

Report to the North Carolina General Assembly

2023 Statewide Year-Over-Year Trends in Achievement: Before, During, and After the Pandemic

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1 Executive Summary

The North Carolina Department of Public Instruction (NCDPI) and SAS Institute Inc. (SAS) collaborated to provide educators, policymakers, and other stakeholders with insight into the path or trajectory of student achievement before, during, and after the pandemic. The purpose of this analysis is to help understand both the initial and remaining impact of the pandemic on student achievement in North Carolina. The analysis focuses on how state achievement changed from one year to the next from 2013 to 2023. This year's report is a continuation of an earlier analysis that focused on achievement from 2013 to 2022.

When discussing the pandemic's impact and students' recovery, a common question is *what constitutes a full recovery*? In other words, how will we know when students have fully recovered from the negative impacts of the pandemic? Although there are several ways to measure students' recovery, this analysis focuses on long- and short-term changes in achievement over time. To provide a comprehensive view of student recovery, this analysis includes two different thresholds for evaluating student recovery. Specifically, student achievement in 2022 and 2023 is compared to a:

- 1. Continuation of the pre-pandemic trend from the 2013–2019 time period.
- 2. Three-year average of the state's achievement observed in the 2017–2019 time period.

The first threshold considers where the state was going *and continues that path*, whereas the second threshold only considers where the state *was* during the years immediately prior to the pandemic.

As part of understanding students' recovery in North Carolina, the analysis produces three key metrics:

- 1. The *Pre-Pandemic Trend* represents the overall trend in achievement for an assessment between 2013 and 2019. This line smooths out the year-to-year variation in the observed achievement for the state.
- The Pandemic Impact represents the extent to which actual achievement in 2021 diverged from the pre-pandemic trend had it continued to 2021. This line represents a counterfactual, or an estimate of what achievement might have been if the prepandemic trend were not disrupted by the pandemic.
- 3. The Distance to a Full Recovery represents the extent to which actual achievement in 2022 and 2023 diverged from a full recovery according to two different thresholds. There are two ways to consider recovery: a continuation of the pre-pandemic trend threshold based on the 2013-19 time frame and a three-year average threshold representing the more immediate time frame of 2017-19. The distance for each threshold is reported separately for 2022 and 2023 to assess the state's recovery after the most recent year of schooling.

Collectively, these three metrics provide a comprehensive picture of the long-term trajectory of year-over-year achievement in North Carolina.

This model uses a more sophisticated approach to determine state achievement than simple averages of scale scores or the percentage of student scoring proficient because it takes into account the year-to-year variation and trends in student achievement that existed prior to the pandemic. The model also adjusts for version changes in the assessments that occurred prior to the pandemic as was the case for EOG Math in 2019.

In addition to providing a more robust estimate of achievement trends, a major advantage of this approach is that it is possible to provide measures of the pandemic impact and recovery for earlier grades because it does not require sufficient predictors by 2019, like the cohort model (which was the basis for the earlier reports provided by NCDPI's Office of Learning and Recovery). Where sufficient data are available, results are provided for EOG Math and Reading in grades 3–8, EOG Science in grades 5 and 8, and EOC Biology, English II, and Math I.

This year-over-year analysis provides the following findings based on statewide trends:

- Prior to the pandemic, achievement was relatively stable for most assessments with small positive or negative trends.
- The pandemic impact estimated in 2021 was negative for all assessments with the exception of English II, and the effect size ranged from medium to large depending on the assessment.
- In 2022, most assessments indicate a recovery that was not sufficient to meet either recovery threshold, though how much pandemic impact remained depended on the threshold.
- In 2023, most assessments indicate further progress toward the recovery thresholds, although all but two assessments (EOG Reading Grade 3 and EOC English II) continue to fall below the threshold.
- As a content area, Math was more negatively impacted by the pandemic than Reading, and Math assessments have a greater distance to the recovery thresholds both one year and two years later than Reading.
- There is considerable variation among schools within the state in terms of the pandemic impact and recovery thresholds. The extent of variation among schools in achievement trends observed prior to the pandemic was, comparatively, modest.

The following sections provide an overview of the model, the information that it provides, and its output.

2 Data

2.1 Data Received

The analysis in this report leveraged student-level assessment data, where available, from 2012-13 through the 2022-23 school year in order to compile a longitudinal data set based on the following assessments:

- EOG Mathematics in grades 3–8
- EOG Reading in grades 3–8
- EOG Science in grades 5 and 8
- EOC Biology, English II, Math 1, and Math 3

The state EOG tests are administered in the spring semester whereas the EOC assessments are typically given at the end of the fall and spring semesters with the occasional summer administration. For each administration, SAS used the following student identifiers, assessment data, and student flags:

- Student Identifiers
 - Student Last Name
 - Student First Name
 - Student Middle Initial
 - Student Date of Birth
 - Student Identification Number
- Assessment Information
 - Scale Score
 - Test Taken
 - Tested Grade
 - Test Semester
 - School Number
 - District Number
 - Administration Window

Note that the model adjusted scores for version changes in the assessments that occurred *prior* to the pandemic as was the case for EOG Math in 2019. However, version changes in many of the Reading assessments occurred during or after the pandemic. As a result, the effects of these changes are confounded with the timing of the pandemic and could not be modeled, so there were no adjustments to those scores.

SAS merged the individual student records over time using an algorithm that incorporated all student identifiers to create a longitudinal database that tracks individual students' performance across grade levels on state assessments each year.

2.2 Business Rules

In creating the longitudinal database, the following business rules were applied regarding student scores.

2.2.1 Missing Grade

In North Carolina, the grade used in the analyses and reporting is the tested grade, not the enrolled grade. If a grade is missing on an early grade or end-of-grade test record, then that record will be excluded from all analyses. The grade is required to include a student's score in the appropriate part of the models.

2.2.2 Duplicate (Same) Scores

If a student has a duplicate score for a particular subject and tested grade in a given testing period in a given school, then the extra score will be excluded from the analysis.

2.2.3 Students with Missing Districts or Schools for Some Scores but Not Others

If a student has a duplicate score with a missing district or school for a particular subject and grade or course in a given testing period, then the duplicate score that has a district and/or school will be included over the duplicate score that has the missing data.

2.2.4 Students with Multiple (Different) Scores in the Same Testing Administration

If a student has multiple scores in the same period for a particular subject and grade or course and the test scores are not the same, then those scores will be excluded from the analysis. If duplicate scores for a particular subject and tested grade in a given testing period are at different schools, then both scores will be excluded from the analysis. Note that if multiple scores are received for grade 3 Reading or Math across years, only the most recent score is used. One exception to this rule is that when a student has both an EOG Math grade 8 score and a Math 1 score in the same year, only their Math 1 score is used in the analysis.

Students with Multiple Grade Levels in the Same Subject in the Same Year

A student should not have different tested grade levels in the same subject in the same year. If that is the case, then the student's records are checked to see whether the data for two separate students were inadvertently combined. If this is the case, then the student data are adjusted so that each unique student is associated with only the appropriate scores. If the scores appear to all be associated with a single unique student, then scores that appear inconsistent are excluded from the analysis.

2.2.5 Students with Records That Have Unexpected Grade Level Changes

If a student skips more than one grade level (e.g., moves from sixth in 2018 to ninth in 2019) or is moved back by one grade or more (i.e., moves from fourth in 2018 to third in 2019) in the same subject, then the student's records are examined to determine whether two separate students were inadvertently combined. If this is the case, then the student data is adjusted so that each unique student is associated with only the appropriate scores. These scores are removed from the analysis if it is the same student. Per NCDPI's decision, the analysis does not remove students with scores that appear to be associated with inconsistent grades. The analysis leaves students in the analysis at the tested grade that EVAAS receives from NCDPI.

