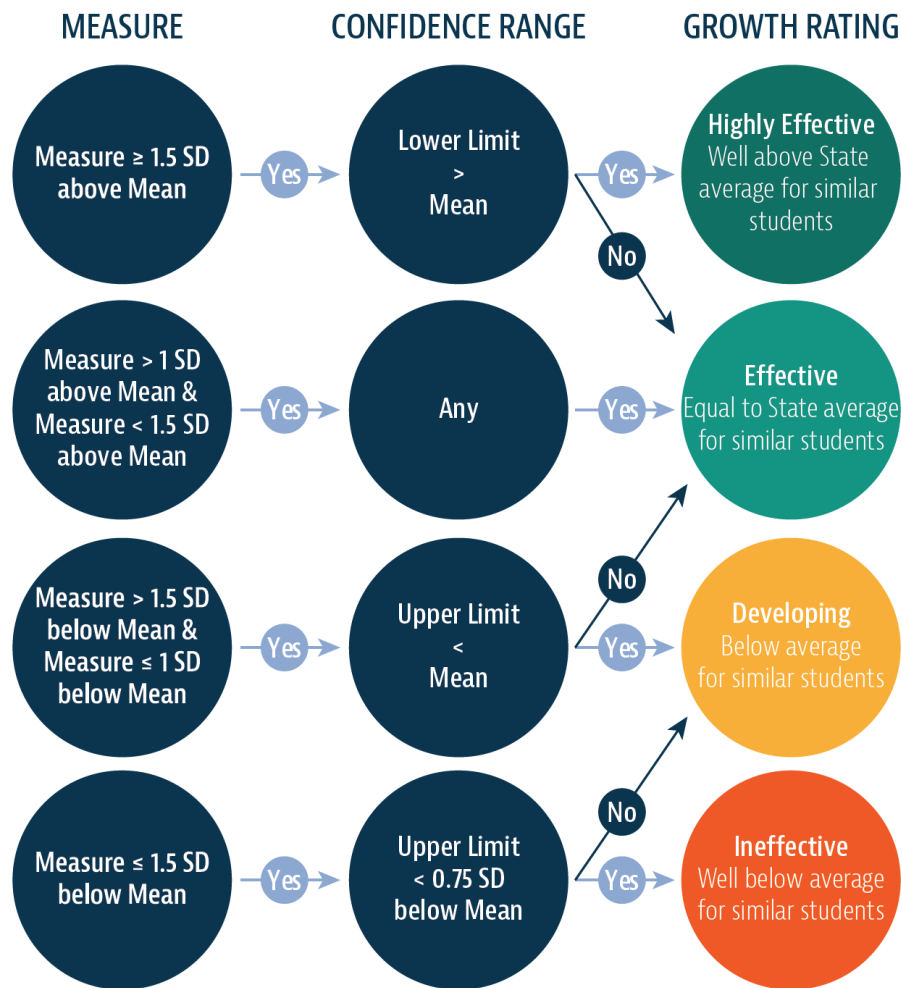
Note: SD = Standard Deviation
Values are rounded to the nearest 0.5 for MGP & 0.1 for GRE.

Note: SD = Standard Deviation
Values are rounded to the nearest 0.5 for MGP for schools and principals, to the nearest whole number for teachers, and to the 0.01 level for GRE measures for schools and principals.

For teachers, schools, and principals of Grades 4-8, the overall adjusted MGP (i.e., the MGP that combines information across all applicable grade levels and subjects) and upper and lower limit MGPs were used to determine growth ratings. To determine the growth rating for a school or principal of Grades 9-12, a growth rating and score for both types of metrics—the MGP measure and the GRE measure—is first found using the process shown in Figure 4. Growth

scores for each Grades 9-12 measure are then averaged together and weighted by the number of students in each measure to find an overall Grades 9-12 growth rating and score.

To determine a final State-provided growth subcomponent rating for schools that and principals who serve Grades 4-8 and Grades 9-12, growth ratings and scores for Grades 4-8 and Grades 9-12 are computed separately and then combined. The Grades 4-8 measure growth rating is determined using the process shown in Figure 4, and an overall Grades 9-12 growth rating and score is determined as described previously. An overall growth subcomponent rating that includes results for both Grades 4-8 and Grades 9-12 students is then computed by averaging Grades 4-8 and Grades 9-12 growth scores by the number of students in each measure and finding the final rating.

Additional details can be found in the resources for educators on the NYSED Growth Measures Toolkits page and in Appendix H.

Results

Results from Growth Models for Grades 4-8

This section provides an overview of the results of 2018/19 growth model estimation. Some comparisons to earlier year growth model results also are included. A pseudo *R*-squared statistic and summary statistics characterizing the SGPs, MGPs, and their precision provide an overview of model fit.

This section focuses on teacher-level and school-level results; additional information on principal-level results is in Appendix J.

Model Fit Statistics for Grades 4-8

The *R*-square value is a statistic commonly used to describe the goodness-of-fit for a regression model. Because the model implemented here is an EiV model, not a least squares regression, we refer to this as a *pseudo R*-square. (See page 24 for more information on the EiV model.) Table 9 presents the pseudo *R*-square values for each grade and subject, computed as the squared correlation between the fitted values and the outcome variable.

Table 9. Grades 4-8 Pseudo R-Squared Values by Grade and Subject

Subject	Grade	Unadjusted Model	Adjusted Model
ELA	4	0.59	0.63
	5	0.66	0.67
	6	0.66	0.68
	7	0.70	0.72
	8	0.69	0.71
Mathematics	4	0.66	0.68
	5	0.71	0.72
	6	0.71	0.72
	7	0.70	0.72
	8	0.63	0.64

Student Growth Percentiles for Grades 4-8

SGPs describe a student's current year score relative to those of other students in the data with similar prior academic histories and other measured characteristics. A student's SGP should not be expected to be higher or lower based on his or her prior-year score. Table 10 shows the correlation between the prior-year scale score and SGP for each grade and subject. These correlations are usually negative as a result of using the EiV approach to account for

measurement variance in the prior-year scale score; the correlation need not be zero. Squaring these values gives the percentage of variation in SGPs explained by prior-year scores for any grade and subject. Although prior-year test scores are generally good predictors of current year test scores, the prior-year test score is a poor predictor of current year SGPs. As shown in Table 10, prior-year test scores explain about 2% to 5% of the variation in Adjusted SGPs. Because SGPs are intended to allow students to show low or high growth no matter their prior performance, this result is as expected.

Table 10. Grades 4-8 Correlation Between Adjusted SGP and Prior-Year Scale Score

Grade	ELA	Mathematics
4	-0.176	-0.213
5	-0.182	-0.167
6	-0.165	-0.154
7	-0.153	-0.24
8	-0.145	-0.238

Mean Growth Percentiles for Grades 4-8

As described earlier in this report, teachers' MGPs are aggregate educator-level statistics, computed as the weighted mean of SGPs for all students associated with a teacher or as the mean for schools or principals. In this section, we provide descriptive statistics on overall or combined MGPs.

For teachers with results for students in both ELA and mathematics, the combined MGP is an average of SGPs for both subjects. For teachers who provide instruction in only one subject, their overall or combined MGP is the same as their subject-specific MGP.

Figure 5 is a histogram of the teacher MGPs in ELA and mathematics for the adjusted model (including demographics). In all grades, the results are approximately normally distributed.

Figure 5. Grades 4-8 Distribution of Teacher MGPs by Grade: ELA and Mathematics
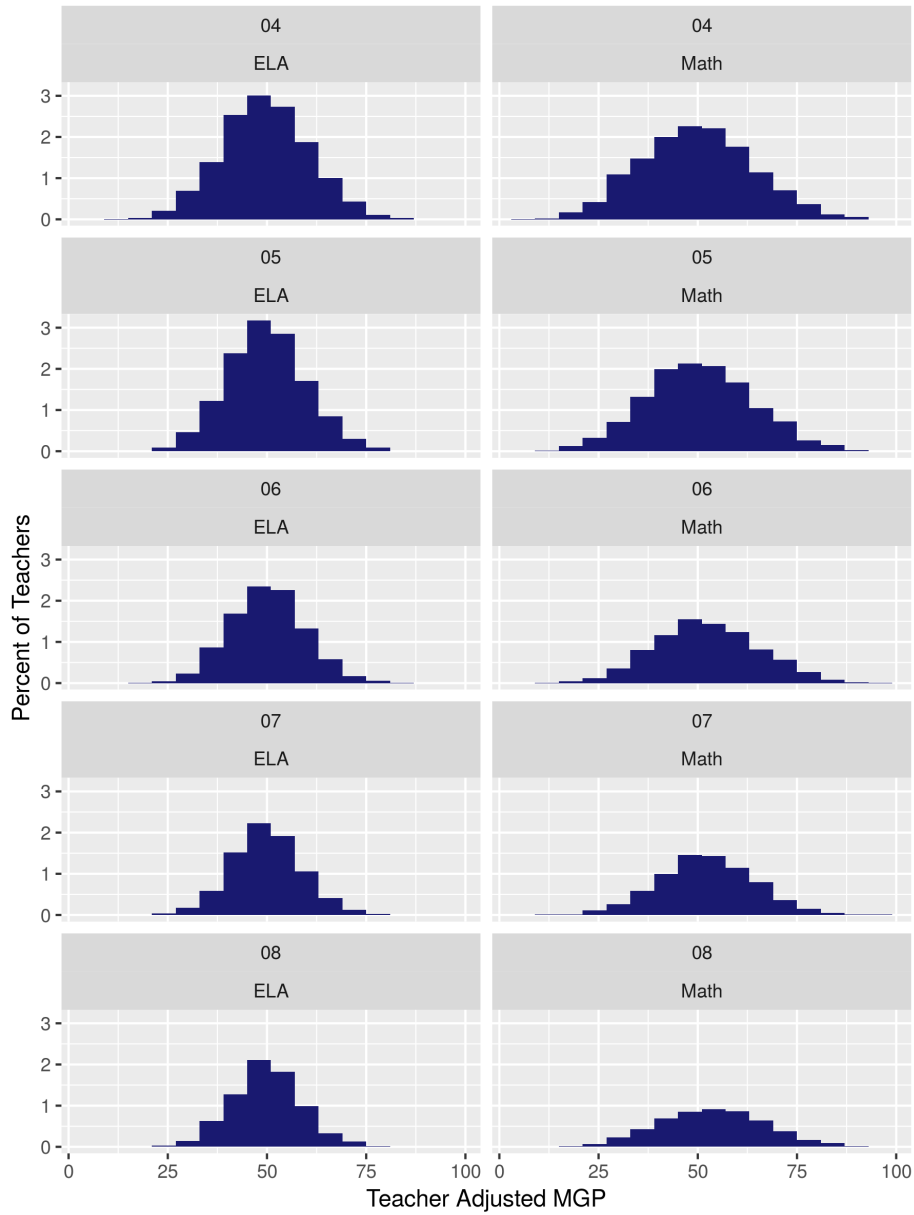
Figure 6 shows that for schools, the results are less widely distributed than for teachers.

Figure 6. Grades 4-8 Distribution of School MGPS
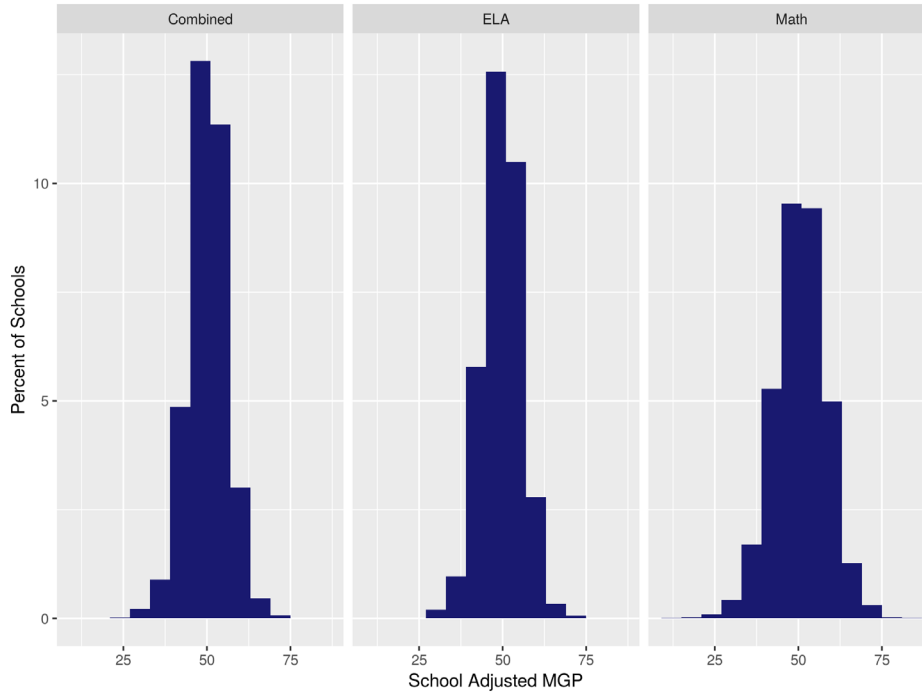
Precision of the MGPs for Grades 4-8

The caterpillar plot in Figure 7 is a random sample of 100 teacher MGPs taken from the 2018/19 data. The MGPs are sorted from lowest to highest, with the corresponding 95% confidence range showing the lower and upper limits of the MGP. Figure 8 shows the same type of plot for schools (where larger underlying samples indicate substantially less variation in the MGP and the error bars are narrower). These figures provide a sample of the distribution of MGPs and a typical confidence range.

Figure 7. Grades 4-8 Overall MGP with 95% Confidence Interval Based on a Random Sample of 100 Teachers
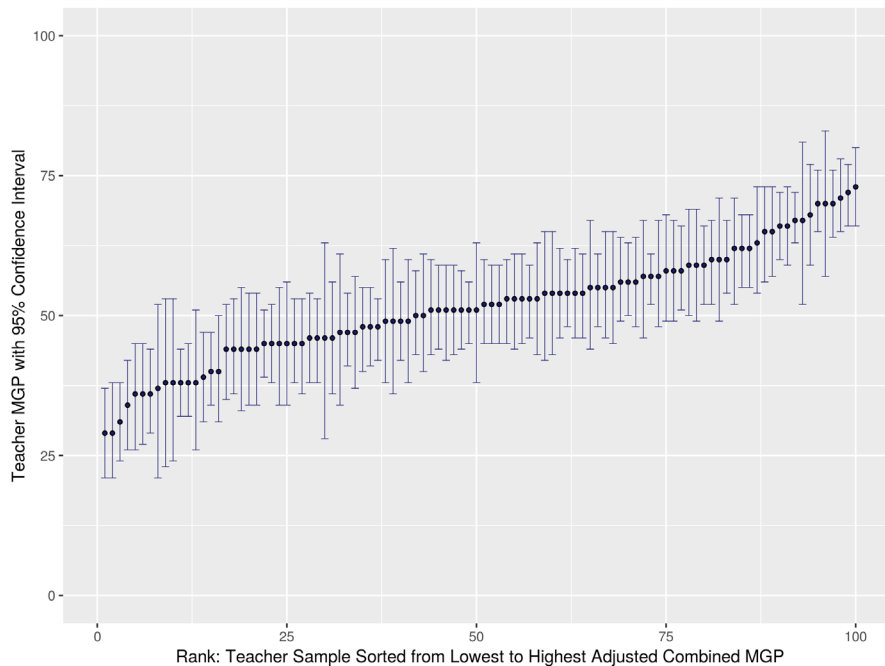
Figure 8. Grades 4-8 Overall MGP with 95% Confidence Interval Based on a Random Sample of 100 Schools
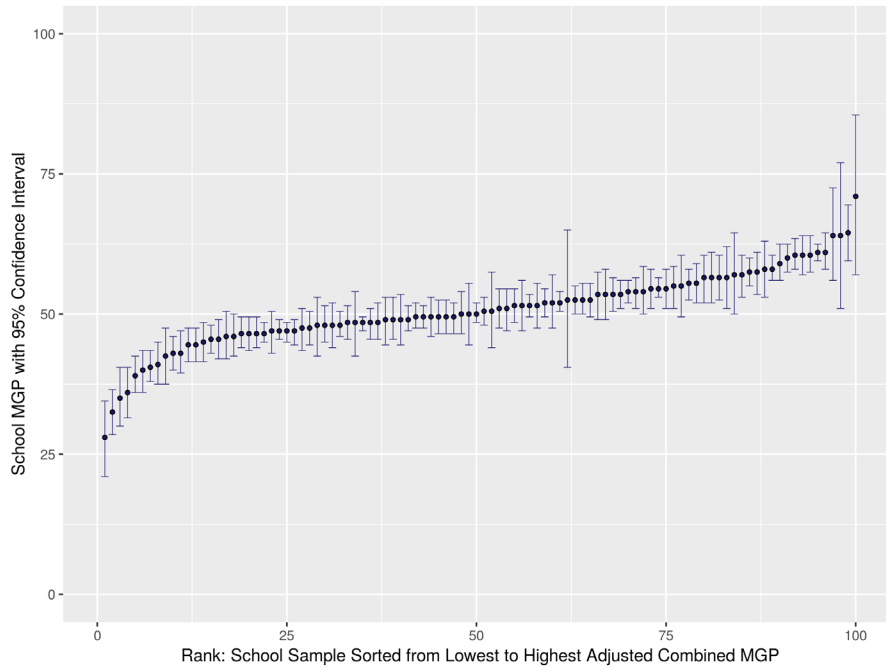
Figure 7 and Figure 8 provide a means to gauge visually the precision of MGPs. However, it also may be useful to examine a reliability statistic to assess the precision of the teacher-level MGPs, specified here as ρ:

$$\rho = 1 - \left(\frac{\bar{\sigma}}{sd\left(\hat{\theta}_j\right)} \right)^2$$

where $\bar{\sigma}$ is the weighted mean standard error of the MGP (weighted by number of SGPs), and $sd\left(\hat{\theta}_j\right)$ is the weighted standard deviation between teacher MGPs (also weighted by number of SGPs). In theory, the highest possible value is one, which would represent complete precision in the measure. When the ratio is zero, the variation in MGPs is explained entirely by sampling variation. Larger values of ρ are associated with more precisely measured MGPs.

Table 11 provides the weighted mean standard errors, the weighted standard deviations, and the values of weighted ρ for the adjusted model by grade and subject using the number of SGPs as weights. The values of the ratio (ρ) quantify imprecision in the estimates. In all grades, the statistics are closer to one than zero, indicating that the differentiation between teachers and schools seen in the measures is not largely related to measurement variance.

Table 11. Grades 4-8 Weighted Mean Standard Errors, Standard Deviation, and Value of ρ by Grade and Subject for Teachers and Schools, Weighted by Number of SGPs

Subject	Grade	Weighted Mean Standard Error	Weighted Standard Deviation	Weighted Reliability Statistic (ρ)
ELA (Teacher)	4	5.242	10.687	0.747
	5	4.955	9.431	0.708
	6	4.041	8.819	0.771
	7	3.817	8.107	0.759
	8	3.780	8.061	0.761
Mathematics (Teacher)	4	5.028	13.750	0.859
	5	4.497	12.998	0.872
	6	3.898	12.354	0.892
	7	3.757	11.508	0.884
	8	4.076	12.761	0.891
All (Schools)	All	1.288	5.047	0.923

Table 12 provides the share of educators whose MGPs are significantly above or below the State mean for that educator type, using the 95% confidence intervals. In all cases, the percentage exceeding the mean is larger than what would be expected by chance alone, indicating the model distinguishes between schools and teachers (2.5% of schools or teachers would be expected to be above and below the mean by chance alone).

Table 12. Grades 4-8 Percentage of Educator MGPs Above or Below Mean at the 95% Confidence Level

Level	Below Mean		Above Mean	
	N	%	N	%
Teacher	7,179	19%	7,891	21%
School	946	26%	1,045	29%

Impact Data Results for Grades 4-8

Table 13 provides the correlations of the combined-subject MGP (or for teachers with only one subject, their single-subject MGP) with five classroom or course characteristics: the three predictor variables at the individual student level NYSED's regulations permit for inclusion in the model and that were selected after discussion with New York's Task Force and other stakeholders—ELL, students with disabilities, and poverty or economic disadvantage—and the

mean prior ELA or mathematics score of the students.[14] Correlations are presented for adjusted MGPs.[15]

Table 13. Grades 4-8 Teacher MGP Correlation with Class or Course Characteristics

School Characteristics	Correlation
ELL students in class or course	0.054
Students with disabilities in class or course	0.106
Economically disadvantaged students in class or course	0.092
Mean prior ELA Z-score	-0.123
Mean prior mathematics Z-score	-0.139

Large correlations between MGP and classroom, course, or school characteristics would indicate systematic relationships between scores and the types of students who teachers and schools serve. A value of 0.10 or less indicates that 1% or less of the variance in MGPs can be predicted with that demographic variable and, therefore, represents results that are essentially zero. In 2018/19, all correlations of MGPs with classroom characteristics have absolute values of 0.123 or lower.

The scatter plots shown in Figure 9 through Figure 13 provide visual representations of the data underlying the correlations for teachers shown in Table 13. Figure 14 through Figure 18 provide similar images of the data underlying the school-level (principal MGP) correlation shown in Table 14.[16]

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[14] For prior scores, the *Z*-score of the scale score is used instead of the actual scale score because many teachers have students in various grades, and the scale scores are not designed to be averaged directly across grades.

[15] The impact of these demographic characteristics on the expected value of students' current test scores used to compute SGPs can be seen through the model coefficients presented in Appendix I. The inclusion of these variables serves to make SGPs for students with different demographic characteristics comparable, given the prior test scores included in the model.

[16] Results disaggregated by grade and subject are shown in Appendix J. The results in this section are combined across grades and subjects.

