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GED® College Readiness Benchmarks and Post-Secondary Success

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GED[®] College Readiness Benchmarks and Post-Secondary Success

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Abstract

In 2016, the GED[®] introduced college readiness benchmarks designed to identify testers who are academically prepared for credit-bearing college coursework. The benchmarks are promoted as awarding college credits or exempting "college-ready" GED[®] graduates from remedial coursework. I show descriptive evidence that those identified as college-ready by these benchmarks enroll and persist in college at significantly higher rates than others who pass the GED[®] exam, but at lower rates than recent graduates with traditional high school diplomas. Regression discontinuity estimates show that crossing a college readiness threshold does not substantially influence testers' college enrollment or persistence during the two years following their first test attempt. Relatedly, I observe little exam retaking by those who fall narrowly short of the minimum college readiness score thresholds. This contrasts strongly with retaking behavior near the lower GED[®] passing threshold that determines eligibility for a high school equivalency credential. Those who narrowly fail a GED[®] subject test are over 100 times more likely to retest than those who fall just short of a college readiness benchmark in the same subject. GED[®] college readiness benchmarks do not currently appear to promote better college outcomes, but in the absence of more detailed test score information they offer a simple heuristic to predict short-run college enrollment and persistence among GED[®] graduates, particularly for those who identify educational gain as a primary reason for testing. The results highlight the promise and challenges associated with building pathways for non-traditional students to earn credit for prior learning.

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1 Introduction

High school equivalency (HSE) credentials offer an alternative pathway for individuals who lack a traditional high school diploma to meet requirements for many jobs and training opportunities in the United States. The GED[®] is the oldest and most popular HSE examination in the United States, awarding passers with a credential that is generally accepted as meeting minimum high school graduation requirements that are a prerequisite for many jobs and nearly all post-secondary institutions nationwide. In 2014, GED Testing Service, LLC. (GEDTS) introduced the 5th edition of the GED[®] exam, transitioning to a predominantly computer-based format, increasing test rigor to align content with national College and Career Readiness Standards, and introducing the GED[®] Honors designation for testers who achieved a subject test score of 170 or higher. The 5th edition of the exam emphasized the GED[®] as a "stepping-stone toward a college classroom or a better career and a family sustaining wage," acknowledging the GED[®] exam's role as "no longer an endpoint for adults, but a springboard for more education, training, and better-paying jobs" (GEDTS, 2014b). In 2016, based on data from early cohorts who passed the revised exam, GEDTS adjusted the overall passing threshold for the exam's subject tests down from 150 to 145, and replaced GED[®] Honors with two new college readiness benchmarks: GED[®] College Ready and GED[®] College Ready Plus Credits for individuals who earned a minimum subject test score of 165 and 175, respectively (Gewertz, 2016).

The college readiness benchmarks were introduced alongside recommendations by the American Council on Education that students who reach a GED[®] College Ready score threshold in a subject be exempt from remedial coursework in that subject, and that those who reach a GED[®] College Ready Plus Credits score threshold be awarded credit for prior learning in the form of lower division college credits in that subject. These recommendations highlight the potential role of HSE credentials as an alternative pathway to post-secondary education, and added the GED[®] to an emerging marketplace of examinations, training programs, and credentials that offer the promise of credit for prior learning in a growing number of post-secondary institutions (Boatman et al., 2020; Klein-Collins et al., 2020; The ACE National Guide, 2021).

Using data from 1.2 million people who took a $\text{GED}^{\mathbb{R}}$ subject test in the United States between 2014 and 2019, this study is the first to describe the modern "college-ready" population of $\text{GED}^{\mathbb{R}}$ testers. This study is also the first to quasi-experimentally examine the causal impact of earning a $\text{GED}^{\mathbb{R}}$ college readiness designation on educational outcomes, linking a random subsample of roughly 30,000 GED[®] testers to National Student Clearinghouse (NSC) college enrollment and degree records. This paper also contributes to the nascent causal literature examining the effectiveness of assessments that award college credit for prior learning.

On average, I find that "college-ready" GED[®] recipients enroll and persist in college at significantly higher rates than others who pass the GED[®] exam but at lower rates than recent graduates with traditional high school diplomas. Estimating impacts a regression discontinuity framework, I find that reaching college readiness benchmarks that are promoted as awarding college credits or exempting "college-ready" GED[®] recipients from remedial coursework do not appear to substantively influence testers' college enrollment or persistence behaviors. I test for heterogeneity in treatment effects by individuals' demographic characteristics, stated motivation for testing, and rates of American Council on Education (ACE)¹ membership among local post-secondary institutions, and I find no consistent evidence that any particular subgroups were differentially impacted by crossing a GED[®] college readiness benchmarks.

Relatedly, testers who fall narrowly short of reaching a GED[®] college readiness threshold exhibit strikingly different retesting behaviors than their peers who fall just short of reaching a minimum subject test *passing* threshold, which testers must meet or exceed to earn a high school equivalency credential. Those who narrowly fail a GED[®] subject test are over 100 times more likely to retest at least once than those who fall just short of earning a college readiness designation in the same subject. This suggests that while GED[®] testers who score near a subject test passing threshold place a high value on earning a credential, higher scoring testers either face constraints to retesting or do not expect similar returns to effort expended toward earning a college readiness designation.

The remainder of this paper is organized as follows: in Section 2, I decribe past research assessing the impact of earning a $\text{GED}^{\textcircled{\text{B}}}$ credential. In Section 3, I provide background on high school equivalency exams in the United States. In Section 4, I describe the data sources and present key descriptive statistics for the sample. In Section 5, I present my empirical strategy and econometric models. I present the main results, heterogeneity analyses, and robustness checks in Section 6, and conclude in Section 7 with a discussion of the results and policy implications.

¹The American Council on Education is a parent organization of GED Testing Service, LLC in a joint venture with Pearson, but ACE does not participate in the creation or production of the GED[®] exam. ACE issues recommendations on credit for prior learning to its member institutions based on third-party evaluations of credentialing and training programs, including the GED[®] exam.

2 Literature Review

Evidence from past thirty years of research on the impacts of the GED[®] on recipients is mixed, but no causal research examines recent cohorts of testers, the most recent edition of the exam, or the GED[®] college readiness designations. Early work comparing the educational and labor market outcomes of exam-certified high school equivalents and traditional high school graduates found that on average, individuals who earned HSE credentials in the 1980s substantially underperformed their peers with traditional high school diplomas in the labor market and educational attainment, even after conditioning on a measure of academic ability. Furthermore, while those with HSE credentials outperformed uncredentialed high school dropouts in the raw data, their outcomes were indistinguishable from uncredentialed dropouts after accounting for years of completed schooling (Cameron & Heckman, 1993).

Several studies, including Cameron and Heckman (1993), used the National Longitudinal Study of Youth (NLSY79), which follows a nationally representative sample of young people who were aged 14-22 in 1979 as they transitioned from adolesence to into adulthood, to compare the educational and labor market outcomes of GED[®] graduates to those of their peers with and without traditional high school diplomas. Studies examining NLSY79 data and other nationally representative longitudinal data sources like High School and Beyond have explored the relationship between GED[®] receipt and life outcomes for a variety of subgroups, including men (Murnane, Willett, & Boudett 1995), women (Boudett, Murnane, & Willett 2000), immigrants (Clark & Jaeger 2006), prisoners (Nuttall, Hollmen, & Staley 2003; Darolia, Mueser, & Cronin, 2021), and students with disabilities (Wagner et al., 2005). While some of these studies found positive changes in earnings trajectories or educational attainment following GED[®] receipt, particularly for individuals with lower measures of baseline academic skill, others found null results, with no clear consensus as to whether, or for whom, HSE credentials improve recipients' life outcomes (for a review, see Heckman, Humphries, & Mader, 2011 or Heckman, Humphries, & Kautz, 2014).

The earliest quasi-experimental work measuring the signaling value of earning a high school equivalency credential by passing the $\text{GED}^{\mathbb{R}}$ focused on men's labor market outcomes and found large increases in wages for marginal $\text{GED}^{\mathbb{R}}$ recipients who were white, but not for others in a sample from 45 states (Tyler, Murnane, & Willett, 2000). However, recent work examining the causal impact of HSE certification on educational and labor market outcomes calls into questions the internal validity of past quasi-experimental analyses that condition on final score (after en-

dogenous retaking) rather than first-attempt GED[®] subject test scores to generate their estimates (including Tyler, Murnane, & Willett, 2000). Jepsen, Mueser, and Troske (2016, 2017) account for the endogeity of exam retaking to measure the impact of passing the GED[®] exam in Missouri using a regression discontinuity framework and find no evidence of lasting educational or labor market returns for marginal HSE credential recipients. Heller and Slungaard Mumma (2019) incorporate these methodological innovations to assess the role of the GED[®] in promoting educational attainment in Massachusetts and find that passing the GED[®] has large, positive effects on the college enrollment and persistence behaviors of public adult education students, but null effects for dropouts who do not participate in adult education.

Given that most studies of the GED[®]'s effects are based on distant cohorts and long-discontinued exams, the generalizability of these studies deserves careful consideration. If differences in exam format, rigor, K-12 schooling, or the larger social and economic context influence selection into the population of exam certified high school equivalents or the way such credentials are viewed and used in the labor market or higher education, it is uncertain how past evidence will generalize over time and space. Since the NLSY79 analyses, two additional HSE exams have come into use at the state and national level, and the GED[®] has been revised twice, transitioning to a computer-based format and introducing two college readiness benchmarks. Jepsen, Mueser, and Troske's (2016, 2017) evaluation relies upon data from Missouri cohorts who took the GED[®] exam between 2001 and 2005, straddling the third edition of the exam (discontinued in 2002) and the fourth edition (discontinued in 2013). Heller and Slungaard Mumma (2019) examine data from Massachusetts GED[®] testers who took the fourth edition of the exam between 2007 and 2013. Additionally, Jepsen, Mueser, and Troske's (2016) methodological innovations cast doubt upon the internal validity of causal estimates from the most recent attempt to examine the causal impacts of passing the GED[®] in a national sample (Tyler, Murnane, & Willett, 2000), which itself followed cohorts who took the exam over 30 years ago.

Historically, theorists have argued that taking and passing the GED[®] is a mixed signal, conveying higher levels of academic skill than the average dropout, but lower levels of non-academic skills than traditional high school graduates (Heckman & Rubinstein, 2001; Araujo, Gottlieb, & Moreira, 2007). However, it is unclear how higher levels of GED[®] certification are viewed by hiring managers, admissions officers, or testers themselves, nor how the national population of exam certified high school equivalents or the overall returns to HSE credentials have changed over the past 10 to 20 years.

3 High School Equivalency Exams in the United States

Over 24 million adults in the United States (nearly 10% of the adult population) do not have a high school diploma or equivalent credential (U.S. Department of Education, 2023). HSE credentials represent the primary "second-chance" pathway for these adults to meet minimum requirements for many jobs or educational opportunities. The GED[®] exam was the first HSE test in the United States, developed in 1942 to provide soldiers returning from World War II with an alternative pathway to finish their secondary education upon their return (GEDTS, 2017). However, demand for alternative credentials extended far beyond the military, and over 20 million people have earned an HSE credential by passing the GED[®] exam since its release (GEDTS, 2014a).²

The modern GED[®] exam comprises a battery of four timed subject tests covering high schoollevel content: Mathematical Reasoning, Reasoning Through Language Arts, Science, and Social Studies. Tests are designed to assess whether testers have developed content mastery and critical thinking skills that are equivalent to the academic skills required to earn a high school diploma. To pass the GED[®] exam, individuals must reach or exceed minimum score thresholds on each subject test.³ Testers who score below the passing threshold on one or more subject tests may retake individual subject tests until they pass and combine their highest scores on each subject test to pass the exam. Rules regarding the timing (e.g., required "cool-down periods" between tests or caps on the number of subject test attempts in a year) and cost of retesting vary by state.

Introduced in 2016, the GED[®] College Ready (GED[®] CR) and College Ready plus credits (GED[®] CR+C) designations are awarded when a student's subject test score crosses higher thresholds: 165 for GED[®] CR and 175 for GED[®] CR+C. These subject-specific benchmarks are designed to identify students whose exam performance indicates that they are academically prepared to succeed in credit-bearing, college-level coursework in that subject. The levels of the GED[®] CR and CR+C benchmarks were set based on recommendations from the American Council on Education (ACE) that meeting a GED[®] CR benchmark "[...]enables the student to be waived from all developmental education requirements or courses and placement testing in that content area..." and that a student who reaches the GED[®] CR+C threshold in a given subject "[...]has demonstrated

²In their 2014 Annual Statistical Report on the GED[®] Test, GEDTS estimates that "[p]assers in 2013 join the approximately 20 million candidates who have passed the GED[®] test in its history," and GED.com advertised "over 20 million graduates and counting" on its homepage in March, 2021.

³Past editions of the GED[®] also required testers to reach a minimum total score summing over all subject test scores, in addition to meeting minumum subtest thresholds. The minimum total score exceeded the sum of the minimum subject test passing thresholds, requiring testers to achieve an average subtest score slightly higher than the minimum passing score for any given subtest to earn a credential.

skills reflective of the appropriate content, scope, and rigor for college credit recommendations" (The ACE National Guide, 2021). GEDTS estimates that reaching a GED[®] CR benchmark is roughly equivalent to scoring in the top quartile of the general U.S. population of graduating high school seniors in that subject, and that in each subject, less than 10% of graduating high school seniors would reach the GED[®] CR+C score threshold (GEDTS, 2016).

For decades, passing the GED[®] was the only widely recognized path to earn an HSE credential in the United States, however in recent years two alternative HSE exams have emerged. The TASC and HiSET exams were developed in 2013 by Data Recognition Corporation (TASC) and ETS (HiSET) in response to the changes introduced with the fifth edition of the GED. In 2014, when the GED[®] released the 5th edition of the exam, transitioning to an exclusively computer-based format, increasing test rigor to align with national College and Career Readiness Standards, and increasing test fees by an average of roughly 70%, some states responded to these changes by seeking out alternative assessments (Auslen, 2013). TASC replaced the GED[®] in both Indiana and New York state ("Move over G.E.D.", 2013), and several states (including Montana, New Hampshire, Massachusetts, Missouri, and Tennessee) adopted the HiSET as their preferred HSE exam (Auslen, 2013). Since 2014, twenty-six states have decided to supplement or replace the GED[®] by offering and accepting one or both of HiSET and TASC as alternative HSE examinations (Data Recognition Corporation, 2020; ETS, 2020b). During some or all of the study period, forty states and the District of Columbia continued to administer the GED[®] exam as an HSE pathway. while ten states did not. Today, forty-eight states and the District of Columbia officially administer the GED[®] exam, with only Iowa and Maine relying exclusively on alterantive HSE assessments. However, nearly all employers and post-secondary institutions nationwide recognize passing the GED[®] exam as satisfactory to meet a high school diploma requirement (GED Testing Service, 2020a, 2020b). The GED[®] is the only national HSE exam that awards explicit college readiness designations, and the only HSE exam with recommendations in The ACE National Guide for using scores in remedial course placement or credit award decisions (The ACE National Guide, 2021). However, HiSET score reports note that testers who reach a qualifying score on any subtest (15 out of 20, versus the minimum subtest passing score of 8 out of 20) have "demonstrated college and career readiness" and "[...] should be able to enroll in credit-bearing course work in the university level" (ETS, 2020a), while the TASC does not provide any equivalent college-readiness benchmarks.