2.2.6 Students with Records at Multiple Schools in the Same Test Period

If a student is tested at two different schools in a given testing period, then the student's records are examined to determine whether two separate students were inadvertently combined. If this is the case, then the student data is adjusted so that each unique student is associated with only the appropriate scores. When students have valid scores at multiple schools in different subjects, all valid scores are used at the appropriate school.

2.2.7 Outliers

Student assessment scores are checked each year to determine whether they are outliers in context with all the other scores in a reference group of scores from the individual student. These reference scores are weighted differently depending on proximity in time to the score in question. Scores are checked for outliers using related subjects as the reference group. For example, when searching for outliers for EOC Math test scores, all EOG and EOC Math subjects are examined simultaneously, and any scores that appear inconsistent, given the other scores for the student, are flagged.

Scores are flagged in a conservative way to avoid excluding any student scores that should not be excluded. Scores can be flagged as either high or low outliers. It should also be noted that test scores within a year, subject and grade are normalized before checking begins. This helps mitigate any unnecessary flagging of outliers due to a year of assessments shifting across the state as might happen in 2021 when the pandemic caused a large, negative shift in the achievement distribution.

This process is part of a data quality procedure to ensure that no scores are used if they were, in fact, errors in the data, and the approach for flagging a student score as an outlier is fairly conservative. Again, students were expected to score lower in 2021 due to the pandemic, and this process is more about flagging data that might be erroneous.

Considerations included in outlier detection are:

- Is the score in the tails of the distribution of scores? Is the score very high or low achieving?
- Is the score "significantly different" from the other scores as indicated by a statistical analysis that compares each score to the other scores?
- Is the score also "practically different" from the other scores? Statistical significance can sometimes be associated with numerical differences that are too small to be meaningful.
- Are there enough scores to make a meaningful decision?

To decide whether student scores are considered outliers, all student scores are first converted into a standardized normal Z-score. Then each individual score is compared to the weighted combination of all the reference scores described above. The difference of these two scores provides a t-value of each comparison. Using this t-value, the models can flag individual scores as outliers.

There are different business rules for the low outliers and the high outliers, and this approach is more conservative when removing a very high-achieving score.

For low-end outliers, the rules are:

- The percentile of the score must be below 50.
- The t-value must be below -3.5 for EOGs in Math and Reading when determining the difference between the score in question and the weighted combination of reference scores (otherwise known as the comparison score). In other words, the score in question must be at least 3.5 standard deviations below the comparison score. For EOC and EOG Science assessments, the t-value must be below -4.0.
- The percentile of the comparison score must be above a certain value. This value depends on the position of the individual score in question but will range from 10 to 90 with the ranges of the individual percentile score.

For high-end outliers, the rules are:

- The percentile of the score must be above 50.
- The t-value must be above 4.5 for EOGs in Math and Reading when determining the difference between the score in question and the reference group of scores. In other words, the score in question must be at least 4.5 standard deviations above the comparison score. For EOC and EOG Science assessments, the t-value must be above 5.0.
- The percentile of the comparison score must be below a certain value. This value depends on the position of the individual score in question but will need to be at least 30 to 50 percentiles below the individual percentile score.
- There must be at least three scores in the comparison score average.

2.2.8 Membership

Students were excluded if they did not meet membership, a designation based on student enrollment at a school and used for accountability purposes.

2.2.9 First Year English Learner

Given the research purpose of the analysis and need to create a comparable student population over time, students were not excluded based on first year English Learner designation.

2.2.10 Exclusion of School or District from the Analysis Based on Years of Data Available

Only schools or districts with at least two years of data prior to the 2019-20 school year and results from the 2020-21 and 2021-22 school years are included in the analysis.

2.2.11 Exclusion of School or District from the Analysis Based on Years of Data Available

Data for a school or district for a given assessment and year is only included in the analysis if there are results for at least 10 students.

3 Methods of Analysis

3.1 Overview

This analysis assesses the pre-pandemic trend, pandemic impact, and recovery trends through a model that accounts for the year-to-year variation and trends in student achievement that existed prior to the pandemic. In contrast to the cohort model, this approach evaluates trends in results for specific assessments over time rather than using individual students' prior testing data. A major advantage of this approach is that it is possible to provide measures of the prepandemic trend, pandemic impact and recovery for earlier grades that are no longer available in the cohort model.

In comparing student achievement in a given content area (such as fifth-grade Math) over the years prior to the pandemic, the average scale score for the state, a district, or a school might change from one year to the next. At the state level, these fluctuations are typically small, although they might be centered around a positive or negative trend line. On the other hand, schools and districts exhibit varying degrees of stability in average scale scores. Statistically, the average scale scores from small schools and districts are more variable over time than large schools and districts. In addition, prior to the pandemic, some schools and districts might exhibit distinct positive trends in average scale scores while other exhibit flat or negative trends. There are reasonable explanations for any variation: the populations of students in a given subject/grade changes from one year to the next; and there might be curricular and instructional strategies in place that influence achievement over time.

Regardless of the source, when assessing the pandemic's impact and recovery trends, it is important to consider this natural variation and the trends in achievement that existed in the state, a district, or a school in the years prior to the pandemic. Doing so provides a more robust estimate of the pandemic's impact and recovery.

The following sections provide a more detailed review of the model and its output.

3.2 Determining Average Achievement

The analysis uses the student assessment data and business rules described in <u>Section 2</u> to build the set of student data. The data follow successive cohorts of students within a specific subject and grade as they pass through the schools over the multiple years, such as 2017 fifth-grade students, 2018 fifth-grade students, 2019 fifth grade-students, etc. For each subject and grade analyzed, the model breaks down variation in student achievement over time into state, school, and student contributions. Specifically, the model estimates the following:

- Statewide pre-pandemic trend in achievement.
- Statewide version effects (when applicable), which is operationalized as an abrupt and persistent change in achievement that occurs the first year a new version of an assessment is administered.
- Statewide pandemic impact, which is operationalized as a deflection of the prepandemic trend in 2021.
- Statewide recovery, which is operationalized as a change in achievement from 2021 to 2022 in last year's report *and* from 2022 to 2023 in this year's report.
- Stochastic cohort-to-cohort variation in achievement observed within all schools included in the analysis. This component is only estimated in the years prior to the pandemic.

- School-specific pre-pandemic trend parameters, version effect parameters, pandemic impact parameters, and recovery parameters that are operationalized as random effects.
- A residual, student-level error term that represents the deviation of a student's scale score value from an expectation derived from all of the components described above.

Generally speaking, the analysis uses a piecewise, hierarchical linear growth model that is estimated separately by subject and grade.¹

In more technical terms, this analysis lets $y_{i,j,t}$ equal the test score (achievement) in a specific subject and grade for student *i* in school *j* at time *t*, where *t* is an index on a set of years that span the pre- and post-pandemic period e.g., $t \in (2017, 2018, ..., 2023)$. The model can be expressed as:

$$y_{i,j,t} = \beta_{0,j} + \beta_{1,j}T + \beta_{2,j}P + \beta_{3,j}R_{2022} + \beta_{4,j}R_{2023} + \gamma_{j,t} + \epsilon_{i,j,t}$$
(1)

With respect to the covariates:

For
$$t \le 2021$$
 let $T = t - 2020$; else $T = 1$. Thus, $T \in (-3, -2, -1, 0, 1, 2, 3)$. (2)

For
$$t \le 2020$$
, let $P = 0$; else $P = 1$. Thus, $P \in (0,0,0,0,1,1,1)$. (3)

For
$$t \le 2021$$
, let $R_{2022} = 0$; else $R = 1$. Thus, $R_{2022} \in (0,0,0,0,0,1,1)$ (4)

For
$$t \le 2022$$
, let $R_{2023} = 0$; else $R = 1$. Thus, $R_{2023} \in (0,0,0,0,0,0,1)$ (5)

 $\beta_{1,j}T$ captures school-specific trends in scale scores observed prior to the pandemic. $\beta_{2,j}P$ represent a deflection from the prior trend associated with the pandemic, $\beta_{3,j}R_{2022}$ represents a change in achievement from 2021 to 2022, and $\beta_{4,j}R_{2023}$ represents a change in achievement from 2022 to 2023. The coding of the *T*, *P*, *R*₂₀₂₂, and *R*₂₀₂₃ variables result in an intercept term, $\beta_{0,j}$ that can be interpreted as the "average" scale score of a student attending school *j* in 2020. Since there was no testing data collected in the spring of the 2019-20 school year in North Carolina, $\beta_{0,j}$ is an extrapolation from each schools' prior trend ($\beta_{1,j}$).