Figure 9. Grades 4-8 Teacher MGP Scores by Percentage of ELL Students in Class or Course

correlation = 0.054
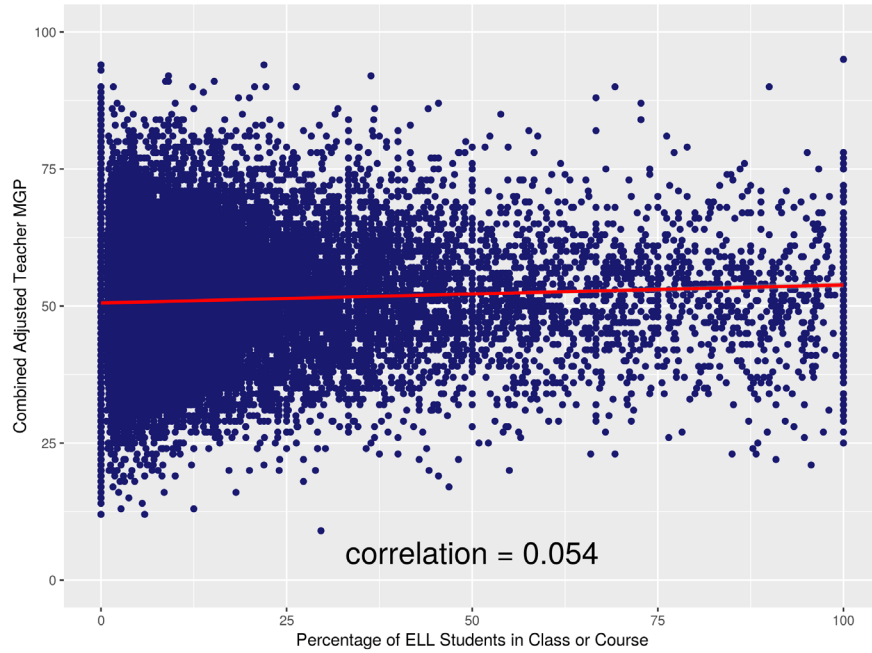
Figure 10. Grades 4-8 Teacher MGP Scores by Percentage of SWD Students in Class or Course

correlation = 0.106
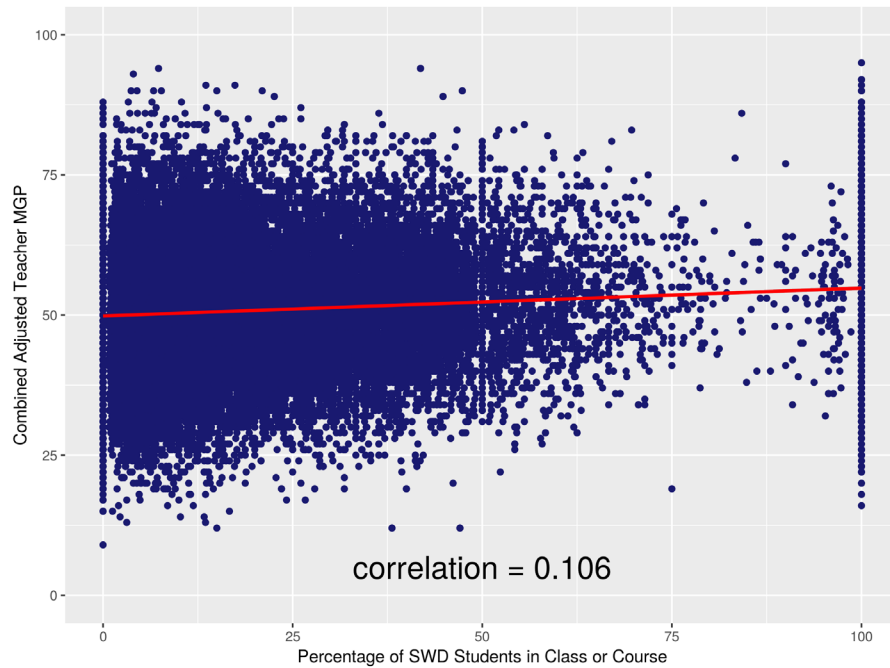
Figure 11. Grades 4-8 Teacher MGP Scores by Percentage of ED Students in Class or Course
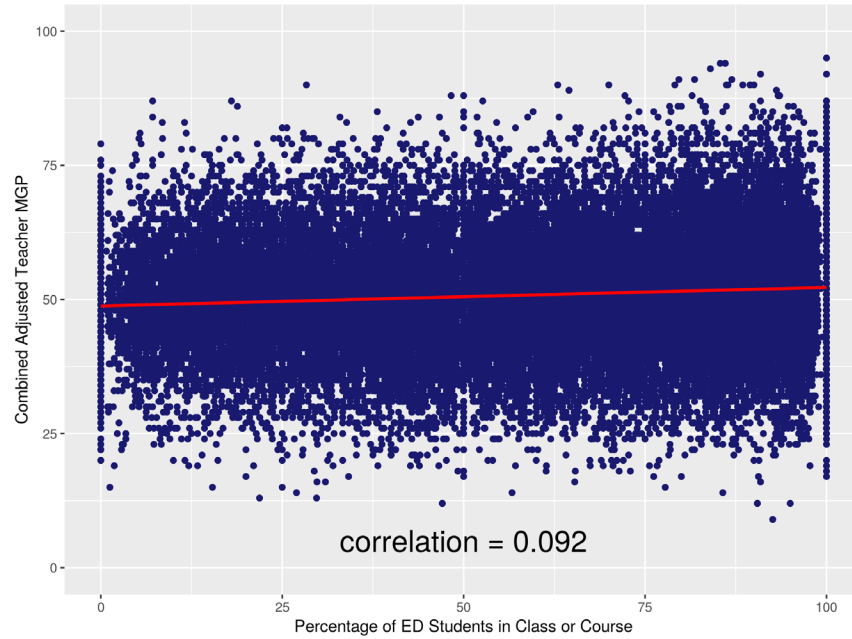
Figure 12. Grades 4-8 Teacher MGP Scores by Mean Prior ELA Z-Score Students in Class or Course
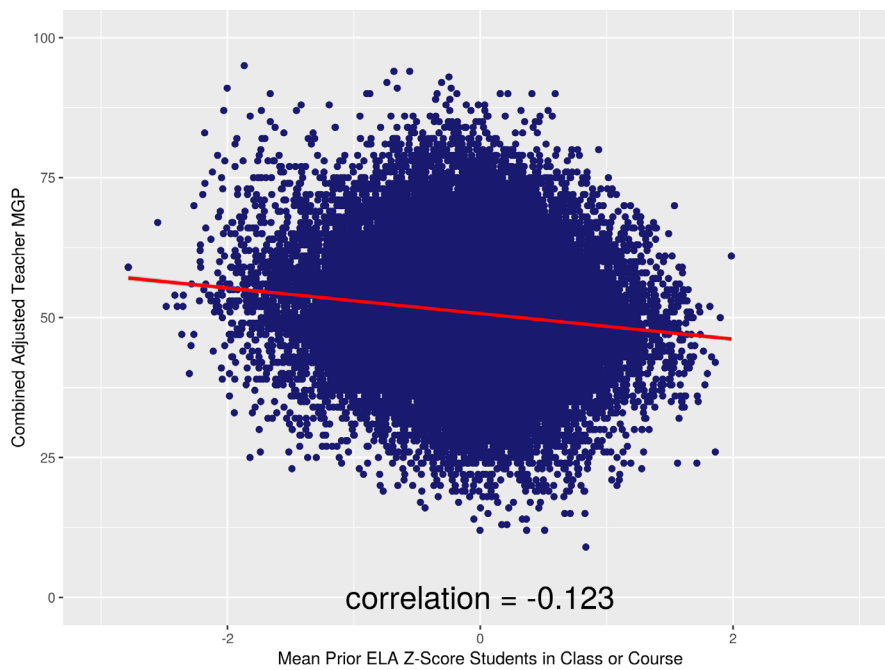
Figure 13. Grades 4-8 Teacher MGP Scores by Mean Prior Mathematics Z-Score Students in Class or Course
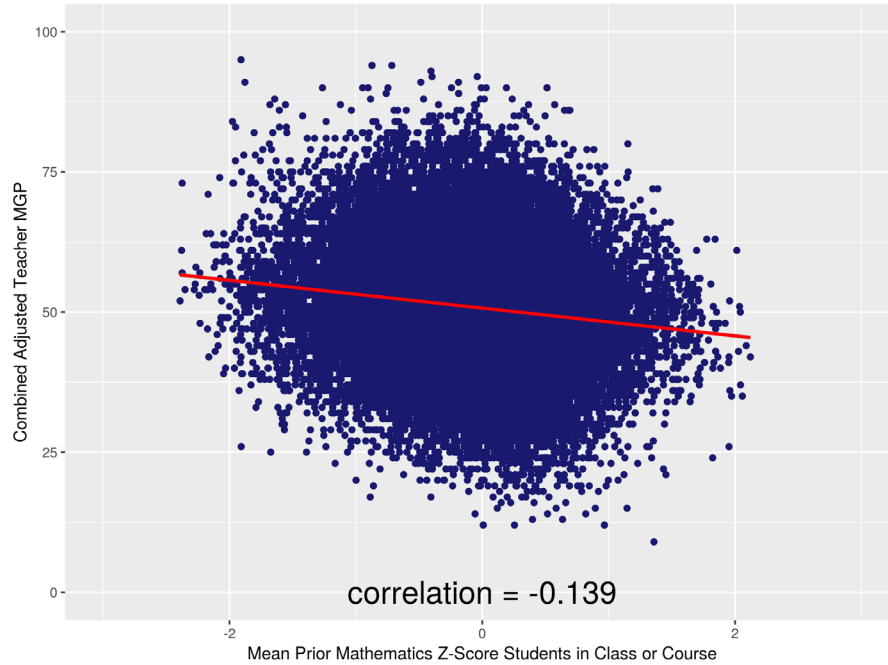
Table 14 provides the observed correlations of school MGPs with the same characteristics presented for teachers but aggregated to the school level. Appendix J contains principal-level correlations.

Table 14. Grades 4-8 School MGP Correlation with School Characteristics

School Characteristics	Correlation
ELL students in school	0.137
Students with disabilities in school	0.07
Economically disadvantaged students in school	0.132
Mean prior ELA Z-score	-0.078
Mean prior mathematics Z-score	-0.097

Figure 14. Grades 4-8 School MGP Scores by Percentage of ELL Students in School
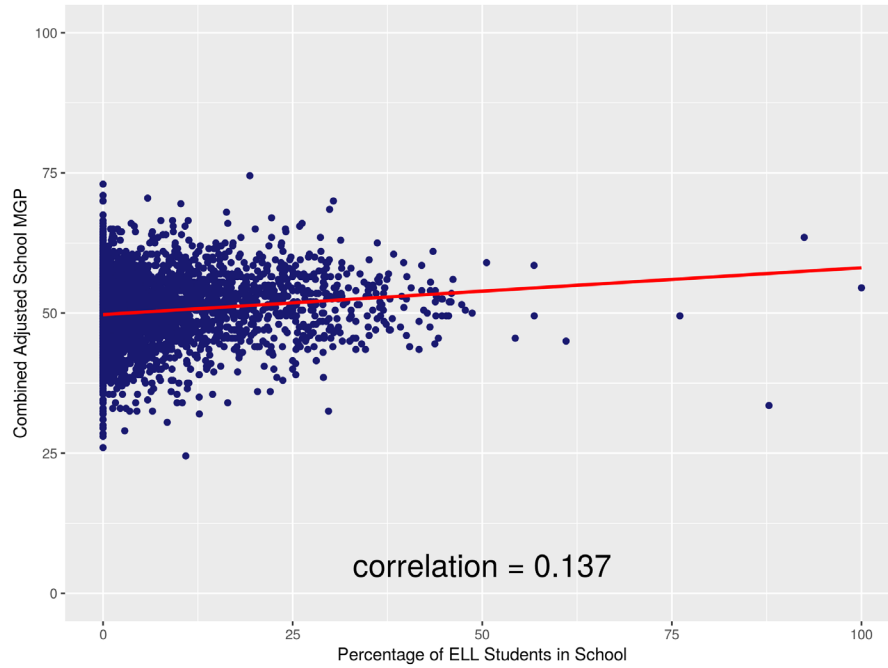
Figure 15. Grades 4-8 School MGP Scores by Percentage of SWD Students in School
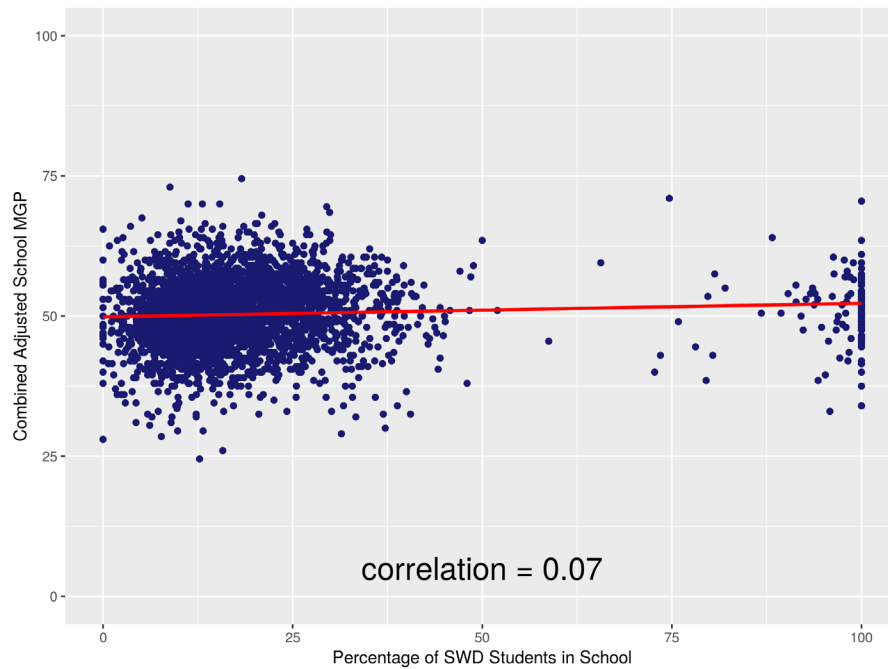
Figure 16. Grades 4-8 School MGP Scores by Percentage of ED Students in School
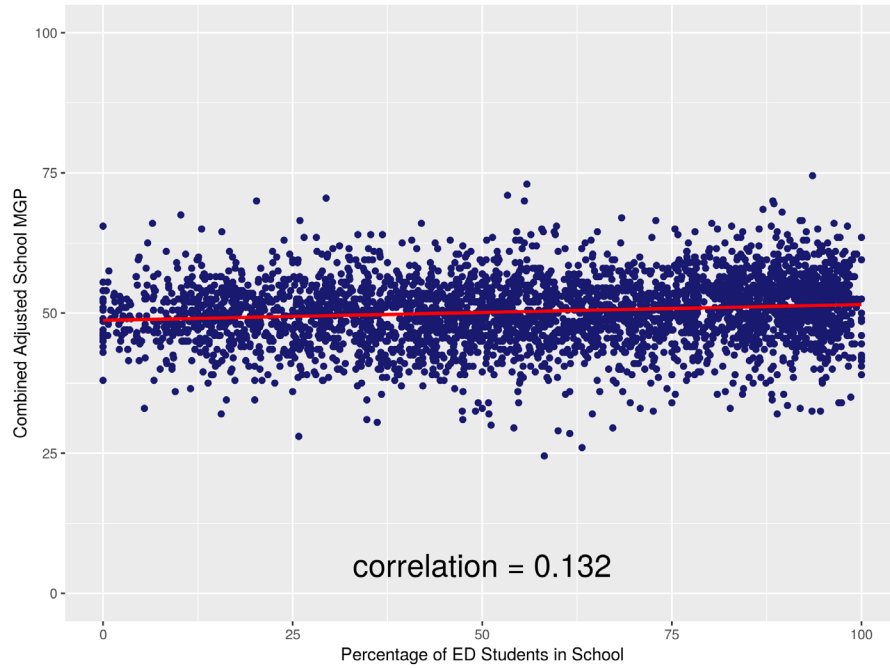
Figure 17. Grades 4-8 School MGP Scores by Mean Prior ELA Z-Score Students in School

Figure 18. Grades 4-8 School MGP Scores by Mean Prior Mathematics Z-Score Students in School
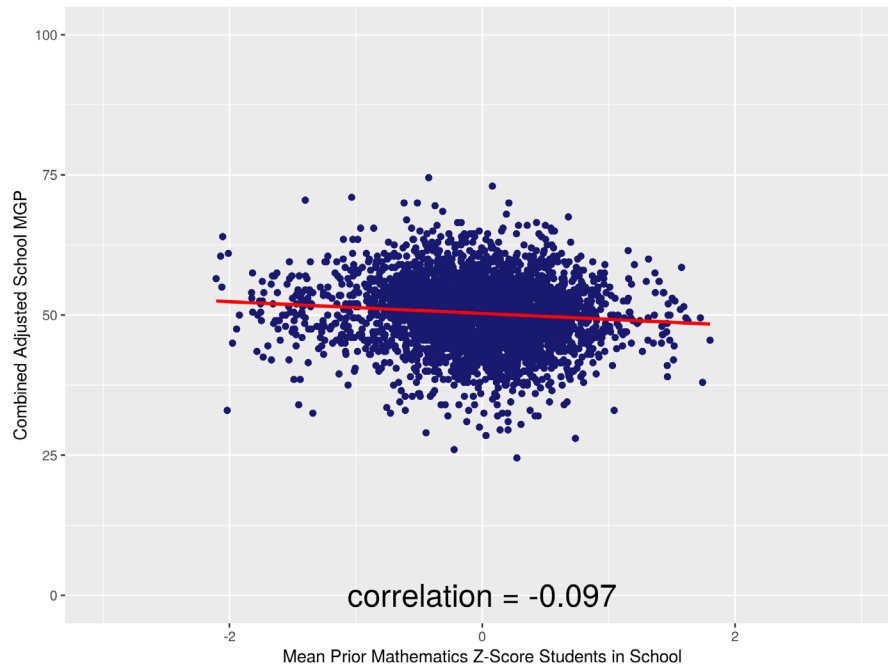
Growth Ratings for Grades 4-8

This section describes the observed distribution of the growth ratings assigned using the rules described earlier in the results section. Table 15 shows the distribution for Grades 4-8 teachers, schools, and principals who serve students in Grades 4-8 (including, for instance, schools serving Grades 4-12) in 2017/18 and 2018/19.

Table 15. Grades 4-8 Teacher, School, and Principal Growth Ratings

School Year	Level	Highly Effective	Effective	Developing	Ineffective
2017/18	Teacher	7%	77%	11%	5%
	Principal	7%	77%	10%	6%
	School	6%	78%	10%	6%
2018/19	Teacher	7%	78%	10%	5%
	Principal	7%	80%	7%	6%
	School	7%	80%	8%	5%

Note: Because of rounding, percentages may not add to 100 percent.

Stability of Growth Ratings for Grades 4-8 across School Years

For teachers who had growth ratings in 2017/18 and 2018/19, Table 16 shows the relationship between ratings across years. Table 17 shows the relationship for school-level MGPs. The results show that the ratings are stable, with about two thirds remaining in the same growth

rating category from year to year. The MGPs have a Pearson correlation coefficient of 0.49 for teachers and a correlation coefficient of 0.44 for schools between 2017/18 and 2018/19. The teacher correlation coefficient is the same as the 2017/18 correlation coefficient and the school correlation coefficient is slightly higher than the 2017/18 correlation coefficient, which was 0.38.

Table 16. Grades 4-8 Teacher Growth Ratings for Teachers Receiving Growth Ratings in Both 2017/18 and 2018/19

		2018/19				
		Highly Effective	Effective	Developing	Ineffective	Total
2017/18	Highly Effective	3%	5%	<1%	<1%	8%
	Effective	5%	63%	6%	2%	77%
	Developing	<1%	7%	2%	1%	10%
	Ineffective	<1%	3%	1%	1%	5%
	Total	8%	78%	9%	5%	100%

Note: Because of rounding, percentages may not add to 100 percent.

Table 17. Grades 4-8 School Growth Ratings for Schools Receiving Growth Ratings in Both 2017/18 and 2018/19

		2018/19				
		Highly Effective	Effective	Developing	Ineffective	Total
2017/18	Highly Effective	2%	4%	<1%	<1%	6%
	Effective	4%	66%	5%	2%	78%
	Developing	<1%	7%	2%	1%	10%
	Ineffective	<1%	3%	1%	1%	6%
	Total	6%	81%	8%	5%	100%

Note: Because of rounding, percentages may not add to 100 percent.

Neutrality of MGPs for Grades 4-8

Given that a primary claim for the use of MGPs is that all educators can demonstrate growth, regardless of the academic starting point of students, it is necessary to determine if there is a strong relationship between MGPs and average prior achievement for students in a school. To that end, Table 18 shows the correlations between MGPs and average prior achievement, which are low to moderate across all grades and subjects. These correlations illustrate that the MGPs are substantially neutral to prior achievement.

Table 18. Correlation Between Adjusted Teacher and School Adjusted MGP and Average Prior Achievement Across Grades and Subjects

Measure of Prior Achievement		Correlation Between Adjusted MGP and Prior Achievement	
Subject	Grade	Teacher	School
ELA	Grade 4	-0.050	-0.081
	Grade 5	-0.132	-0.155
	Grade 6	-0.127	-0.173
	Grade 7	-0.148	-0.130
	Grade 8	-0.132	-0.120
Mathematics	Grade 4	-0.114	-0.135
	Grade 5	-0.104	-0.093
	Grade 6	-0.134	-0.184
	Grade 7	-0.211	-0.186
	Grade 8	-0.158	-0.095

Results from Growth Models for Grades 9-12

This section provides the results for the Grades 9-12 models using 2018/19 Regents Exam data.

Model Fit Statistics for Grades 9-12 Models

Table 19 shows the *R*-squared values for the MGP models based on ELA and Algebra Regents Exam data.

Table 19. Grades 9-12 Pseudo R-Squared Values

School year	Subject	Adjusted	Unadjusted
2018/19	Algebra 1	0.49	0.42
	ELA	0.60	0.50
2017/18	Algebra 1	0.48	0.41
	ELA	0.60	0.51

The GRE model is not a linear model, so we do not provide pseudo *R*-squared values; instead, we evaluate the behavior of the model using impact data.

Correlation of Combined MGP with GRE Results

For Grades 9-12 in 2017/18, the correlation between a school's combined MGP and GRE results was 0.48, which may indicate that these two measures capture different aspects of student growth (one reason both measures were computed for Grades 9-12 schools and principals).

Fraction of Students Included in Measures

On average, the GRE measure includes a larger percentage of students in a Grades 9-12 school than does the combined MGP measure. Table 20 shows the percentages of students included in each measure.

Table 20. Grades 9-12 Weighted Average Percentage of Students Included, weight is number of students attributed to each school.

Measure	Weighted Mean Fraction of Students in a School Included
MGP (ELA/Algebra 1)	35%
GRE	69%

Distribution of MGPs and GRE Scores for Grades 9-12

Figure 19 shows the distribution of combined school MGPs for Grades 9—12—that is, MGPs that combine information across SGPs in Algebra and ELA. The distribution is approximately normal.