4 Data and Descriptive Statistics

Data for this analysis comes from GEDTS, which maintains a database of all test attempts by individuals who take any GED[®] subject test. Data for this project is limited to subject test attempts completed between January 1, 2014 and December 31, 2019 anywhere in the United States. The final data set comprises 5.34 million subject test scores from 1.26 million unique individuals. During test administration, GEDTS collects a rich set of information from testers about their demographics, test-preparation strategies, primary reason for pursuing a GED[®] credential, primary reason for not completing a traditional high school diploma, and labor force status. This data is linked to subject test scores for all testers who provided such information.

Each year, GEDTS contracted with National Student Clearinghouse (NSC) to link a random sample of individuals who earned an HSE credential by passing the GED[®] exam (so-called "GED[®] graduates") at least one year prior to college enrollment data and updates records for previously matched cohorts. Between 2014 and 2020, 31,959 GED[®] graduates' records were linked to NSC's college enrollment database. For all analyses, I restrict the sample to individuals who first attempted a GED[®] subject test after the announcement of the GED[®] CR and GED[®] CR+C benchmarks on January 26, 2016 and who took the exam in English outside of a correctional facility. Implementing these sample restrictions yields an analysis sample of 644,615 unique testers, with 15,262 linked to NSC college outcome data.

Table 1 reports summary statistics for the full sample of $\text{GED}^{\textcircled{\text{B}}}$ testers as well as those that form the analytic samples. Column (1) summarizes the characteristics of all $\text{GED}^{\textcircled{\text{B}}}$ testers in the United States between 2014 and 2019, and Column (2) summarizes the same characteristics for the full analytic sample after implementing the sample restrictions described above. Column (3) summarizes the characteristics of the population of $\text{GED}^{\textcircled{\text{R}}}$ recipients in the full analytic sample who were randomly sampled to be matched to NSC's college enrollment database.⁴ Column (4) summarizes the characteristics of the random sample of $\text{GED}^{\textcircled{\text{R}}}$ recipients from the analytic sample whose records were sent to National Student Clearinghouse (the matched analytic sample). Column (5) tests whether the samples in Columns (3) and (4) are statistically distinguishable along the observable dimensions considered in the analysis. Columns (6) and (7) summarize the char-

⁴Note that this is not the full population of individuals in the analytic sample who passed the GED[®] exam. In 2018, only GED[®] graduates who met the original 5th edition GED[®] subject test passing score threshold (150) in all subjects were sampled for NSC matching (including those who initially scored below 150 in one or more subjects, but exceeded that threshold after retesting).

acteristics of all GED[®] testers in the full analytic sample who earned a GED[®] College Ready or GED[®] College Ready Plus Credits designation in any subject. Appendix Table 1 reports counts of GED[®] testers, completers, and passers by year.

4.1 The GED[®] College Ready Population

The GED[®] CR and GED[®] CR+C benchmarks identify a high-performing subset of GED[®] testers. To earn a GED[®] CR designation a given subject over the period I consider, an individual had to score in the top 7%-16% of all GED[®] testers in that subject, and only the top 2%-4% earned a GED[®] CR+C designation in a given subject. Out of 644,615 individuals in the full analytic sample, 144,228 (22%) qualified as GED[®] CR in any subject, and 38,995 (6%) earned a GED[®] CR+C designation in any subject. Appendix Table 2 reports the proportion of the analytic sample who reached the minimum scores to qualify for GED[®] CR and GED[®] CR+C designations on each subject test, and Appendix Table 3 reports the number of individuals who achieved GED[®] CR, GED[®] CR+C, or GED[®] Honors by subject and year. Columns (6) and (7) of Table 1 show that GED[®] CR and GED[®] CR+C designees are more likely to cite education as their primary motivation for testing, pay for test preparation, prepare using online resources, study independently, or be homeschooled, and less likely to be female, Black, Hispanic, attend adult education classes, or study for the exam than the general population who take or pass the exam.

Figures 1a-1d show the distribution of subject test scores by an individual's category of GED[®] performance in all other subjects, plotting separate distributions for testers (a) who failed at least one other subject tests, (b) passed all other subject tests without earning a GED[®] CR or CR+C designation in any other subject, (c) earned a GED[®] CR designation in at least one other subject, and (d) who earned a GED[®] CR+C designation in at least one other subject. The distribution of test scores will be mechanically higher for GED[®] CR and GED[®] CR+C testers in their CR or CR+C subject, but not other subjects. Since Figures 1a-1d define each college ready population leaving out their score in that subject⁵, the rightward shift at each level of GED[®] performance shows that testers who achieve GED[®] CR and GED[®] CR+C designations score higher in all subjects. On average, college readiness in one content area predicts general knowledge as measured by performance on other GED[®] subject tests.

Figure 2a reports rates of college enrollment and persistence for GED[®] CR and GED[®] CR+C

⁵E.g., a tester's classification in Figure 1a, which plots the distribution of Mathematical Reasoning subject test scores, is based on her performance on the Reasoning through Language Arts, Science, and Social Studies subject tests only.

graduates compared with GED[®] graduates who do not qualify for a college readiness designation and a non-representative, national sample of 1.4 million 2016 high school graduates from National Student Clearinghouse (NSCRC, 2019). Overall, 30.4% of GED[®] graduates were observed as enrolled in a post-secondary institution for at least one quarter. However, this rate falls to 26.3% for those who did not reach any GED[®] CR benchmark, rising to 34.1% for those who did, and 41.0% for GED[®] graduates who reached the GED[®] CR+C score threshold in at least one subject. While GED[®] CR and GED[®] CR+C testers are drawn from the right tail of test score distribution, their rates of college going still lag behind the average traditional high school graduate in the NSC sample. This holds true if we restrict the comparison sample to graduates of traditional public schools, public charter schools, or private schools, and even if we limit the comparison sample to graduates of traditional public schools NSC defines as "Low-Income", where at least 50% of students are free and reduced-price lunch eligibile (NSCRC, 2019). Similarly, conditional on enrolling within one year of passing the exam, Figure 2a shows that GED[®] CR and GED[®] CR+C are more likely to re-enroll than their peers who pass the GED[®] but do not earn a college readiness designation, but still lag behind the comparison sample of traditional high school graduates in their rates of persistence from year one to year two of college.

Some GED[®] testers take the exam because of a desire to leave school, while others may view the exam as a pathway back into formal education. If I limit the sample to GED[®] gradutes who cite "educational gain" as their primary reason for testing, rates of college-going rise for all groups by roughly 30% and rates of persistence increase only slightly, but the basic patterns remain: college outcomes increase by GED[®] college readiness category, but group means stay below the average traditional HS graduate in the comparison sample, even for GED[®] CR+C designees.

Similarly, we might expect to find differences in college outcomes by GED[®] CR or GED[®] CR+C content area, e.g., since subject tests assess different dimensions of skill and could correlate with college major or coursetaking choices that vary in their completion rates and expected wage returns. Splitting out GED[®] CR scores by subject reveals a slight enrollment advantage for those who earn a GED[®] CR or GED[®] CR+C designation in Mathematical Reasoning versus other subjects, but no differences in rates of persistence by subject within the GED[®] CR or GED[®] CR+C categories. In all subjects, a GED[®] college readiness designation predicts higher rates of college enrollment and persistence relative to individuals who do not achieve that level of college readiness designation in any subject, but lower rates of enrollment or persistence than the average traditional HS graduate in the NSC comparison sample.

5 Empirical Strategy

5.1 Regression Discontinuity Analysis

To examine the impact of earning a GED[®] CR or GED[®] CR+C designation in a given subject, I use a regression discontinuity framework to predict college outcomes and retaking behavior of individuals at the GED[®] CR and GED[®] CR+C qualifying thresholds from above and below these thresholds for each subject, interpreting the difference in predictions as the causal impact of crossing a given college readiness threshold. To estimate returns to different categories of GED[®] performance, I construct regression discontinuity estimators that identify Intent-to-Treat (ITT) and Local Average Treatment Effect (LATE) estimates of the impacts of achieving a GED[®] CR or GED[®] CR+C designation in each subject. I cannot assess the impact of earning such a designation on outcomes by simply comparing the outcomes of those who earn a college readiness designation to those who do not, because these estimates would be confounded by unobserved dimensions of academic and non-academic skills that are correlated with an individual's subject test score and GED[®] CR or GED[®] CR+C classification.

The RD design leverages quasi-random variation in the distance of an individual's first score on a given subject test relative to the precise location of a qualifying score threshold to identify the impact of earning a GED[®] CR or GED[®] CR+C designation in that subject. The impact of earning a GED[®] CR or CR+C designation is estimated by comparing the difference in the predicted outcomes for individuals at each college readiness threshold from above and below those thresholds for each subject test based on the score they earn on their first attempt of a given subject test.

For the RD estimates to accurately measure the impact of earning a GED[®] CR or GED[®] CR+C designation on our outcomes of interest, it must be true that individuals cannot systematically manipulate the location of their subject test score relative to the qualifying threshold and that any underlying difference in the probability of earning a college readiness designation in the neighborhood of the qualifying threshold is captured by the smooth polynomial trends on GED[®] subject test scale scores on either side of the threshold, with the remaining variation uncorrelated with unmeasured ability. Additionally, any differences in predicted college outcomes from above and below the college readiness threshold must be explained only by the difference in probability of earning a college readiness designation that is caused by the location of an individual's subject

test score relative to the qualifying threshold, net of error and the smooth polynomial trends on either side of the threshold.

Following Jepsen, Mueser, and Troske (2016, 2017), I use an individual's first attempt on each GED[®] subject test as the running variable to generate regression discontinuity estimates of the impact of attaining GED[®] college readiness designations on college outcomes and exam retaking. Using the test-taker's first score (as opposed to final score) addresses concerns about endogenous retesting behavior that could bias regression discontinuity estimates (Jepsen, Mueser, & Troske, 2016, 2017). While the qualifying thresholds are common knowledge, testers are unable to systematically manipulate their scores to reach the passing, GED[®] CR, or GED[®] CR+C thresholds on any particular attempt as the mapping of raw scores to scaled scores is unknown to testers and individuals only learn their scaled score after completing a subject test. Since sufficiently motivated and talented individuals may retake subject tests until they achieve a desired score, I exclude all subject test attempts after an individual's first attempt from the analysis.

I estimate the first-stage impact of reaching the minimum qualifying threshold on one's first attempt for each $\text{GED}^{\mathbb{R}}$ college readiness threshold in each subject. At the $\text{GED}^{\mathbb{R}}$ CR threshold for subject j, this can be modeled as:

$$CR_{ij} = \alpha_0 + \alpha_{1j}AboveCR_{ij} + \sum_{d=1}^{D} \alpha_{bjd}[BelowCR_{ij}(Score_{ij} - 165)]^d + \sum_{d=1}^{D} \alpha_{ajd}[AboveCR_{ij}(Score_{ij} - 165)]^d + \alpha_{2j}X_i + \theta_t + \psi_s + v_{ijts}$$

$$(1)$$

Where $AboveCR_{ij}$ and $BelowCR_{ij}$ respectively indicate whether individual *i* scored above or below the GED[®] CR qualifying threshold on subject test *j*. Interacting each with an individual's distance from that threshold ($Score_{ij} - 165$) for subject *j* allows the slope of the relationship between earning a GED[®] CR designation in a given subject and that GED[®] subject test score to vary on either side of the qualifying threshold (i.e., α_{ajd} versus α_{bjd}). X_i is a vector of covariates including age at first HSE attempt, an indicator for gender, and mutually exclusive and exhaustive sets of indicators for race/ethnicity as well as status in the labor force. θ_t represents a set of fixed effects for the year and quarter an individual first attempted a GED[®] subject test, and ψ_s is a set of fixed effects for state of residence. The coefficient α_{1j} identifies the discontinuity in the probability individuals eventually earn a GED[®] CR designation in subject *j* at the qualifying threshold. In my preferred specification, I set D = 1 to control for linear trends above and below the threshold, choose optimal bandwidths using the methods outlined in Calonico et al. (2019), and weight observations using a triangular kernel.⁶

Similarly, we can model the relationship between our college outcomes of interest and GED[®] subject test score at the threshold using the same model, but replacing the dependent variable with each outcome of interest, Y_i :

$$Y_{i} = \beta_{0} + \beta_{1j}AboveCR_{ij} + \sum_{d=1}^{D} \beta_{bjd}[BelowCR_{ij}(Score_{ij} - 165)]^{d} + \sum_{d=1}^{D} \beta_{ajd}[AboveCR_{ij}(Score_{ij} - 165)]^{d} + \beta_{2j}X_{i} + \theta_{t} + \psi_{s} + \epsilon_{ijts}$$

$$(2)$$

The coefficient β_{1j} can be thought of as an "intent to treat" estimate of crossing the GED[®] CR threshold on one's first attempt of a given subject test on outcomes of interest. The Wald estimator, τ_j , which estimates the impact of actually earning a GED[®] CR designation on college outcomes is the ratio of β_{1j} to α_{1j}

$$\tau_j = \frac{\beta_{1j}}{\alpha_{1j}} \tag{3}$$

We can reformulate the fuzzy RD specification in an instrumental variables (IV) framework (Hahn, Todd, & Van der Klaauw, 2001; Imbens & Lemieux, 2008). To generate two-stage least squares (2SLS) estimates of the impact of earning a GED[®] CR designation on college outcomes, I estimate the predicted value of CR_{ij} from equation (1) and use this predicted value to estimate equation (4) below for each subject test j. Once $\widehat{CR_{ij}}$ is estimated from the predicted values in equation (1), we estimate the causal impact of earning a GED[®] CR designation on college outcomes from:

$$Y_{i} = \gamma_{0} + \gamma_{1j}\widehat{CR_{ij}} + \sum_{d=1}^{D} \gamma_{bjd} [Below CR_{ij}(Score_{ij} - 165)]^{d} + \sum_{d=1}^{D} \gamma_{ajd} [Above CR_{ij}(Score_{ij} - 165)]^{d} + \gamma_{2j}X_{i} + \theta_{t} + \psi_{s} + e_{ijts}$$

$$(4)$$

Importantly, all independent variables other than the predicted value $\widehat{CR_{ij}}$ enter the first and ⁶See section 6.4 below for a discussion of sensitivity to these and other modeling parameters. second stage of the 2SLS regression exactly as in equations (1) and (2), so the coefficient τ_j from equation (3) is numerically identical to γ_{1j} in equation (4), identifying the local impact of earning a college readiness designation in subject test j. Note that as the first-stage coefficient, $\alpha_{1j} \to 1$, the ITT and LATE estimates will converge, $\tau_j \to \beta_{1j}$.