The school-specific growth parameters are random effects centered around state-wide averages. Specifically,

$$\beta_{0,j} = \tau_{00} + \mu_{0,j},\tag{6}$$

$$\beta_{1,j} = \tau_{10} + \mu_{1,j},\tag{7}$$

$$\beta_{2,i} = \tau_{20} + \mu_{2,i} \tag{8}$$

$$\beta_{3,i} = \tau_{30} + \mu_{3,i} \tag{9}$$

$$\beta_{4,j} = \tau_{40} + \mu_{4,j} \tag{10}$$

¹ See, for example: Willett, J. B., Singer, J. D., & Martin, N. C. 1998. "The Design and Analysis of Longitudinal Studies of Development and Psychopathology in Context: Statistical Models and Methodological Recommendations." *Development and Psychopathology* 10(2), 395-426.

A final school-level random effect, $\gamma_{j,t}$, represents idiosyncratic variation of the school means around the school-specific growth trajectories captured by the random growth parameters from above. In the remainder of this document, "cohort error" refers to the $\gamma_{j,t}$ term and the model assumes that it is sampled from a univariate normal distribution with time-constant variance.

$$\gamma_{j,t} \sim N(0,\omega^2) \tag{11}$$

While the random growth parameters capture systematic changes in a school's average scale score as a function of the prior trend, pandemic impact, and recovery parameters, $\gamma_{j,t}$ captures the influence of unobserved and time-varying factors that are unique to that year and entirely independent of *T*, *P*, *R*₂₀₂₂ and *R*₂₀₂₃.

Theoretically, the inclusion of $\gamma_{j,t}$ allows the model to control for idiosyncratic, and unobserved events impacting achievement that would have taken place, even if there are no other modeled influences on achievement (e.g., the impact of the pandemic). However, such an adjustment is based on the very strong assumption of independent cohort error. It would also lead to a very conservative estimate of school-level pandemic impact and recovery estimates since such effects would be confounded with any cohort error that coincided with pandemic and recovery periods. Thus, for $t \in 2021, 2022, 2023$ the model assumes $\omega^2 = 0$. Effectively, this assigns all 2021, 2022, and 2023 school deviations to the statewide pandemic setback and achievement target respectively. Although this risks overstating school variation in the magnitude of pandemic setbacks and achievement targets, it makes the analysis more sensitive to detecting school-specific effects. However, by allowing for non-zero $\gamma_{j,t}$ variance during the period prior to the pandemic, the risk is reduced of over adjusting for school-specific pre-pandemic trends and version effects.

All school-level random effects are assumed to be sampled from a multivariate normal distribution:

$$\left[\mu_{0,j}, \mu_{1,j}, \mu_{2,j}, \mu_{3,j}, \mu_{4,j}, \gamma_{j,2017}, \gamma_{j,2018}, \gamma_{j,2019}, \gamma_{j,2021}, \gamma_{j,2022}, \gamma_{j,2023}\right] \sim N(\mathbf{0}, \mathbf{G})$$
(12)

where,

	λ_0^2	$\lambda^2_{0,1}$	$\lambda^2_{0,2}$	$\lambda^2_{0,3}$	$\lambda^2_{0,4}$	0	0	0	0	0	0	ן0	
	$\lambda^2_{1,0}$	λ_1^2	$\lambda_{1,2}^2$	$\lambda_{1,3}^2$	$\lambda_{1,4}^2$	0	0	0	0	0	0	0	
	$\lambda^2_{2,0}$	$\lambda_{1,1}^2$	λ_3^2	$\lambda^2_{2,3}$	$\lambda^2_{2,4}$	0	0	0	0	0	0	0	
	$\lambda^2_{3,0}$	$\lambda^2_{3,1}$	$\lambda^2_{3,2}$	λ_3^2	$\lambda^2_{3,4}$	0	0	0	0	0	0	0	
	$\lambda^2_{4,0}$	$\lambda^2_{4,1}$	$\lambda^2_{4,2}$	$\lambda^2_{4,3}$	λ_4^2	0	0	0	0	0	0	0	
=	0	0	0	0	0	ω^2	0	0	0	0	0	0	(13)
	0	0	0	0	0	0	ω^2	0	0	0	0	0	(10)
	0	0	0	0	0	0	0	ω^2	0	0	0	0	
	0	0	0	0	0	0	0	0	0	0	0	0	
	0	0	0	0	0	0	0	0	0	0	0	0	
	0	0	0	0	0	0	0	0	0	0	0	0	
	LO	0	0	0	0	0	0	0	0	0	0	۲0	

The last two rows and columns of the **G** matrix is included to emphasize the assumption of zero variance in the cohort error term discussed previously.

 $\epsilon_{i,j,t}$ is a within-school residual. To account for changes in the within school variances over time and school, the model estimates $t \times j$ specific variances. Let *N* equal the total sample size. Conditional on the random effects above, the model assumes that the within-school random effect is sampled from a *N*-dimensional multivariate normal distribution.

$$\epsilon_{i,j,t} \sim N(\mathbf{0}, \mathbf{R}) \sim N\left(\begin{bmatrix} 0\\0\\\vdots\\0 \end{bmatrix} , \begin{bmatrix} \sigma_{j,t}^2 & 0 & \dots & 0\\0 & \sigma_{j,t}^2 & 0 & \vdots\\\vdots & 0 & \ddots & 0\\0 & \dots & 0 & \sigma_{j,t}^2 \end{bmatrix} \right)$$
(14)

3.3 Model in Reduced Form

Before discussing estimation, it is useful to rewrite the model in reduced form, bracketing the fixed and random components of the model.

$$y_{i,j,t} = [\tau_{00}(1) + \tau_{10}T + \tau_{20}P + \tau_{30}R] + [\mu_{0,j}(1) + \mu_{1,j}T + \mu_{2,j}P + \mu_{3,j}R + \gamma_{j,t}(1) + \epsilon_{i,j,t}]$$
(15)

The reduced form expression leads to a much simplified, matrix expression:

$$\mathbf{y} = \mathbf{X}\boldsymbol{\tau} + \mathbf{Z} \begin{bmatrix} \boldsymbol{\mu} \\ \boldsymbol{\gamma} \end{bmatrix} + \boldsymbol{\epsilon}$$
(16)

with,

$$\begin{bmatrix} \boldsymbol{\mu} \\ \boldsymbol{\gamma} \end{bmatrix} \sim N(\mathbf{0}, \mathbf{G}) \tag{17}$$

and,

$$\boldsymbol{\epsilon} \sim N(\mathbf{0}, \mathbf{R}) \tag{18}$$

When the within-school variance is homoscedastic, **R** can be written as an $N \times N$ identity matrix multiplied by a scalar: **R** = I σ^2 .