Figure 19. Grades 9-12 Distribution of School MGP
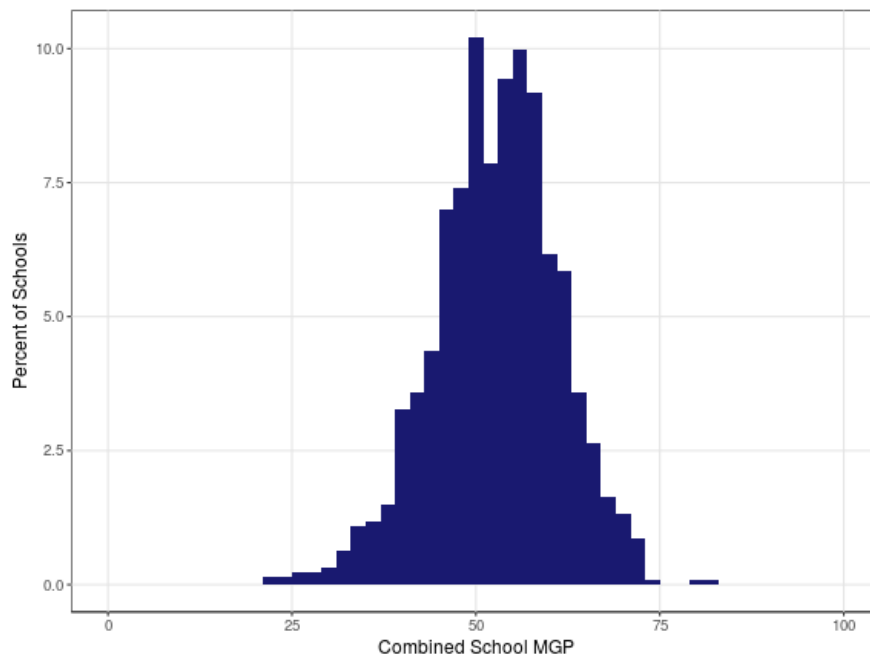
The GRE model reports results as the number of Regents Exams that the average student in a school will pass compared with the number passed by similar students. For example, a GRE score of 0.25 would indicate that, on average, students in that school pass one quarter of a Regents Exam more than do similar students. Over 4 years of high school, this rate per year

would add up to one additional Regents Exam passed by each student. Figure 20 is a histogram of the GRE results, which are somewhat skewed relative to the normal distribution.

Figure 20. Grades 9-12 Distribution of School GRE Scores
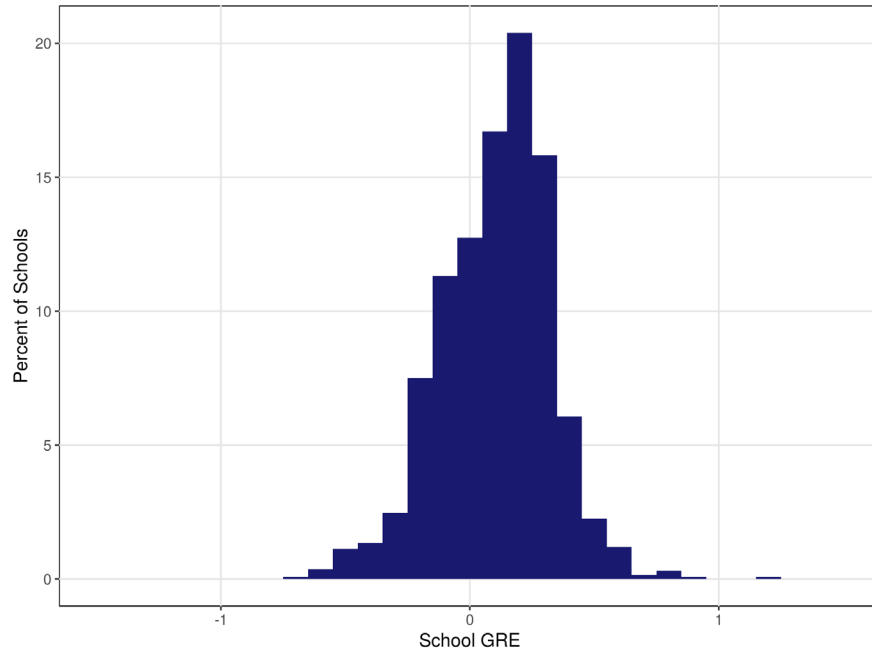
Precision of the Measures for Grades 9-12

The caterpillar plot in Figure 21 shows 100 randomly selected school MGPs and their confidence interval, giving a sense of the precision of the estimates. A second caterpillar plot in Figure 22 shows the GRE measure values and the associated confidence intervals. In both plots, it is apparent that typical confidence intervals are small relative to the overall dispersion in the measures themselves.

Figure 21. Grades 9-12 Caterpillar Plot of School MGPs
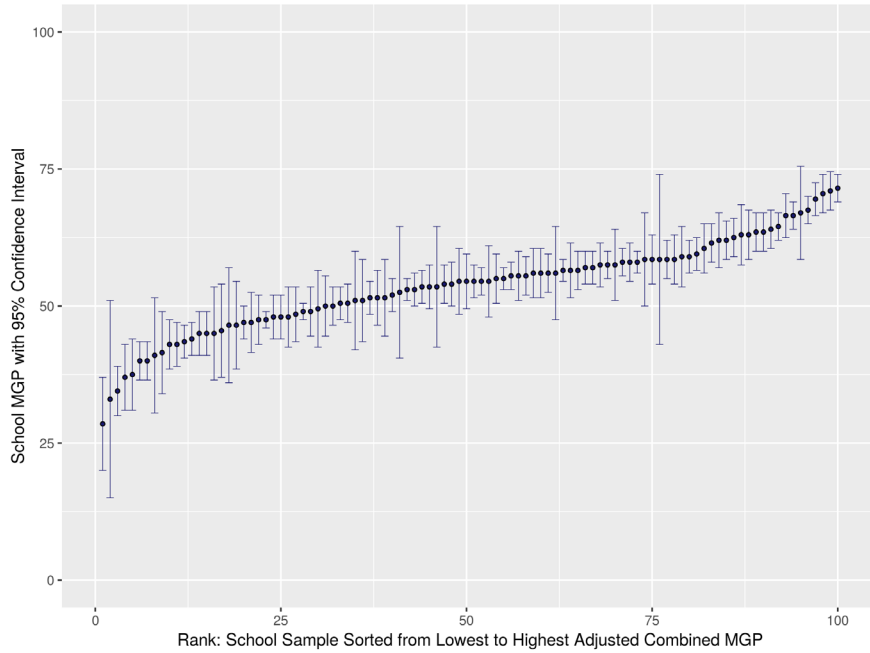
Figure 22. Grades 9-12 Caterpillar Plot of School GRE Scores
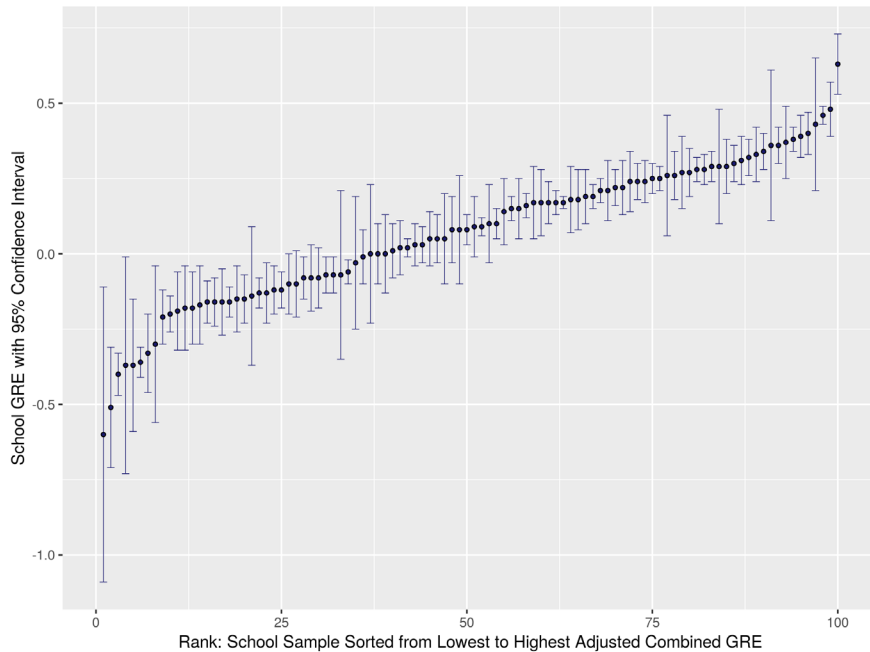
Table 21 shows the share of Grades 9-12 schools whose scores are significantly different from the mean (their confidence intervals on the caterpillar plot do not cross the average value).

Once again, the share exceeds what would be expected by chance alone, indicating that the model is able to distinguish among schools.

Table 21. Percentage of Grades 9-12 School Measures Above or Below the Mean at the 95% Confidence Interval

Measure	Below Mean	Above Mean
MGP	27%	33%
GRE	31%	34%

The weighted reliability (ρ) statistic, which was introduced earlier as a measure of the precision of the MGP measure, is shown in Table 22 for both the GRE and MGP adjusted models for Grades 9–12 models. In both cases, the statistics are much closer to one than zero, indicating that the differentiation between schools seen in the measures is not largely the result of measurement variance.

Table 22. Grades 9-12 Weighted Mean Standard Errors, Standard Deviation, and Value of ρ, Weighted by Number of SGPs

Measure	Weighted Mean Standard Error	Weighted Standard Deviation	Weighted Reliability Statistic (ρ)
MGP	1.609	7.558	0.945
GRE	0.034	0.206	0.964

Impact Data Results for Grades 9-12

Table 23 shows the correlations for the MGP and GRE adjusted models with several school-level demographic variables.[17] Several correlations for the GRE model are larger than 0.10 in absolute value. For example, schools that have a higher percentage of students with disabilities or lower achieving students receive lower GRE scores on average. For the MGP model, the correlation between the school MGP and Grade 8 student test scores is around 0.3, suggesting about nine percent of the variation in MGPs is explained by students' prior scores (the percent of variation explained is equal to the square of the correlation in Table 23). Appendix J shows correlations of school characteristics with principal-level MGPs.

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[17] Note that for Grades 9—12 models, prior scores are all from Grade 8 but are not all equated. Thus, they are all standardized by year and assessment before being used to compute the correlations shown in this section.

Table 23. Grades 9-12 School MGP Correlation with Demographic Characteristics

	MGP	GRE
ELL students in school	0.011	-0.056
Students with disabilities in school	-0.165	-0.264
Economically disadvantaged students in school	0.016	-0.102
Mean Grade 8 ELA score	0.348	0.449
Mean Grade 8 mathematics score	0.273	0.462

Figure 23 through Figure 27 plot these data for MGP results, and Figure 28 through Figure 32 plot these data for GRE results. Note that the demographic correlations are higher for the GRE than for the MGP measures. However, note that there is variation in school-level results at all levels of average prior achievement (as seen in the following figures), suggesting that individual schools over a wide range of characteristics can demonstrate strong results.

Figure 23. Grades 9-12 School MGP Scores by Percentage of ELL Students in School
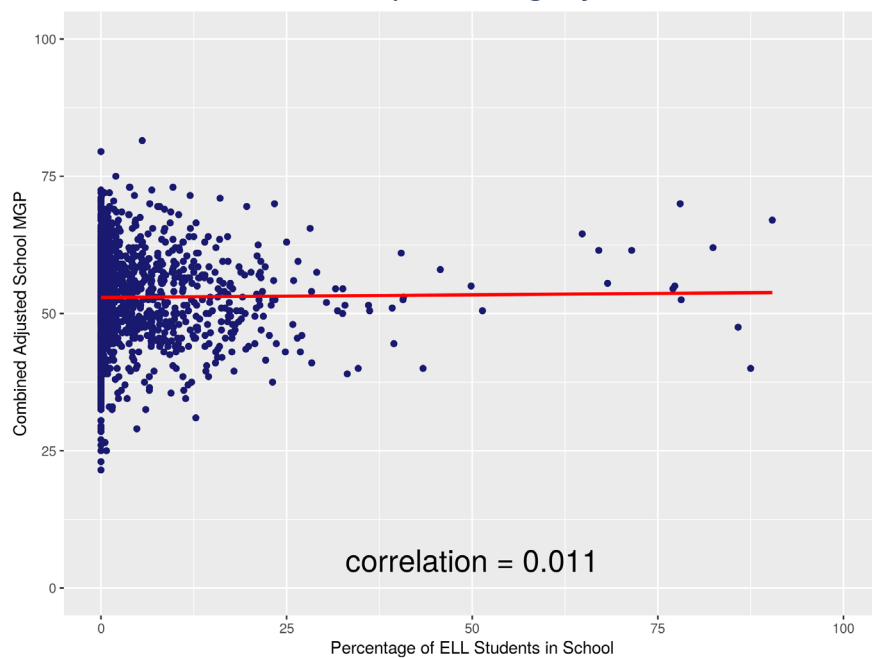
Figure 24. Grades 9-12 School MGP Scores by Percentage of SWD Students in School
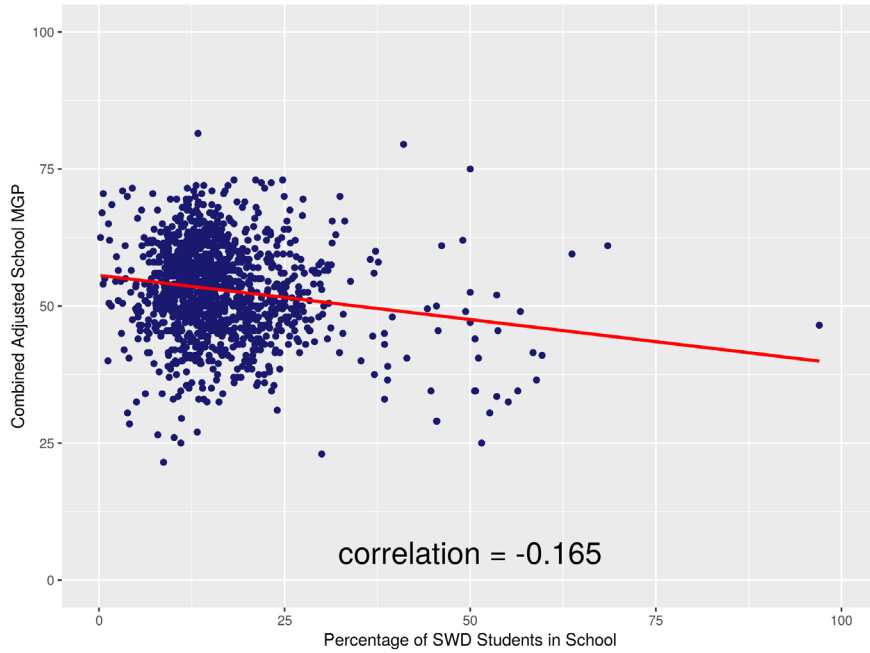
Figure 25. Grades 9-12 School MGP Scores by Percentage of ED Students in School
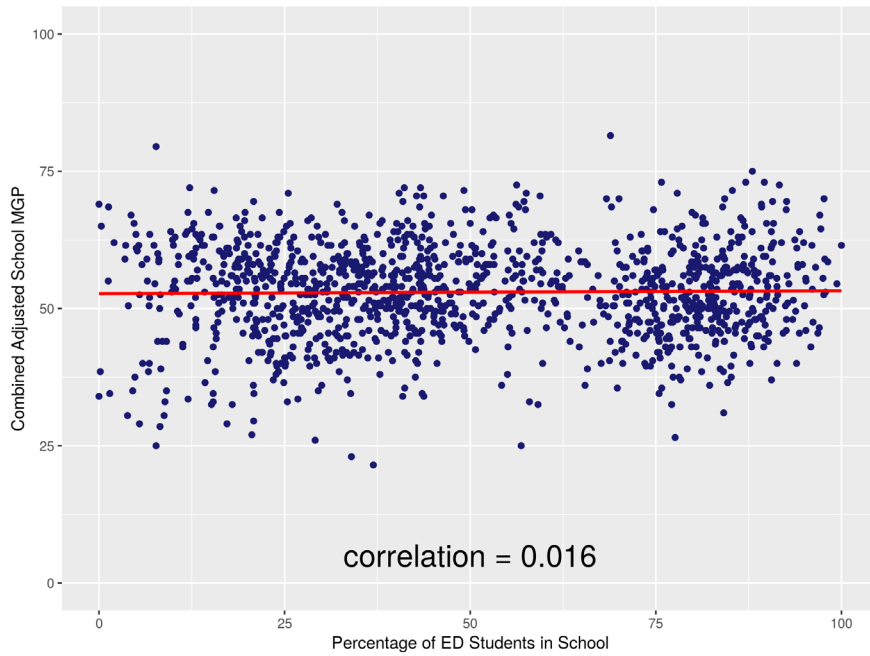
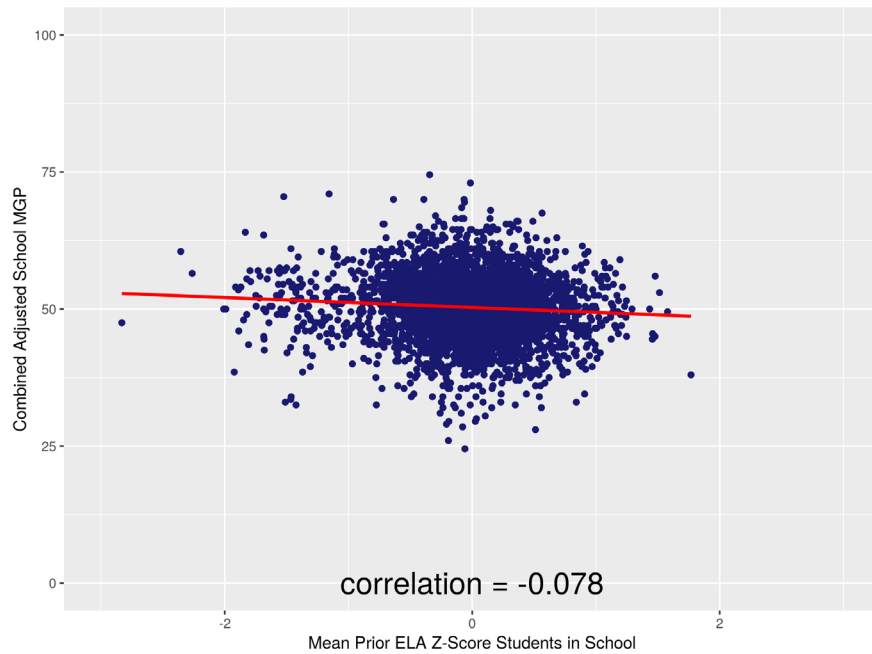*Figure 26. Grades 9-12 School MGP Scores by Mean Grade 8 ELA Z-Score in School*
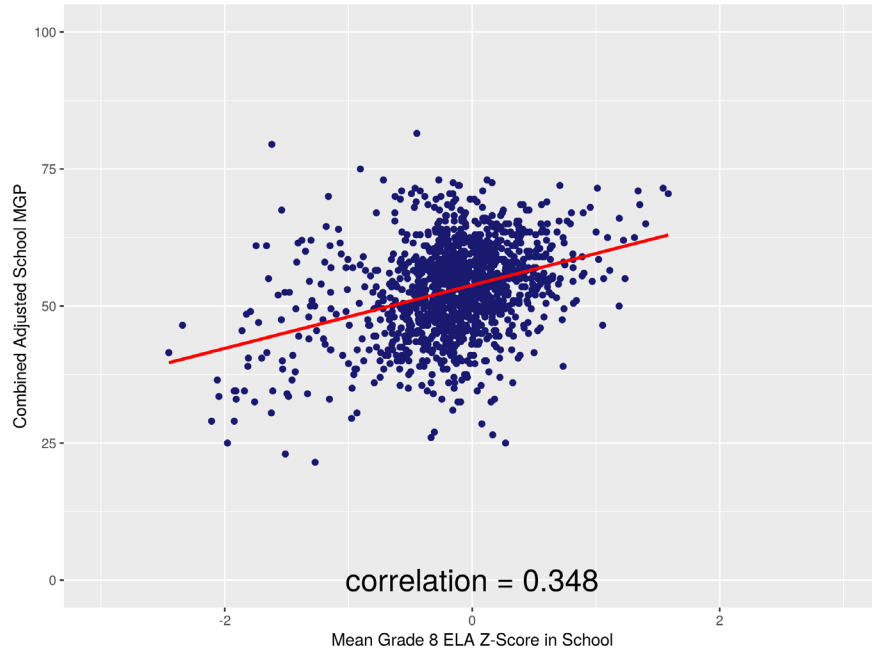
Figure 27. Grades 9-12 School MGP Scores by Mean Grade 8 Mathematics Z-Score in School
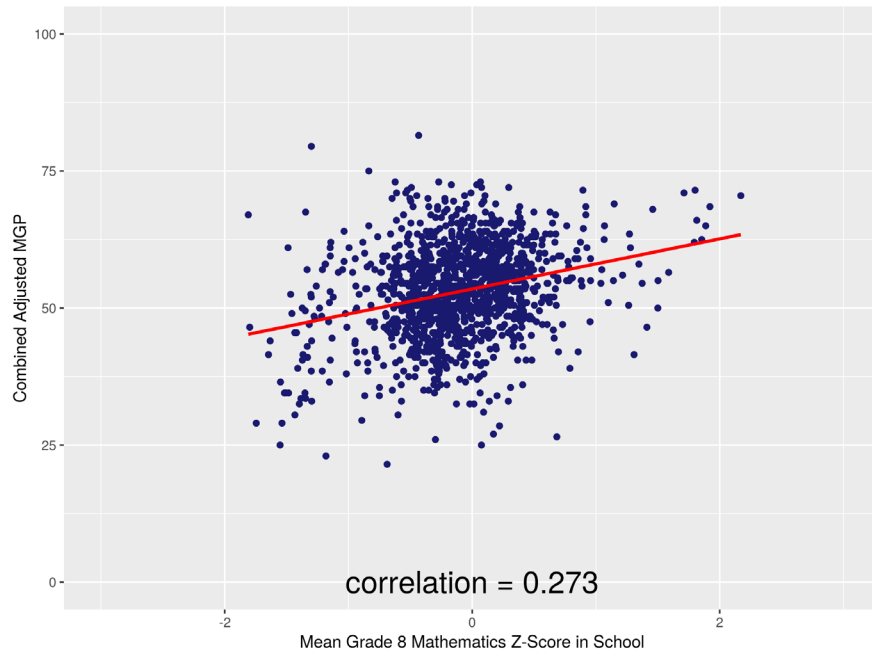
Figure 28. Grades 9-12 School GRE Scores by Percentage of ELL Students in School
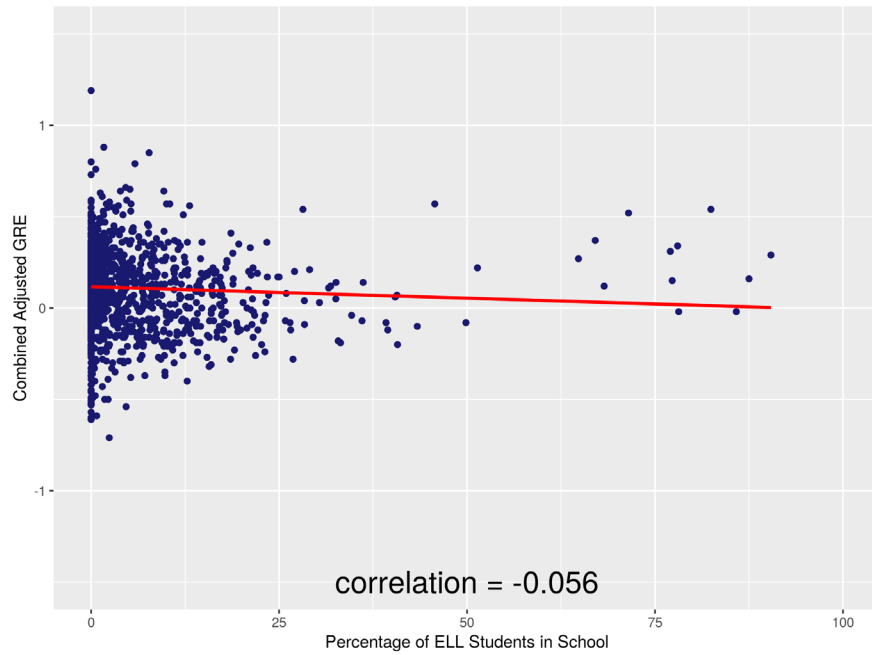
Figure 29. Grades 9-12 School GRE Scores by Percentage of SWD Students in School
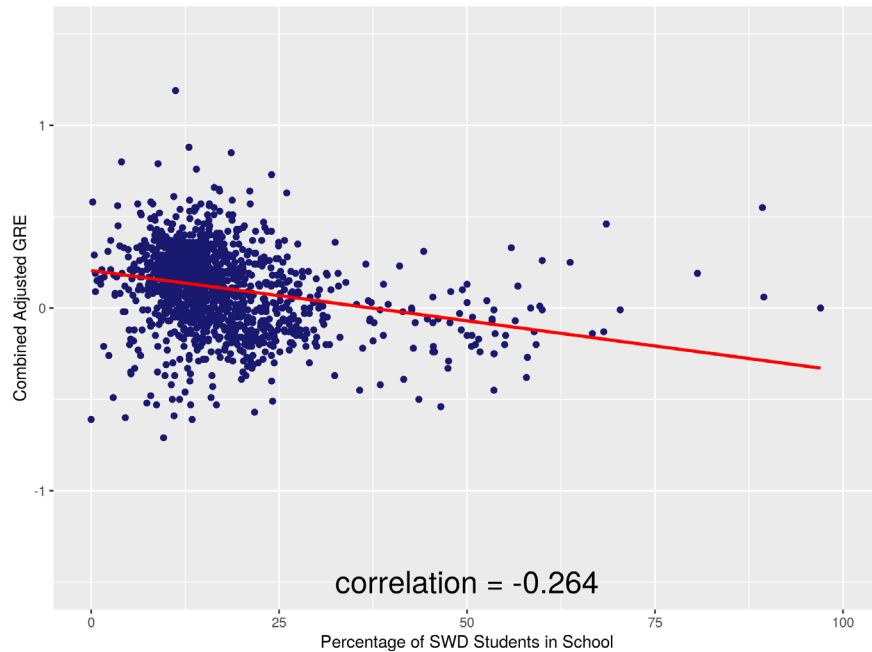
Figure 30. Grades 9-12 School GRE Scores by Percentage of ED Students in School
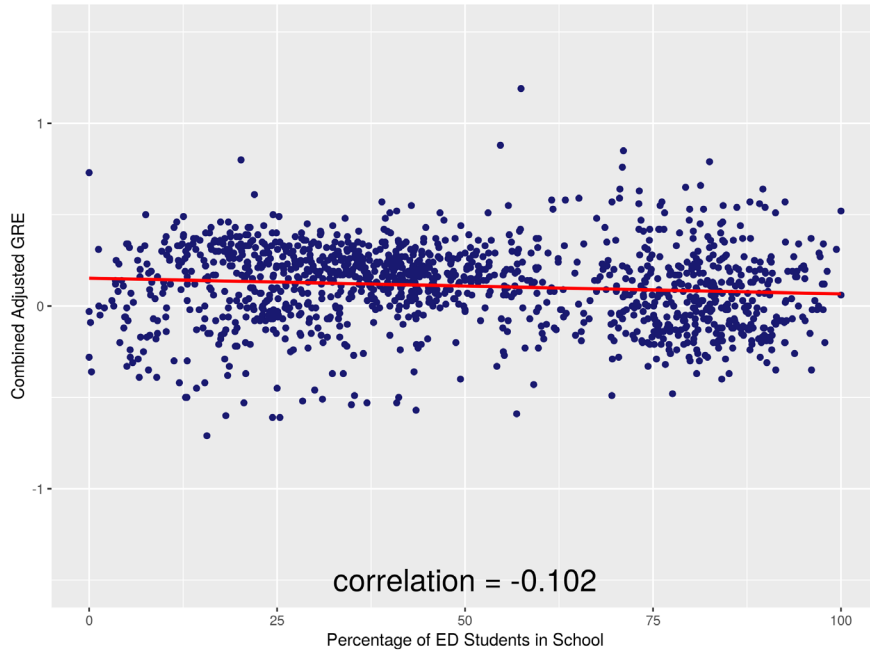
Figure 31. Grades 9-12 School GRE Scores by Mean Grade 8 ELA Z-Score in School
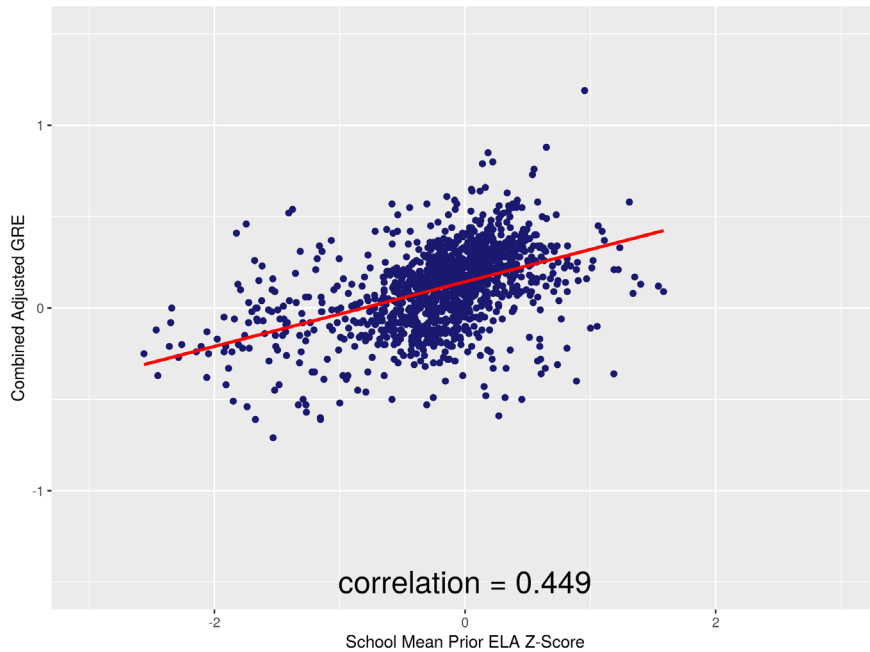
Figure 32. Grades 9-12 School GRE Scores by Mean Grade 8 Mathematics Z-Score in School