This framework can easily be adapted to estimate the local impact of crossing the GED[®] CR+C or overall passing threshold rather than the GED[®] CR threshold by (a) replacing CR_{ij} in equations (1) and (4) with a variable $CR + C_{ij}$ or $PassEver_{ij}$ that respectively measure whether individual *i* ever crosses the GED[®] CR+C qualifying threshold in subject *j* or earns a GED[®] credential; and (b) updating ($Score_{ij} - 165$) in equations (1), (2), and (4) to match the relevant qualifying threshold [($Score_{ij} - 175$) for GED[®] CR+C, and ($Score_{ij} - 145$) for the passing the GED[®] exam]. I consider impacts on college outcomes and exam retaking using each of the subject tests and college readiness designations (i.e., estimated at the GED[®] CR and GED[®] CR+C thresholds for each of the four subject tests).

Similarly, if we replace the outcome variable Y_i with elements from the vector of baseline covariates X_i , we can use equation (2) to test for the baseline equivalence of predicted observable characteristics from either side of the GED[®] CR and GED[®] CR+C thresholds:

$$X_{i} = \lambda_{0} + \lambda_{1j}AboveCR_{ij} + \sum_{d=1}^{D} \lambda_{bjd}[BelowCR_{ij}(Score_{ij} - 165)]^{d} + \sum_{d=1}^{D} \lambda_{ajd}[AboveCR_{ij}(Score_{ij} - 165)]^{d} + \lambda_{2j}X_{i} + \theta_{t} + \psi_{s} + \epsilon_{ijts}$$

$$(5)$$

This functions like a balance test in a randomized trial, assessing whether the RD model identifies unexpected discontinuities in observable characteristics that should vary smoothly through the GED[®] CR and GED[®] CR+C thresholds. Results of this balance test at the GED[®] CR and GED[®] CR+C thresholds in each subject test are presented in Table 2. Across 424 tests (testing for discontinuities in 53 characteristics at 2 CR thresholds in each of 4 subjects), I find 38 significant differences at the 10% level, and 15 significant differences at the 5% level, close to what one would expect by random chance if observable characteristics were, in fact, distributed smoothly through the thresholds.

6 Main Results

6.1 First-Stages

Panel A of Table 3 presents first-stage estimates of the relationship between an individual's first $GED^{\textcircled{R}}$ subject test score in a given subject and their eventual $GED^{\textcircled{R}}$ CR or $GED^{\textcircled{R}}$ CR+C attainment in that subject. Unlike the overall passing threshold, where predicted rates of exam passing are only 4% to 19% higher from above versus below each subject test passing threshold, Panel A of Table 3 shows that an individual's first attempt score almost perfectly predicts whether they ever qualify as $GED^{\textcircled{R}}$ CR or $GED^{\textcircled{R}}$ CR+C in that subject.⁷

This is explained by the large differences, shown in Panel D of Table 3, in rates of exam retaking at the passing threshold, but not at the GED[®] CR or GED[®] CR+C thresholds, where impacts on retaking are close to zero, as I will discuss in section 6.3 below. Because the first stage estimates for GED[®] college readiness in a given subject are practically (and in most cases, statistically) indistinguishable from one, I will treat this RD analysis as a sharp RD and report ITT estimates from equation (2) as the main results rather than LATE estimates from equation (4) that rescale the estimated discontinuity by the difference in rates of eventual GED[®] CR and GED[®] CR+C qualification. The subject-specific LATE estimates are qualitatively and statistically indistinguishable from the corresponding ITT estimates in all cases.

However, if one conceives of the GED[®] college readiness designations as functioning as a general signal of academic preparedness and not primarily as a subject-specific signal or path out of noncredit bearing remedial coursework in that subject, we can reformulate the first stage equation with a measure of college readiness in *any* subject as the dependent variable (i.e., replace CR_{ij} with $CR_i = max(CR_{ij}, j \in \{Math, ELA, Sci, SS\})$ in equations (1) and (4). In this case, the first stage impact of crossing any particular subject test college readiness threshold ranges from 24% to 60%, implying that depending on the specific threshold in question, the LATE estimates will be about 1.6 to 4 times larger than ITT estimates in their absolute value.⁸

⁷See Panel A of Appendix Table 8 for first-stage estimates of differences in ever passing the $\text{GED}^{\textcircled{B}}$ exam at each subject test threshold.

⁸See Appendix Table 9 for LATE estimates using this model that correspond to the main ITT estimates in Table 4, which are discussed in Section 6.2.

6.2 College Outcomes

Table 4 presents RD estimates estimated from equation (2) of the impact of crossing each subject's $GED^{\textcircled{R}}$ CR and $GED^{\textcircled{R}}$ CR+C thresholds on measures of college enrollment and persistence. Figures 3a-3h (Enrollment) and 4a-4h (Persistence) provide a graphical representation of these results. Table 5 presents estimates of the impact of crossing these college readiness thresholds on enrollment in each of the first 8 quarters following an individual's first $GED^{\textcircled{R}}$ subject test attempt in any subject. Each subject test presents an opportunity for individuals to achieve a $GED^{\textcircled{R}}$ CR or $GED^{\textcircled{R}}$ CR+C designation in that subject, and the RD estimates reported in Tables 4 and Table 5 represent local impact of earning a $GED^{\textcircled{R}}$ CR or $GED^{\textcircled{R}}$ CR+C designation in a given subject using data from testers whose first attempt subject test score is in the neighborhood of the minimum CR or CR+C score to predict college outcomes from above and below the qualifying threshold.

I find no consistent evidence that crossing the GED[®] CR or GED[®] CR+C threshold and earning a GED[®] college readiness designation influences individuals' short-term college enrollment or persistence behavior. Out of the 80 estimates I consider in Tables 4 and 5 (for 10 outcomes at 2 college readiness thresholds over 4 subject tests), 5 are statistically distinguishable from zero at the 10% level and just over half (55%) of point estimates are positive, close to what one would expect by random chance in the case of truly null effects. While this analysis is underpowered to detect small effects ($\beta_{1j} < 10\%$), we can confidently rule out moderately sized or large effects of the GED[®] college readiness designations on college outcomes for the marginally qualifying tester. The only point estimates that are consistently distinguishable from zero suggest a *negative* effect of GED[®] CR+C qualification in Social Studies on both college enrollment and persistence. The following sections explore explanations for the overall pattern of null results, examining subject test retaking behavior and exploring potential heterogeneity in impacts.

6.3 Subject test retaking

Considering exam retaking as an outcome allows me to contrast the behavioral impacts of crossing the GED[®] CR and GED[®] CR+C thresholds with the impacts of crossing the overall passing threshold (\geq 145 during the period I consider) on the probability that an individual retakes a given subject test. Using subject test scores in Mathematical Reasoning as the running variable, Figures 5a-5c show the striking contrast in the change in rates of retesting at the the overall passing threshold (Figure 5a) versus the GED[®] CR (Figure 5b) or GED[®] CR+C thresholds (Figure 5c). Other subject tests exhibit the same pattern (see Appendix Figures 3a-3i).

Among GED[®] testers who narrowly fail a given subject test by scoring within three points of passing on their first attempt, 70% to 80% retest at least once in that subject, while nearly no one retakes a subject test they narrowly pass. This corresponds to the point estimates in the first row in Panel D of Table 3, where RD estimates predict marginal passers to be 72 to 81 percentage points less likely to retake a subject test than their peers who fall just short of passing. Strikingly, the remaining point estimates in panels C and D of Table 3 are very close to zero, as less than 0.5% of testers who fall narrowly short of reaching a GED[®] CR or GED[®] CR+C threshold on any subject test retest in that subject. This suggests that while lower-performing GED[®] testers in a given subject are willing and able to exert substantial effort to ensure they earn a credential, higher performers either are not willing to exert effort to ensure they earn a GED[®] CR or GED[®] CR+C designation or face administrative obstacles to retesting. Indeed, while testers who fail a subject test may retake that test—generally free of cost—those who pass a subject test but want to improve their score to reach a college readiness benchmark must pay an additional retesting fee (though this fee may be discounted), and in some states or localities, they may be discouraged or disallowed from retaking a subject test they have previously passed.⁹ In the absence of such obstacles, retesting behavior is consistent with GED[®] testers placing a high value on earning a credential, but not expecting similar returns to effort expended toward earning a college readiness designation.

6.4 Heterogeneity

I examine four pathways that could lead to heterogeneity in treatment effects. First, I explore whether treatment effects vary by subject or the point in the overall skill distribution where a particular GED[®] CR or GED[®] CR+C qualifying threshold is located. Second, I examine treatment effect heterogeneity in demographic subgroups that are commonly collected in administrative datasets, covering gender, race, and baseline employment status. Next, I examine whether a tester's stated primary reason for taking the exam predicts the effect of crossing a college readiness threshold. Finally, as a proxy for differences in rates of institutional take up, I construct a measure of the intensity of ACE membership of local post-secondary institutions at the state level and estimate impacts separately for testers in high versus low ACE membership states.

 $^{^{9}}$ Costs of retesting and official subject test retaking rules vary by state. However, among individuals who passed a given subject test on their first attempt, I do observe some subject test retaking in raw subject test score data between 2016 and 2019 in each state that offers the GED[®] exam.

6.4.1 Subject, Skill Level, and Benefits

By estimating treatment effects separately at both college readiness thresholds for each the four subject-tests separately the main results in Table 4 allow me to explore two dimensions of heterogeneity in treatment effects. First, each subject test confers a GED[®] CR or GED[®] CR+C in that subject, and earning a college readiness designation in one subject may affect individuals differently than in another. For example, twice as many students at two-year colleges are assigned to remedial coursework in math (nearly 60%) than are required to complete remedial coursework related to English or reading, and students are more likely to fail to complete a remedial course sequence in math than any other subject (Chen, 2016; Kozakowski, 2019). This could mean that the ACE recommendation that individuals who earn a GED[®] CR designation in Mathematical Reasoning be granted a waiver from developmental coursework in math makes that GED[®] CR designation more consequential than earning a GED[®] CR designation in other subjects. Furthermore, ACE recommendations for the number of lower-division credits institutions award to students who earn a GED[®] CR+C designation varies by subject, suggesting that institutions award only one credit in English for students who qualify as GED[®] CR+C in Reasoning Through Language Arts but three credits in math, science, or social studies for reaching the corresponding GED [®] CR+C thresholds. However, I do not find any systematic differences in treatment effects by GED[®] CR or GED[®] CR+C subject. The only consistently statistically significant estimates are negative impacts of crossing the GED[®] CR+C Social Studies threshold on college outcomes. While it is possible that earning a GED[®] CR+C designation in social studies discourages recipients from enrolling or persisting in college, there are no similar effects in any other subject test threhold or at the GED[®] CR threshold in social studies. Additionally, these significant coefficients are not robust to standard corrections that account for multiple hypothesis tests (Bonferroni, 1935; Holm, 1979).

Additionally, the GED[®] CR and GED[®] CR+C score thresholds are binding at different parts of the score distribution for each subject test. For example, the marginal tester who earned a GED[®] CR designation in Math scored in the 93rd percentile, while the marginal tester who earned a GED[®] CR designation in Social Studies only scored in the 84th percentile. There is much less variation in the precise location of the GED[®] CR+C qualifying threshold, where testers had to score in roughly the 98th percentile in Math, ELA, or Science, and the 96th percentile in Social Studies to qualify. Examining the effects of crossing the GED[®] CR thresholds in each subject does not suggest a strong link between the location of a subject's qualifying threshold in the overall score distribution and the estimated impact of earning a college readiness designation in that subject, with the correlation between percentile rank at a subject test CR or CR+C threshold and the estimated impact of crossing that threshold ranging from -0.76 for impacts on persistance at the $GED^{\textcircled{R}}$ CR thresholds to +0.91 for impacts on enrollment at the $GED^{\textcircled{R}}$ CR+C thresholds.

Finally, the qualifying thresholds for the different categories of college readiness themselves—GED[®] CR versus GED[®] CR+C—both (a) affect testers at different parts of the overall score distribution and (b) come with different ACE recommendations for how institutions should interpret and use that college readiness designation. If the primary benefit of earning a college readiness designation is the exemption from remedial coursework, then we might expect the GED[®] CR benchmarks to have larger effects, because they identify a slightly lower-performing (albeit still high-performing) subset of GED[®] graduates who may be less likely to pass out of remedial coursework in the absence of earning a GED[®] CR designation. However, if the primary benefit is earning credits for prior learning, which is exclusive to the GED[®] CR+C designation, we might expect crossing the GED[®] CR+C threshold to yield larger benefits. In practice, I find that there is no consistently positive or negative relationship between crossing the GED[®] CR or GED[®] CR+C thresholds.

6.4.2 Demographic Subgroups

In columns (1) through (4) of Tables 6a and 6b, I divide the sample into demographic subgroups based on gender and under-represented minority identity. If individuals from under-represented groups are less likely to view themselves as "college ready" or more likely to experience self doubt or imposter syndrome, then receiving a positive, private signal of their ability or academic preparedness could motivate these individuals to enroll in college or persist once enrolled. Contrary to this prediction, I find no consistent differences in the impact of reaching a GED[®] CR or GED[®] CR+C threshold by gender or membership in an under-represented minority group. Taken at face-value, the point estimates suggest that earning a GED [®] CR+C designation in Science may be more impactful for women and under-represented minorities, both groups that are under-represented in STEM fields, but the confidence intervals are large, and a similar advantage does not appear consistently in Math or at the GED [®] CR threshold.

6.4.3 Motivation for Testing

In columns (7) and (8) of Tables 6a and 6b, I divide the sample into demographic subgroups based on a tester's stated primary reason for taking the GED, comparing those who identify Educational Gain as a primary reason versus all other reasons (Personal, Military, Work, Special Requirement). Comparing group means in Figures 2b and 2c reveals substantial differences in rates of college enrollment across all categories of college readiness by reason for testing (Education vs. any other reason). Since individuals who are motivated by educational gain appear to drive college enrollment among GED graduates, they may be more likely to seek out ways to use GED[®] CR or GED[®] CR+C benefits, while individuals who identify other reasons for testing may be unlikely or unable to enroll in college as they pursue other goals. Conversely, if individuals who are motivated by educational gain are more committed to enrolling in college, they may have higher rates of enrollment regardless of GED[®] CR qualification, and individuals who took the exam for other reasons may be more sensitive to unexpectedly learning that they qualify for GED[®] CR or GED[®] CR+C benefits that could make college appear more accessible or less costly. Despite the large differences in rates of college enrollment by reason for testing in Figure 2, the GED[®] CR designations do not appear to exert systematically different impacts on college outcomes for testers whose primary reason for testing is educational gain versus other reasons.