For a heteroskedastic within-school variance, $\mathbf{R} = \mathbf{W}^{-\frac{1}{2}}\mathbf{I}\mathbf{W}^{-\frac{1}{2}}\sigma^2$, where **W** is a diagonal "weight" matrix:

$$\mathbf{W} = \begin{bmatrix} \frac{w_{i,j,t}}{N^{-1} \sum w_{i,j,t}} & 0 & \dots & 0\\ 0 & \frac{w_{i,j,t}}{N^{-1} \sum w_{i,j,t}} & 0 & \vdots\\ \vdots & 0 & \ddots & 0\\ 0 & \dots & 0 & \frac{w_{i,j,t}}{N^{-1} \sum w_{i,j,t}} \end{bmatrix}$$
(19)

where $w_{i,j,t} = \frac{m^{-1} \Sigma \hat{\sigma}_{i,j,t}^2}{\hat{\sigma}_{i,j,t}^2}$, and $\hat{\sigma}_{i,j,t}^2$ is the sample variance for the j^{th} school. Estimation is carried out using the HPMIXED procedure in SAS.

It should be noted that in continuing the analysis this year with the 2022-23 data, the model reestimated achievement for each year since having the additional year of data can refine estimates reported in 2022. Changes are typically very small at the state level but can be more noticeable for individual LEAs. Results that show graphs use the re-estimated values from 2023. To avoid confusion in comparing this year's results with last year's, results that show actual numbers, percentages, or effect sizes pull forward the metrics reported in 2022.

3.4 Four Key Metrics (Post-Estimation)

The specification of the model described above does not directly produce all the state- and school-level estimates of interest. However, these estimates can be obtained as linear combinations of the fixed and random model parameters described above. Although the relevant linear combinations can be represented as matrix calculations, a visual representation of the estimates using the example below is more straight forward.





In the graph above, the information can be interpreted as follows:

- The light blue shading indicates the time period for the **Pre-Pandemic Trend Line**. The orange shading indicates the time period for the **Pandemic Impact**, and the purple shading indicates the time period for the **Recovery Thresholds**.
- Average Scale Score (purple dots) represents the state's average achievement and is reported in scale scores. It is based on the average of school-level achievement. As a reminder, the average shown in these figures are adjusted for version changes.
- Achievement Trajectory (solid black line) is the path achievement in the state took since 2013. Visible as abrupt changes in direction, the trajectory is broken into three distinct parts:
 - Pre-pandemic Trend Line (2013 to 2020)

- Pandemic Response (2020 to 2021)
- Recovery Response (2021 to 2022 and 2022 to 2023)
- Extended Trend Line (dashed black line) is the Pre-Pandemic Trend Line, extended into the Pandemic and Recovery periods. It represents an estimate of state achievement had the state's pre-pandemic trend line (shown by the black solid line) continued into the post-COVID period (2021 to 2023).
- **3-Year-Average Threshold (dashed teal line)** is the three-year average achievement threshold representing the state's average achievement from 2017, 2018, and 2019.

In this report, the information in Figure 1 is summarized in six key metrics constructed from the estimated model parameters.

- The slope of the **Pre-Pandemic Trend Line** or the average year to year change in achievement over the years prior to the pandemic.
- The **Pandemic Impact** or the distance from the achievement trajectory and the extended trend line threshold evaluated at the 2021 time point.
- 2022 Distance to Full Recovery (extended trend line) or the distance from the achievement trajectory and the extended trendline threshold evaluated at the 2022 time point.
- 2023 Distance to Full Recovery (extended trend line) or the distance from the achievement trajectory and the extended trendline threshold evaluated at the 2023 time point.
- 2022 Distance to Full Recovery (3-year-average) or the distance from the achievement trajectory and the three-year average threshold evaluated at the 2022 time point.
- 2023 Distance to Full Recovery (3-year-average) or the distance from the achievement trajectory and the three-year average threshold evaluated at the 2023 time point.

Each metric has a state-wide version and a school-level version. The state-level version can be interpreted as the average over the school-level versions. For the state-level version, this report contains point estimates for each metric. The report does not contain estimated metrics for individual schools. Rather, the focus of the report is on characteristics of the school-level distributions for each metric.

3.5 Model Features

The model's features detailed above summarize changes in achievement over the pre- and post-pandemic period at both the state and school levels. Pandemic impact and recovery-related estimates reported herein account for trends in achievement observed prior the pandemic (i.e., pre-pandemic trends). These state and school pre-pandemic trends provide the basis for a valid and transparent counterfactual, especially in cases where *measured* student achievement is changing rapidly over the years prior to the pandemic. Furthermore, even though the pandemic disrupted testing administration at the end of 2020 in most locales, the extrapolation of the pre-pandemic trend provides estimates of what 2020 achievement would likely have been had it been observed. This is important because the pandemic did not impact in-person schooling until late in the 2019-2020 school year, and the observed level of achievement in that year would have otherwise constituted a natural baseline for estimating the impact of the pandemic. When the pre-pandemic trend is allowed to continue indefinitely, it provides both the basis for estimating the magnitude of the setback as well as a threshold for a full recovery. However, its relevance as counterfactual diminishes over time.

Another important feature of the model relates to the treatment of schools as random effects. In principle, the model could have treated school-level model parameters as fixed effects by fitting the model to each school's data individually. However, this could produce wildly imprecise estimates of school-level pandemic and recovery effects, especially in cases where the number of students in a school is small. In contrast, by assuming a parametric form for school-level model parameters, school-level estimates are also informed by associations estimated at the state level. This leads to estimates with smaller standard errors than would otherwise be obtained if each school's data were considered in isolation.

It is important to note that the enhanced precision obtained from random effect models come at the cost of increased reliance on parametric assumptions. Model fit was assessed visually for each subject and grade at the state level to see whether the pre-pandemic trend was being captured. When violations are detected that call into question the validity of the approach when applied to a particular assessment, this is noted in the report and remedial steps are taken when appropriate. Violations could be in the form of non-linear pre-pandemic trends or lack of data causing model fit issues.

Additionally with random effects, the school-specific estimates can be compared to the statewide averages or fixed effects upon which they are centered. They could also be compared to a value of zero, which, would indicate no pandemic setback or a flat pre-pandemic trend.

Finally, the model provides a comprehensive and concise summary of between and within school variation in student achievement in a manner that addresses a broad range of research questions.

For the state-level output, the model answers the following questions:

- For a specific subject/grade, what was the statewide trend in the school achievement in the years prior to the pandemic?
- Accounting for that trend, what was the pandemic impact on school-level achievement statewide in 2021?
- In 2022 and 2023, how far are schools (on average) from fully recovering the drop in achievement observed during the pandemic, and how much does that distance depend on the recovery threshold?

For school-level output, the model answers the following questions:

- How much variation is seen in the pre-pandemic trend of schools?
- Is there evidence that schools were differentially impacted by the pandemic?
- What proportion of schools have full recovered by 2022 and by 2023, and how much does it depend on the threshold used?

The <u>Results</u> section provides an overview of the modeling outputs and how to interpret them with these questions in mind.

3.6 Conversion of Metrics from Scale Score Units to Effect Size Units

In order to facilitate comparisons across assessments that employ different scaling units, the metrics are converted from scale score units to effect size units by dividing each metric estimate by the standard deviation of the scale-score in 2019. In all cases, this converts the metric into a standardized distance or "effect size." Effect sizes can also be classified as small, medium, or large to assist with interpretation. Various researchers have offered thoughts on what defines a small, medium, and large effect size.

- Cohen describes 0.20 as small, 0.50 as medium, and 0.80 as large (Cohen, Jacob. Statistical Power Analysis for the Behavioral Sciences. 2nd ed. Mahwah, NJ: Lawrence Erlbaum, 1988).
- Hattie describes an effect size of 0.40 as the average seen across all interventions, and 0.40 as the "hinge point" (Hattie, John, *Visible Learning: A Synthesis of Over 800 Meta-Analyses Relating to Achievement.* London: Routledge, 2008).
- Kraft suggested < 0.05 as small, 0.05 to 0.20 as medium, and > 0.20 as large based on the distributions of effect sizes and changes in achievement (Kraft MA. "Interpreting Effect Sizes of Education Interventions." *Educational Researcher*. 2020; 49 (4):241-253).