correlation = 0.462
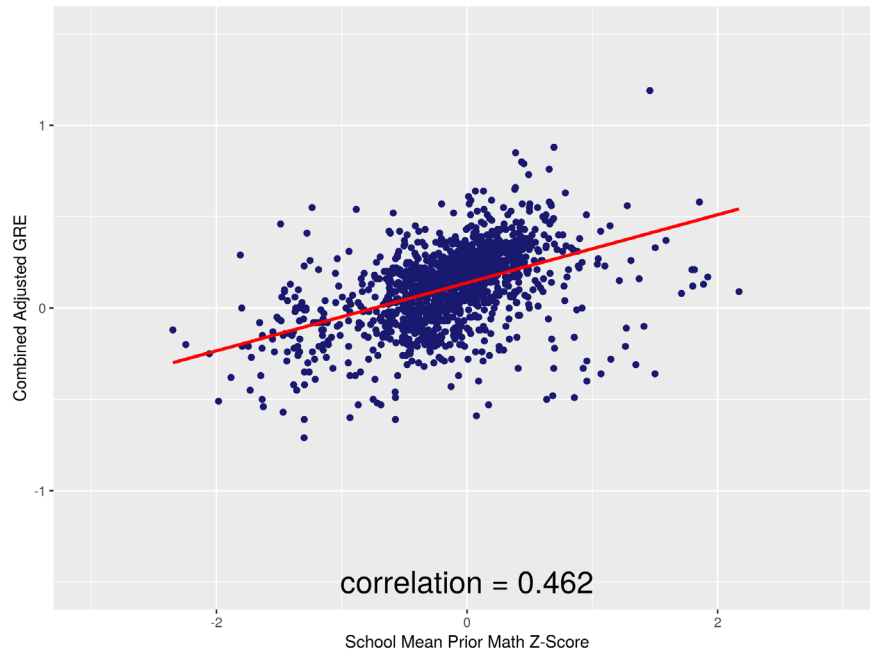
Growth Ratings for Schools with Grades 9-12

Table 24 shows the distribution of growth ratings for schools and principals of all schools serving Grades 9-12 (including schools that may serve other grades, such as Grades 4-8) for 2017/18 and 2018/19.

Table 24. Grades 9-12 School and Principal Growth Ratings

School Year	Level	Highly Effective	Effective	Developing	Ineffective
2017/18	Principal	1%	84%	11%	3%
	School	2%	85%	11%	3%
2018/19	Principal	3%	78%	15%	4%
	School	3%	79%	14%	3%

Note: Because of rounding, percentages may not add to 100 percent.

Table 25 shows the relationship between school ratings across years. The results show that the ratings are stable, with nearly 80% of schools remaining in the same growth rating category from year to year.

Table 25. Grades 9-12 School Growth Ratings for Schools Receiving Growth Ratings in Both 2017/18 and 2018/19

		2018/19				
		Highly Effective	**Effective**	**Developing**	**Ineffective**	**Total**
2017/18	**Highly Effective**	1%	1%	<1%	<1%	2%
	Effective	2%	73%	9%	1%	85%
	Developing	<1%	5%	4%	2%	11%
	Ineffective	<1%	2%	1%	<1%	3%
	Total	3%	80%	14%	3%	100%

Note: Because of rounding, percentages may not add to 100 percent.

Growth Ratings for Schools and Principals Serving Grades 4-8 and Grades 9-12

Some schools receive separate growth ratings for Grades 4-8 and Grades 9-12.

Table 26 shows growth ratings for schools that serve only Grades 4-8 (4-8 only), schools that serve Grades 9-12 only (9-12 only), schools that serve Grades 4-12 and receive both 4-8 and 9-12 growth ratings (4-8 and 9-12), and all schools that received a growth rating (all schools).

Table 27 shows similar information for principals.

Table 26. Growth Ratings for Schools

	Model	**Highly Effective**	**Effective**	**Developing**	**Ineffective**	**Number of Schools**
4-8 Growth Rating	4-8 Only	7%	81%	8%	5%	3,192
	4-8 and 9-12	7%	76%	8%	9%	404
	All Schools	7%	80%	8%	5%	3,596
9-12 Growth Rating	4-8 and 9-12	4%	84%	9%	3%	404
	9-12 Only	3%	77%	17%	3%	931
	All Schools	3%	79%	14%	3%	1,335
Overall Growth Rating	4-8 and 9-12	2%	85%	12%	1%	404
	All Schools	5%	80%	10%	4%	4,527

Note: Because of rounding, percentages may not add to 100 percent.

Table 27. Growth Ratings for Principals

	Model	Highly Effective	Effective	Developing	Ineffective	Number of Principals
4-8 Growth Rating	4-8 Only	7%	81%	7%	6%	3,284
	4-8 and 9-12	7%	76%	7%	10%	444
	All Principals	7%	80%	7%	6%	3,728
9-12 Growth Rating	4-8 and 9-12	3%	82%	9%	5%	444
	9-12 Only	3%	79%	15%	3%	888
	All Principals	3%	80%	13%	4%	1,332
Overall Growth Rating	4-8 and 9-12	2%	84%	14%	0%	444
	All Principals	6%	81%	9%	5%	4,616

Note: Because of rounding, percentages may not add to 100 percent.

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Appendix A. Variables Included in the Adjusted Model

Academic History Variables

- **Prior Achievement**
  - For Grades 4-8 growth measures, up to three years of prior achievement scores in the same subject were included (except for Grades 4 and 5, where fewer years of data were available). Students without scores from the immediate prior grade level in the immediate prior year were excluded from analysis. In addition, the immediate prior grade-level score in the other subject (for ELA models, the mathematics score; for mathematics models, the ELA score) was included if available.
  - For Grades 9-12 growth measures, scores from Grade 7 and Grade 8 assessments (if available) in ELA and mathematics were used as predictors. For the MGP measure, students must have had at least one score from Grade 7 or Grade 8 in the same subject (for the Algebra Regents model, from the Grade 7 or Grade 8 mathematics test; for the ELA Regents models, from the Grade 7 or Grade 8 ELA test). For the GRE measure, to be included in the analysis, students must have had at least one Grade 7 or Grade 8 score in either mathematics or ELA.
- **Retained in Grade (Grades 4-8 Growth Measures Only).** This variable is a yes or no variable that indicates whether a student was retained in grade in one of the two years preceding the most recent school year for students above Grade 4 (e.g., if a student was in Grade 5, Grade 5 again, and then Grade 6). Because students must have an immediate prior score from the prior grade, students who were retained in grade between 2017/18 and 2018/19 were not included in the model (e.g., students with data from Grade 6 in 2017/18 and Grade 6 in 2018/19). This variable was computed based on students' tested grade in the assessment score file.
- **Mean Prior Score.** This variable is intended to account for differences in learning environments that are made up of students with disparate levels of incoming achievement.
  - For Grades 4-8 growth measures, the average immediate prior same-subject achievement on the State test of all students attributed to a teacher in the current year was included in the model (e.g., the average prior ELA achievement of all students in a teacher's class/course was included in ELA models).
  - For Grades 9-12 growth measures, average Grade 8 achievement of the schools' students when they were in Grade 8 was included in each model. For the MGP measure, average Grade 8 achievement of the schools' students when they were in Grade 8 at the school level in the same subject (for the Algebra Regents

model, from the Grade 8 mathematics test; for the ELA Regents model, from the Grade 8 ELA test) was used. For the GRE measure, average Grade 8 achievement at the school level in mathematics and ELA was used. Note that separate mean prior variables were used for Common Core–aligned and prior versions of State assessments.

- **Range Around Mean Prior Score (Grades 4-8 Growth Measures Only).** Classrooms or courses with the same average prior score may differ in the range of prior scores, and students may have different growth trajectories based on being in classrooms or courses with more widely varying prior scores than those with more closely bunched prior scores. In other words, students' peers may affect students not only through their average ability but also through the diversity of ability levels in the classroom or course. This aggregate-level variable is an indicator of the magnitude of difference in prior achievement in a teacher's class or course, calculated as the interquartile range of prior test scores—that is, the distance between the 25th and the 75th percentile of prior performance in the class or course. This variable was calculated using prior achievement scores in the same subject in a teacher's class or course. For example, for the ELA model, the interquartile range of prior scores in ELA in a teacher's class or course was used in the model.

- **New to School in Nonarticulation Year.** This student-level variable is intended to account for differences among students who enroll in a school at a different grade level than the typical entering year for most students. For example, a student enrolls as a seventh grader in a school that serves Grades 6-8 when most other students entered the school at Grade 6, or for students in a Grades 9-12 school, a student enters in a grade other than Grade 9. To compute this variable for the Grades 4-8 model, a student's tested school and grade in 2018/19 was compared with their prior tested school and the range of grades served in the school. For Grades 9-12 models, enrollment data from 2017/18 and 2018/19 were compared.

- **Years Since Entering Ninth Grade (Grades 9-12 Growth Measures Only).** This variable is intended to account for differences among students related to when they take Regents Exams, rather than using a student's grade level (because student grade assignment is affected by credit accumulation and Regents Exams are taken in many different grades). For example, a student who takes the Algebra Regents Exam in his third year after entering Grade 9 has a different academic history than a student who takes the exam in his first year as a ninth grader. This variable is used as an alternative to the "retained in grade" variable used in Grades 4-8 analysis as a way to compare students with similar kinds of academic histories. To compute this variable, the Grade 9 entry date provided on an enrollment file was used.

- **Count of Prior Required Regents Exams (Grades 9-12 Measures Only).** This variable captures the number of Regents Exams in the five traditionally required subject areas[18] that students have passed before the current year (in this case, 2018/19) for Grades 9-12 MGP models. To compute this variable, we reviewed Regents assessment score files back to 2010/11.

Students with Disabilities Variables

- **Student with Disability Status.** A yes or no variable is used for each student to indicate the student has an individualized education program (IEP). This variable was derived directly from the assessment score file, representing data that districts reported to the State.

- **Student with Disability Spending Less Than 40% Time in General Education Settings.** This variable is intended to account for differences among special education students in terms of the intensity or type of services received. According to Individuals with Disabilities Education Act (IDEA) requirements, students should be enrolled in the least restrictive environment (LRE) appropriate for their learning needs. This variable identifies students who spend less than 40% of their time in a general education setting (who may have a disability requiring more specialized or intensive services). This variable was derived directly from the assessment score file, representing data that districts reported to the State.

- **Percentage of Students with Disabilities.** This variable is intended to account for differences in the learning environment for courses or schools serving different proportions of special education students. The variable was defined as the percentage of students identified as having a disability in the class or course for Grades 4-8 growth measures and the percentage of students identified as having a disability in the school for Grades 9-12 measures.

English Language Learner Variables

- **ELL Status.** This variable is a yes or no variable for each student to indicate whether he or she is an ELL student. This variable was derived directly from the assessment score file, representing data that districts reported to the State. Part 154 of Commissioner's Regulations defines students with limited English proficiency as students who, by reason of foreign birth or ancestry, speak or understand a language other than English and speak or understand little or no English, and require support in order to become proficient in English and are identified pursuant to Section 154-2.3 of this Subpart.

- **NYSESLAT Scores.** This variable is intended to account for differences in the English language proficiency of students identified as ELLs by controlling directly for their prior-year NYSESLAT scores. For Grades 9-12 models, NYSESLAT scores from Grade 7/8 forms were used. Three versions of NYSESLAT scores are used in Grades 9-12 models. Separate

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[18] See footnote 2 for details on the change in graduation requirements beginning in the 2014/15 school year.

predictor variables were included for NYSESLAT scores from 2011/12 and earlier (when two separate scale scores for Listening/Speaking and Reading/Writing were used); 2012/13 and 2013/14 (when there was a single scale score); and another, different single scale score version for 2014/15 and later. For Grades 4-8 models, NYSESLAT scores from the immediate prior year were used.

- **Percentage of ELL Students.** This variable is intended to account for differences in the learning environment for courses or schools serving diverse proportions of ELL students. The variable was defined as the percentage of students identified as ELL in the class or course for Grades 4-8 growth measures and the percentage of students identified as ELL in the school for Grades 9-12 measures.

Economically Disadvantaged Variables

- **Economic Disadvantage (Poverty).** A yes or no variable for each student indicates whether the student is identified as economically disadvantaged based on eligibility for a variety of State economic assistance programs. This flag was set to yes for students whose families participate in economic assistance programs, such as the free or reduced-price lunch programs, Social Security Insurance, food stamps, foster care, refugee assistance, earned income tax credit, the Home Energy Assistance Program, Safety Net Assistance, the Bureau of Indian Affairs, or Temporary Assistance for Needy Families, based on district-provided information. This variable was derived directly from the assessment score file, representing data that districts reported to the State.
- **Percentage of Economically Disadvantaged Students.** This variable is intended to account for differences in the learning environment for courses or schools serving diverse proportions of economically disadvantaged students. The variable was defined as the percentage of students identified as economically disadvantaged in the class or course for Grades 4-8 growth measures and percentage of students identified as economically disadvantaged in the school for Grades 9-12 measures.

Appendix B. Technical Advisory Committee Members

Member	Affiliation[1]
Dan Goldhaber	University of Washington
Hamilton Lankford	State University of New York at Albany
Daniel F. McCaffrey	ETS/RAND
Jonah Rockoff	Columbia University
Tim R. Sass	Georgia State University
Douglas Staiger	Dartmouth College
Marty West	Harvard University
James A. Wyckoff	University of Virginia

1: Affiliations are shown as of the time of the Technical Advisory Group's meetings with New York State in 2012 and 2013.

Appendix C. Grades 4-8 Data Processing Overview

The process used to convert the raw data to results runs through six standardized processes for both the 4-8 and 9-12 results. The process and raw data files used to produce the 4-8 results are explained in greater detail below.

Raw Data

All historical and current data files transferred from NYSED. In addition to EA's standard raw data QC process, we conducted an additional quality control check this year where EA and NYSED separately confirmed the file size and number of rows in each file transferred. This ensured that the files were complete and there would be no missing data. The raw data files that were used in the production of 4-8 growth results this year include:

1. **Assessment and CSEM (2018/19, 2017/18, 2016/17, and 2015/16)** – Student-level results on the state 3-8 assessments and CSEMs.
2. **New York State English as a Second Language Achievement Test (NYSESLAT)** – Assessment to determine an English language learner's English language proficiency level.
3. **Directory** – Listing of all New York State Public and Nonpublic Schools.
4. **Teacher Student Course** – Students linked to each teachers' classroom used to attribute students to teachers.
5. **Staff Assignment** – Students linked to programs that principals oversee including the start and end dates.
6. **Enrollment (Algebra 8 Continuous Enrollment and BOCES Enrollment)** –Students that were enrolled on Basic Educational Data System (BEDS) day and during the test administration period.