6.4.4 ACE Membership

Finally, I use the prevalence of ACE membership in a state as a proxy for the likelihood a qualifying tester could find a local institution that had adopted ACE recommendations for the GED[®] college readiness benchmarks regarding credit for prior learning. In columns (5) and (6) of Tables 6a and 6b, I split the sample by whether the state where an individual took their first GED[®] subject test is above or below the median of a measure of ACE membership intensity.

To construct my index of ACE membership intensity, I measure the proportion of the collegeaged population in a GED[®] tester's state that is enrolled in ACE institutions and the proportion of post-secondary enrollment in that state that occurs in ACE member institutions. I standardize each measure to have a mean of zero and standard deviation of one, and sum these standardized components to form an index of ACE membership intensity. ACE is a partner in the publication of the GED[®] exam and issues recommendations about how to interpret and use GED[®] CR and GED[®] CR+C qualifications. If member institutions are more likely to accept or use these qualifications in thier remediation or credit award decisions, then testers in states with relatively more ACE membership may be able to more easily access benefits for earning a GED[®] CR or GED[®] CR+C designation.

While I do find a handful of point estimates are statistically distinguishable from zero and, in

three cases, from their complement, there is no consistent pattern in the direction or subject in which impacts vary by ACE membership intensity.

6.5 Robustness Checks

6.5.1 Alternative Models

Appendix Tables 4, 5, and 6 present alternative RD estimates of the impact of crossing GED[®] CR and GED[®] CR+C thresholds on college outcomes, respectively using alternative local polynomial trends, adding alternative sets of covariates (including no covariates), and using the robust and bias-corrected local-polynomial methods described in Calonico et al. (2019).

Another important consideration in RD analyses is the choice of bandwidth for the analysis sample. For the main analyses and balance tests, I calculate and use optimal bandwidths using the methods described in Calonico et al. (2019). In Appendix Figures 2a and 2b and Appendix Tables 7a and 7b, I present alternative RD estimates for all analyses in Table 4 using all possible bandwidths between 2 to 9 points¹⁰ in addition to the Calonico et al. (2019) optimal bandwidths that are used in the main analysis.

With the exception of RD estimates at the GED[®] CR+C Social Studies threshold, nearly every choice of bandwidth, covariates, degree of polynomial trend, and bias-correction yields null estimates of treatment effects for both college outcomes in all subjects. The largest bandwidths suggest negative effects on college persistence from crossing the GED[®] CR+C Mathematical Reasoning threshold and higher order polynomials yield large negative effects from crossing the GED[®] CR+C threshold in Science. Notably, no modeling decision yields positive, statistically significant estimates in any subject, threshold, or outcome.

6.5.2 Density Checks

Appendix Figures 1a-1d plot the density of observations at each point of the discrete subject test score distribution within 5 points of either college readiness threshold. Each figure is labeled with the p-values from tests of score manipulation at each threshold.¹¹ While the tests, which measure the smoothness of the subject test score density function at the qualifying threshold, sometimes

¹⁰This range represents all possible symmetric integer bandwidths that can be used without encountering potential non-linearities caused by marginal eligibity for the other $\text{GED}^{\textcircled{B}}$ college readiness designation.

 $^{^{11}}$ I use a test developed by Frandsen (2017) that builds on the methods described by McCrary (2008) to develop correctly-sized tests for smoothness of the density function in regression discontinuity analyses that use a discrete running variable.

reject the null hypothesis that the distribution is smooth, in all cases where the test rejects the null hypothesis of a smooth distribution, the change in density at the threshold runs counter to the expected change in densities that would be caused by intentional manipulation. In these cases, the density of observations *falls* at the college readiness threshold rather than rising, which would be the expected effect if individuals manipulating their score placement to qualify for a given GED[®] CR or GED[®] CR+C designation. As such, I interpret the density plots and tests as showing no evidence of score manipulation, but hypothesize that the bunching observed near some subject test score thresholds are psychometric artifacts that arise from the coarseness of the mapping from raw scores to scale scores, especially toward the tails of the test score distributions where the GED[®] CR and GED[®] CR+C thresholds are located.

7 Discussion

The introduction of the GED[®] CR and CR+C college readiness designations foreshadowed a future where the GED[®] expanded its role as an alternative pathway to post-secondary education for college-ready testers, helping this group of high-achieving HSE credential holders learn and signal their ability, avoid costly remediation, and potentially earn transferable college credits. Four years later, I find that while testers who earned GED[®] CR and GED[®] CR+C designations enrolled and persisted in college at higher rates than other GED[®] graduates, rates of college enrollment and persistence for GED[®] CR and GED[®] CR+C testers were indistinguishable from that of GED[®] testers who fell narrowly short of qualifying for these designations. Currently, the GED[®] college readiness benchmarks predict—but do not cause—better college outcomes.

There are many potential explanations for the null effects of the college readiness designations on college outcomes.¹² First, waivers from developmental coursework, a primary benefit afforded to testers who earn GED[®] CR and GED[®] CR+C designations at institutions that follow ACE recommendations, may be of little practical value for this population. The GED[®] CR and GED[®] CR+C benchmarks are difficult to attain, representing the 86th-98th percentiles of the GED[®] subject test score distributions. Testers who score near a college readiness threshold in a given subject test may already be likely to pass out of remediation on placement exams like the Accuplacer in that subject, substantially reducing the utility of a waiver from developmental education or

 $^{^{12}}$ In addition to the explanations discussed below, one limitation of this study's analysis of college outcomes is statistical power, as I was only able to attempt to match < 5% of GED graduates from 2016-2019 to college enrollment records. Matching a larger set of GED testers, including those who never pass the exam, would improve the precision of the estimated effects and open new lines of research.

placement exams.¹³

Similarly, the high bar to qualify means that testers who earn GED[®] CR and GED[®] CR+C designations may be less sensitive to receiving a positive private signal of academic preparedness. While, such a signal could motivate or instill confidence in individuals whose skills are truly on the margin of academic preparedness, the GED[®] CR and GED[®] CR+C thresholds may be sufficiently high in the score distribution that marginally qualifying testers believe they are ready for college with or without the additional signal provided by a GED[®] CR designation.

Finally, low rates of institutional takeup and informational constraints may limit the ability of students to find institutions where they can use their GED[®] CR and GED[®] CR+C benefits, further muting any potential impacts. Anecdotally, only a small percentage of institutions nationwide are believed to have adopted a policy of using the GED[®] CR and GED[®] CR+C thresholds in their course placement and credit award decisions, and there is no up-to-date, centralized database or common signalling mechanism to help testers identify participating institutions. Conversely, the College Board, whose College Level Examination Program (CLEP) offers similar credit benefits to those promised by the GED[®] CR+C designation provides a detailed, searchable database that allows students to identify institution-specific policies regarding qualifying score thresholds and credits awarded (if any) for each of the 34 CLEP exams in over 2,900 partner institutions (College Board, 2021).¹⁴

While some estimate that roughly 400 institutions consider the GED[®] college readiness designations in their course placement decisions, very few institutions awarded college credits to students who earned a GED[®] CR+C designation (ACE CREDIT College & University Partnerships, 2018; "Colleges that accept the GED", 2020). Before retiring their College and University Partnerships Database in 2018, the American Council on Education identified only 26 colleges and universities that awarded college credits to students who earned a GED[®] College Ready Plus Credits designation (ACE CREDIT College & University Partnerships, 2018).¹⁵ Indeed, even in states with the

¹³In math, the subject in which college students are most often assigned to non-credit bearing remedial coursework, only the top 7% of GED[®] testers earned a GED[®] CR or GED[®] CR+C designation. If an individual's probability of being assigned to remedial coursework in a given subject is decreasing in their subject test score, an exemption from developmental math coursework would likely be of limited use even for those who marginally qualify for a GED[®] CR designation in Mathematical Reasoning, unless more than 93% of GED[®] testers have a substantial probability of requiring remediation. Additionally, in social studies, the subject where the most students earn a GED[®] CR or GED[®] CR+C designation, few institutions or programs offer or require developmental coursework to enroll in introductory college-level courses.

¹⁴Notably, CLEP exams have been offered for over 50 years, potentially explaining their substantially more mature infrastructure for institutional partnerships to award credit for prior learning.

¹⁵In conversations with the author, representatives from ACE and GEDTS expressed their belief that relying exclusively the institutions listed in the defunct ACE CUP database likely substantially underestimates of the number

highest rates of ACE and ACE CUP membership, earning a GED[®] CR or GED[®] CR+C designation does not appear to influence college outcomes and exam retaking rates remain vanishingly small just below the college readiness thresholds.

If many qualifying GED[®] graduates find they are unable to redeem their GED[®] CR+C test score for college credits at their preferred post-secondary institution, this could also have a discouragement effect for those testers, potentially counterbalancing any benefits to GED[®] CR+C designees who do succeed in redeeming their scores for college credits, as advertised, at participating institutions. Experimental research in psychology shows that individuals who invest effort toward a goal, like studying for the GED[®] to earn the benefits of a college readiness designation, experience regret and disappointment if their effort goes unrewarded, which could lead them to abandon activities related to that effort or degrade trust in associated institutions (Van Dijk, Van Der Pligt, & Zeelenberg, 1999; Tzieropoulos et al., 2011).

Moreover, if institutional quality is negatively correlated with $\text{GED}^{\textcircled{B}}$ CR or CR+C adoption, students could be drawn to lower quality institutions, where they are less likely to receive other supports, in order to use their $\text{GED}^{\textcircled{B}}$ CR designation to avoid remedial coursework or redeem $\text{GED}^{\textcircled{B}}$ CR+C credits, again providing potential countervailing pressure against the benefits of receiving college credit in acknowledgement of a high $\text{GED}^{\textcircled{B}}$ subject test score. Using data from a Massachusetts scholarship program, Cohodes and Goodman (2014) show that policies that subsidize higher education can reduce educational attainment when institutional quality is negatively correlated with subsidy eligibility.

Evaluating credit for prior learning (CPL) initiatives in 72 institutions with active CPL initiatives, Klein-Collins and colleagues (2020) find that earning credit for prior learning is associated with substantially higher rates of post-secondary credential completion after controlling for baseline observable characteristics using propensity score matching methods. This analysis suggests that offering CPL through recognizing credentials like GED[®] CR and GED[®] CR+C can be mutually beneficial for students and institutions, making participating students substantially more likely to complete degrees or certificates, reducing time to degree, and generating more tuition revenue for institutions—despite the foregone earnings from credits awarded directly via CPL programs—as participating students earn more non-CPL credits and are more likely to remain enrolled through

of institutions that, in practice, accept GED[®] CR designations as a waiver out of remedial coursework or awarded credit to students who earned a GED[®] CR+C designation. However, they could not provide an estimate the true number of participating institutions, acknowledging that the difficulty of collecting and updating detailed institutional adoption data is one reason the database project ended.

credential completion. However, underlying this promising evidence is the stark reality that even in a sample of institutions with active CPL initiatives, only roughly 1 in 25 adult students was able to leverage their prior non-military educational or professional experience to earn transferable college credits.¹⁶ Examining community and technical colleges, where the majority of matriculating college-ready GED graduates enroll, the rate of take-up for non-military CPL programs falls to only 2% of students, suggesting that universal adoption of ACE recommendations for GED [®] CR+C credit awards could meaningfully increase access to CPL, but that substantial barriers to such widespread institutional adoption and individual participation in CPL programs remain.

In light of this evidence and other promising evaluations of CPL programs (e.g., Boatman et al., 2020), my finding that earning a GED[®] CR or GED[®] CR+C designation does not appear to influence college outcomes both demands further exploration and represents a substantial opportunity to strengthen programs to support college-ready GED[®] graduates. GED[®] stands out as a CPL pathway serving a target population of more disadvantaged, non-traditional students, making CPL awarded through GED[®] performance a potentially potent driver of educational equity, but also potentially requiring that GED college readiness be complemented by other supports to promote college matriculation and persistance for eligible GED[®] graduates. The GED[®] college readiness benchmarks are designed to identify academic preparedness, but do not measure non-academic skills that may be necessary for post-secondary success and are generally found to be negatively correlated with pursuing a high school equivalency credential through examinations like the GED[®] rather than completing a traditional high school diploma (Heckman & Rubinstein, 2001). Recent evidence suggests that counselors exert a strong influence on high school students' post-secondary outcomes, particularly for lower-achieving students (Mulhern, 2020); future research may assess whether college counselors, admissions officers, advisors, or other mentors play a similarly important role in helping college ready GED[®] graduates access program benefits and earn CPL.

Regardless of eligible students' support systems or underlying skills and abilities, they are unlikely to benefit from earning a GED [®] CR or GED [®] CR+C designation if local institutions do not recognize these credentials or information about where credentials are accepted is unclear or difficult to ascertain. In addition to growing and deepening institutional partnerships, GEDTS must develop clear and easily navigable processes to help eligible students and their mentors identify those

¹⁶Credit for military training is the most common source of credit for prior learning in the sample, but contrasts sharply with exam-based CPL programs like the $\text{GED}^{\textcircled{B}}$ college readiness benchmarks or CLEP in terms of the eligible population and method of certification. Including credit for military training increases the overall rate of CPL receipt to 11% in the sample.

institutions that will consider GED[®] CR and GED[®] CR+C qualifications to understand what benefits students are eligible to receive as well as how and where these benefits can be accessed. For the GED[®] exam to reach its potential as an alternative pathway to post-secondary success, GEDTS must expand the visibility, reach, and utility of its college readiness designations.

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Figures



Figure 1: GED[®] subject test score distributions by performance in other subjects

Notes: Each graph plots the density of $\text{GED}^{\textcircled{8}}$ subject test scores by an individual's category of performance in all other subjects, as indicated in the graph's legend. N=644,615.



Figure 2: College outcomes by GED[®] performance versus a national sample of traditional HS grads

(a) Full matched analytic sample







(c) Subgroup: Reason for testing = Other





(d) Full matched sample, by $\operatorname{GED}^{\textcircled{R}}$ CR subject



(e) Full matched sample, by $\operatorname{GED}^{\textcircled{B}}$ CR+C subject



Figure 3: RD plots, college enrollment at the GED[®] CR threshold by subject test



Figure 3 (con't): RD plots, college enrollment at the GED[®] CR+C threshold by subject test

Notes: Each graph plots rates of college enrollment above and below the indicated GED[®] subject test score thresholds, as well as lines of best fit and 95% confidence interval on either side of the threshold. College enrollment is a binary variable that takes on the value 1 if an individual was observed in National Student Clearinghouse (NSC) data as enrolled in college within two years of their first GED[®] subject test attempt and zero otherwise. Bandwidth and effective sample size vary by subject test threshold, determined using the optimal bandwidth calculations described in Calonico et al. (2019); see Table 4 for the effective number of observations that contribute to each estimate. Distance from each threshold is modeled as a linear function with triangular kernel weights and separate slopes above and below each threshold, as in Equation (2). Points are scaled relative to the size of the sample that earned that subject test score.