Most researchers agree that what constitutes a small, medium, or large effect size should be considered in the context of previous research conducted in a relevant field or substantive area of research. To place the findings in the context of the educational intervention literature, this report uses Kraft's definitions. This enables researchers to characterize a distance to a full recovery metric in terms of effect of a typical intervention in education (e.g., a tutoring program). For example, recent studies have indicated that double-dose Math instruction (where students have an extra period of Math) provided over the course of the entire year yields, on average, a positive effect size of about 0.20.²

Some of the results are also color-coded according to Kraft's definitions as shown below to assist with interpretation.

Color	Effect Size	Definition
	Large Negative	Student effect size is less than or equal to -0.20
	Medium	Student effect size is less than or equal to -0.05 but greater than
	Negative	-0.20
	Small Negative	Student effect size is less than 0.0 but greater than -0.05
	Small Positive	Student effect size is greater than 0.0 but less than +0.05
	Medium Positive	Student effect size is greater than or equal to +0.05 but less than
		+0.20
	Large Positive	Student effect size is greater than or equal to +0.20

² In an interview with "The 74," Harvard researcher Tom Kane referenced studies by Stephen W. Raudenbush and Takako Nomi's in Chicago schools and Eric Taylor's similar research in Miami-Dade schools. <u>https://www.the74million.org/article/harvard-economist-offers-gloomy-forecast-on-reversing-pandemic-learning-loss/</u>

4 Results

The model's three key metrics (Pre-Pandemic Trend, Pandemic Impact, and Distance to a Full Recovery) track the educational achievement of North Carolina students. Collectively, these metrics offer a comprehensive context to understand their collective achievement trajectory before, during, and after the pandemic.

These trajectories are summarized in two different formats: summary tables across assessments and visualizations for a specific assessment.

4.1 Summary Tables Across Assessments

- Table that shows the pre-pandemic trend, pandemic impact, and distance to each recovery threshold in 2022 and 2023 as an effect size for the state by assessment.
- Bar charts that show the pre-pandemic trend, pandemic impact, and distance to each recovery threshold in 2022 and 2023 as an effect size for the state by assessment. This is the same information as described in the previous bullet point but in a different format.
- Table that shows the percentage of schools in the state that obtained the following benchmarks by assessment:
 - The school's 2022 and 2023 achievement level met or exceeded the recovery threshold based on the three-year average achievement.
 - The school's 2022 and 2023 achievement level met or exceeded the recovery threshold based on the extended trend.
- Table that shows the percentage of schools in the state that obtained the following benchmarks by assessment:
 - The school's achievement level improved from 2021 to 2022 and from 2022 to 2023.
 - The schools' achievement level improved from 2021 to 2022 but declined from 2022 to 2023.
 - The school's achievement declined from 2021 to 2022 but improved from 2022 to 2023.
 - The school's achievement level declined from 2021 to 2022 and from 2022 to 2023.
 - The school's achievement level improved from 2021 to 2023.

4.2 Visualizations for a Specific Assessment

- Graph that shows the year-over-year trends and impacts from the pandemic for a specific assessment. This information is explained in <u>Section 3.3</u> and it is also summarized across assessments in the table showing pre-pandemic trends, pandemic impact, and distance to each recovery threshold and the bar chart described in <u>Section 4.1</u>.
- Table that shows the sample size (number of schools and students) and percentage of students meeting or exceeding proficiency for a specific assessment.
- Boxplot that shows the school-level distributions for the pre-pandemic trend, pandemic impact, and recovery thresholds for a specific assessment in 2022 and 2023.
- Table that shows the percentage of schools meeting recovery thresholds in 2022 and 2023. This information is summarized across assessments in 1.c.ii and 1.c.iii described above.

4.3 Reflections

When reviewing the results, consider the following questions for reflection:

- During the pre-pandemic period, was achievement increasing, decreasing, or holding steady?
- Assuming the Pre-Pandemic Trend had continued, what was the average magnitude of the pandemic's impact on state achievement across different assessments?
- Which assessments had the largest negative impact from the pandemic?
- In comparison to the Pre-Pandemic Trend extended to 2022, what is the continued gap in achievement?
- In comparison to the achievement threshold based on the most recent three-years prior to the pandemic, what is the continued gap in achievement?
- For each assessment, how does the magnitude of the pandemic compare to the remaining impact? Which assessments have closed the gap in achievement, and which are furthest from a full recovery?

5 Appendix: Charts and Tables for the Statewide Year-Over-Year Results

Charters and tables are presented for the following:

- 1. Statewide Pre-Pandemic Trend, Pandemic Impact, and Distance to Recovery
- 2. Statewide Pre-Pandemic Trend and Pandemic Impact by Assessment
- 3. Statewide Recovery (Extended Trend Threshold): by Year and Assessment
- 4. Statewide Recovery (3-Year Average Threshold): by Year and Assessment
- 5. Percentage of Schools that Met Recovery Thresholds (2022 and 2023)
- 6. Percentage of Schools by Recovery Trajectory Typology and Assessment
- 7. EOG Math 3 Year-Over-Year Trends and Impacts from the Pandemic
- 8. EOG Math 3 School-level Distributions on 6 Key Metrics
- 9. EOG Math 4 Year-Over-Year Trends and Impacts from the Pandemic
- 10. EOG Math 4 School-level Distributions on 6 Key Metrics
- 11. EOG Math 5 Year-Over-Year Trends and Impacts from the Pandemic
- 12. EOG Math 5 School-level Distributions on 6 Key Metrics
- 13. EOG Math 6 Year-Over-Year Trends and Impacts from the Pandemic
- 14. EOG Math 6 School-level Distributions on 6 Key Metrics
- 15. EOG Math 7 Year-Over-Year Trends and Impacts from the Pandemic
- 16. EOG Math 7 School-level Distributions on 6 Key Metrics
- 17. EOG Math 8 Year-Over-Year Trends and Impacts from the Pandemic
- 18. EOG Math 8 School-level Distributions on 6 Key Metrics
- 19. EOC Math 1 Year-Over-Year Trends and Impacts from the Pandemic
- 20. EOC Math 1 School-level Distributions on 6 Key Metrics
- 21. EOG Reading 3 Year-Over-Year Trends and Impacts from the Pandemic
- 22. EOG Reading 3 School-level Distributions on 6 Key Metrics
- 23. EOG Reading 4 Year-Over-Year Trends and Impacts from the Pandemic
- 24. EOG Reading 4 School-level Distributions on 6 Key Metrics
- 25. EOG Reading 5 Year-Over-Year Trends and Impacts from the Pandemic
- 26. EOG Reading 5 School-level Distributions on 6 Key Metrics
- 27. EOG Reading 6 Year-Over-Year Trends and Impacts from the Pandemic
- 28. EOG Reading 6 School-level Distributions on 6 Key Metrics
- 29. EOG Reading 7 Year-Over-Year Trends and Impacts from the Pandemic
- 30. EOG Reading 7 School-level Distributions on 6 Key Metrics
- 31. EOG Reading 8 Year-Over-Year Trends and Impacts from the Pandemic
- 32. EOG Reading 8 School-level Distributions on 6 Key Metrics

- 33. EOC English 2 Year-Over-Year Trends and Impacts from the Pandemic
- 34. EOC English 2 School-level Distributions on 6 Key Metrics
- 35. EOG Science 5 Year-Over-Year Trends and Impacts from the Pandemic
- 36. EOG Science 5 School-level Distributions on 6 Key Metrics
- 37. EOG Science 8 Year-Over-Year Trends and Impacts from the Pandemic
- 38. EOG Science 8 School-level Distributions on 6 Key Metrics
- 39. EOC Biology Year-Over-Year Trends and Impacts from the Pandemic
- 40. EOC Biology School-level Distributions on 6 Key Metrics