Standard Data

Raw data are transformed into a standardized format that 1) facilitates the processing of raw data through business rules and 2) can be interpreted by other analysts. Throughout this process, raw data modifications are catalogued, all observations are maintained, and variable names are standardized.

Input Sets

Most of the business rules in data processing are applied in transition from standard data to input sets. Input sets are the data sets that are used to estimate the regression models. Students that will ultimately be excluded from the model are retained in the input sets with an exclusion reason flag activated. These exclusion reasons, which describe students excluded from the growth results for teachers, schools, and principals, are investigated as part of the process of producing input sets.

Modeling

The statistical models are computed using the input sets in the modeling phase and the output is analyzed using a diagnostics tool that examines coefficients, residual mean squared error, student predictions, HOSS/LOSS, and other key metrics.

Aggregation

Results from the modeling phase are combined to create teacher, principal, and school level metrics, such as Mean Growth Percentile, for each level. This step also includes examining aggregate diagnostic measures such as neutralities, reliability, and sample size.

Output

After the aggregation step, the rules for HEDI points and ratings are applied and the final files are created for NYSED and parsed for each district.

Appendix D. Grades 4-8 Item Description Used in Analysis

The teacher-student-course linkage file includes information about courses taught to students. The item description provides information about which courses are relevant to State tests. Table D 1 shows the records used for growth model analysis. Students enrolled in Algebra I (course code 02052CC), Geometry (course code 02072CC), or Algebra II (course code 02056CC) who take Grades 6-8 mathematics assessments are included in the analysis.

Table D 1. Relevant Grades 4-8 Item Descriptions

Item Descriptions
Grade 3 ELA
Grade 3 Mathematics
Grade 4 ELA
Grade 4 Mathematics
Grade 5 ELA
Grade 5 Mathematics
Grade 6 ELA
Grade 6 Mathematics
Grade 7 ELA
Grade 7 Mathematics
Grade 8 ELA
Grade 8 Mathematics

Appendix E. Model Derivation

The following describes a general case of the growth model described in this report. In New York State in 2014-15, there were no indicator variables included for specific educators, so the Z and D matrix in the following are always zero in every entry.

To describe how the model accounts for measurement variance, we first re-express the true score regression as follows:

$$y_t^* = X\beta + \sum_{r=1}^{L} y_{t-r}^* \gamma_{t-r} + Z\theta + e$$

We use $*$ to denote the variables without measurement variance. For convenience, the matrices are defined as $W = \{X, y_{t-1}, y_{t-2}, \ldots, y_{t-L}\}$, $W^* = \{X, y_{t-1}^*, y_{t-2}^*, \ldots, y_{t-L}^*\}$, and $\delta' = \{\beta', \gamma'\}$. Label the matrix of measurement variance disturbances U for disturbances associated with $y_{t-1}, y_{t-2}, \ldots, y_{t-L}$ and label the vector of measurement disturbances with the dependent variable, y_t, v, hence $y_t = y_t^* + v$. Let U have the same dimensions as W, but only the final L columns of U are nonzero, so $W = W^* + U$. If those disturbances were observed, the parameters $\{\delta', \theta'\}$ can be estimated using Henderson's methods (1953) by solving the following mixed model equations:

$$\begin{pmatrix} W^{*\prime}\Omega^{-1}W^* & W^{*\prime}\Omega^{-1}Z \\ Z'\Omega^{-1}W^* & Z'\Omega^{-1}Z + D^{-1} \end{pmatrix} \begin{pmatrix} \delta \\ \theta \end{pmatrix} = \begin{pmatrix} W'\Omega^{-1}y_t^* \\ Z'\Omega^{-1}y_t^* \end{pmatrix}$$

The matrix D is made up of Q diagonal blocks, one for each level in the hierarchy. Each diagonal is constructed as $\sigma_q^2 I_q$, where I_q is an identity matrix with dimensions equal to the number of units at level q, and σ_q^2 is the estimated variance of the random effects among units at level q. When concatenated diagonally, the square matrix D has dimension $m = \sum_{q=1}^{Q} J_q$.

Two complications intervene. First, we cannot observe U; second, the unobservable nature of this term, along with the heterogeneous measurement variance in the dependent variable, renders this estimator inefficient.

Addressing the first issue, on expansion we see that

$$W'\Omega^{-1}W = (W^{*\prime} + U')\Omega^{-1}(W^* + U) = W^{*\prime}\Omega^{-1}W^* + U'\Omega^{-1}W^* + W^{*\prime}\Omega^{-1}U + U'\Omega^{-1}U$$

Taking expectation over the measurement error distributions and treating the true score matrix, W^*, as fixed, we have

$$E(W'\Omega^{-1}W) = E\big((W^{*\prime} + U')\Omega^{-1}(W^* + U)\big) = W^{*\prime}\Omega^{-1}W^* + E(U'\Omega^{-1}U)$$

We also have $Z'\Omega^{-1}W^* = E(Z'\Omega^{-1}W)$, with the expectation taken over the measurement error distributions associated with observed W, and $\begin{pmatrix} W'\Omega^{-1}y_t^* \\ Z'\Omega^{-1}y_t^* \end{pmatrix} = E\begin{pmatrix} W'\Omega^{-1}y_t \\ Z'\Omega^{-1}y_t \end{pmatrix}$, with the expectation taken over the measurement error distributions associated with observed y_t.

Addressing the second issue, both the right-side and left-side variables in the model equation measured with variance contribute to the heteroscedasticity. Although the correction $U'\Omega^{-1}U$ eliminates the bias caused by measurement variance associated with the independent variables, we still do not have a variance-free measure of y for any time period. Therefore, the residual is made up of

$$\bar{\bar{y}} - W'\delta = -U'\delta + v + e$$

where $\bar{\bar{y}} = y - z\tilde{\theta}$, and $\tilde{\theta}$ is the conditional mean of the random effects. The residual variance of any given observation is:

$$\sigma_{ti}^2 = \sigma_e^2 + \sigma_v^2 + \sum_{r=1}^{L} \delta_{t-r}^2 \sigma_{u,t-r(i)}^2$$

where $\sigma_{u,t-r(i)}^2$ is the known measurement variance of r prior test scores. Now, let Ω be a diagonal matrix of dimension N with diagonal elements σ_{ti}^2.

We can now define the mixed model equations as follows:

$$\begin{pmatrix} E(W'\Omega^{-1}W) - E(U'\Omega^{-1}U) & E(W'\Omega^{-1}Z) \\ E(Z'\Omega^{-1}W) & Z'\Omega^{-1}Z + D^{-1} \end{pmatrix}\begin{pmatrix} \delta \\ \theta \end{pmatrix} = E\begin{pmatrix} W'\Omega^{-1}y_t \\ Z'\Omega^{-1}y_t \end{pmatrix}$$

Using observed scores and measurement error variance, the mixed model equations are redefined as follows:

$$\begin{pmatrix} W'\Omega^{-1}W - E(U'\Omega^{-1}U) & W'\Omega^{-1}Z \\ Z'\Omega^{-1}W & Z'\Omega^{-1}Z + D^{-1} \end{pmatrix}\begin{pmatrix} \delta \\ \theta \end{pmatrix} = \begin{pmatrix} W'\Omega^{-1}y_t \\ Z'\Omega^{-1}y_t \end{pmatrix}$$

Observed Values for $E(U'\Omega^{-1}U)$

As indicated, U is unobserved, so solving the mixed model equation cannot be computed unless U is replaced with some observed values. First, the mixed model equations are redefined as

$$\begin{pmatrix} W'\Omega^{-1}W - S & W'\Omega^{-1}Z \\ Z'\Omega^{-1}W & Z'\Omega^{-1}Z + D^{-1} \end{pmatrix}\begin{pmatrix} \delta \\ \theta \end{pmatrix} = \begin{pmatrix} W'\Omega^{-1}y_t \\ Z'\Omega^{-1}y_t \end{pmatrix}$$

where S is a diagonal "correction" matrix with dimensions $p \times p$ accounting for measurement variance in the predictor variables ($p = p_X + L$), and p_X is the column dimension of X.

The matrix S is used in lieu of $E(U'\Omega^{-1}U)$ based on the following justification. Recall that we previously defined Ω as $diag(\sigma_{t1}^2, \sigma_{t2}^2, \dots, \sigma_{tN}^2)$, and the matrix of unobserved disturbances is

$$U = \begin{bmatrix} 0_{pX} & 0 \\ 0 & U_L \end{bmatrix}$$

where 0_{pX} is a matrix of dimension of p_X with elements of 0, and

$$U_L = \begin{bmatrix} u_{11} & u_{12} & \cdots & u_{1L} \\ u_{21} & u_{22} & \cdots & u_{2L} \\ \vdots & \vdots & \ddots & \vdots \\ u_{N1} & u_{N2} & \cdots & u_{NL} \end{bmatrix}$$

The theoretical result of the matrix operation yields the following symmetric matrix:

$$U_L'\Omega^{-1}U_L = \begin{bmatrix} \sum_{i=1}^{N} \frac{1}{\sigma_{ti}^2} u_{i1}^2 & \sum_{i=1}^{N} \frac{1}{\sigma_{ti}^2} u_{i1}u_{i2} & \cdots & \sum_{i=1}^{N} \frac{1}{\sigma_{ti}^2} u_{i1}u_{iL} \\ \sum_{i=1}^{N} \frac{1}{\sigma_{ti}^2} u_{i1}u_{i2} & \sum_{i=1}^{N} \frac{1}{\sigma_{ti}^2} u_{i2}^2 & \cdots & \sum_{i=1}^{N} \frac{1}{\sigma_{ti}^2} u_{i2}u_{iL} \\ \vdots & \vdots & \ddots & \vdots \\ \sum_{i=1}^{N} \frac{1}{\sigma_{ti}^2} u_{i1}u_{iL} & \sum_{i=1}^{N} \frac{1}{\sigma_{ti}^2} u_{i2}u_{iL} & \cdots & \sum_{i=1}^{N} \frac{1}{\sigma_{ti}^2} u_{iL}^2 \end{bmatrix}$$

The theoretical result is limited only because we do not observe u_{ip} because it is latent. However, $E(u_{ip}u_{ip}) = \sigma_{ip}^2$, where σ_{ip}^2 is taken as the square of the conditional standard error of measurement for student i. The theoretical result also simplifies because variances of measurement on different variables are by expectation uncorrelated: $E(u_{ip}u_{ip'}) = 0$ when $p \neq p'$.

Because the conditional standard error of measurement varies for each student i and the off-diagonals can be ignored, let S be

$$S = diag\left(0, \dots, 0, \sum_{i=1}^{N} \frac{1}{\sigma_{ti}^2} \sigma_{u,t-1(i)}^2, \sum_{i=1}^{N} \frac{1}{\sigma_{ti}^2} \sigma_{u,t-2(i)}^2, \dots, \sum_{i=1}^{N} \frac{1}{\sigma_{ti}^2} \sigma_{u,t-L(i)}^2\right)$$

where $\sigma_{u,j(i)}^2$ denotes the measurement variance for the jth $- j = (1,2,\dots,L) -$ variable measured with variance.

Appendix F. Interpolating Standard Errors of Measurement at the Lowest and Highest Obtainable Scale Scores

The linear model used to produce student-level predictions $\hat{y}_i$ can cause these predictions to fall outside the boundaries of the defined scale score. Let the floor and ceiling in the data be denoted as η_f and η_c, respectively. It is, therefore, possible that $\hat{y}_i < \eta_f$ or $\eta_c < \hat{y}_i$. However, the observed score can never fall outside these bounds.

When a prediction falls outside the boundaries of the scale score, it can cause bias in the statistics used to characterize a student, teacher, principal, or school. This phenomenon seems to occur as a result of the large conditional standard errors of measurement at the extreme scores, $csem(\hat{\theta}_i)$. The following procedure is implemented to deal with these large standard errors.

Interpolation Procedure for Conditional Standard Errors of Lowest and Highest Obtainable Scale Scores

Interpolate new conditional standard errors of measurement as the "nearest neighbor" of any extreme value. Thus, at an $M = 2$ cutoff, for the highest obtainable scale score (HOSS) and the score immediately below the HOSS, the SEM associated with the score two below the HOSS would be used. Similarly, the lowest obtainable scale score (LOSS) and the score immediately above the LOSS would have the SEM associated with the score two above the LOSS. As M increases, more points are included, and the point they are set to moves toward the middle of the scale score distribution.

Implement the linear regression using the following steps:

Step 1.   Run the regression without modification.
Step 2.   Verify that $\eta_f \leq \hat{y}_i \leq \eta_c$ for all i.
Step 3.   If the inequality in Step 2 is true, stop; the run is complete. Otherwise, continue to Step 4.
Step 4.   Set $M = 1$ and update the SEMs of the exact HOSS and LOSS scores.
Step 5.   Use the updated $csem(\hat{\theta}_i)$ in lieu of the standard error of the LOSS and HOSS in the test score data.
Step 6.   Run the growth model.
Step 7.   Verify the inequality in Step 2; if it holds, stop updating. If it does not hold, increase M by 1 and return to Step 5.

If this method does not result in the inequality in Step 2 being met after $M = 7$ (i.e., after running with $M = 7$), then simply take the most recent run that did converge, set $\hat{y}_i = \eta_c$ where $\hat{y}_i > \eta_c$, and $\hat{y}_i = \eta_f$ where $\hat{y}_i < \eta_f$. For the predicted variance, use the predicted variance of the closest estimate where the inequality in Step 6 does hold.

Appendix G. Grades 9-12 Data Processing Overview

The process used to convert the raw data to results for 9-12 runs through the same six standardized processes outlined in Appendix C for 4-8. The raw data files used to produce the 9-12 results are explained in greater detail below.

Raw Data

The raw data files that were used in the production of 9-12 growth results this year include:

1. **Regents Assessment Files (2010/11 – 2018/19)** – Student-level results on the Regents Examinations.
2. **Assessment and CSEM (2008/09 – 2018/19)** – Student-level results on the state 3—8 assessments and CSEMs.
3. **New York State English as a Second Language Achievement Test (NYSESLAT)** – Assessment to determine an English language learner's English language proficiency level.
4. **Directory** – Listing of all New York State Public and Nonpublic Schools and the grades served.
5. **Staff Assignment** – Students linked to programs that principals oversee including the start and end dates.
6. **Enrollment (2015/16 – 2018/19)** – Students that were enrolled on Basic Educational Data System (BEDS) day and during the test administration period and demographics information.

Appendix H. Assigning HEDI Ratings and Points

HEDI ratings are assigned according to Figure H 1, shown in the body of the report but repeated here for reference.

Figure H 1. HEDI Rating Rules
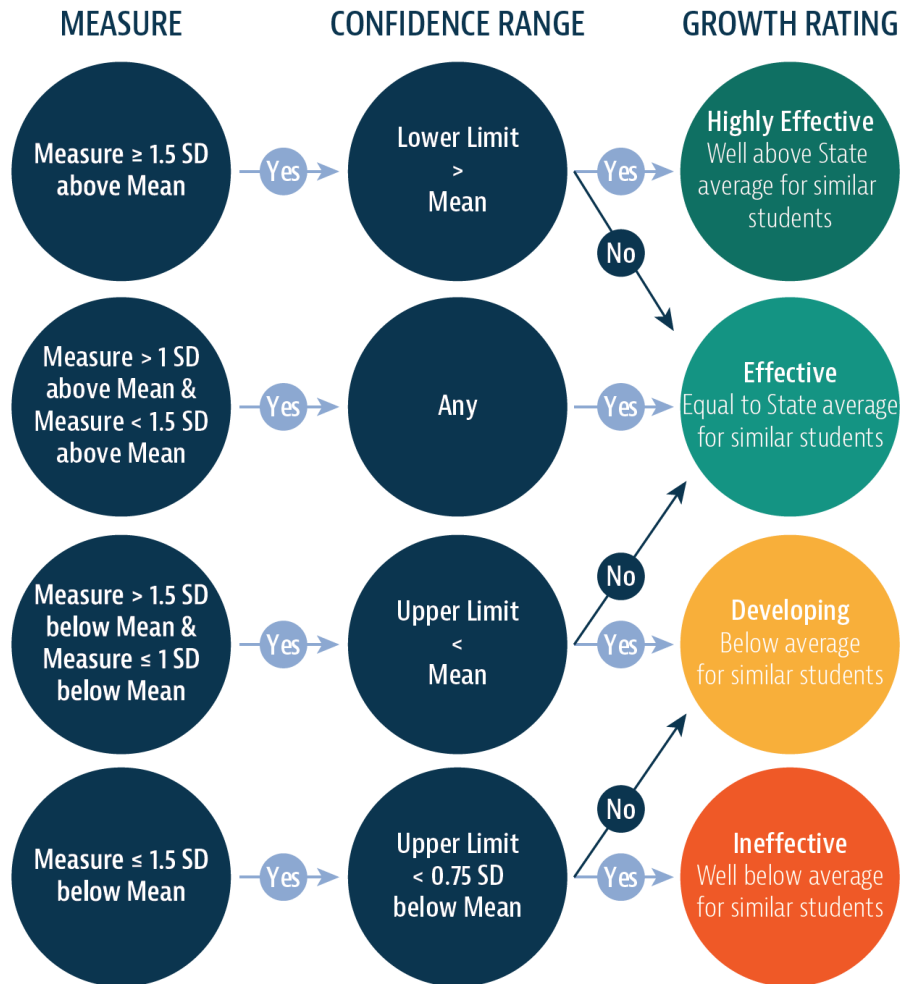
Note: SD = Standard Deviation
Values are rounded to the nearest 0.5 for MGP & 0.1 for GRE.

HEDI ratings are assigned in Grades 4-8 for the combined MGPs (pooled across Grades 4-8 ELA and Grades 4-8 mathematics), in Grades 9-12 for the combined growth model (pooled across ELA and Algebra), and in Grades 9-12 for the GRE model. Values used in 2018/19 to assign HEDI ratings for teachers are shown in Figure H 1; for schools, in Table H 2; and for principals, in Table H 3.

Table H 1. Teacher HEDI Rating Values

Measure	Grades 4-8 Growth Model
Mean	50.897
Standard Deviation	10.632
Highly Effective	67 ≤ MGP and confidence range lower limit > 51
Effective	67 ≤ MGP and confidence range lower limit ≤ 51
Effective	40 < MGP < 67
Effective	35 < MGP ≤ 40 and confidence range upper limit ≥ 51
Developing	35< MGP ≤ 40 and confidence range upper limit < 51
Developing	MGP ≤ 35 and confidence range upper limit ≥ 43
Ineffective	MGP ≤ 35 and confidence range upper limit < 43

Table H 2. School HEDI Rating Values

Measure	Grades 4-8 Growth Model	Grades 9-12 Growth Model	Grades 9-12 GRE Model
Mean	50.320	52.971	0.110
Standard Deviation	5.888	8.499	0.222
Highly Effective	59.0 ≤ MGP and confidence range lower limit > 50.5	66.0 ≤ MGP and confidence range lower limit > 53.0	0.44 ≤ GRE and confidence range lower limit > 0.11
Effective	59.0 ≤ MGP and confidence range lower limit ≤ 50.5	66.0 ≤ MGP and confidence range lower limit ≤ 53.0	0.44 ≤ GRE and confidence range lower limit ≤ 0.11
Effective	44.5 < MGP < 59.0	44.5 < MGP < 66.0	-0.11 < GRE < 0.44
Effective	41.5 < MGP ≤ 44.5 and confidence range upper limit ≥ 50.5	40.0 < MGP ≤ 44.5 and confidence range upper limit ≥ 53.0	-0.22 < GRE ≤ -0.11 and confidence range upper limit ≥ 0.11
Developing	41.5 < MGP ≤ 44.5 and confidence range upper limit < 50.5	40.0 < MGP ≤ 44.5 and confidence range upper limit < 53.0	-0.22 < GRE ≤ -0.11 and confidence range upper limit < 0.11
Developing	MGP ≤ 41.5 and confidence range upper limit ≥ 46.0	MGP ≤ 40.0 and confidence range upper limit ≥ 46.5	GRE ≤ -0.22 and confidence range upper limit ≥ -0.06
Ineffective	MGP ≤ 41.5 and confidence range upper limit < 46.0	MGP ≤ 40.0 and confidence range upper limit < 46.5	GRE ≤ -0.22 and confidence range upper limit < -0.06

Table H 3. Principal HEDI Rating Values

Measure	Grades 4-8 Growth Model	Grades 9-12 Growth Model	Grades 9-12 GRE Model
Mean	50.346	53.134	0.120
Standard Deviation	5.896	8.458	0.217
Highly Effective	59.0 ≤ MGP and confidence range lower limit > 50.5	66.0 ≤ MGP and confidence range lower limit > 53.0	0.44 ≤ GRE and confidence range lower limit > 0.12
Effective	59.0 ≤ MGP and confidence range lower limit ≤ 50.5	66.0 ≤ MGP and confidence range lower limit ≤ 53.0	0.44 ≤ GRE and confidence range lower limit ≤ 0.12
Effective	44.5 < MGP < 59.0	44.5 < MGP < 66.0	-0.10 < GRE < 0.44
Effective	41.5 < MGP ≤ 44.5 and confidence range upper limit ≥ 50.5	40.5 < MGP ≤ 44.5 and confidence range upper limit ≥ 53.0	-0.21< GRE ≤ -0.10 and confidence range upper limit ≥ 0.12
Developing	41.5 < MGP ≤ 44.5 and confidence range upper limit < 50.5	40.5 < MGP ≤ 44.5 and confidence range upper limit < 53.0	-0.21 < GRE ≤ -0.10 and confidence range upper limit < 0.12
Developing	MGP ≤ 41.5 and confidence range upper limit ≥ 46	MGP ≤ 40.5 and confidence range upper limit ≥ 47.0	GRE ≤ -0.21 and confidence range upper limit ≥ -0.04
Ineffective	MGP ≤ 41.5 and confidence range upper limit < 46	MGP ≤ 40.5 and confidence range upper limit < 47.0	GRE ≤ -0.21 and confidence range upper limit < -0.04

Starting from the highest MGP or GRE score in a HEDI category, educators are awarded HEDI points so that those with the highest value on the metric (MGP or GRE) in the rating category receive the highest score. The HEDI growth scores associated with HEDI ratings are shown in Table H 4.