Figure 4: RD plots, college persistence at the GED[®] CR threshold by subject test



Figure 4 (con't): RD plots, college persistence at the GED[®] CR+C threshold by subject test

Notes: Each graph plots rates of three semester persistence above and below the indicated GED[®] subject test score thresholds, as well as lines of best fit and 95% confidence interval on either side of the threshold. Three semester persistence is a binary variable that takes on the value 1 if an individual was observed in National Student Clearinghouse (NSC) data as enrolled in college for at least six quarters within two years of their first GED[®] subject test attempt and zero otherwise. Bandwidth and effective sample size vary by subject test threshold, determined using the optimal bandwidth calculations described in Calonico et al. (2019); see Table 4 for the effective number of observations that contribute to each estimate. Distance from each threshold is modeled as a linear function with triangular kernel weights and separate slopes above and below each threshold, as in Equation (2). Points are scaled relative to the size of the sample that earned that subject test score.



Figure 5: RD plots, subject test retaking by threshold (Mathematical Reasoning)

Notes: This figure plots rates of exam retaking above and below the indicated $\text{GED}^{\textcircled{B}}$ subject test score thresholds, as well as a line of best fit and 95% confidence interval on either side of the threshold. Exam retaking is a binary variable that takes on the value 1 if an individual was observed taking a $\text{GED}^{\textcircled{B}}$ subject test in the indicated content area more than once and zero otherwise. Bandwidth and effective sample size vary by subject test threshold, determined using the optimal bandwidth calculations described in Calonico et al. (2019); see panel D of Table 3 for the effective number of observations that contribute to each estimate. Distance from each threshold is modeled as a linear function with triangular kernel weights and separate slopes above and below each threshold, as in Equation (2). Points are scaled relative to the size of the sample that earned that subject test score.

Tables

 Table 1: Summary Statistics

Characteristic	All Testers (1)	Analytic Sample (2)	$\begin{array}{c} \hline \text{Passed} \\ \text{GED}^{\textcircled{R}} \\ (3) \end{array}$	NSC Sample (4)	P-Value (3)=(4) (5)	$\overline{\begin{array}{c} \operatorname{GED}^{\textcircled{R}} \\ \operatorname{CR} \\ (6) \end{array}}$	$ \begin{array}{c} GED^{\textcircled{R}} \\ CR+C \\ (7) \end{array} $
A. Demographics							
Age at first test attempt	26.1	25.1	22.7	22.7	.251	23.2	23.2
Male	.565	.500	.572	.564	.662	.593	.631
Asian	.024	.028	.028	.029	.237	.039	.060
Black	.191	.167	.074	.080	.064	.050	.030
Hispanic	.255	.224	.172	.168	.303	.145	.112
Native American	.017	.017	.012	.013	.788	.010	.007
White	.451	.486	.625	.625	.998	.664	.694
Multi-Racial	.040	.048	.059	.058	.058	.062	.064
Other Race	.022	.030	.029	.030	.605	.031	.032
Employed	.399	.468	.462	.459	.478	.467	.450
Unemployed	.324	.346	.342	.343	.824	.329	.308
Permanently Disabled	.011	.013	.007	.006	.256	.008	.009
Retired	.005	.004	.002	.002	.566	.003	.003
B. Baseline Survey Data							
Motivation: Personal	.295	.266	.218	.219	.896	.198	.160
Motivation: Work	.202	.226	.217	.220	.125	.221	.207
Motivation: Education	.423	.457	.517	.512	.136	.537	.593
Motivation: Special Requirement	.057	.021	.015	.016	.169	.013	.010
Motivation: Military	.024	.029	.034	.031	.493	.032	.029
Paid for Test Prep	.395	.464	.476	.473	.899	.499	.525
Prepared via Practice Test	.356	.413	.403	.405	.728	.378	.341
Prepared via Books	.269	.243	.248	.248	.413	.258	.267
Prepared via Online Resources	.133	.168	.196	.193	.927	.215	.241
Prepared via Audio Resources	.009	.010	.008	.009	.599	.008	.007
Prepared via Television	.004	.005	.004	.003	.168	.003	.003
Prepared via Mobile App	.032	.044	.052	.052	.607	.056	.055
Prepared via Social Network	.021	.026	.024	.023	.858	.023	.021
Prepared via Other Resources	.021	.023	.024	.024	.574	.024	.024
Prepared via Adult Education	.434	.374	.295	.299	.528	.256	.189
Studied for Exam	.833	.816	.786	.788	.875	.769	.732
Studied at Test Prep Center	.288	.334	.284	.298	.102	.252	.192
Studied at Testing Center	.032	.037	.027	.028	.669	.022	.015
Studied at HS Program	.092	.109	.093	.095	.329	.080	.061
Studied at CC Program	.058	.069	.058	.058	.776	.053	.045
Studied at Workplace Program	.015	.018	.012	.013	.319	.010	.007
Studied at Online Program	.089	.113	.128	.124	.441	.134	.137
Studied at Military Program	.005	.005	.004	.003	.036	.003	.002

Studied at Correctional Facility	.157	.023	.015	.014	.390	.011	.007
Studied via Tutoring	.061	.070	.063	.066	.452	.061	.053
Studied in Self-Directed Setting	.266	.322	.385	.380	.814	.417	.466
Studied at Other	.029	.030	.023	.023	.959	.019	.014
Highest grade $= 9$ or lower	.231	.214	.173	.174	.547	.168	.155
Past Year Income Above Median	.359	.423	.409	.405	.690	.423	.405
Reason for no diploma: Academic	.437	.428	.411	.408	.607	.397	.363
Reason for no diploma: Personal	.717	.720	.714	.719	.201	.699	.653
Reason for no diploma: Other	.069	.055	.051	.051	.627	.052	.060
Reason for no diploma: Homeschool	.041	.051	.080	.078	.479	.097	.137
Reason for no diploma: Int'l	.033	.031	.027	.026	.696	.034	.046
C. Missing Data Indicators							
Missing Race Data	.09	.075	.081	.086	.012	.088	.106
Missing Gender Data	.034	.047	.054	.058	.013	.063	.082
Missing Labor Force Status	.019	.006	.006	.006	.453	.005	.003
Missing Motivation Data	.019	.007	.006	.006	.363	.005	.003
Missing Educational Attainment	.073	.054	.049	.049	.577	.054	.060
Missing Reason for No Diploma	.019	.006	.006	.006	.418	.005	.003
D. Sample Exclusion Parameters							
First attempt after $1/26/2016$.696	1	1	1		1	1
Tested in English	.955	1	1	1		1	1
Tested in a Correctional Facility	.188	0	0	0		0	0
E. GED [®] Test Score Data							
Ever Passed GED [®]	.621	.636	1	1		.899	.925
$\operatorname{GED}^{\mathbb{R}}$ CR - Math	.065	.073	.157	.150	.551	.277	.516
$\operatorname{GED}^{\mathbb{R}}$ CR - ELA	.117	.127	.285	.275	.110	.521	.734
$\operatorname{GED}^{\mathbb{R}}$ CR - Science	.110	.134	.305	.297	.033	.523	.747
$\operatorname{GED}^{\mathbb{R}}$ CR - Soc. Stud.	.137	.160	.349	.337	.031	.641	.869
GED [®] CR - Any Subject	.197	.224	.379	.533	.134	1	1
$\operatorname{GED}^{\mathbb{R}}$ CR+C - Math	.016	.020	.044	.040	.515	.076	.276
$\operatorname{GED}^{\mathbb{R}}$ CR+C - ELA	.019	.022	.055	.053	.251	.091	.338
$GED^{\mathbb{R}}$ CR+C - Science	.015	.020	.048	.047	.650	.077	.282
$GED^{\mathbb{R}}$ CR+C - Soc. Stud.	.034	.042	.098	.084	.027	.167	.611
$\operatorname{GED}^{\mathbb{R}}$ CR+C - Any Subject	.049	.060	.163	.151	.196	.270	1
1st Math Subj. Test Score	149.7	150.4	157.1	156.7	.467	159.5	166.6
1st ELA Subj. Test Score	152.9	153.2	160.5	160.1	.004	164.4	170.1
1st Science Subj. Test Score	153.8	154.9	161.3	161.2	.094	164.6	169.8
1st Soc. Stud. Subj. Test Score	153.0	154.0	161.7	161.3	.004	166.7	174.9
Observations	$1,\!258,\!509$	$644,\!615$	$320,\!918$	$15,\!262$		$144,\!228$	$38,\!995$

Notes: Columns (1) through (4), (6), and (7) report the mean value of the covariate listed each row for the sample indicated in each column. Column (5) reports the p-value from a test of equivalence between the means in column (3) and column (4).

 Table 2: Balance Tests

	Mathematical	Reasoning		
	Reasoning	Thru. ELA	Science	Soc. Stud.
A. RD Estimates GED [®] College Ready	v Threshold (Subse	core>165)		
Age at first test attempt	.0542	.1392	.7435*	.5059
O	(.6553)	(.5441)	(.4172)	(1.0355)
Male	.0149	0105	0216	1311
	(.0585)	(.0413)	(.0294)	(.0822)
Asian	0185	0356*	0241	0004
	(.0265)	(.0192)	(.0165)	(.0255)
Black	.0053	.0032	0084	.0611
	(.029)	(.0182)	(.014)	(.0407)
Hispanic	.0135	0034	.0232	.0236
	(.0431)	(.0274)	(.0218)	(.0453)
Native American	- 0103	0044	0059	0085
	(0132)	(0071)	(0063)	(0146)
White	0099	0245	0115	- 0401
W III 00	(.0636)	(0.0210)	(0317)	(0797)
Multi-Bacial	- 0043	0105	- 0019	- 0/38
Wulti-Itaciai	(0305)	(0178)	(0144)	(043)
Other Bace	(.0505)	(.0170)	- 0065	(.043)
Other Race	(0108)	(0147)	(0107)	(0350)
Employed	(.0150)	(.0147)	(.0107) 0317	(.0555)
Employed	(0733)	(051)	(0205)	(0821)
Unomployed	(.0735)	(.031)	(.0293)	(.0621)
Unemployed	(067)	0080	(0226)	(0235)
Permanently Disabled	(.007)	(.0462)	(.0280)	(.0773)
Fermanentry Disabled	.0032	009	(0047)	0172
Mativation, Damanal	(.01)	(.0078)	(.0047)	(.0100)
Motivation: reisonal	0023	.0497	(0.00000000000000000000000000000000000	0507
Mativation, Work	(.0401)	(.0311)	(.0241)	(.0058)
WOIVATION: WOIK	.0204	.0429	(0.026)	.0524
Mativation, Education	(.0477)	(.0525)	(.020)	(.0000)
Motivation: Education	0140	(0.0092)	(0210)	.0423
Mativation, Chasial Dequipement	(.0013)	(.0392)	(.0322)	(.0012)
Motivation: Special Requirement	.0172	0072	0008	0195
	(.0102)	(.009)	(.0009)	(.0244)
Motivation: Minitary	0117	0423	(0092)	0459
Daid for Tost Drop	(.0252)	(.018)	(.0104)	(.0547)
Paid for Test Prep	0047	.0122	.0233	0340
Durant late Duration Trat	(.0095)	(.0400)	(.0559)	(.0929)
Prepared via Practice Test	0273	.0293	(0021^{+1})	0184
Durana laria Daala	(.0559)	(.0404)	(.0275)	(.0812)
Prepared via Books	0018	.007	.0818	.0902
	(.0587)	(.0349)	(.0283)	(.0665)
Prepared via Online Resources	012	.0404	.0017	$.12(^{-1})$
Duran and aris M 1.1 A	(160.)	(.0307)	(.0230)	(.0047)
Frepared via Mobile App	.0201	0545^{TT}	.0003	0471
	(.0314)	(.0257)	(.0144)	(.0435)
Prepared via Social Network	.0239	0111	0001	.0083
	(.0154)	(.0132)	(.0091)	(.0205)

	Prepared via Other Resources	.0074	0015	0267	0351
	•	(.0176)	(.0105)	(.0173)	(.0319)
	Prepared via Adult Education	.0006	.0255	0316	0743
	-	(.056)	(.0437)	(.0267)	(.0713)
	Studied for Exam	0071	.0112	.0118	0.0345
		(.0528)	(.0293)	(.0243)	(.0681)
	Studied at Test Prep Center	.0424	.0421	0484*	0565
	Stanta di Lott Lop Contra	(.0507)	(.0428)	(.0277)	(.0703)
	Studied at Testing Center	.0042	.0159	.0085	.026
	Stuarou at Tosting Contor	(0163)	(0107)	(009)	(0188)
	Studied at HS Program	0148	0182	0104	0452
	Studied at HS I regram	(0306)	(0212)	(0156)	(0462)
	Studied at CC Program	- 0332	- 0048	- 0059	015
	Studied at CC 1 logram	(0.0002)	(0165)	(0132)	(0365)
	Studied at Workplace Program	(.0212) 0.0254**	0100	(.0132)	(.0505)
	Studied at Workplace I logram	(0204)	(0005)	(0129)	(0042)
	Studied at Online Program	(.0099)	(.0095)	(.0007)	(.0042)
	Studied at Olime Program	.0201	.0010	.0204	.0722
		(.0404)	(.0259)	(.0187)	(.0545)
	Studied at Correctional Facility	0103	0077	.0051	.003
		(.0145)	(.0084)	(.0069)	(.0105)
	Studied via Tutoring	0134	.0261	.0064	.0814**
		(.0238)	(.0186)	(.0156)	(.0393)
	Studied in Self-Directed Setting	0307	0096	.022	.0479
		(.0553)	(.0377)	(.0311)	(.0801)
	Studied at Other	0293	0088	.005	0101
		(.022)	(.0122)	(.0087)	(.0242)
	Highest grade $= 9$ or lower	0241	.0033	.0344	0343
		(.0477)	(.0289)	(.0221)	(.0624)
	Past Year Income Above Median	0756	.0989**	.034	.0001
		(.0638)	(.0495)	(.0316)	(.0806)
	Reason for no diploma: Academic	0846	03	.0568*	.0678
		(.0627)	(.0387)	(.0297)	(.0801)
	Reason for no diploma: Personal	.027	.0333	.0051	031
		(.0598)	(.0347)	(.0267)	(.074)
	Reason for no diploma: Other	0082	0182	0026	.0022
		(.0282)	(.0167)	(.0132)	(.0353)
	Reason for no diploma: Homeschool	0291	.0345*	0264	.0082
		(.0433)	(.0205)	(.0174)	(.0455)
	Reason for no diploma: Int'l	.0075	0232*	0016	0103
	-	(.0221)	(.0133)	(.0098)	(.0288)
В.	RD Estimates GED [®] College Ready T	hreshold (Subs	core > 175)	~ /	. ,
	Age at first test attempt	1.1608	.2026	2119	.2033
	0 1	(2.0977)	(1.0501)	(1.3659)	(.6599)
	Male	1117	0259	.0707	.0117
		(.2188)	(.0782)	(.0694)	(.0444)
	Asian	.0374	0381	0063	0324*
		(.1129)	(.0349)	(.0496)	(.0173)
	Black	0431	.0058	.0873*	.0103
		(1199)	(0286)	(0451)	(025)
	Hispanic	- 0859	0269	- 0156	- 0482
	mponio	(1238)	(047)	(0441)	(0.364)
	Native American	(1200)	_ 0012	(10441)	- 0055
	TAULIC MININGH	(0302)	0013	(0222)	0033
	White	(.0302)	0707	(.0222)	0453
	** 11100	1431	.0191	0012	.0400