[See next section for Appendix]

Statewide Pre-Pandemic Trend, Pandemic Impact, and Distance to Recovery Reported as Effect Size

Assessment		Pre-Pandemic Trend	Pandemic Impact	Reco Extended Tre	overy: nd Threshold	Reco 3-Year Avera	overy: ge Threshold
Subject	Grade-Level	2013-2019	2021	2022	2023	2022	2023
Math	EOG Math 3	+0.01	-0.54	-0.26	-0.19	-0.21	-0.13
	EOG Math 4	+0.00	-0.50	-0.23	-0.15	-0.23	-0.14
	EOG Math 5	+0.00	-0.49	-0.27	-0.18	-0.24	-0.14
	EOG Math 6	+0.01	-0.47	-0.27	-0.24	-0.22	-0.18
	EOG Math 7	+0.00	-0.40	-0.27	-0.24	-0.24	-0.20
	EOG Math 8	-0.01	-0.48	-0.19	-0.16	-0.25	-0.23
	EOC Math 1	+0.05	-0.48	-0.31	-0.29	-0.10	-0.04
Reading	EOG Reading 3	-0.02	-0.18	-0.00	+0.03	-0.10	-0.08
	EOG Reading 4	-0.01	-0.24	-0.15	-0.07	-0.22	-0.17
	EOG Reading 5	-0.01	-0.15	-0.12	-0.07	-0.18	-0.16
	EOG Reading 6	-0.00	-0.17	-0.15	-0.12	-0.18	-0.16
	EOG Reading 7	-0.01	-0.16	-0.14	-0.12	-0.21	-0.21
	EOG Reading 8	-0.01	-0.11	-0.06	-0.04	-0.13	-0.13
	EOC English 2	-0.01	+0.10	+0.14	+0.14	+0.06	+0.04
Science	EOG Science 5	+0.03	-0.55	-0.36	-0.35	-0.20	-0.16
	EOG Science 8	+0.03	-0.32	-0.23	-0.33	-0.09	-0.15
	EOC Biology	+0.01	-0.33	-0.21	-0.16	-0.15	-0.09

Note: For a detailed look at the findings for a particular assessment, go to the corresponding page containing the detailed results for that assessment.

Color	Effect Size	Definition
	Large Negative	Student effect size is less than or equal to -0.20
	Medium Negative	Student effect size is less than or equal to05 but greater than -0.20
	Small Negative	Student size is less than 0 but greater than -0.05
	Small Positive	Student size is greater than 0 but less than +0.05
	Medium Positive	Student size is greater than or to equal to +0.05 but less than +0.20
	Large Positive	Student size is greater than or equal to +0.20

Statewide Pre-Pandemic Trend and Pandemic Impact by Assessment Reported as Effect Size



Statewide Recovery (Extended Trend Threshold): by Year and Assessment Reported as Effect Size



Statewide Recovery (3-Year Average Threshold): by Year and Assessment Reported as Effect Size



Percentage of Schools that Met Recovery Thresholds (2022 and 2023) by Assessment

	Assessment	Recovery Thresho	ld: Extended Trend	Recovery Threshold: 3-Year Average		
Subject	Grade-Level	2022	2023	2022	2023	
Math	EOG Math 3	13.37%	23.21%	13.58%	27.18%	
	EOG Math 4	13.08%	26.30%	9.35%	22.03%	
	EOG Math 5	8.48%	21.52%	8.98%	23.70%	
	EOG Math 6	8.18%	14.37%	10.58%	14.94%	
	EOG Math 7	5.92%	12.24%	3.90%	11.06%	
	EOG Math 8	23.78%	29.76%	11.28%	17.95%	
	EOC Math 1	11.84%	15.67%	30.78%	43.46%	
Reading	EOG Reading 3	48.49%	59.55%	18.86%	25.62%	
	EOG Reading 4	11.32%	27.15%	2.67%	8.60%	
	EOG Reading 5	13.86%	28.19%	5.10%	8.26%	
	EOG Reading 6	8.46%	13.06%	3.81%	8.18%	
	EOG Reading 7	13.44%	18.44%	4.77%	4.13%	
	EOG Reading 8	25.48%	35.68%	11.57%	11.84%	
	EOC English 2	85.40%	83.02%	71.99%	63.29%	
Science	EOG Science 5	5.60%	11.16%	21.19%	28.62%	
	EOG Science 8	9.37%	7.05%	31.63%	23.69%	
	EOC Biology	13.18%	24.91%	18.49%	33.22%	

Percentage of Schools by Recovery Trajectory Typology and Assessment

Assessment				Overall Improvement		
Subject	Grade-Level	Mult-year Improvement (2022 and 2023)	Improve (2022) Then Decline (2023)	Decline (2022) Then Improve (2023)	Mult-year Decline (2022 and 2023)	2021 to 2023
Math	EOG Math 3	65.15%	28.60%	4.68%	1.56%	94.82%
	EOG Math 4	73.63%	23.67%	2.20%	0.50%	96.23%
	EOG Math 5	73.99%	19.86%	5.43%	0.72%	96.52%
	EOG Math 6	61.35%	35.34%	2.87%	0.43%	96.12%
	EOG Math 7	59.14%	32.01%	5.90%	2.95%	89.38%
	EOG Math 8	51.81%	42.05%	5.35%	0.79%	91.02%
	EOC Math 1	56.35%	30.29%	9.90%	3.46%	89.71%
Reading	EOG Reading 3	56.21%	40.95%	1.92%	0.92%	95.39%
	EOG Reading 4	66.31%	19.83%	11.37%	2.49%	91.33%
	EOG Reading 5	33.99%	23.55%	32.32%	10.14%	67.32%
	EOG Reading 6	38.16%	22.53%	28.84%	10.47%	73.74%
	EOG Reading 7	26.40%	30.97%	24.78%	17.85%	58.70%
	EOG Reading 8	37.33%	30.43%	17.99%	14.24%	65.22%
	EOC English 2	18.87%	34.48%	19.73%	26.93%	46.31%
Science	EOG Science 5	55.94%	34.86%	6.30%	2.90%	90.51%
	EOG Science 8	23.99%	59.22%	7.20%	9.60%	64.17%
	EOC Biology	56.06%	25.61%	14.71%	3.63%	85.12%

Year-Over-Year Trends and Impacts from the Pandemic EOG Math 3

Achievement Over Time Scale Score School Year

Sample Size and % Proficient

		Year									
Statistic	2013	2014	2015	2016	2017	2018	2019	2020	2021	2022	2023
# Schools	1,334	1,352	1,372	1,391	1,407	1,419	1,421		1,421	1,421	1,409
# Students	93,795	100,765	104,916	108,808	109,682	113,101	109,237		95,729	101,307	101,980
% Proficient	50.1%	64.3%	63.4%	65.7%	64.1%	64.5%	65.7%		44.6%	56.4%	59.5%



School-level Distributions on 6 Key Metrics EOG Math 3



School-Level Distributions for the Pre-Pandemic Trend, Pandemic Impact, and Recovery Thresholds

Recovery Threshold Year	Percentage of Schools Meeting Recovery Thresholds in 2022	Percentage of Schools Meeting Recovery Thresholds in 2023
Extended Trend	13.37%	23.21%
3-Year Average	13.58%	27.18%

Year-Over-Year Trends and Impacts from the Pandemic EOG Math 4



Statistic	2013	2014	2015	2016	2017	2018	2019	2020	2021	2022	2023
# Schools	1,330	1,343	1,365	1,383	1,402	1,422	1,422		1,422	1,422	1,407
# Students	101,770	94,958	104,335	106,923	111,229	116,005	114,399		97,821	102,655	102,852
% Proficient	50.7%	57.2%	57.2%	57.7%	58.4%	56.9%	58.5%		37.3%	49.0%	53.9%