Table H 4. Cut points for HEDI Scores

HEDI Rating[a]	HEDI Score Points	
	Minimum	Maximum
Highly Effective	18	20
Effective	15	17
Developing	13	14
Ineffective	0	12

a: Pursuant to Education Law §3012-d and Subpart 30-3 of the Rules of the Board of Regents, all districts will use the same scoring ranges, determined by the Commissioner, for evaluations conducted pursuant to this subpart.

Scores lower than the highest score are assigned so that at any HEDI score point, the number of educators with that HEDI score or higher is proportional to (or larger than) the proportion of score points in that category that are at least as large as the score point. For example, the HEDI

rating Highly Effective is associated with HEDI score points 18, 19, and 20. For educators who receive a HEDI rating of Highly Effective, at least one third of them will receive 20 HEDI score points, and at least two thirds of them will receive 19 or 20 HEDI score points.

The tables that follow display the observed minimum and maximum MGP and GRE scores for the Grades 4-8 and 9-12 MGP and GRE models.

Table H 5. Grades 4-8 Teacher, School, and Principal MGP HEDI Point Distribution

HEDI Points	Teacher		School		Principal	
	Minimum	Maximum	Minimum	Maximum	Minimum	Maximum
0	9	22	24.5	31.0	24.5	32.0
1	23	24	32.0	33.5	32.5	34.5
2	25	25	34.0	35.5	35.0	36.0
3	26	26	36.0	36.5	36.5	37.0
4	27	27	37.0	37.0	37.5	37.5
5	28	28	37.5	37.5	38.0	38.0
6	29	29	38.0	38.0	38.5	38.5
7	30	30	38.5	38.5	39.0	39.0
8	31	31	39.0	39.0	39.5	39.5
9	32	32	39.5	39.5	40.0	40.0
10	33	33	40.0	40.0	40.5	40.5
11	34	34	40.5	40.5	41.0	41.0
12	35	35	41.0	41.0	41.5	42.0
13	28	37	33.0	42.5	36.5	42.5
14	38	40	43.0	44.0	43.0	44.0
15	36	47	41.5	48.5	42.5	48.5
16	48	54	49.0	52.5	49.0	52.5
17	55	68	53.0	63.5	53.0	59.5
18	67	68	59.0	59.5	59.0	59.5
19	69	72	60.0	61.5	60.0	62.0
20	73	95	62.0	74.5	62.5	74.5

Table H 6. Grades 9-12 School and Principal MGP HEDI Point Distribution

HEDI Points	School		Principal	
	Minimum	Maximum	Minimum	Maximum
0	21.5	26.0	23.0	25.0
1	26.5	29.5	26.5	28.5
2	30.5	31.0	29.0	31.5
3	32.5	33.5	32.5	32.5
4	34.0	34.5	33.0	34.0
5	35.0	35.5	34.5	35.0
6	36.0	36.5	35.5	35.5
7	37.0	37.0	36.0	37.0
8	37.5	38.0	37.5	37.5
9	38.5	38.5	38.0	38.5
10	39.0	39.0	39.0	39.0
11	39.5	39.5	39.5	39.5
12	40.0	40.0	40.0	40.0
13	33.0	41.5	33.0	42.5
14	42.0	44.0	43.0	45.0
15	40.5	50.5	41.5	51.0
16	51.0	56.5	51.5	57.0
17	57.0	67.0	57.5	67.0
18	66.0	67.0	66.0	67.0
19	67.5	69.5	67.5	69.5
20	70.0	81.5	70.0	81.5

Table H 7. Grades 9-12 School and Principal GRE HEDI Point Distribution

HEDI Points	School		Principal	
	Minimum	Maximum	Minimum	Maximum
0	-0.71	-0.59	-0.70	-0.59
1	-0.57	-0.52	-0.53	-0.50
2	-0.51	-0.48	-0.49	-0.47
3	-0.46	-0.42	-0.43	-0.40

HEDI Points	School		Principal	
	Minimum	Maximum	Minimum	Maximum
4	-0.40	-0.38	-0.39	-0.36
5	-0.37	-0.36	--0.35	-0.32
6	-0.35	-0.32	-0.31	-0.30
7	-0.31	-0.30	-0.29	-0.29
8	-0.29	-0.29	-0.28	-0.27
9	-0.28	-0.27	-0.26	-0.25
10	-0.26	-0.26	-0.24	-0.24
11	-0.25	-0.25	-0.23	-0.22
12	-0.24	-0.22	-0.21	-0.21
13	-0.49	-0.17	-0.49	-0.16
14	-0.16	-0.11	-0.15	-0.10
15	-0.21	0.08	-0.20	0.09
16	0.09	0.22	0.10	0.22
17	0.23	0.55	0.23	0.55
18	0.44	0.49	0.44	0.48
19	0.50	0.56	0.49	0.55
20	0.57	1.19	0.56	1.19

When an educator has at least 16 attributed students or student scores in only one of the three growth measures (Grades 4-8 MGP, Grades 9-12 MGP, and Grades 9-12 GRE), then the HEDI rating and score based on that growth measure serves as the educator's final HEDI rating and score. However, most Grades 9-12 principals and schools have multiple HEDI ratings and scores (Grades 9-12 MGP and Grades 9-12 GRE), and some principals and schools may have multiple HEDI ratings and scores if they serve students in Grades 4-8 and 9-12. HEDI ratings and scores from the 9-12 MGP and 9-12 GRE models are first combined to create an overall Grades 9-12 HEDI rating and score. HEDI ratings and scores from the Grades 4-8 MGP model and the overall Grades 9-12 HEDI rating and score are then combined to obtain a final overall rating.

To combine HEDI ratings and scores, we used the following procedure, pooling all educators at a given level (principals or schools) across the State into a single group and using only their appropriate HEDI score from the column labeled "HEDI Score Points" in Table H 4.

> Step 1.   Find the aggregate HEDI growth score using the following equation:

$$G = \frac{n_A G_A + n_B G_B}{n_A + n_B}$$

where G is the growth score, n is the number of students attributed to a school, the subscript A is one of the two HEDI scores being combined, and the subscript B is the other HEDI score being combined. If either of the HEDI scores is not assigned because n was not at least 16, simply set G equal to the assigned HEDI score and continue. For example, if only n_A is greater than or equal to 16 ($n_A \geq 16, n_B < 16$), then $G = G_A$.

The same also holds if A and B are switched in the example. Also, if neither HEDI scores was assigned ($n_A < 16, n_B < 16$), set G to missing and do not include in the final HEDI score.

Step 2.   Round G to the nearest integer. This integer is the HEDI score for the combination.

Step 3.   For *all* principals and schools, assign a final HEDI rating by using the cut points table, assigning the HEDI rating associated with each school or principal's final rounded HEDI point value (G from Step 2) based on the column labeled "HEDI Score Points".

Step 4.   Every principal and school with two HEDI ratings and scores to combine is assigned a 3012-d HEDI rating and score by applying the rules for assigning scores described previously to the unrounded value of G found in Step 1. 3012-d ratings are then reported only to educators in relevant districts.

Appendix I. Model Coefficients

The tables that follow display regression model coefficients (labeled as "Effects") for the New York growth model in each grade and subject. For the Grades 4-8 model and the Grades 9-12 MGP model, these model coefficients represent the predicted change in current year test scores for one unit of change in each variable shown in the table, holding other variables constant. For example, in Table I 2, holding all other variables constant, the predicted change in a student's current year ELA test score given a one-point increase in a student's prior grade ELA test score is 0.675. The interpretation of a one-unit change varies by variable type. For yes/no variables, model coefficients represent the predicted change in current year test scores given a change from no to yes. For example, in Table I 2, holding all other variables constant, the predicted difference in a student's current year ELA test score if the student has a disability (versus a student without a disability) is -2.285 points. Missing flags also are yes/no variables set to yes if the noted variable is missing and no otherwise. Variables that are percentages are on a scale from 0 to 100 and represent the change in prediction for a single percentage point increase.

Because the GRE model has a different form (an ordered logistic regression) than the MGP model, GRE model coefficients (labeled as "Estimates") are not interpretable as linear changes in the outcome given a one-unit change in a predictor. Instead, the predicted number of Regents Exams passed varies according to the equations in the section titled "Comparative Growth in Regents Exams Passed Model." For example, in Table I 29, because the coefficient is positive, an increase in the Grade 8 ELA scale scores from 2012/13 and subsequent years is associated with a higher number of GRE Exams passed in the current year. Larger positive coefficients indicate larger predicted increases in the number of Regents Exams passed in the current year per unit change in the predictor variable. Predictor variables with fewer than 10 cases in the GRE models were dropped from analysis.

Because of the differences in model and variable types, it is important to keep in mind that effect sizes cannot be compared directly across different types of variables.

Table I 1. Grade 4 ELA Model Coefficients, Unadjusted Model

Effect Name	Effect	Standard Error	*p*-value
Constant Term	82.236	1.163	0.000
Prior-Grade ELA Scale Score	0.861	0.002	0.000

Table I 2. Grade 4 ELA Model Coefficients, Adjusted Model

Effect Name	Effect	Standard Error	*p*-value
Constant Term	50.069	2.558	0.000
Mean Prior Score	0.057	0.004	0.000
Range Around Prior Score	-0.002	0.005	0.655
Prior-Grade ELA Scale Score	0.675	0.004	0.000
Prior-Grade Mathematics Scale Score	0.140	0.004	0.000
Missing Flag: Prior-Grade Mathematics Scale Score	82.967	2.212	0.000
Grades 3-4 NYSESLAT Scale Score	0.085	0.003	0.000
Missing Flag: Grades 3-4 NYSESLAT Scale Scores	26.796	1.077	0.000
New to School	-0.434	0.111	0.000
Missing Flag: Percentage Variables	33.646	2.768	0.000
General Education < 40% (LRE3)	-2.352	0.222	0.000
ELLs	3.230	0.246	0.000
Percentage of ELLs	0.027	0.002	0.000
Economically Disadvantaged	-0.903	0.076	0.000
Percentage of Economically Disadvantaged	-0.008	0.002	0.000
Students with Disabilities	-2.825	0.106	0.000
Percentage of Students with Disabilities	0.018	0.002	0.000

Table I 3. Grade 5 ELA Model Coefficients, Unadjusted Model

Effect Name	Effect	Standard Error	*p*-value
Constant Term	115.105	1.868	0.000
Prior-Grade ELA Scale Score	0.729	0.004	0.000
Two-Grades-Prior ELA Scale Score	0.153	0.002	0.000
Missing Flag: Two-Grades-Prior ELA Scale Score	46.008	0.700	0.000

Table I 4. Grade 5 ELA Model Coefficients, Adjusted Model

Effect Name	Effect	Standard Error	*p*-value
Constant Term	111.426	2.731	0.000
Mean Prior Score	-0.070	0.004	0.000
Range Around Prior Score	-0.001	0.005	0.911
Prior-Grade ELA Scale Score	0.648	0.005	0.000
Prior-Grade Mathematics Scale Score	0.112	0.004	0.000
Missing Flag: Prior-Grade Mathematics Scale Score	65.665	2.220	0.000
Retained in Grade	-0.538	0.292	0.065
Two-Grades-Prior ELA Scale Score	0.138	0.002	0.000
Missing Flag: Two-Grades-Prior ELA Scale Score	41.760	0.704	0.000
Grades 3-4 NYSESLAT Scale Score	0.113	0.004	0.000
Missing Flag: Grades 3-4 NYSESLAT Scale Scores	33.877	1.177	0.000
New to School	-0.160	0.119	0.177
Missing Flag: Percentage Variables	-44.889	2.760	0.000
General Education < 40% (LRE3)	-1.105	0.223	0.000
ELLs	3.043	0.273	0.000
Percentage of ELLs	0.001	0.003	0.740
Economically Disadvantaged	-0.261	0.078	0.001
Percentage of Economically Disadvantaged	-0.019	0.002	0.000
Students with Disabilities	-1.081	0.108	0.000
Percentage of Students with Disabilities	-0.011	0.002	0.000

Table I 5. Grade 6 ELA Model Coefficients, Unadjusted Model

Effect Name	Effect	Standard Error	*p*-value
Constant Term	99.725	2.122	0.000
Prior-Grade ELA Scale Score	0.737	0.005	0.000
Three-Grades-Prior ELA Scale Score	0.073	0.002	0.000
Missing Flag: Three-Grades-Prior ELA Scale Score	22.922	0.753	0.000
Two-Grades-Prior ELA Scale Score	0.110	0.003	0.000
Missing Flag: Two-Grades-Prior ELA Scale Score	32.511	0.878	0.000

Table I 6. Grade 6 ELA Model Coefficients, Adjusted Model

Effect Name	Effect	Standard Error	*p*-value
Constant Term	44.881	3.383	0.000
Mean Prior Score	0.046	0.005	0.000
Range Around Prior Score	0.021	0.007	0.001
Prior-Grade ELA Scale Score	0.637	0.006	0.000
Prior-Grade Mathematics Scale Score	0.118	0.003	0.000
Missing Flag: Prior-Grade Mathematics Scale Score	68.505	1.985	0.000
Retained in Grade	-2.588	0.377	0.000
Three-Grades-Prior ELA Scale Score	0.062	0.002	0.000
Missing Flag: Three-Grades-Prior ELA Scale Score	19.564	0.757	0.000
Two-Grades-Prior ELA Scale Score	0.101	0.003	0.000
Missing Flag: Two-Grades-Prior ELA Scale Score	29.628	0.856	0.000
Grades 5-6 NYSESLAT Scale Score	0.082	0.005	0.000
Missing Flag: Grades 5-6 NYSESLAT Scale Scores	22.742	1.483	0.000
New to School	0.503	0.138	0.000
Missing Flag: Percentage Variables	27.685	3.198	0.000
General Education < 40% (LRE3)	0.362	0.235	0.123
ELLs	1.700	0.358	0.000
Percentage of ELLs	0.020	0.003	0.000
Economically Disadvantaged	-0.868	0.081	0.000
Percentage of Economically Disadvantaged	0.006	0.002	0.000
Students with Disabilities	-0.992	0.114	0.000
Percentage of Students with Disabilities	-0.005	0.002	0.059

Table I 7. Grade 7 ELA Model Coefficients, Unadjusted Model

Effect Name	Effect	Standard Error	*p*-value
Constant Term	119.786	1.809	0.000
Prior-Grade ELA Scale Score	0.723	0.004	0.000
Three-Grades-Prior ELA Scale Score	0.069	0.002	0.000
Missing Flag: Three-Grades-Prior ELA Scale Score	20.975	0.723	0.000
Two-Grades-Prior ELA Scale Score	0.084	0.002	0.000
Missing Flag: Two-Grades-Prior ELA Scale Score	23.935	0.696	0.000

Table I 8. Grade 7 ELA Model Coefficients, Adjusted Model

Effect Name	Effect	Standard Error	*p*-value
Constant Term	12.881	2.923	0.000
Mean Prior Score	0.109	0.005	0.000
Range Around Prior Score	0.005	0.006	0.399
Prior-Grade ELA Scale Score	0.597	0.005	0.000
Prior-Grade Mathematics Scale Score	0.152	0.003	0.000
Missing Flag: Prior-Grade Mathematics Scale Score	89.574	1.989	0.000
Retained in Grade	-1.594	0.343	0.000
Three-Grades-Prior ELA Scale Score	0.056	0.002	0.000
Missing Flag: Three-Grades-Prior ELA Scale Score	17.397	0.722	0.000
Two-Grades-Prior ELA Scale Score	0.082	0.002	0.000
Missing Flag: Two-Grades-Prior ELA Scale Score	23.494	0.678	0.000
Grades 5-6 NYSESLAT Scale Score	0.097	0.004	0.000
Missing Flag: Grades 5-6 NYSESLAT Scale Scores	28.413	1.304	0.000
New to School	-0.463	0.117	0.000
Missing Flag: Percentage Variables	68.499	2.828	0.000
General Education < 40% (LRE3)	-0.467	0.222	0.035
ELLs	2.540	0.297	0.000
Percentage of ELLs	0.027	0.003	0.000
Economically Disadvantaged	-0.003	0.072	0.966
Percentage of Economically Disadvantaged	0.030	0.002	0.000
Students with Disabilities	-0.485	0.101	0.000
Percentage of Students with Disabilities	0.025	0.002	0.000

Table I 9. Grade 8 ELA Model Coefficients, Unadjusted Model

Effect Name	Effect	Standard Error	*p*-value
Constant Term	148.611	1.825	0.000
Prior-Grade ELA Scale Score	0.685	0.004	0.000
Three-Grades-Prior ELA Scale Score	0.037	0.002	0.000
Missing Flag: Three-Grades-Prior ELA Scale Score	11.179	0.564	0.000
Two-Grades-Prior ELA Scale Score	0.099	0.003	0.000
Missing Flag: Two-Grades-Prior ELA Scale Score	27.894	0.733	0.000

Table I 10. Grade 8 ELA Model Coefficients, Adjusted Model

Effect Name	Effect	Standard Error	*p*-value
Constant Term	77.756	2.806	0.000
Mean Prior Score	0.045	0.004	0.000
Range Around Prior Score	-0.002	0.006	0.768
Prior-Grade ELA Scale Score	0.579	0.005	0.000
Prior-Grade Mathematics Scale Score	0.137	0.003	0.000
Missing Flag: Prior-Grade Mathematics Scale Score	81.368	2.093	0.000
Retained in Grade	-1.160	0.322	0.000
Three-Grades-Prior ELA Scale Score	0.037	0.002	0.000
Missing Flag: Three-Grades-Prior ELA Scale Score	11.052	0.565	0.000
Two-Grades-Prior ELA Scale Score	0.090	0.003	0.000
Missing Flag: Two-Grades-Prior ELA Scale Score	25.424	0.734	0.000
Grades 7-8 NYSESLAT Scale Score	0.091	0.004	0.000
Missing Flag: Grades 7-8 NYSESLAT Scale Scores	26.048	1.269	0.000
New to School	-0.095	0.137	0.488
Missing Flag: Percentage Variables	29.641	2.756	0.000
General Education < 40% (LRE3)	-0.075	0.225	0.739
ELLs	1.965	0.317	0.000
Percentage of ELLs	0.018	0.003	0.000
Economically Disadvantaged	-0.225	0.072	0.002
Percentage of Economically Disadvantaged	0.034	0.002	0.000
Students with Disabilities	-0.391	0.101	0.000
Percentage of Students with Disabilities	0.006	0.002	0.011

Table I 11. Grade 4 Mathematics Model Coefficients, Unadjusted Model

Effect Name	Effect	Standard Error	*p*-value
Constant Term	36.566	1.109	0.000
Prior-Grade Mathematics Scale Score	0.940	0.002	0.000

Table I 12. Grade 4 Mathematics Model Coefficients, Adjusted Model

Effect Name	Effect	Standard Error	*p*-value
Constant Term	50.035	2.362	0.000
Mean Prior Score	0.001	0.004	0.899
Range Around Prior Score	-0.016	0.005	0.001
Prior-Grade ELA Scale Score	0.078	0.004	0.000
Missing Flag: Prior-Grade ELA Scale Score	47.299	2.162	0.000
Prior-Grade Mathematics Scale Score	0.837	0.004	0.000
Grades 3-4 NYSESLAT Scale Score	0.016	0.003	0.000
Missing Flag: Grades 3-4 NYSESLAT Scale Scores	4.152	0.880	0.000
New to School	-0.780	0.108	0.000
Missing Flag: Percentage Variables	-2.835	2.541	0.265
General Education < 40% (LRE3)	-1.648	0.223	0.000
ELLs	0.556	0.230	0.016
Percentage of ELLs	0.017	0.002	0.000
Economically Disadvantaged	-0.868	0.074	0.000
Percentage of Economically Disadvantaged	-0.025	0.002	0.000
Students with Disabilities	-1.644	0.102	0.000
Percentage of Students with Disabilities	-0.005	0.002	0.021

Table I 13. Grade 5 Mathematics Model Coefficients, Unadjusted Model

Effect Name	Effect	Standard Error	*p*-value
Constant Term	115.991	1.692	0.000
Prior-Grade Mathematics Scale Score	0.750	0.004	0.000
Two-Grades-Prior Mathematics Scale Score	0.113	0.002	0.000
Missing Flag: Two-Grades-Prior Mathematics Scale Score	34.155	0.553	0.000

Table I 14. Grade 5 Mathematics Model Coefficients, Adjusted Model

Effect Name	Effect	Standard Error	*p*-value
Constant Term	101.927	2.441	0.000
Mean Prior Score	0.022	0.004	0.000
Range Around Prior Score	-0.016	0.005	0.001
Prior-Grade ELA Scale Score	0.037	0.003	0.000
Missing Flag: Prior-Grade ELA Scale Score	22.139	1.926	0.000
Prior-Grade Mathematics Scale Score	0.705	0.005	0.000
Retained in Grade	-1.879	0.267	0.000
Two-Grades-Prior Mathematics Scale Score	0.108	0.002	0.000
Missing Flag: Two-Grades-Prior Mathematics Scale Score	32.673	0.562	0.000
Grades 3-4 NYSESLAT Scale Score	0.033	0.003	0.000
Missing Flag: Grades 3-4 NYSESLAT Scale Scores	7.781	0.888	0.000
New to School	-0.374	0.109	0.001
Missing Flag: Percentage Variables	12.718	2.405	0.000
General Education < 40% (LRE3)	-1.901	0.214	0.000
ELLs	0.039	0.239	0.872
Percentage of ELLs	0.020	0.002	0.000
Economically Disadvantaged	-0.250	0.071	0.000
Percentage of Economically Disadvantaged	-0.006	0.001	0.000
Students with Disabilities	-1.384	0.098	0.000
Percentage of Students with Disabilities	0.004	0.002	0.061

Table I 15. Grade 6 Mathematics Model Coefficients, Unadjusted Model

Effect Name	Effect	Standard Error	*p*-value
Constant Term	191.300	1.676	0.000
Prior-Grade Mathematics Scale Score	0.597	0.004	0.000
Three-Grades-Prior Mathematics Scale Score	0.053	0.002	0.000
Missing Flag: Three-Grades-Prior Mathematics Scale Score	16.415	0.544	0.000
Two-Grades-Prior Mathematics Scale Score	0.116	0.002	0.000
Missing Flag: Two-Grades-Prior Mathematics Scale Score	34.122	0.610	0.000