	(.2126)	(.0736)	(.1019)	(.0477)
Multi-Racial	.0602	.0101	.0568	.0108
	(.0923)	(.0419)	(.0738)	(.0251)
Other Race	.1423*	0707	0185	.0142
	(.0809)	(.046)	(.024)	(.0198)
Employed	.3367	0578	0303	061
	(.2424)	(.0757)	(.0652)	(.0472)
Unemployed	2002	.0022	0779	.0482
	(.242)	(.0673)	(.1004)	(.0539)
Permanently Disabled	0	0074	0904	.003
·		(.0134)	(.0596)	(.0055)
Motivation: Personal	.0901	0807	.0371	0233
	(.1574)	(.0541)	(.0792)	(.0329)
Motivation: Work	2652	.0362	.0297	.0107
	(.2018)	(.0626)	(.1026)	(.0372)
Motivation: Education	.1699	.012	0548	0189
	(2275)	(0585)	(1343)	(0433)
Motivation: Special Requirement	0105	- 0198	- 0317	- 0094
Notivation. Special Requirement	(015)	(0152)	(0278)	(0089)
Motivation: Military	- 0485	(.0102) 033/*	- 034	0307**
Woolvation. Wintary	(0537)	(0105)	(0.004)	(0172)
Daid for Test Prop	(.0007)	(.0135)	(.0217)	0065
I ald for fest frep	(2820)	(0037)	.0288	(052)
Dropanad via Drastica Test	(.2039)	(.0622)	(.1364)	(.052)
Prepared via Practice Test	1075	1033	0005	0024
	(.1904)	(.0705)	(.0049)	(.0410)
Prepared via Books	.1084	.0184	0313	00/*
	(.1247)	(.0655)	(.123)	(.0389)
Prepared via Online Resources	.1417	0724	.0365	.0308
	(.148)	(.0657)	(.0499)	(.0384)
Prepared via Mobile App	.0176	077*	.0045	0127
	(.0637)	(.0398)	(.0643)	(.019)
Prepared via Social Network	.0229	0018	.0397*	0029
	(.0923)	(.0243)	(.0235)	(.0183)
Prepared via Other Resources	0561	.0078	.0098	.0049
	(.0495)	(.0166)	(.0204)	(.0128)
Prepared via Adult Education	.1774	.0174	.1036	0125
	(.1201)	(.0542)	(.1013)	(.0327)
Studied for Exam	.0005	0182	0778	0107
	(.1967)	(.0661)	(.1136)	(.0385)
Studied at Test Prep Center	.1203	.0014	.0295	005
	(.1469)	(.055)	(.0491)	(.0354)
Studied at Testing Center	0054	042	.0104	0014
	(.0548)	(.0305)	(.0271)	(.0093)
Studied at HS Program	0645	0067	.1055**	.0123
-	(.1194)	(.0352)	(.046)	(.0223)
Studied at CC Program	.0855	0112	.0188	0219
-	(.0666)	(.0247)	(.0271)	(.0157)
Studied at Workplace Program	0	0107	.0112	.0098
1 0		(.0133)	(.0165)	(.0082)
Studied at Online Program	0798	0907 [*]	0305	0028
	(.1582)	(.0538)	(.0716)	(.0296)
Studied at Correctional Facility	0137	0265	0174	.0041
· · · · · · · · · · · · · · · · · · ·	(.0167)	(.0180)	(.0141)	(.0075)
Studied via Tutoring	0987	.0175	.0338	0497**
8				

	(.1190)	(.0301)	(.0278)	(.0199)
udied in Self-Directed Setting	.0433	0258	2363*	0201
-	(.1508)	(.0814)	(.1326)	(.0441)
udied at Other	0339	.0054	.0229	.0041
	(.0579)	(.0168)	(.0169)	(.0128)
ghest grade $= 9$ or lower	.1338	0385	.0431	0539
	(.1332)	(.0597)	(.0515)	(.0356)
st Year Income Above Median	.2428	0653	1411	0113
	(.2131)	(.0648)	(.1229)	(.0442)
eason for no diploma: Academic	2559	0191	.0380	0268
-	(.2195)	(.0654)	(.101)	(.0416)
ason for no diploma: Personal	.0629	0585	.0303	.0510
	(.2273)	(.0689)	(.1226)	(.0394)
eason for no diploma: Other	.0487	0211	0004	.0307
	(.1166)	(.0346)	(.0369)	(.0199)
ason for no diploma: Homeschool	.0453	.0974*	.0316	0415
	(.1324)	(.0522)	(.0778)	(.0306)
eason for no diploma: Int'l	.0272	.0043	0344	0073
-	(.1365)	(.0227)	(.0278)	(.0143)
eason for no diploma: Personal eason for no diploma: Other eason for no diploma: Homeschool eason for no diploma: Int'l	$\begin{array}{c} .0629\\ (.2273)\\ .0487\\ (.1166)\\ .0453\\ (.1324)\\ .0272\\ (.1365)\end{array}$	$\begin{array}{c}0585\\ (.0689)\\0211\\ (.0346)\\ .0974^{*}\\ (.0522)\\ .0043\\ (.0227)\end{array}$	$\begin{array}{c} .0303\\ (.1226)\\0004\\ (.0369)\\ .0316\\ (.0778)\\0344\\ (.0278)\end{array}$	$\begin{array}{c} .0510\\ (.0394\\ .0307\\ (.0199\\041\\ (.0300\\007\\ (.0143)\end{array}$

Notes: This table reports coefficients testing whether discontinuities exist in the covariate indicated in each row at the GED[®] College Ready (Panel A) or GED[®] College Ready Plus Credits (Panel B) thresholds. Distance from each threshold is modeled as a linear function with separate slopes above and below the threshold, as in Equation (5). Bandwidth varies by covariate and subject test threshold, determined using the optimal bandwidth calculations described in Calonico et al. (2019). * = p < 0.1, ** = p < 0.05, *** = p < 0.01. To conserve space, I have omitted balance tests for all missing data variables as well as any baseline characteristics that characterize less than one percent of the sample in column (2) of Table 1. The sample is limited to the NSC sample described in column (4) of Table 1, N=15,262.

	Mathematical	Reasoning		Social
	Reasoning	thru. ELA	Science	Studies
	(1)	(2)	(3)	(4)
A. Discontinuities in subject-specific col	llege readiness de	esignations (1	NSC Sample)
GED [®] CR	.9972***	.9939***`	.9990***	1.0007***
	(.0021)	(.0028)	(.0007)	(.0007)
	7153	6993	9336	4694
GED [®] CR+C	1^{***}	1^{***}	.9985***	1^{***}
			(.0019)	
	1099	403	761	2198
B. Discontinuities in any college readine	ess designation (1	NSC Sample)		
$\operatorname{GED}^{\mathbb{R}}$ CR	.2429***	.2878***	$.3546^{***}$.3878***
	(.0239)	(.0196)	(.0183)	(.0238)
	4147	6993	8262	6846
$GED^{(\mathbb{R})} CR + C$.6008***	.4789***	.4086***	.5722***
	(.1802)	(.0838)	(.0852)	(.0444)
	369	725	824	1244
C. Discontinuities in Retaking Behavior	(NSC Sample)			
At $\text{GED}^{\mathbb{R}}$ CR Threshold (165)	0132	0050	.0003	.0016
	(.0096)	(.0043)	(.0019)	(.0012)
	2048	4916	4389	3782
At GED [®] CR+C Threshold (175)	0	.0053	0013	0
		(.0054)	(.0018)	
	369	725	761	1701
D. Discontinuities in Retaking Behavior	· (Full Sample)			
At Passing Threshold (145)	8076***	7536***	7644***	7230***
	(.0053)	(.0048)	(.0054)	(.0094)
	156435	161160	123635	86947
At $\text{GED}^{\mathbb{R}}$ CR Threshold (165)	.0010	0018**	0004	0003
	(.0026)	(.0007)	(.0012)	(.0012)
	14208	94616	34397	55170
At $\text{GED}^{\mathbb{R}}$ CR+C Threshold (175)	0077	0052**	0031	0009
	(.0066)	(.0024)	(.0019)	(.0015)
Observations	6474	12039	13355	22748

Table 3: First-Stage Estimates

Notes: This table reports coefficients testing whether discontinuities exist in the probability individuals ever reach the credentialling outcomes listed in each row (Panels A and B) or retakes that subtest (Panels C and D). RD models include all demographic covariates listed in Panel A of Table 1 and their corresponding missing data indicators, as well as fixed effects for quarter of first GED[®] subject test attempt and state of residence. Distance from each threshold is modeled as a linear function with triangular kernel weights and separate slopes above and below the threshold, as in Equations (1) and (2). Bandwidth varies by outcome and subject test threshold, determined using the optimal bandwidth calculations described in Calonico et al. (2019). * = p < 0.1, ** = p < 0.05, *** = p < 0.01.

	Mathematical	Reasoning		Social
	Reasoning	thru. ELA	Science	Studies
	(1)	(2)	(3)	(4)
A. Impact of crossing GED [®]	CR threshold			
Enrolled within 2 years	.009	008	.007	.063
	(.054)	(.036)	(.027)	(.078)
	2048	3781	6050	2887
Persisted ≥ 1.5 years	005	007	.013	.044
	(.045)	(.026)	(.021)	(.038)
	2048	3781	4389	3782
B. Impact of crossing GED [®]	^O CR+C thresho	old by subject		
Enrolled within 2 years	.106	.014	.146	100*
	(.225)	(.062)	(.106)	(.053)
	369	1338	824	1701
Persisted ≥ 1.5 years	033	.004	053	089**
	(.181)	(.059)	(.108)	(.040)
	369	997	761	1701

Table 4: RD estimates of impacts on college outcomes (ITT)

Notes: This table reports coefficients testing whether discontinuities exist in the probability individuals ever achieve the college outcomes indicated in each row. RD models include all demographic covariates listed in Panel A of Table 1 and their corresponding missing data indicators, as well as fixed effects for quarter of first GED[®] subject test attempt and state of residence. Distance from each threshold is modeled as a linear function with triangular kernel weights and separate slopes above and below the threshold, as in Equation (2). Bandwidth varies by outcome and subject test threshold, determined using the optimal bandwidth calculations described in Calonico et al. (2019). * = p < 0.1, ** = p < 0.05, *** = p < 0.01.

	RD estimates	by subject (G	ED [®] CR	threshold)	RD estimates b	y subject (GF	ED [®] CR+	C threshold)
Quarter Since 1st GED [®] Attempt	Mathematical Reasoning	Reasoning Thru. ELA	Science	Soc. Stud.	Mathematical Reasoning	Reasoning Thru. ELA	Science	Soc. Stud.
-	(1)	(2)	(3)	(4)	(5)	(9)	(2)	(8)
0 (Q of 1st test)	016	.012	.012	.043	.064	.041	.061	051
	(.035)	(.019)	(.015)	(039)	(.060)	(.049)	(990.)	(.033)
	2048	3781	6050	2887	771	297	824	1244
1	.042	017	.026	620.	.031	.027	.024	050
	(.044)	(.028)	(.020)	(.055)	(.194)	(.049)	(.109)	(.036)
	2048	3781	6050	2887	369	1338	761	1701
2	.059	016	008	.094	.131	.046	011	050
	(.044)	(.031)	(.023)	(.063)	(.126)	(.055)	(.115)	(.040)
	2048	3781	6050	2887	490	1338	761	1701
3	.030	016	.008	$.127^{**}$.102	.011	005	043
	(.046)	(.032)	(.022)	(.064)	(.189)	(.064)	(.115)	(.039)
	2048	3781	6050	2887	369	297	761	2198
4	049	043	.024	.076	.075	05	.086	032
	(.051)	(.031)	(.025)	(.064)	(.196)	(.065)	(.114)	(.039)
	2048	3781	4389	2887	369	297	761	2198
5	015	027	.019	.059	101	.031	.046	039
	(.049)	(.031)	(.025)	(.049)	(.129)	(.053)	(.112)	(.041)
	2048	3781	4389	3782	490	1338	761	1701
9	064	008	.020	.062	14	.087	.024	074*
	(.061)	(.027)	(.024)	(.061)	(.129)	(.064)	(.108)	(.041)
	1612	4916	6050	2887	490	266	761	1701
7	.065	015	.016	.071	.104	.037	.056	039
	(.047)	(.029)	(.022)	(.064)	(.18)	(.063)	(.102)	(.038)
	1909	3484	5667	2683	335	606	752	2030
8	.059	.001	.020	016	.222	045	.019	086**
	(.049)	(.029)	(.022)	(.074)	(.186)	(.056)	(.136)	(.039)
Observations	1709	3157	5184	2416	301	1085	624	1833
Notes: This table report each row (relative to the indicated GED [®] CR thre demographic covariates lis	s coefficients testing quarter in which th sholds, while the rig sted in Panel A of 7	whether discont ey took their firs cht half reports e Pable 1 and theii	inuities exis st GED [®] su stimated di	t in the probab ubject test). Th scontinuities at ling missing dat	ility individuals we ne left half of the t the indicated GED ta indicators, as we	re enrolled in co able reports esti ® CR+C thresh all as fixed effect	illege in the mated disco tolds. RD m s for quarter	quarter listed in ntinuities at the odels include all : of first GED [®]
subject test attempt and threshold. Bandwidth var $(300, 0.1, 0.1, 0.1, 0.1, 0.1, 0.1, 0.1, 0$	state of residence. I ries by outcome and $a^{***} - a^{**} - a^{***} - a^{$	Distance from ea l subject test thn	ch threshold eshold, det	l is modeled as ermined using t	a quadratic functic he optimal bandwi	on with separate dth calculations	slopes abov described in	e and below the Calonico et al.
(2019). = p < 0.1, = 1	p < 0.00, $= p < 0$	J.UI.						