School-level Distributions on 6 Key Metrics EOG Math 4



School-Level Distributions for the Pre-Pandemic Trend, Pandemic Impact, and Recovery Thresholds

Recovery Threshold Year	Percentage of Schools Meeting Recovery Thresholds in 2022	Percentage of Schools Meeting Recovery Thresholds in 2023
Extended Trend	13.08%	26.30%
3-Year Average	9.35%	22.03%

Year-Over-Year Trends and Impacts from the Pandemic EOG Math 5



Sample Size and % Proficient

		Year									
Statistic	2013	2014	2015	2016	2017	2018	2019	2020	2021	2022	2023
# Schools	1,290	1,306	1,326	1,346	1,366	1,390	1,391		1,392	1,392	1,380
# Students	100,296	101,266	96,808	104,013	108,037	115,614	115,781		99,642	103,033	103,320
% Proficient	51.3%	59.7%	58.9%	61.1%	60.8%	59.4%	61.6%		41.7%	50.8%	54.9%



averaged over all schools in the state (2017-2019).

School-level Distributions on 6 Key Metrics EOG Math 5



School-Level Distributions for the Pre-Pandemic Trend, Pandemic Impact, and Recovery Thresholds

Recovery Threshold Year	Percentage of Schools Meeting Recovery Thresholds in 2022	Percentage of Schools Meeting Recovery Thresholds in 2023			
Extended Trend	8.48%	21.52%			
3-Year Average	8.98%	23.70%			

Year-Over-Year Trends and Impacts from the Pandemic EOG Math 6



Sample Size and % Proficient

		Year									
Statistic	2013	2014	2015	2016	2017	2018	2019	2020	2021	2022	2023
# Schools	612	628	651	665	682	708	707		709	709	696
# Students	101,644	100,409	104,283	97,413	105,031	112,463	115,567		102,085	103,666	101,846
% Proficient	42.2%	50.1%	50.0%	53.2%	53.7%	52.4%	60.1%		40.1%	49.7%	51.5%



averaged over all schools in the state (2017-2019).

School-level Distributions on 6 Key Metrics EOG Math 6



School-Level Distributions for the Pre-Pandemic Trend, Pandemic Impact, and Recovery Thresholds

Recovery Threshold Year	Percentage of Schools Meeting Recovery Thresholds in 2022	Percentage of Schools Meeting Recovery Thresholds in 2023		
Extended Trend	8.18%	14.37%		
3-Year Average	10.58%	14.94%		

Year-Over-Year Trends and Impacts from the Pandemic EOG Math 7



Sample Size and % Proficient

	Year										
Statistic	2013	2014	2015	2016	2017	2018	2019	2020	2021	2022	2023
# Schools	598	614	628	646	660	690	690		692	692	678
# Students	102,880	104,775	106,074	106,983	99,877	110,758	112,925		104,100	108,426	103,909
% Proficient	41.9%	49.4%	48.6%	50.1%	51.1%	51.9%	59.7%		42.3%	48.3%	49.5%



averaged over all schools in the state (2017-2019).

School-level Distributions on 6 Key Metrics EOG Math 7



School-Level Distributions for the Pre-Pandemic Trend, Pandemic Impact, and Recovery Thresholds

Recovery Threshold Year	Percentage of Schools Meeting Recovery Thresholds in 2022	Percentage of Schools Meeting Recovery Thresholds in 2023		
Extended Trend	5.92%	12.24%		
3-Year Average	3.90%	11.06%		

Year-Over-Year Trends and Impacts from the Pandemic EOG Math 8

Achievement Over Time Scale Score School Year

Sample Size and % Proficient

		Year									
Statistic	2013	2014	2015	2016	2017	2018	2019	2020	2021	2022	2023
# Schools	579	581	603	616	630	654	651		656	656	635
# Students	68,291	71,090	78,480	77,354	78,515	72,915	76,213		72,788	79,350	76,381
% Proficient	22.5%	30.8%	30.1%	30.3%	31.5%	29.6%	36.8%		15.9%	26.1%	27.2%



School-level Distributions on 6 Key Metrics EOG Math 8



School-Level Distributions for the Pre-Pandemic Trend, Pandemic Impact, and Recovery Thresholds

Recovery Threshold Year	Percentage of Schools Meeting Recovery Thresholds in 2022	Percentage of Schools Meeting Recovery Thresholds in 2023		
Extended Trend	23.78%	29.76%		
3-Year Average	11.28%	17.95%		

Year-Over-Year Trends and Impacts from the Pandemic EOC Math 1

Achievement Over Time



Sample Size and % Proficient

		Year									
Statistic	2013	2014	2015	2016	2017	2018	2019	2020	2021	2022	2023
# Schools	944	972	984	1,002	1,023	1,042	1,027		1,056	1,056	1,040
# Students	110,902	109,401	113,645	119,750	117,614	117,369	114,851		110,358	119,127	121,771
% Proficient	37.5%	55.0%	51.8%	52.8%	56.7%	57.2%	56.4%		39.3%	45.9%	48.4%



School-level Distributions on 6 Key Metrics EOC Math 1



School-Level Distributions for the Pre-Pandemic Trend, Pandemic Impact, and Recovery Thresholds

Recovery Threshold Year	Percentage of Schools Meeting Recovery Thresholds in 2022	Percentage of Schools Meeting Recovery Thresholds in 2023		
Extended Trend	11.84%	15.67%		
3-Year Average	30.78%	43.46%		

Year-Over-Year Trends and Impacts from the Pandemic EOG Reading 3





Sample Size and % Proficient

		Year									
Statistic	2013	2014	2015	2016	2017	2018	2019	2020	2021	2022	2023
# Schools	1,334	1,352	1,372	1,391	1,407	1,419	1,421		1,421	1,421	1,409
# Students	93,292	100,010	104,902	108,788	109,672	113,128	109,292		96,014	101,419	102,032
% Proficient	48.3%	63.8%	60.2%	58.5%	58.5%	56.4%	57.2%		45.1%	46.5%	48.0%



School-level Distributions on 6 Key Metrics EOG Reading 3



School-Level Distributions for the Pre-Pandemic Trend, Pandemic Impact, and Recovery Thresholds

Recovery Threshold Year	Percentage of Schools Meeting Recovery Thresholds in 2022	Percentage of Schools Meeting Recovery Thresholds in 2023		
Extended Trend	48.49%	59.55%		
3-Year Average	18.86%	25.62%		

Year-Over-Year Trends and Impacts from the Pandemic EOG Reading 4

Achievement Over Time Scale Score School Year

Sample Size and % Proficient

		Year											
Statistic	2013	2014	2015	2016	2017	2018	2019	2020	2021	2022	2023		
# Schools	1,330	1,343	1,365	1,383	1,402	1,422	1,422		1,422	1,422	1,407		
# Students	101,029	94,633	104,351	106,964	111,252	116,046	114,441		98,082	102,794	102,875		
% Proficient	46.6%	58.5%	59.7%	58.2%	57.2%	56.4%	54.9%		45.7%	49.5%	53.2%		



averaged over all schools in the state (2017-2019).