Table I 16. Grade 6 Mathematics Model Coefficients, Adjusted Model

Effect Name	Effect	Standard Error	*p*-value
Constant Term	155.837	2.697	0.000
Mean Prior Score	0.041	0.004	0.000
Range Around Prior Score	0.009	0.006	0.103
Prior-Grade ELA Scale Score	0.100	0.003	0.000
Missing Flag: Prior-Grade ELA Scale Score	59.397	1.906	0.000
Prior-Grade Mathematics Scale Score	0.521	0.004	0.000
Retained in Grade	-3.505	0.324	0.000
Three-Grades-Prior Mathematics Scale Score	0.044	0.002	0.000
Missing Flag: Three-Grades-Prior Mathematics Scale Score	14.076	0.545	0.000
Two-Grades-Prior Mathematics Scale Score	0.101	0.002	0.000
Missing Flag: Two-Grades-Prior Mathematics Scale Score	29.483	0.606	0.000
Grades 5-6 NYSESLAT Scale Score	0.024	0.003	0.000
Missing Flag: Grades 5-6 NYSESLAT Scale Scores	4.910	0.996	0.000
New to School	0.392	0.120	0.001
Missing Flag: Percentage Variables	25.126	2.442	0.000
General Education < 40% (LRE3)	0.641	0.207	0.002
ELLs	-0.801	0.292	0.006
Percentage of ELLs	0.005	0.003	0.070
Economically Disadvantaged	-0.307	0.069	0.000
Percentage of Economically Disadvantaged	-0.014	0.002	0.000
Students with Disabilities	-1.970	0.096	0.000
Percentage of Students with Disabilities	-0.007	0.002	0.001

Table I 17. Grade 7 Mathematics Model Coefficients, Unadjusted Model

Effect Name	Effect	Standard Error	*p*-value
Constant Term	61.355	2.175	0.000
Prior-Grade Mathematics Scale Score	0.861	0.005	0.000
Three-Grades-Prior Mathematics Scale Score	0.023	0.002	0.000
Missing Flag: Three-Grades-Prior Mathematics Scale Score	7.411	0.598	0.000
Two-Grades-Prior Mathematics Scale Score	0.053	0.003	0.000
Missing Flag: Two-Grades-Prior Mathematics Scale Score	15.736	0.778	0.000

Table I 18. Grade 7 Mathematics Model Coefficients, Adjusted Model

Effect Name	Effect	Standard Error	*p*-value
Constant Term	13.095	2.896	0.000
Mean Prior Score	0.065	0.004	0.000
Range Around Prior Score	0.026	0.006	0.000
Prior-Grade ELA Scale Score	0.111	0.003	0.000
Missing Flag: Prior-Grade ELA Scale Score	65.712	1.901	0.000
Prior-Grade Mathematics Scale Score	0.761	0.006	0.000
Retained in Grade	-2.620	0.347	0.000
Three-Grades-Prior Mathematics Scale Score	0.019	0.002	0.000
Missing Flag: Three-Grades-Prior Mathematics Scale Score	6.580	0.596	0.000
Two-Grades-Prior Mathematics Scale Score	0.045	0.003	0.000
Missing Flag: Two-Grades-Prior Mathematics Scale Score	13.283	0.765	0.000
Grades 5-6 NYSESLAT Scale Score	0.023	0.003	0.000
Missing Flag: Grades 5-6 NYSESLAT Scale Scores	4.642	1.104	0.000
New to School	-1.151	0.119	0.000
Missing Flag: Percentage Variables	41.199	2.664	0.000
General Education < 40% (LRE3)	-1.730	0.246	0.000
ELLs	0.272	0.286	0.343
Percentage of ELLs	0.017	0.003	0.000
Economically Disadvantaged	-0.042	0.071	0.552
Percentage of Economically Disadvantaged	0.010	0.002	0.000
Students with Disabilities	-1.294	0.102	0.000
Percentage of Students with Disabilities	0.005	0.002	0.055

Table I 19. Grade 8 Mathematics Model Coefficients, Unadjusted Model

Effect Name	Effect	Standard Error	*p*-value
Constant Term	35.605	3.484	0.000
Prior-Grade Mathematics Scale Score	0.917	0.007	0.000
Three-Grades-Prior Mathematics Scale Score	0.000	0.003	0.972
Missing Flag: Three-Grades-Prior Mathematics Scale Score	0.443	0.838	0.597
Two-Grades-Prior Mathematics Scale Score	0.061	0.004	0.000
Missing Flag: Two-Grades-Prior Mathematics Scale Score	17.919	1.064	0.000

Table I 20. Grade 8 Mathematics Model Coefficients, Adjusted Model

Effect Name	Effect	Standard Error	*p*-value
Constant Term	-9.387	4.332	0.030
Mean Prior Score	0.073	0.007	0.000
Range Around Prior Score	0.062	0.010	0.000
Prior-Grade ELA Scale Score	0.065	0.005	0.000
Missing Flag: Prior-Grade ELA Scale Score	39.187	2.773	0.000
Prior-Grade Mathematics Scale Score	0.844	0.009	0.000
Retained in Grade	-4.281	0.448	0.000
Three-Grades-Prior Mathematics Scale Score	0.001	0.003	0.748
Missing Flag: Three-Grades-Prior Mathematics Scale Score	1.058	0.837	0.206
Two-Grades-Prior Mathematics Scale Score	0.058	0.004	0.000
Missing Flag: Two-Grades-Prior Mathematics Scale Score	16.622	1.070	0.000
Grades 7-8 NYSESLAT Scale Score	0.027	0.005	0.000
Missing Flag: Grades 7-8 NYSESLAT Scale Scores	5.333	1.435	0.000
New to School	-1.728	0.188	0.000
Missing Flag: Percentage Variables	44.942	4.086	0.000
General Education < 40% (LRE3)	-0.985	0.319	0.002
ELLs	-0.030	0.404	0.940
Percentage of ELLs	0.015	0.004	0.000
Economically Disadvantaged	0.609	0.101	0.000
Percentage of Economically Disadvantaged	0.006	0.002	0.014
Students with Disabilities	-0.632	0.136	0.000
Percentage of Students with Disabilities	-0.011	0.003	0.000

Table I 21. Grades 9-12, Algebra Model Coefficients, Unadjusted Model

Effect Name	Effect	Standard Error	*p*-value
Constant Term	-368.210	46.174	0.000
Grade 7 Mathematics Scale Score 2011/12 and Prior	0.014	0.018	0.443
Missing Flag: Grade 7 Mathematics Scale Score 2011/12 and Prior	15.922	11.517	0.167
Grade 7 Mathematics Scale Score 2012/13 to 2016/17	0.153	0.002	0.000
Missing Flag: Grade 7 Mathematics Scale Score 2012/13 to 2016/17	45.863	0.617	0.000
Grade 8 ELA Scale Score 2011/12 and Prior	-0.050	0.092	0.584
Missing Flag: Grade 8 ELA Scale Score 2011/12 and Prior	-30.322	58.030	0.601
Grade 8 ELA Scale Score 2012/13 to 2016/17	0.000	0.003	0.902
Missing Flag: Grade 8 ELA Scale Score 2012/13 to 2016/17	2.284	0.778	0.003
Grade 8 ELA Scale Score 2017/18 and Subsequent	0.026	0.003	0.000
Missing Flag: Grade 8 ELA Scale Score 2017/18 and Subsequent	16.747	1.725	0.000
Grade 8 Mathematics Scale Score 2011/12 and Prior	0.144	0.055	0.009
Missing Flag: Grade 8 Mathematics Scale Score 2011/12 and Prior	95.777	34.954	0.006
Grade 8 Mathematics Scale Score 2012/13 to 2016/17	0.187	0.005	0.000
Missing Flag: Grade 8 Mathematics Scale Score 2012/13 to 2016/17	52.317	1.384	0.000
Grade 8 Mathematics Scale Score 2017/18 and Subsequent	0.405	0.005	0.000
Missing Flag: Grade 8 Mathematics Scale Score 2017/18 and Subsequent	240.879	2.783	0.000

Table I 22. Grades 9-12, Algebra Model Coefficients, Adjusted Model

Effect Name	Effect	Standard Error	*p*-value
Cohort 1	7.490	0.378	0.000
Cohort 2	2.858	0.347	0.000
Cohort 3	-0.030	0.348	0.932
Cohort 4	-0.736	0.353	0.037
Count of Prior Required Regents Exams = 0	-223.979	70.231	0.001
Count of Prior Required Regents Exams = 1	-221.434	70.231	0.002
Count of Prior Required Regents Exams = 2	-220.471	70.231	0.002
Count of Prior Required Regents Exams = 3	-219.716	70.234	0.002
Count of Prior Required Regents Exams = 4	-219.444	70.242	0.002
Count of Prior Required Regents Exams = 5	-219.810	70.238	0.002
Grade 7 ELA Scale Score 2011/12 and Prior	-0.044	0.034	0.198
Missing Flag: Grade 7 ELA Scale Score 2011/12 and Prior	-26.523	21.940	0.227
Grade 7 ELA Scale Score 2012/13 to 2016/17	0.008	0.002	0.000
Missing Flag: Grade 7 ELA Scale Score 2012/13 to 2016/17	3.377	0.568	0.000

Grade 7 Mathematics Scale Score 2011/12 and Prior	0.018	0.022	0.416
Missing Flag: Grade 7 Mathematics Scale Score 2011/12 and Prior	13.725	14.271	0.336
Grade 7 Mathematics Scale Score 2012/13 to 2016/17	0.101	0.003	0.000
Missing Flag: Grade 7 Mathematics Scale Score 2012/13 to 2016/17	29.144	0.753	0.000
Grade 8 ELA Scale Score 2011/12 and Prior	-0.082	0.123	0.503
Missing Flag: Grade 8 ELA Scale Score 2011/12 and Prior	-55.502	77.419	0.473
Grade 8 ELA Scale Score 2012/13 to 2016/17	-0.018	0.003	0.000
Missing Flag: Grade 8 ELA Scale Score 2012/13 to 2016/17	-4.285	0.811	0.000
Grade 8 ELA Scale Score 2017/18 and Subsequent	0.021	0.003	0.000
Missing Flag: Grade 8 ELA Scale Score 2017/18 and Subsequent	14.147	2.024	0.000
Grade 8 Mathematics Scale Score 2011/12 and Prior	0.132	0.073	0.071
Missing Flag: Grade 8 Mathematics Scale Score 2011/12 and Prior	87.798	46.388	0.058
Grade 8 Mathematics Scale Score 2012/13 to 2016/17	0.178	0.005	0.000
Missing Flag: Grade 8 Mathematics Scale Score 2012/13 to 2016/17	46.469	1.364	0.000
Grade 8 Mathematics Scale Score 2017/18 and Subsequent	0.393	0.005	0.000
Missing Flag: Grade 8 Mathematics Scale Score 2017/18 and Subsequent	233.647	2.962	0.000
Mean Prior Grade 8 Mathematics 2011/12 and Prior	0.001	0.001	0.370
Missing Flag: Mean Prior Grade 8 Mathematics 2011/12 and Prior	0.919	0.784	0.241
Mean Prior Grade 8 Mathematics 2012/13 to 2016/17	-0.008	0.003	0.006
Missing Flag: Mean Prior Grade 8 Mathematics 2012/13 to 2016/17	-0.611	0.873	0.484
Mean Prior Grade 8 Mathematics 2017/18 and Subsequent	-0.017	0.005	0.001
Missing Flag: Mean Prior Grade 8 Mathematics 2017/18 and Subsequent	-10.308	3.153	0.001
NYSESLAT LS Scale Score 2011/12 and Prior	0.062	0.080	0.440
NYSESLAT RW Scale Score 2011/12 and Prior	-0.110	0.103	0.287
NYSESLAT Scale Score 2012/13 and 2013/14	-0.015	0.014	0.313
NYSESLAT Scale Score 2012/13 and 2014/15	0.043	0.002	0.000
Missing Flag: NYSESLAT Scale Scores 2011/12 and Prior	-32.404	38.998	0.406
Missing Flag: NYSESLAT Scale Scores 2012/13 and 2013/14	-13.286	12.065	0.271
Missing Flag: NYSESLAT Scale Scores 2014/15 and Later	11.506	0.638	0.000
New to School After Grade 9	-0.091	0.138	0.509
Missing Flag: School Percentage Variables	-14.084	0.376	0.000
General Education < 40% (LRE3)	-3.691	0.207	0.000
ELLs	-0.075	0.170	0.657
School Percentage of ELLs	0.015	0.003	0.000

Economically Disadvantaged	-0.490	0.052	0.000
School Percentage of Economically Disadvantaged	-0.095	0.001	0.000
Students with Disabilities	-2.675	0.065	0.000
School Percentage of Students with Disabilities	-0.120	0.003	0.000

Table I 23. Grades 9-12, ELA Model Coefficients, Unadjusted Model

Effect Name	Effect	Standard Error	*p*-value
Constant Term	-739.928	44.785	0.000
Grade 7 ELA Scale Score 2011/12 and Prior	0.153	0.031	0.000
Missing Flag: Grade 7 ELA Scale Score 2011/12 and Prior	107.287	19.588	0.000
Grade 7 ELA Scale Score 2012/13 to 2016/17	0.176	0.002	0.000
Missing Flag: Grade 7 ELA Scale Score 2012/13 to 2016/17	51.997	0.441	0.000
Grade 8 ELA Scale Score 2011/12 and Prior	0.375	0.095	0.000
Missing Flag: Grade 8 ELA Scale Score 2011/12 and Prior	253.432	58.722	0.000
Grade 8 ELA Scale Score 2012/13 to 2016/17	0.135	0.002	0.000
Missing Flag: Grade 8 ELA Scale Score 2012/13 to 2016/17	41.614	0.569	0.000
Grade 8 ELA Scale Score 2017/18 and Subsequent	0.515	0.018	0.000
Missing Flag: Grade 8 ELA Scale Score 2017/18 and Subsequent	325.631	11.098	0.000
Grade 8 Mathematics Scale Score 2011/12 and Prior	-0.012	0.057	0.833
Missing Flag: Grade 8 Mathematics Scale Score 2011/12 and Prior	-12.868	36.785	0.726
Grade 8 Mathematics Scale Score 2012/13 to 2016/17	0.081	0.002	0.000
Missing Flag: Grade 8 Mathematics Scale Score 2012/13 to 2016/17	26.316	0.479	0.000
Grade 8 Mathematics Scale Score 2017/18 and Subsequent	0.053	0.020	0.008
Missing Flag: Grade 8 Mathematics Scale Score 2017/18 and Subsequent	28.542	12.174	0.019

Table I 24. Grades 9-12, ELA Model Coefficients, Adjusted Model

Effect Name	Effect	Standard Error	*p*-value
Cohort 1	-1.167	1.044	0.264
Cohort 2	2.316	0.357	0.000
Cohort 3	3.007	0.350	0.000
Cohort 4	1.038	0.356	0.004
Count of Prior Required Regents Exams = 0	-574.032	60.946	0.000
Count of Prior Required Regents Exams = 1	-569.646	60.948	0.000
Count of Prior Required Regents Exams = 2	-565.230	60.948	0.000
Count of Prior Required Regents Exams = 3	-560.768	60.948	0.000
Count of Prior Required Regents Exams = 4	-561.521	60.948	0.000
Count of Prior Required Regents Exams = 5	-559.521	60.952	0.000

Grade 7 ELA Scale Score 2011/12 and Prior	0.118	0.036	0.001
Missing Flag: Grade 7 ELA Scale Score 2011/12 and Prior	78.972	22.675	0.000
Grade 7 ELA Scale Score 2012/13 to 2016/17	0.129	0.002	0.000
Missing Flag: Grade 7 ELA Scale Score 2012/13 to 2016/17	36.958	0.476	0.000
Grade 7 Mathematics Scale Score 2011/12 and Prior	0.018	0.020	0.382
Missing Flag: Grade 7 Mathematics Scale Score 2011/12 and Prior	12.436	12.963	0.337
Grade 7 Mathematics Scale Score 2012/13 to 2016/17	0.019	0.002	0.000
Missing Flag: Grade 7 Mathematics Scale Score 2012/13 to 2016/17	6.357	0.491	0.000
Grade 8 ELA Scale Score 2011/12 and Prior	0.295	0.109	0.007
Missing Flag: Grade 8 ELA Scale Score 2011/12 and Prior	201.301	66.648	0.003
Grade 8 ELA Scale Score 2012/13 to 2016/17	0.106	0.002	0.000
Missing Flag: Grade 8 ELA Scale Score 2012/13 to 2016/17	32.233	0.535	0.000
Grade 8 ELA Scale Score 2017/18 and Subsequent	0.452	0.018	0.000
Missing Flag: Grade 8 ELA Scale Score 2017/18 and Subsequent	276.732	11.113	0.000
Grade 8 Mathematics Scale Score 2011/12 and Prior	-0.049	0.064	0.441
Missing Flag: Grade 8 Mathematics Scale Score 2011/12 and Prior	-42.110	40.387	0.297
Grade 8 Mathematics Scale Score 2012/13 to 2016/17	0.013	0.002	0.000
Missing Flag: Grade 8 Mathematics Scale Score 2012/13 to 2016/17	3.953	0.529	0.000
Grade 8 Mathematics Scale Score 2017/18 and Subsequent	-0.056	0.020	0.006
Missing Flag: Grade 8 Mathematics Scale Score 2017/18 and Subsequent	-42.510	12.227	0.001
Mean Prior Grade 8 ELA 2011/12 and Prior	0.007	0.002	0.001
Missing Flag: Mean Prior Grade 8 ELA 2011/12 and Prior	3.649	1.344	0.007
Mean Prior Grade 8 ELA 2012/13 to 2016/17	0.044	0.004	0.000
Missing Flag: Mean Prior Grade 8 ELA 2012/13 to 2016/17	20.527	2.932	0.000
Mean Prior Grade 8 ELA 2017/18 and Subsequent	-0.040	0.005	0.000
Missing Flag: Mean Prior Grade 8 ELA 2017/18 and Subsequent	-22.546	2.986	0.000
NYSESLAT LS Scale Score 2011/12 and Prior	0.088	0.099	0.373
NYSESLAT RW Scale Score 2011/12 and Prior	-0.063	0.094	0.500
NYSESLAT Scale Score 2012/13 and 2013/14	0.056	0.015	0.000
NYSESLAT Scale Score 2012/13 and 2014/15	0.079	0.003	0.000
Missing Flag: NYSESLAT Scale Scores 2011/12 and Prior	18.198	47.211	0.700
Missing Flag: NYSESLAT Scale Scores 2012/13 and 2013/14	46.571	12.882	0.000
Missing Flag: NYSESLAT Scale Scores 2014/15 and Later	21.803	0.952	0.000
New to School After Grade 9	0.175	0.128	0.171
Missing Flag: School Percentage Variables	-15.206	2.737	0.000
General Education < 40% (LRE3)	-3.517	0.367	0.000

ELLs	-1.678	0.216	0.000
School Percentage of ELLs	0.071	0.004	0.000
Economically Disadvantaged	-0.746	0.058	0.000
School Percentage of Economically Disadvantaged	-0.041	0.001	0.000
Students with Disabilities	-5.309	0.080	0.000
School Percentage of Students with Disabilities	-0.028	0.004	0.000

The tables that follow are for the unadjusted and adjusted GRE models by Year in School. Note: a dash (—) indicates that the coefficient was either excluded because there were fewer than 10 students with that variable or for perfect collinearity.