Table 5: RD estimates of impacts on college enrollment by quarter since 1st $GED^{\textcircled{B}}$ subject test attempt

Subject	Outcome	Male (1)	Female (2)	URM (3)	Not URM (4)	Above Median ACE (5)	Below Median ACE (6)	Motivation Education (7)	Motivation Other (8)
Math	Enrolled within 2 years	047 (.071)	.08 (.096) 717	$.006$ $(.129)$ $_{376}$.036 (.069) 1296	.009 (.088) 841	.007 (069) 707	.021 (.075) 1179	.044 (.074) 857
Math	Persisted ≥ 1.5 years	(.056)	(.089). (.089)	.152152 (.117) 376	0.035. (0.057)	01 01 (.071) 841	.005 .005 (.06) 1207	(.071)	$.081^{\circ}$ (.044) .857
ELA	Enrolled within 2 years	(.05)	.011 $(.06)$	096 (.088)	$ \begin{array}{c} 0 \\ (.044) \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0$	050 (.056) 1540	.021. $.021$. $(.047)$	021 (.052)	.022 (.047)
ELA	Persisted ≥ 1.5 years	1940 014 (.035) 1948	0.000. 0.10 . $(.043)$	000 035 (.066) 686	$^{2434}_{024}$ (.031) $^{2434}_{034}$	1.049057057 (.040)	028 (.034)	2023 038 (.042) 2020	.030 .030 (.028) 1733
Science	Enrolled within 2 years	029 (.035) 3431	.057 (.047) 2245	.013 (.065) 1110	2321. 013 . $(.034)$ 3904	2505073* (.041) 2505	$.062^{+}+$ (.036) 3545	.022. $(.04)$	006 (.034) 2813
Science	Persisted ≥ 1.5 years	022 (.026) 2479	$.069^{*}$ (.039) 1645	022 (.051) 790	.022 (.026) 2847	$\begin{array}{c} 0 \\ (.032) \\ 1827 \end{array}$.023 $(.028)$ 2562	.013 $(.035)$ 2334	.021 (.02) 2029
Soc. Stud.	Enrolled within 2 years	.033 $(.101)$ 1629	.225 (.137) 1102	.027 (.194) 619	.055 (.102) 1798	.026 (.118) 1187	.091 (.107) 1700	.073 (.117) 1525	.019 (.094) 1344
Soc. Stud. Observations	Persisted ≥ 1.5 years	.053 (.045) 2152	.061 (.072) 1421	043 (.097) 801	.06 (.047) 2364	.013 $(.06)$ 1557	.076 (.049) 2225	.032 (.066) 2006	.04 (.033) 1756
Notes: This table each row for the s listed in Panel A state of residence the threshold, as described in Calo	le reports coefficients testing who subgroups indicated in each colu of Table 1 and their correspondi \therefore Distance from each threshold in Equation (2). Bandwidth v mico et al. (2019). * = $p < 0.1$,	ether disco imn at the ing missin is modele aries by c $^{**} = p < 0$	ontinuities e GED [®] CR g data indic d as a linee utcome and 0.05, *** = l	exist in the exist in the threshold threshold at the state of the transformer is the transformer in the subject $0 < 0.01$.	probability in indicated in e ell as fixed effe t with triangul test threshold,	dividuals ev ach row. RJ ects for quar ar kernel we determined	er achieve th O models inc ter of first C sights and so i using the o	he college outcor lude all demogr JED [®] subject t eparate slopes a optimal bandwi	mes indicated in aphic covariates est attempt and bove and below dth calculations

						Above Median	Below Median	Motivation	Motivation
Subject	Outcome	Male (1)	Female (2)	URM (3)	Not URM (4)	ACE (5)	ACE (6)	Education (7)	Other (8)
Math	Enrolled within 2 years	.231	.203	.622	.074	.980***	329+	.104	141
	2	(.283)	(.382)	(1.008)	(.336)	(.348)	(.267)	(.295)	(.298)
		220	120	00	243	141	228	234	134
Math	Persisted ≥ 1.5 years	008	.086	264	225	.190	265	207	065
		(.236)	(.368)	(.570)	(.263)	(.306)	(.244)	(.261)	(.270)
		220	120	09	243	141	228	234	134
ELA	Enrolled within 2 years	.071	042	.125	.089	040	.078	015	056
		(60.)	(260.)	(.176)	(.076)	(260.)	(.080)	(.082)	(.085)
		679	545	182	895	562	276	786	549
ELA	Persisted ≥ 1.5 years	$.137^{*}$	116+	.227	.077	163*	$.125^{+}$.048	139^{**}
		(.076)	(.102)	(.162)	(.071)	(960.)	(.075)	(.089)	(.069)
		495	418	136	660	422	575	592	403
Science	Enrolled within 2 years	.102	$.360^{*}$	$.622^{*}$.005+	.082	.175	.116	.163
		(.135)	(.192)	(.345)	(.132)	(.165)	(.146)	(.148)	(.144)
		515	238	101	553	330	494	489	333
Science	Persisted ≥ 1.5 years	072	.192	.342	117	01	034	036	112
		(.130)	(.203)	(.371)	(.131)	(.182)	(.133)	(.165)	(.118)
		477	217	91	510	301	460	451	308
Soc. Stud.	Enrolled within 2 years	115*	021	172	019	186**	028	160^{**}	050
		(070.)	(.098)	(.165)	(.066)	(.082)	(.071)	(.075)	(.073)
		1016	532	233	1121	669	1002	951	743
Soc. Stud.	Persisted ≥ 1.5 years	114**	013	014	053	073	110**	153^{**}	044
		(.050)	(070)	(.108)	(.050)	(.062)	(.052)	(.063)	(.044)
Observations		1016	532	233	1121	669	1002	951	743
Notes: This takes each row for the listed in Panel A	ole reports coefficients testing when the subgroups indicated in each colure of Table 1 and their correspond	Thether discont turn at the (ding missing	ontinuities e GED® CR+ 3 data indic	xist in the -C threshol ators, as w	probability inc d indicated in ϵ ell as fixed effe	lividuals eve each row. Rl cts for quart	er achieve th D models inc ter of first G	ne college outcor clude all demogr TED [®] subject to	nes indicated in aphic covariates est attempt and
threshold, as in the second	e. Distance from each threshold Equation (2). Bandwidth varies (2010) $* - \frac{1}{2} \sim 0.1 * * - \frac{1}{2} \sim 0.0$	l is modeled by outcome $0.05^{***} = i$	as a linear and subject	function w ot test thre	ith triangular l shold, determir	ternel weigh ted using the	tts and separ e optimal b	rate slopes abov andwidth calcul	e and below the ations described

Appendix Figures



Appendix Figure 1: Observation density by subject test score

Notes: Each histogram plots the discrete density of score observations in each GED[®] subject test. Scores are restricted to an individual's first attempt of each subject test. Red lines mark the minimum scores required to earn the GED[®] CR (165) and GED[®] CR+C (175) designations. P-values beside each threshold report the results of a test for the smoothness of the density function, accounting for the discreteness of the distribution using the methods described in Frandsen (2017) with the bound coefficient set to k = 0.05.



Appendix Figure 2: RD estimates using alternative bandwidths

Notes: Each graph plots coefficients for the RD estimates from Table 4 using the optimal bandwidth, as well as RD estimates using all possible symmetric integer bandwidths between 3 and 9 (GED[®] CR) or between 3 and 10 (GED[®] CR+C). Bandwiths are reported in parentheses next to each point estimate. RD models include all demographic covariates listed in Panel A of Table 1 and their corresponding missing data indicators, as well as fixed effects for quarter of first GED[®] subject test attempt and state of residence. Distance from each threshold is modeled as a linear function with triangular kernel weights and separate slopes above and below the threshold, as in Equation (2).



Appendix Figure 3: RD plots, subject test retaking by threshold (Reasoning through Lang. Arts)









Appendix Figure 3 (cont'): RD plots, subject test retaking by threshold (Social Studies)

Notes: Each graph plots rates of exam retaking above and below the indicated GED[®] subject test score thresholds, as well as lines of best fit and 95% confidence interval on either side of the threshold. Exam retaking is a binary variable that takes on the value 1 if an individual was observed taking a GED[®] subject test in the indicated content area more than once and zero otherwise. Bandwidth and effective sample size vary by subject test threshold, determined using the optimal bandwidth calculations described in Calonico et al. (2019); see panel D of Table 3 for the effective number of observations that contribute to each estimate. Distance from each threshold is modeled as a linear function with triangular kernel weights and separate slopes above and below each threshold, as in Equation (2). Points are scaled relative to the size of the sample that earned that subject test score.

Appendix Tables

	Attempted	Completed	Passed	Passed	Passed
	At Least 1	All 4	All 4	All 4	All 4
	Subj. Test	Subj. Tests	(145+)	(150+)	At Time
	(1)	(2)	(3)	(4)	(5)
2014	$172,\!258$	98,434	72,955	$58,\!481$	58,481
2015	$244,\!193$	$138,\!917$	$112,\!660$	90,824	90,824
2016	$302,\!169$	$193,\!351$	$153,\!673$	69,248	$151,\!117$
2017	299,727	$192,\!527$	$151,\!175$	66,000	$151,\!175$
2018	$299,\!668$	$193,\!406$	$154,\!956$	$64,\!397$	$154,\!956$
2019	$291,\!057$	183,779	$145,\!082$	$63,\!163$	$145,\!082$

Appendix Table 1: GED[®] Counts by Year

Note: Cells record the number of GED[®] testers who met the criteria described in each column in the year listed in each row. Individuals who attempted subject tests in multiple years are counted as having attempted at least 1 subject test in each year they take a subject test. Individuals who complete all four subject tests are only recorded as completers until they cross the indicated passing threshold for the first time, and individuals who pass all four subject tests are only counted as passing the indicated threshold in the first year they pass that threshold in all four subject tests (i.e., the small number of testers who retake a subject test in subsequent years after passing all four subject tests are recorded as attempting at least 1 subject test in that year, but not as completing or passing the exam in subsequent years). Columns (3) and (4) record the number of people who passed all four subject tests with a minimum score of 145 or 150 for the first time in that year (including retakes). Column (5) records the number of people who met the relevant passing criteria at the time they took the $\text{GED}^{\textcircled{R}}$ exam. Note that this count corresponds to the counts in column (3) or column (4) for all years except 2016, when the passing standard was revised on January 26. For 2016, testers whose last subject test attempt occured before January 26 are recorded as being subject to the 150 passing standard, while individuals who took their last subject test after January 26 are recorded as being subject to the revised 145 passing standard. N=1,258,509.

Subject Test	$\operatorname{GED}^{\textcircled{R}}\operatorname{CR}$	$\operatorname{GED}^{(\!\!R\!)}\operatorname{CR}+\operatorname{C}$
Mathematical Reasoning	7.26%	2.00%
Reasoning through Lang. Arts	12.66%	2.22%
Science	13.41%	1.98%
Social Studies	16.03%	4.17%
Any Subject	22.37%	6.05%

Appendix Table 2: Percent of testers meeting college readiness subject test score thresholds

Note: N=644,615.

	2014	2015	2016	2017	2018	2019
	(1)	(2)	(3)	(4)	(5)	(6)
A. GED College Ready or higher (S	Subject 7	Test Scor	re = 165	+)		
Mathematical Reasoning	7942	12312	11797	10814	10520	10414
Reasoning through Lang. Arts	23171	24326	19737	20508	21596	20553
Science	13466	18175	23102	22136	20827	17264
Social Studies	18672	23419	29345	27799	26233	24275
Any Subject	34392	43383	46428	45902	44668	41238
B. GED College Ready Plus Credit	s(175+)	1				
Mathematical Reasoning	1786	2583	3119	2459	2911	2935
Reasoning through Lang. Arts	3479	3382	3199	3503	3765	3329
Science	1873	2432	3407	3072	2808	2418
Social Studies	4493	5316	7464	6229	5780	7403
Any Subject	7779	9262	11746	10838	10813	11505

Appendix Table 3: GED CR, CR+C, and Honors Counts by Year

Note: Cells report the number of individuals who met the minimum requirements for the designation described in each panel in the subject and year described in the corresponding row and column. Note that the GED[®] College Ready and College Ready Plus Credits designations were not awarded until January 26, 2016. Between January 1, 2014 and January 25, 2016, testers who achieved a score of at least 170 in a given subject test earned a GED[®] Honors designation. N=1,258,509.

		Mathematical	Reasoning		
	Trend	Reasoning	Thru. ELA	Science	Soc. Stud.
		(1)	(2)	(3)	(4)
A. Impact of crossing GED ^(F)	[®] CR thresh	old by subject			
Enrolled within 2 years	Linear	.009	008	.007	.063
0		(.054)	(.036)	(.027)	(.078)
		2048	3781	6050	2887
Enrolled within 2 years	Quadratic	003	028	007	.100
-	-	(.087)	(.086)	(.05)	(.128)
		2943	3781	6050	3782
Enrolled within 2 years	Cubic	010	051	259	.108
		(.208)	(.157)	(.177)	(.229)
		2943	4916	4389	4694
Peristed ≥ 1.5 years	Linear	005	007	.013	.044
		(.045)	(.026)	(.021)	(.038)
		2048	3781	4389	3782
Peristed ≥ 1.5 years	Quadratic	069	.003	.014	.050
		(.085)	(.043)	(.034)	(.084)
		2048	4916	6050	3782
Peristed ≥ 1.5 years	Cubic	140	045	097	.034
		(.150)	(.114)	(.128)	(.154)
		2943	4916	4389	4694
B. Impact of crossing GED [®]	CR+C three	eshold			
Enrolled within 2 years	Linear	.106	.014	.146	100*
		(.225)	(.062)	(.106)	(.053)
		369	1338	824	1701
Enrolled within 2 years	Quadratic	.176	.014	.173	113
		(.354)	(.118)	(.141)	(.071)
		490	1338	1591	2198
Enrolled within 2 years	Cubic	818	.001	451	096
		(.684)	(.303)	(.400)	(.137)
		490	1338	1591	2383
Peristed ≥ 1.5 years	Linear	033	.004	053	089**
		(.181)	(.059)	(.108)	(.040)
		369	997	761	1701
Peristed ≥ 1.5 years	Quadratic	.051	.061	446**	104**
		(.311)	(.133)	(.210)	(.052)
		490	1338	824	2198
Peristed ≥ 1.5 years	Cubic	.424	.102	-1.32**	130
		(.534)	(.252)	(.563)	(.097)
Observations		771	1338	1471	2383

Appendix Table 4: RD estimates using alternative local polynomial trends

Notes: This table re-estimates the main results from Table 4 varying the degree of the local polynomial used to model the relationship between the outcomes of interest and the running variable on either side of each GED[®] college readiness threshold. The "Trend" column indicates the functional form of the local polynomial used to generate each set of RD estimates. * = p < 0.1, ** = p < 0.05, *** = p < 0.01.