School-level Distributions on 6 Key Metrics EOG Reading 4



School-Level Distributions for the Pre-Pandemic Trend, Pandemic Impact, and Recovery Thresholds

Recovery Threshold Year	Percentage of Schools Meeting Recovery Thresholds in 2022	Percentage of Schools Meeting Recovery Thresholds in 2023		
Extended Trend	11.32%	27.15%		
3-Year Average	2.67%	8.60%		

Year-Over-Year Trends and Impacts from the Pandemic EOG Reading 5



Sample Size and % Proficient

	Year										
Statistic	2013	2014	2015	2016	2017	2018	2019	2020	2021	2022	2023
# Schools	1,290	1,306	1,326	1,346	1,366	1,390	1,391		1,392	1,392	1,380
# Students	99,532	100,850	96,853	104,109	108,104	115,754	115,897		99,707	103,219	103,374
% Proficient	42.3%	56.7%	53.9%	55.5%	56.5%	52.5%	52.2%		42.9%	44.2%	45.9%



averaged over all schools in the state (2017-2019).

School-level Distributions on 6 Key Metrics EOG Reading 5



School-Level Distributions for the Pre-Pandemic Trend, Pandemic Impact, and Recovery Thresholds

Recovery Threshold Year	Percentage of Schools Meeting Recovery Thresholds in 2022	Percentage of Schools Meeting Recovery Thresholds in 2023		
Extended Trend	13.86%	28.19%		
3-Year Average	5.10%	8.26%		

Year-Over-Year Trends and Impacts from the Pandemic EOG Reading 6



Sample Size and % Proficient

	Year										
Statistic	2013	2014	2015	2016	2017	2018	2019	2020	2021	2022	2023
# Schools	612	628	651	665	682	708	707		709	709	697
# Students	101,071	100,074	104,369	97,491	105,148	112,560	115,734		102,192	103,769	101,928
% Proficient	50.0%	60.4%	58.5%	59.6%	61.4%	60.8%	58.7%		45.8%	46.2%	47.7%



School-level Distributions on 6 Key Metrics EOG Reading 6



School-Level Distributions for the Pre-Pandemic Trend, Pandemic Impact, and Recovery Thresholds

Recovery Threshold Year	Percentage of Schools Meeting Recovery Thresholds in 2022	Percentage of Schools Meeting Recovery Thresholds in 2023			
Extended Trend	8.46%	13.06%			
3-Year Average	3.81%	8.18%			

Year-Over-Year Trends and Impacts from the Pandemic EOG Reading 7



Sample Size and % Proficient

	Year											
Statistic	2013	2014	2015	2016	2017	2018	2019	2020	2021	2022	2023	
# Schools	598	614	628	646	660	690	690		692	692	678	
# Students	102,425	104,491	106,124	107,091	99,981	110,907	113,061		104,391	108,580	104,014	
% Proficient	51.3%	60.8%	57.4%	59.2%	58.8%	59.9%	57.6%		47.1%	47.7%	48.5%	



averaged over all schools in the state (2017-2019).

School-level Distributions on 6 Key Metrics EOG Reading 7



School-Level Distributions	for the Pre-Pandemic T	rend, Pandemic Impa	ct, and Recover	y Thresholds
		,		

Recovery Threshold Year	Percentage of Schools Meeting Recovery Thresholds in 2022	Percentage of Schools Meeting Recovery Thresholds in 2023			
Extended Trend	13.44%	18.44%			
3-Year Average	4.77%	4.13%			

Year-Over-Year Trends and Impacts from the Pandemic EOG Reading 8



Sample Size and % Proficient

	Year											
Statistic	2013	2014	2015	2016	2017	2018	2019	2020	2021	2022	2023	
# Schools	598	605	627	640	655	681	680		683	683	667	
# Students	100,759	103,168	108,364	106,857	107,897	104,108	110,902		104,818	110,725	108,284	
% Proficient	43.9%	57.4%	54.8%	54.4%	54.2%	54.1%	54.7%		48.7%	50.0%	50.1%	



averaged over all schools in the state (2017-2019).

School-level Distributions on 6 Key Metrics EOG Reading 8



School-Level Distributions for the Pre-Pandemic Trend, Pandemic Impact, and Recovery Thresholds

Recovery Threshold Year	Percentage of Schools Meeting Recovery Thresholds in 2022	Percentage of Schools Meeting Recovery Thresholds in 2023		
Extended Trend	25.48%	35.68%		
3-Year Average	11.57%	11.84%		

Year-Over-Year Trends and Impacts from the Pandemic EOC English 2

Achievement Over Time



Sample Size and % Proficient

	Year											
Statistic	2013	2014	2015	2016	2017	2018	2019	2020	2021	2022	2023	
# Schools	507	527	535	552	569	583	588		589	589	583	
# Students	102,939	105,637	110,858	113,554	114,655	114,690	115,502		105,629	111,860	115,002	
% Proficient	52.9%	63.4%	60.3%	59.2%	60.9%	59.7%	59.4%		58.2%	57.9%	58.1%	



School-level Distributions on 6 Key Metrics EOC English 2



School-Level Distributions for the Pre-Pandemic Trend, Pandemic Impact, and Recovery Thresholds

Recovery Threshold Year	Percentage of Schools Meeting Recovery Thresholds in 2022	Percentage of Schools Meeting Recovery Thresholds in 2023
Extended Trend	85.40%	83.02%
3-Year Average	71.99%	63.29%

Year-Over-Year Trends and Impacts from the Pandemic EOG Science 5

Achievement Over Time



Sample Size and % Proficient

	Year										
Statistic	2013	2014	2015	2016	2017	2018	2019	2020	2021	2022	2023
# Schools	1,290	1,306	1,326	1,346	1,366	1,390	1,391		1,392	1,392	1,380
# Students	100,100	101,340	96,909	104,070	108,137	115,752	115,879		99,623	103,207	103,401
% Proficient	48.5%	67.4%	65.9%	72.2%	70.4%	68.4%	71.4%		53.3%	62.6%	64.5%



School-level Distributions on 6 Key Metrics EOG Science 5



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Recovery Threshold Year	Percentage of Schools Meeting Recovery Thresholds in 2022	Percentage of Schools Meeting Recovery Thresholds in 2023		
Extended Trend	5.60%	11.16%		
3-Year Average	21.19%	28.62%		

Year-Over-Year Trends and Impacts from the Pandemic EOG Science 8

Achievement Over Time



Sample Size and % Proficient

	Year										
Statistic	2013	2014	2015	2016	2017	2018	2019	2020	2021	2022	2023
# Schools	598	605	626	640	654	681	679		683	683	667
# Students	100,977	103,392	108,326	106,844	107,875	104,067	110,808		104,272	110,523	108,242
% Proficient	62.7%	74.8%	74.3%	75.2%	76.7%	76.4%	78.8%		70.3%	73.5%	70.2%



School-level Distributions on 6 Key Metrics EOG Science 8



School-Level Distributions for the Pre-Pandemic	Trend, Pandemic Impac	t, and Recovery Thresholds
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Recovery Threshold Year	Percentage of Schools Meeting Recovery Thresholds in 2022	Percentage of Schools Meeting Recovery Thresholds in 2023
Extended Trend	9.37%	7.05%
3-Year Average	31.63%	23.69%

Year-Over-Year Trends and Impacts from the Pandemic EOC Biology

Achievement Over Time



Sample Size and % Proficient

	Year										
Statistic	2013	2014	2015	2016	2017	2018	2019	2020	2021	2022	2023
# Schools	513	521	539	545	561	578	581		584	584	578
# Students	101,914	103,035	108,121	111,611	109,511	111,993	110,258		100,284	106,746	111,203
% Proficient	47.3%	55.9%	54.3%	55.9%	56.2%	58.2%	59.2%		45.2%	51.7%	53.6%



School-level Distributions on 6 Key Metrics EOC Biology



School-Level Distributions for the Pre-Pandemic Trend, Pandemic Impact, and Recovery Thresholds

Recovery Threshold Year	Percentage of Schools Meeting Recovery Thresholds in 2022	Percentage of Schools Meeting Recovery Thresholds in 2023
Extended Trend	13.18%	24.91%
3-Year Average	18.49%	33.22%