Table I 25. Grades 9-12, GRE, Year in School 1 Model Coefficients, Unadjusted Model

Effect Name	Estimate	Standard Error
Intercept 1	-32.278	0.288
Intercept 2	-33.730	0.290
Intercept 3	-38.074	0.294
Intercept 4	-40.375	0.296
Intercept 5	-43.000	0.331
Grade 7 ELA Scale Score 2012/13 to 2016/17	0.010	0.000
Missing Flag: Grade 7 ELA Scale Score 2012/13 to 2016/17	2.942	0.087
Grade 7 Mathematics Scale Score 2012/13 to 2016/17	0.014	0.000
Missing Flag: Grade 7 Mathematics Scale Score 2012/13 to 2016/17	4.093	0.074
Grade 8 ELA Scale Score 2017/18 and Subsequent	0.016	0.000
Missing Flag: Grade 8 ELA Scale Score 2017/18 and Subsequent	9.265	0.273
Grade 8 Mathematics Scale Score 2017/18 and Subsequent	0.029	0.000
Missing Flag: Grade 8 Mathematics Scale Score 2017/18 and Subsequent	17.554	0.273

Table I 26. Grades 9-12, GRE, Year in School 1 Model Coefficients, Adjusted Model

Effect Name	Estimate	Standard Error
Intercept 1	-46.224	1.199
Intercept 2	-47.763	1.200
Intercept 3	-52.179	1.201
Intercept 4	-54.476	1.202
Intercept 5	-57.100	1.211
Count of Prior Required Regents Exams = 0	2.658	0.728
Count of Prior Required Regents Exams = 1	2.566	0.728
Count of Prior Required Regents Exams = 2	2.450	0.728
Count of Prior Required Regents Exams = 3	2.124	0.729
Count of Prior Required Regents Exams = 4	2.093	0.735
Grade 7 ELA Scale Score 2012/13 to 2016/17	0.009	0.000
Missing Flag: Grade 7 ELA Scale Score 2012/13 to 2016/17	2.971	0.092
Grade 7 Mathematics Scale Score 2012/13 to 2016/17	0.014	0.000
Missing Flag: Grade 7 Mathematics Scale Score 2012/13 to 2016/17	3.922	0.080
Grade 8 ELA Scale Score 2017/18 and Subsequent	0.017	0.000
Missing Flag: Grade 8 ELA Scale Score 2017/18 and Subsequent	9.841	0.288
Grade 8 Mathematics Scale Score 2017/18 and Subsequent	0.031	0.000

Missing Flag: Grade 8 Mathematics Scale Score 2017/18 and Subsequent	18.191	0.289
Missing Flag: Mean Prior Grade 8 ELA 2011/12 and Prior	0.039	0.360
Mean Prior Grade 8 ELA 2011/12 and Prior	0.000	0.001
Missing Flag: Mean Prior Grade 8 ELA 2012/13 to 2016/17	-1.274	83.626
Mean Prior Grade 8 ELA 2012/13 to 2016/17	0.017	0.001
Missing Flag: Mean Prior Grade 8 ELA 2017/18 and Subsequent	6.276	1.300
Mean Prior Grade 8 ELA 2017/18 and Subsequent	0.014	0.002
Missing Flag: Mean Prior Grade 8 Mathematics 2011/12 and Prior	1.135	0.238
Mean Prior Grade 8 Mathematics 2011/12 and Prior	0.002	0.000
Missing Flag: Mean Prior Grade 8 Mathematics 2012/13 to 2016/17	-4.857	83.648
Mean Prior Grade 8 Mathematics 2012/13 to 2016/17	-0.030	0.001
Missing Flag: Mean Prior Grade 8 Mathematics 2017/18 and Subsequent	1.580	0.905
Mean Prior Grade 8 Mathematics 2017/18 and Subsequent	0.002	0.002
NYSESLAT Scale Score 2012/13 and 2014/15	0.014	0.001
Missing Flag: NYSESLAT Scale Scores 2014/15 and Later	4.093	0.211
General Education < 40% (LRE3)	-0.269	0.049
ELLs	-0.007	0.052
School Percentage of ELLs	-0.006	0.001
Economically Disadvantaged	-0.311	0.013
School Percentage of Economically Disadvantaged	-0.007	0.000
Students with Disabilities	-0.064	0.017
School Percentage of Students with Disabilities	-0.023	0.001

Table I 27. Grades 9-12, GRE, Year in School 2 Model Coefficients, Unadjusted Model

Effect Name	Estimate	Standard Error
Intercept 1	-6.788	0.060
Intercept 2	-8.037	0.061
Intercept 3	-9.713	0.062
Intercept 4	-12.267	0.065
Intercept 5	-15.274	0.081
Intercept 6	-18.635	0.275
Grade 7 ELA Scale Score 2012/13 to 2016/17	0.005	0.000
Missing Flag: Grade 7 ELA Scale Score 2012/13 to 2016/17	1.712	0.076
Grade 7 Mathematics Scale Score 2012/13 to 2016/17	0.006	0.000
Missing Flag: Grade 7 Mathematics Scale Score 2012/13 to 2016/17	1.830	0.066
Grade 8 ELA Scale Score 2012/13 to 2016/17	0.005	0.000
Missing Flag: Grade 8 ELA Scale Score 2012/13 to 2016/17	1.710	0.075
Grade 8 Mathematics Scale Score 2012/13 to 2016/17	0.014	0.000
Missing Flag: Grade 8 Mathematics Scale Score 2012/13 to 2016/17	3.784	0.062

Table I 28. Grades 9-12, GRE, Year in School 2 Model Coefficients, Adjusted Model

Effect Name	Effect	Standard Error
Intercept 1	-6.963	0.704
Intercept 2	-8.335	0.704
Intercept 3	-10.114	0.704
Intercept 4	-12.705	0.704
Intercept 5	-15.705	0.706
Intercept 6	-19.065	0.753
Count of Prior Required Regents Exams = 0	1.818	0.160
Count of Prior Required Regents Exams = 1	2.770	0.160
Count of Prior Required Regents Exams = 2	2.800	0.159
Count of Prior Required Regents Exams = 3	2.272	0.160
Count of Prior Required Regents Exams = 4	1.708	0.167
Grade 7 ELA Scale Score 2012/13 to 2016/17	0.003	0.000
Missing Flag: Grade 7 ELA Scale Score 2012/13 to 2016/17	1.195	0.079
Grade 7 Mathematics Scale Score 2012/13 to 2016/17	0.003	0.000
Missing Flag: Grade 7 Mathematics Scale Score 2012/13 to 2016/17	0.839	0.070
Grade 8 ELA Scale Score 2012/13 to 2016/17	0.004	0.000
Missing Flag: Grade 8 ELA Scale Score 2012/13 to 2016/17	1.481	0.076
Grade 8 Mathematics Scale Score 2012/13 to 2016/17	0.011	0.000
Missing Flag: Grade 8 Mathematics Scale Score 2012/13 to 2016/17	2.724	0.066
Missing Flag: Mean Prior Grade 8 ELA 2011/12 and Prior	0.662	0.318
Mean Prior Grade 8 ELA 2011/12 and Prior	0.001	0.001
Missing Flag: Mean Prior Grade 8 ELA 2012/13 to 2016/17	5.090	0.867
Mean Prior Grade 8 ELA 2012/13 to 2016/17	0.008	0.001
Missing Flag: Mean Prior Grade 8 ELA 2017/18 and Subsequent	-6.978	0.902
Mean Prior Grade 8 ELA 2017/18 and Subsequent	-0.011	0.001
Missing Flag: Mean Prior Grade 8 Mathematics 2011/12 and Prior	1.367	0.216
Mean Prior Grade 8 Mathematics 2011/12 and Prior	0.002	0.000
Missing Flag: Mean Prior Grade 8 Mathematics 2012/13 to 2016/17	-13.041	238.481
Mean Prior Grade 8 Mathematics 2012/13 to 2016/17	-0.004	0.001
Missing Flag: Mean Prior Grade 8 Mathematics 2017/18 and Subsequent	1.018	0.750
Mean Prior Grade 8 Mathematics 2017/18 and Subsequent	0.002	0.001
NYSESLAT Scale Score 2012/13 and 2014/15	0.011	0.001
Missing Flag: NYSESLAT Scale Scores 2014/15 and Later	2.888	0.180
New to School After Grade 9	-0.255	0.023
General Education < 40% (LRE3)	-0.451	0.068
ELLs	-0.340	0.049
School Percentage of ELLs	-0.009	0.001
Economically Disadvantaged	-0.166	0.012

School Percentage of Economically Disadvantaged	-0.002	0.000
Students with Disabilities	-0.084	0.016
School Percentage of Students with Disabilities	-0.030	0.001

Table I 29. Grades 9-12, GRE, Year in School 3 Model Coefficients, Unadjusted Model

Effect Name	Estimate	Standard Error
Intercept 1	-2.752	0.058
Intercept 2	-3.985	0.059
Intercept 3	-6.040	0.060
Intercept 4	-8.301	0.062
Intercept 5	-10.704	0.076
Intercept 6	-13.249	0.175
Intercept 7	-15.251	0.451
Intercept 8	-16.860	1.002
Grade 7 ELA Scale Score 2012/13 to 2016/17	0.005	0.000
Missing Flag: Grade 7 ELA Scale Score 2012/13 to 2016/17	1.453	0.070
Grade 7 Mathematics Scale Score 2012/13 to 2016/17	0.001	0.000
Missing Flag: Grade 7 Mathematics Scale Score 2012/13 to 2016/17	0.669	0.070
Grade 8 ELA Scale Score 2012/13 to 2016/17	0.001	0.000
Missing Flag: Grade 8 ELA Scale Score 2012/13 to 2016/17	0.660	0.079
Grade 8 Mathematics Scale Score 2012/13 to 2016/17	0.010	0.000
Missing Flag: Grade 8 Mathematics Scale Score 2012/13 to 2016/17	2.453	0.066

Table I 30. Grades 9-12, GRE, Year in School 3 Model Coefficients, Adjusted Model

Effect Name	Estimate	Standard Error
Intercept 1	-4.480	5.891
Intercept 2	-6.236	5.891
Intercept 3	-8.832	5.891
Intercept 4	-11.184	5.891
Intercept 5	-13.589	5.891
Intercept 6	-16.133	5.893
Intercept 7	-18.135	5.908
Intercept 8	-19.744	5.975
Count of Prior Required Regents Exams = 0	1.054	0.048
Count of Prior Required Regents Exams = 1	2.709	0.049
Count of Prior Required Regents Exams = 2	4.043	0.047
Count of Prior Required Regents Exams = 3	4.262	0.044
Count of Prior Required Regents Exams = 4	1.940	0.044
Grade 7 ELA Scale Score 2012/13 to 2016/17	0.003	0.000

Missing Flag: Grade 7 ELA Scale Score 2012/13 to 2016/17	0.844	0.075
Grade 7 Mathematics Scale Score 2012/13 to 2016/17	-0.002	0.000
Missing Flag: Grade 7 Mathematics Scale Score 2012/13 to 2016/17	-0.408	0.075
Grade 8 ELA Scale Score 2012/13 to 2016/17	0.001	0.000
Missing Flag: Grade 8 ELA Scale Score 2012/13 to 2016/17	0.544	0.082
Grade 8 Mathematics Scale Score 2012/13 to 2016/17	0.008	0.000
Missing Flag: Grade 8 Mathematics Scale Score 2012/13 to 2016/17	1.769	0.071
Missing Flag: Mean Prior Grade 8 ELA 2011/12 and Prior	0.259	0.317
Mean Prior Grade 8 ELA 2011/12 and Prior	0.001	0.001
Missing Flag: Mean Prior Grade 8 ELA 2012/13 to 2016/17	5.065	0.496
Mean Prior Grade 8 ELA 2012/13 to 2016/17	0.008	0.001
Missing Flag: Mean Prior Grade 8 ELA 2017/18 and Subsequent	-2.093	0.795
Mean Prior Grade 8 ELA 2017/18 and Subsequent	-0.001	0.001
Missing Flag: Mean Prior Grade 8 Mathematics 2011/12 and Prior	-0.090	0.209
Mean Prior Grade 8 Mathematics 2011/12 and Prior	0.000	0.000
Missing Flag: Mean Prior Grade 8 Mathematics 2012/13 to 2016/17	-16.619	72.031
Mean Prior Grade 8 Mathematics 2012/13 to 2016/17	-0.007	0.001
Missing Flag: Mean Prior Grade 8 Mathematics 2017/18 and Subsequent	-0.735	0.720
Mean Prior Grade 8 Mathematics 2017/18 and Subsequent	-0.003	0.001
NYSESLAT Scale Score 2012/13 and 2013/14	0.000	0.007
NYSESLAT Scale Score 2012/13 and 2014/15	0.008	0.001
Missing Flag: NYSESLAT Scale Scores 2012/13 and 2013/14	1.085	5.847
Missing Flag: NYSESLAT Scale Scores 2014/15 and Later	2.117	0.186
New to School After Grade 9	0.084	0.036
General Education < 40% (LRE3)	-0.246	0.084
ELLs	-0.260	0.050
School Percentage of ELLs	0.001	0.001
Economically Disadvantaged	-0.058	0.012
School Percentage of Economically Disadvantaged	-0.003	0.000
Students with Disabilities	-0.264	0.018
School Percentage of Students with Disabilities	-0.019	0.001

Table I 31. Grades 9-12, GRE, Year in School 4 Model Coefficients, Unadjusted Model

Effect Name	Estimate	Standard Error
Intercept 1	2.939	0.086
Intercept 2	1.349	0.086
Intercept 3	-0.063	0.088
Intercept 4	-1.511	0.096
Intercept 5	-3.378	0.142
Intercept 6	-6.337	0.507
Intercept 7	-7.030	0.712
Grade 7 ELA Scale Score 2012/13 to 2016/17	-0.006	0.000
Missing Flag: Grade 7 ELA Scale Score 2012/13 to 2016/17	-1.346	0.097
Grade 7 Mathematics Scale Score 2012/13 to 2016/17	0.000	0.000
Missing Flag: Grade 7 Mathematics Scale Score 2012/13 to 2016/17	-0.248	0.101
Grade 8 ELA Scale Score 2012/13 to 2016/17	-0.005	0.000
Missing Flag: Grade 8 ELA Scale Score 2012/13 to 2016/17	-1.382	0.093
Grade 8 Mathematics Scale Score 2012/13 to 2016/17	-0.002	0.000
Missing Flag: Grade 8 Mathematics Scale Score 2012/13 to 2016/17	-0.699	0.095

Table I 32. Grades 9-12, GRE, Year in School 4 Model Coefficients, Adjusted Model

Effect Name	Estimate	Standard Error
Intercept 1	-5.060	2.290
Intercept 2	-7.183	2.290
Intercept 3	-8.749	2.290
Intercept 4	-10.222	2.291
Intercept 5	-12.091	2.293
Intercept 6	-15.049	2.344
Intercept 7	-15.743	2.397
Count of Prior Required Regents Exams = 0	1.232	0.041
Count of Prior Required Regents Exams = 1	2.841	0.039
Count of Prior Required Regents Exams = 2	3.631	0.034
Count of Prior Required Regents Exams = 3	3.687	0.028
Count of Prior Required Regents Exams = 4	2.690	0.021
Grade 7 ELA Scale Score 2012/13 to 2016/17	0.001	0.000
Missing Flag: Grade 7 ELA Scale Score 2012/13 to 2016/17	0.292	0.115
Grade 7 Mathematics Scale Score 2012/13 to 2016/17	0.002	0.000
Missing Flag: Grade 7 Mathematics Scale Score 2012/13 to 2016/17	0.565	0.116
Grade 8 ELA Scale Score 2012/13 to 2016/17	-0.001	0.000
Missing Flag: Grade 8 ELA Scale Score 2012/13 to 2016/17	-0.136	0.110
Grade 8 Mathematics Scale Score 2012/13 to 2016/17	0.002	0.000

Missing Flag: Grade 8 Mathematics Scale Score 2012/13 to 2016/17	0.340	0.109
Missing Flag: Mean Prior Grade 8 ELA 2011/12 and Prior	-0.483	0.516
Mean Prior Grade 8 ELA 2011/12 and Prior	-0.001	0.001
Missing Flag: Mean Prior Grade 8 ELA 2012/13 to 2016/17	4.498	0.971
Mean Prior Grade 8 ELA 2012/13 to 2016/17	0.012	0.002
Missing Flag: Mean Prior Grade 8 ELA 2017/18 and Subsequent	3.084	1.139
Mean Prior Grade 8 ELA 2017/18 and Subsequent	0.006	0.002
Missing Flag: Mean Prior Grade 8 Mathematics 2011/12 and Prior	-0.012	0.317
Mean Prior Grade 8 Mathematics 2011/12 and Prior	0.000	0.000
Missing Flag: Mean Prior Grade 8 Mathematics 2012/13 to 2016/17	-3.396	1.033
Mean Prior Grade 8 Mathematics 2012/13 to 2016/17	-0.007	0.001
Missing Flag: Mean Prior Grade 8 Mathematics 2017/18 and Subsequent	-3.644	1.118
Mean Prior Grade 8 Mathematics 2017/18 and Subsequent	-0.006	0.002
NYSESLAT Scale Score 2012/13 and 2013/14	0.001	0.002
NYSESLAT Scale Score 2012/13 and 2014/15	0.001	0.001
Missing Flag: NYSESLAT Scale Scores 2012/13 and 2013/14	0.410	2.029
Missing Flag: NYSESLAT Scale Scores 2014/15 and Later	-0.009	0.223
New to School After Grade 9	0.565	0.053
General Education < 40% (LRE3)	-0.041	0.094
ELLs	-0.067	0.057
School Percentage of ELLs	0.011	0.001
Economically Disadvantaged	-0.009	0.020
School Percentage of Economically Disadvantaged	0.003	0.001
Students with Disabilities	0.065	0.023
School Percentage of Students with Disabilities	-0.006	0.002

Table I 33. Grades 9-12, GRE, Year in School 5+ Model Coefficients, Unadjusted Model

Effect Name	Estimate	Standard Error
Intercept 1	-6.802	4.396
Intercept 2	-7.978	4.396
Intercept 3	-9.289	4.397
Intercept 4	-10.608	4.398
Intercept 5	-12.812	4.415
Intercept 6	-14.604	4.509
Grade 7 ELA Scale Score 2011/12 and Prior	0.004	0.004
Missing Flag: Grade 7 ELA Scale Score 2011/12 and Prior	3.445	2.431
Grade 7 ELA Scale Score 2012/13 to 2016/17	0.001	0.001
Missing Flag: Grade 7 ELA Scale Score 2012/13 to 2016/17	0.483	0.345
Grade 7 Mathematics Scale Score 2011/12 and Prior	0.004	0.002
Missing Flag: Grade 7 Mathematics Scale Score 2011/12 and Prior	2.093	1.374
Grade 7 Mathematics Scale Score 2012/13 to 2016/17	-0.003	0.001
Missing Flag: Grade 7 Mathematics Scale Score 2012/13 to 2016/17	-1.014	0.345
Missing Flag: Grade 8 ELA Scale Score 2011/12 and Prior	0.189	5.169
Grade 8 ELA Scale Score 2011/2012 and Prior	0.001	0.008
Grade 8 ELA Scale Score 2012/13 to 2016/17	-0.003	0.001
Missing Flag: Grade 8 ELA Scale Score 2012/13 to 2016/17	-0.842	0.294
Missing Flag: Grade 8 Mathematics Scale Score 2011/12 and Prior	1.664	3.216
Grade 8 Mathematics Scale Score 2011/2012 and Prior	0.002	0.005
Grade 8 Mathematics Scale Score 2012/13 to 2016/17	-0.001	0.001
Missing Flag: Grade 8 Mathematics Scale Score 2012/13 to 2016/17	-0.389	0.293

Table I 34. Grades 9-12, GRE, Year in School 5+ Model Coefficients, Adjusted Model

Effect Name	Estimate	Standard Error
Intercept 1	-1.999	6.988
Intercept 2	-3.315	6.988
Intercept 3	-4.679	6.988
Intercept 4	-6.008	6.989
Intercept 5	-8.213	7.000
Intercept 6	-10.005	7.059
Count of Prior Required Regents Exams = 0	2.315	0.236
Count of Prior Required Regents Exams = 1	3.521	0.230
Count of Prior Required Regents Exams = 2	3.951	0.227
Count of Prior Required Regents Exams = 3	4.216	0.224
Count of Prior Required Regents Exams = 4	3.601	0.223
Grade 7 ELA Scale Score 2011/12 and Prior	0.005	0.004
Missing Flag: Grade 7 ELA Scale Score 2011/12 and Prior	4.274	2.822

Grade 7 ELA Scale Score 2012/13 to 2016/17	0.002	0.001
Missing Flag: Grade 7 ELA Scale Score 2012/13 to 2016/17	0.401	0.386
Grade 7 Mathematics Scale Score 2011/12 and Prior	0.001	0.002
Missing Flag: Grade 7 Mathematics Scale Score 2011/12 and Prior	0.684	1.486
Grade 7 Mathematics Scale Score 2012/13 to 2016/17	-0.002	0.001
Missing Flag: Grade 7 Mathematics Scale Score 2012/13 to 2016/17	-0.671	0.371
Missing Flag: Grade 8 ELA Scale Score 2011/12 and Prior	0.474	5.964
Grade 8 ELA Scale Score 2011/2012 and Prior	0.001	0.010
Grade 8 ELA Scale Score 2012/13 to 2016/17	-0.001	0.001
Missing Flag: Grade 8 ELA Scale Score 2012/13 to 2016/17	-0.322	0.330
Missing Flag: Grade 8 Mathematics Scale Score 2011/12 and Prior	1.883	3.450
Grade 8 Mathematics Scale Score 2011/2012 and Prior	0.001	0.005
Grade 8 Mathematics Scale Score 2012/13 to 2016/17	0.001	0.001
Missing Flag: Grade 8 Mathematics Scale Score 2012/13 to 2016/17	0.141	0.323
Missing Flag: Mean Prior Grade 8 ELA 2011/12 and Prior	-0.384	1.640
Mean Prior Grade 8 ELA 2011/12 and Prior	-0.001	0.003
Missing Flag: Mean Prior Grade 8 ELA 2012/13 to 2016/17	-5.930	29.875
Mean Prior Grade 8 ELA 2012/13 to 2016/17	0.003	0.005
Missing Flag: Mean Prior Grade 8 ELA 2017/18 and Subsequent	-5.283	3.388
Mean Prior Grade 8 ELA 2017/18 and Subsequent	-0.007	0.006
Missing Flag: Mean Prior Grade 8 Mathematics 2011/12 and Prior	-0.873	1.081
Mean Prior Grade 8 Mathematics 2011/12 and Prior	-0.001	0.002
Missing Flag: Mean Prior Grade 8 Mathematics 2012/13 to 2016/17	0.824	1.920
Mean Prior Grade 8 Mathematics 2012/13 to 2016/17	0.001	0.004
Missing Flag: Mean Prior Grade 8 Mathematics 2017/18 and Subsequent	0.813	3.624
Mean Prior Grade 8 Mathematics 2017/18 and Subsequent	0.000	0.006
NYSESLAT LS Scale Score 2011/12 and Prior	0.007	0.008
NYSESLAT RW Scale Score 2011/12 and Prior	-0.010	0.009
NYSESLAT Scale Score 2012/13 and 2013/14	-0.003	0.003
Missing Flag: NYSESLAT Scale Scores 2011/12 and Prior	-2.115	4.118
Missing Flag: NYSESLAT Scale Scores 2012/13 and 2013/14	-3.116	2.149
New to School After Grade 9	0.808	0.129
General Education < 40% (LRE3)	-0.332	0.186
ELLs	-0.151	0.142
School Percentage of ELLs	0.000	0.004
Economically Disadvantaged	0.033	0.069
School Percentage of Economically Disadvantaged	0.001	0.002
Students with Disabilities	0.249	0.070
School Percentage of Students with Disabilities	0.004	0.004

Appendix J. Additional Impact Correlation Tables (Grades 4-8 by Grade and Subject and Grades 4-8 and 9-12 Principal)

Table J 1. Principal Impact Correlations by Grade for ELA[19]

Grade	Percent ELL	Percent SWD	Percent ED	Mean Prior Scale Score
4	0.108	0.069	0.082	-0.081
5	0.087	0.09	0.031	-0.17
6	0.013	-0.027	0.021	-0.19
7	0.055	0.126	0.094	-0.131
8	0.074	0.056	0.086	-0.116

Table J 2. Principal Impact Correlations by Grade for Mathematics[19]

Grade	Percent ELL	Percent SWD	Percent ED	Mean Prior Scale Score
4	0.115	0.051	0.062	-0.128
5	0.074	0.055	0.031	-0.096
6	0.06	0.035	0.102	-0.172
7	0.133	0.116	0.129	-0.19
8	0.029	0.029	0.089	-0.108

Table J 3. Principal Impact Correlations[19]

Model	Percent ELL	Percent SWD	Percent ED	Mean Prior ELA	Mean Prior Mathematics
4-8 MGP	0.085	0.063	0.09	-0.06	-0.069
9-12 MGP	0.034	-0.205	0.055	0.342	0.283
9-12 GRE	-0.047	-0.273	-0.099	0.431	0.45

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[19] Correlations for 4-8 models were calculated between principals' 4-8 MGPs and school-level percent demographics and mean prior scores.