		Mathematical	Reasoning		
	Covariates	Reasoning	Thru. ELA	Science	Soc. Stud.
		(1)	(2)	(3)	(4)
A. Impact of crossing GED ⁽⁾	³ CR threshold by s	ubject			
Enrolled within 2 years	None	.007	013	004	.081
Ŭ		(.054)	(.037)	(.028)	(.077)
Enrolled within 2 years	Demographics	.009	008	.007	.063
, i i i i i i i i i i i i i i i i i i i		(.054)	(.036)	(.027)	(.078)
Enrolled within 2 years	All baseline chars	.021	.001	.013	.048
U U		(.052)	(.035)	(.026)	(.076)
Persist ≥ 1.5 years	None	010	013	.007	.052
, , , , , , , , , , , , , , , , , , ,		(.046)	(.026)	(.022)	(.038)
Persist ≥ 1.5 years	Demographics	005	007	.013	.044
-		(.045)	(.026)	(.021)	(.038)
Persist ≥ 1.5 years	All baseline chars	001	002	.016	.034
-		(.044)	(.025)	(.021)	(.038)
		2048	3781	4389	3782
B. Impact of crossing GED ^(f)	© CR+C threshold b	y subject			
Enrolled within 2 years	None	.020	.007	.135	098*
		(.222)	(.063)	(.105)	(.054)
Enrolled within 2 years	Demographics	.106	.014	.146	100*
		(.225)	(.062)	(.106)	(.053)
Enrolled within 2 years	All baseline chars	001	025	.167	089*
		(.214)	(.059)	(.104)	(.052)
Persist ≥ 1.5 years	None	063	.002	044	088**
		(.187)	(.060)	(.106)	(.039)
Persist ≥ 1.5 years	Demographics	033	.004	053	089**
		(.181)	(.059)	(.108)	(.040)
Persist ≥ 1.5 years	All baseline chars	061	024	069	072*
		(.176)	(.057)	(.108)	(.038)
Observations		369	997	761	1701

Appendix Table 5: RD estimates using alternative covariate sets

Notes: This table re-estimates the main results from Table 4 varying the set of covariates included in the RD model. The "Covariates" column indicates which covariates were included to generate for each set of RD estimates. "Demographics" includes all covariates in Panel A of Table 1 as well as their corresponding missing data indicators. "All baseline chars" includes all covariates in Panels A, B, and C of Table 1. * = p < 0.1, ** = p < 0.05, *** = p < 0.01.

Outcome	Method	Mathematical Reasoning (1)	Reasoning Thru. ELA (2)	Science (3)	Social Studies (4)
A. Impact of crossing GED ⁽⁾	³ CR threshold by subject	:t			
Enrolled within 2 years	Conventional	.009	008	.007	.063
U U		(.050)	(.035)	(.027)	(.072)
		2048	3781	6050	2887
Enrolled within 2 years	Bias-Corrected	.002	001	.005	.082
		(.050)	(.035)	(.027)	(.072)
		2048	3781	6050	2887
Enrolled within 2 years	Robust Bias-Corrected	.002	001	.005	.082
		(.062)	(.047)	(.035)	(.087)
		2048	3781	6050	2887
Persisted ≥ 1.5 years	Conventional	005	007	.013	.044
		(.041)	(.026)	(.021)	(.037)
		2048	3781	4389	3782
Persisted ≥ 1.5 years	Bias-Corrected	020	001	.015	.049
		(.041)	(.026)	(.021)	(.037)
		2048	3781	4389	3782
Persisted ≥ 1.5 years	Robust Bias-Corrected	020	001	.015	.049
		(.052)	(.035)	(.028)	(.045)
		2048	3781	4389	3782
B. Impact of crossing $\text{GED}^{\textcircled{e}}$	[®] CR+C threshold by sub	oject			
Enrolled within 2 years	Conventional	.106	.014	.146	100**
		(.176)	(.059)	(.092)	(.049)
		369	1338	824	1701
Enrolled within 2 years	Bias-Corrected	.169	.033	.173*	116**
		(.176)	(.059)	(.092)	(.049)
		369	1338	824	1701
Enrolled within 2 years	Robust Bias-Corrected	.169	.033	.173	116**
		(.212)	(.074)	(.110)	(.058)
		369	1338	824	1244
Persisted ≥ 1.5 years	Conventional	033	004	053	089**
		(.138)	(.055)	(.094)	(.037)
		369	997	761	1701
Persisted ≥ 1.5 years	Bias-Corrected	017	.014	096	099***
		(.138)	(.055)	(.094)	(.037)
		369	997	761	1701
Persisted ≥ 1.5 years	Robust Bias-Corrected	017	.014	096	099**
		(.170)	(.069)	(.116)	(.043)
Observations		369	997	761	1701

Notes: This table re-estimates the main results from Table 4 using the robust, bias-corrected local polynomial methods described in Calonico et al. (2019). Local linear regression is used to construct the point estimator. Local quadratic regression is used to construct the bias correction. * = p < 0.1, ** = p < 0.05, *** = p < 0.01.

						B	andwith				
Subject	Threshold	Outcome	Optimal	bw=2	3	4	ъ	9	2	×	6
Math	CR (165)	Enrolled within 2 years	600.	015	.011	.017	.026	.028	.027	.023	.024
			(.054)	(.071)	(.053)	(.04)	(.036)	(.033)	(.03)	(.028)	(.027)
			2048	1612	2048	2943	3480	4147	5067	6052	7153
Math	CR+C (175)	Enrolled within 2 years	.106	.117	.071	047	057	064	07	067	063
			(.225)	(.22)	(.142)	(60.)	(.083)	(.07)	(.064)	(.059)	(.054)
			369	369	490	771	837	1099	1305	1527	1952
ELA	CR (165)	Enrolled within 2 years	008	019	004	005	01	011	008	009	012
			(.036)	(.047)	(.035)	(.029)	(.026)	(.024)	(.022)	(.021)	(.02)
			3781	2697	3781	4916	5985	6993	8004	8991	9872
\mathbf{ELA}	CR+C (175)	Enrolled within 2 years	.014	.003	.014	.015	.002	008	02	027	031
			(.062)	(960.)	(.071)	(.058)	(.051)	(.047)	(.044)	(.041)	(.039)
			1338	725	697	1338	1705	2089	2533	3047	3562
Science	CR (165)	Enrolled within 2 years	200.	023	600.	000.	.004	.007	.01	.01	.012
			(.027)	(.05)	(.03)	(.026)	(.025)	(.022)	(.02)	(.019)	(.018)
			6050	3164	4389	6050	6501	8262	9336	10274	11050
Science	CR+C (175)	Enrolled within 2 years	.146	.129	.152	.087	.082	.064	.059	.042	.034
			(.106)	(.129)	(.104)	(.062)	(.057)	(.05)	(.048)	(.044)	(.042)
			824	761	824	1471	1591	2325	2405	3336	3832
Soc. Stud.	CR (165)	Enrolled within 2 years	.063	.064	.065	.017	.003	003	002	004	005
			(.078)	(.078)	(.056)	(.033)	(.027)	(.025)	(.023)	(.021)	(.02)
			2887	2887	3782	4694	5902	6846	7760	8535	9666
Soc. Stud.	CR+C (175)	Enrolled within 2 years	1*	122	103*	08*	069	059	063	055	051
			(.053)	(.086)	(.056)	(.046)	(.042)	(.042)	(.039)	(.036)	(.035)
			1701	824	1244	1701	2198	2383	3105	3558	4031
Notes: This t_{δ} optimal bandw. ** = $p < 0.05$,	the reports the metric idth, as well as RI $^{***} = p < 0.01.$	ain enrollment estimates from $\overline{\mathbf{T}}_{\mathbf{k}}$) estimates using all possible sy	able 4 at GE /mmetric inte	D [®] CR or sger bandv	r GED [®] C widths bet	R+C sub ween 2 an	ject test th d 9 points	nreshold in in the inc	ldicated in licated co	l each row lumns. * =	using the = $p < 0.1$,

Appendix Table 7a: RD estimates using alternative bandwidths (Enrollment)

Appendix Table 7b: RD estimates using alternative bandwidths (Persistence)

						В	andwith				
Subject	Threshold	Outcome	Optimal	bw=2	3	4	5	9	7	8	6
Math	CR (165)	Persisted ≥ 1.5 years	005	024	.014	.035	.039	.033	.036	$.037^{*}$	$.036^{*}$
			(.045)	(.053)	(.04)	(.031)	(.028)	(.025)	(.023)	(.022)	(.021)
			2048	1612	2048	2943	3480	4147	5067	6052	7153
Math	CR+C (175)	Persisted ≥ 1.5 years	033	05	158	081	085	102*	109**	107**	11**
			(.181)	(.182)	(.118)	(220.)	(.071)	(.06)	(.054)	(.05)	(.046)
			369	369	490	771	837	1099	1305	1527	1952
\mathbf{ELA}	CR (165)	Persisted ≥ 1.5 years	007	-000	006	014	015	014	013	015	018
			(.026)	(.034)	(.025)	(.021)	(.019)	(.017)	(.016)	(.015)	(.014)
			3781	2697	3781	4916	5985	6993	8004	8991	9872
\mathbf{ELA}	CR+C (175)	Persisted ≥ 1.5 years	.004	.021	.004	011	021	02	024	027	033
			(.059)	(620.)	(.059)	(.048)	(.043)	(.039)	(.036)	(.034)	(.032)
			266	725	2000	1338	1705	2089	2533	3047	3562
Science	CR (165)	Persisted ≥ 1.5 years	.013	.014	.013	000.	.004	.002	001	001	.001
			(.021)	(.034)	(.021)	(.019)	(.018)	(.016)	(.015)	(.014)	(.013)
			4389	3164	4389	6050	6501	8262	9336	10274	11050
Science	CR+C (175)	Persisted ≥ 1.5 years	053	06	.003	.041	.056	.034	.02	0	006
			(.108)	(.105)	(.084)	(.052)	(.047)	(.041)	(.039)	(.036)	(.035)
			761	761	824	1471	1591	2325	2405	3336	3832
Soc. Stud.	CR (165)	Persisted ≥ 1.5 years	.044	.046	.044	.031	.03	.027	.026	.025	$.025^{*}$
			(.038)	(.051)	(.037)	(.023)	(.019)	(.018)	(.016)	(.015)	(.015)
			3782	2887	3782	4694	5902	6846	2760	8535	9666
Soc. Stud.	CR+C (175)	Persisted ≥ 1.5 years	089**	133**	092**	077**	071**	065**	061**	053*	049*
			(.04)	(.063)	(.042)	(.035)	(.032)	(.031)	(.029)	(.027)	(.026)
			1701	824	1244	1701	2198	2383	3105	3558	4031
Notes: This t_{ϵ} bandwidth, as *** = $p < 0.01$.	able reports the me well as RD estima	ain persistence estimates from ttes using all possible symme	1 Table 4 at C tric integer b	HED® CR o	or GED [®] C between 2 a	R+C subjec and 9 points	ct test thres s in the ind	hold indicat icated colur	ted in each r mns. $^* = p$	ow using th $< 0.1, ** =$	the optimal $p < 0.05$,

	Mathematical	Reasoning		
	Reasoning	Thru. ELA	Science	Soc. Stud.
	(1)	(2)	(3)	(4)
A. Discontinuities in	n exam passing d	& subject-spec	ific CR des	signations
Passed $\text{GED}^{\mathbb{R}}$.1940***	.1190***	.0379***	.1003***
	(.0056)	(.0053)	(.0101)	(.0061)
	156435	195403	89855	184052
GED $^{\textcircled{R}}$ CR	.9956***	.9982***	.9993***	1^{***}
	(.0016)	(.0008)	(.0007)	
	26382	72753	53996	13251
GED $^{\textcircled{R}}$ CR+C	1^{***}	.9985***	.9982***	1^{***}
		(.0008)	(.0010)	
	1522	22342	6110	5529
B. Discontinuities in	n any college rea	diness designa	tion	
GED $^{\textcircled{R}}$ CR	.2784***	.3975***	.4016***	.5379***
	(.0075)	(.0101)	(.0101)	(.0123)
	47743	51431	53996	55170
GED $^{\textcircled{R}}$ CR+C	.5250***	.5277***	.3916***	.6881***
	(.0242)	(.0152)	(.0237)	(.0076)
	3170	16676	11411	10285

Appendix Table 8: Additional Full-Sample First Stage RD Estimates

Notes: This table reports coefficients testing whether discontinuities exist in the probability individuals ever reach the credentialling outcomes listed in each row. RD models include all demographic covariates listed in Panel A of Table 1 and their corresponding missing data indicators, as well as fixed effects for quarter of first GED[®] subject test attempt and state of residence. Distance from each threshold is modeled as a linear function with triangular kernel weights and separate slopes above and below the threshold, as in Equation (1). Bandwidth varies by outcome and subject test threshold, determined using the optimal bandwidth calculations described in Calonico et al. (2019). * = p < 0.1, ** = p < 0.05, *** = p < 0.01.

	Mathematical	Reasoning		
	Reasoning	Thru. ELA	Science	Soc. Stud.
	(1)	(2)	(3)	(4)
A. Impact of earning any Gl	ED [®] CR designa	ation		
Enrolled within 2 years	.035	031	.021	.224
	(.206)	(.143)	(.078)	(.272)
	2048	3781	6050	2887
Persisted ≥ 1.5 years	017	028	.037	.136
	(.167)	(.104)	(.061)	(.118)
	2048	3781	4389	3782
B. Impact of earning any Gl	$ED^{\mathbb{R}} CR + C des$	ignation		
Enrolled within 2 years	.183	.026	.359	170*
	(.361)	(.11)	(.259)	(.090)
	369	1338	824	1701
Persisted ≥ 1.5 years	057	.008	132	150**
	(.285)	(.105)	(.256)	(.067)
Observations	369	997	761	1701

Appendix Table 9: RD estimates of any CR or CR+C designation (LATE)

Notes: This table reports two-stage least squares instrumental variable estimates of the marginal impact of earning any GED[®] CR or GED[®] CR+C designation on college outcomes. RD models instrument for earning any college readiness designation with an indicator for whether an individual reached a given GED[®] CR or GED[®] CR+C threshold on his or her first GED[®] subject test score in that subject, controlling for distance from the threshold and its interaction with being above the threshold. RD models include all demographic covariates listed in Panel A of Table 1 and their corresponding missing data indicators, as well as fixed effects for quarter of first GED[®] subject test attempt and state of residence. Distance from each threshold is modeled as a linear function with triangular kernel weights and separate slopes above and below the threshold, as in Equation (2). Bandwidth varies by outcome and subject test threshold, determined using the optimal bandwidth calculations described in Calonico et al. (2019). * = p < 0.1, ** = p < 0.05, *** = p < 0.01.