

1 **Longitudinal analysis of a diversity support program in biology: a national call for**
2 **further assessment**

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25 **Abstract**

26 National calls to improve the performance and persistence of students from historically
27 underrepresented backgrounds in science have led to a surge of research on inclusive,
28 evidence-based teaching methods. Less work has examined the effects of diversity support
29 initiatives that improve campus climate and community cohesion. Here, we examine whether
30 participation in the Biology Scholars Program (BSP) at Cornell University— a diversity
31 support program at a prominent university—affects underrepresented racial minority (URM)
32 student performance. We found BSP participants are less academically prepared when they
33 enter college, but typically have similar GPAs to their non-BSP counterparts upon
34 graduation, thereby closing achievement gaps. Although the BSP appears to help URM
35 students, we cannot assert that the BSP alone is responsible for these effects; future work
36 should isolate effective strategies that contribute to student success. In response to these
37 results, we lay out strategies that support programs could implement to maximize positive
38 impacts.

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41 **Keywords:** STEM equity, science diversity program, Biology Scholars Program, minority
42 students

43

44 **Introduction**

45 Minority demographics are underrepresented in science, technology, engineering, and
46 mathematics (STEM) disciplines (Landivar 2013), highlighting the need for effective
47 approaches that promote and retain student diversity (Brewer and Smith 2011).
48 Underrepresented racial minority (URM) students in the United States include African
49 American, Hispanic, Pacific Islander, and Native American undergraduates, and each
50 demographic faces significant inequity before and upon entering university. Social challenges
51 that disproportionately affect URM students include transitioning to college (Cooper et al.
52 2005), feelings of exclusion (Hurtado and Ruiz 2012), stereotype threat (Cohen and Garcia
53 2008, Steele 1997), and discrimination (Milkman et al. 2015). Within the classroom, URM
54 students are more likely to struggle in large introductory science classes (Alexander et al.
55 2009) due to inadequate high school preparation and limited opportunities to interact with
56 instructors (Hurtado et al. 2011). A negative learning environment can undermine self-
57 efficacy, which reduces the number of URM students who enter STEM majors and complete
58 a STEM degree (Olson and Riordan 2012). The gap in racial and ethnic demographic
59 representation also widens as students progress through the STEM pathway and enter the
60 workforce. For example, although 10.8 percent of the total workforce in the United States
61 was black or African American in 2011, they held only 6.4 percent of STEM jobs. Similarly,
62 14.9 percent of the total workforce identified as Hispanic or Latino, but they held only 6.5
63 percent of STEM jobs (Landivar 2013). Initiatives supporting URM students in higher
64 education thus require creative practices rather than the replication of past practices that
65 have yet to achieve the desired goal of improving racial and ethnic diversity in STEM.

66 One way that many campuses have tried to promote and retain URM students in
67 STEM is through diversity support programs that focus on aspects of student life outside
68 the classroom. Although few URM support programs have identified specific strategies that
69 improve student performance or other quantitative metrics of success, a handful of
70 programs have been successful in their efforts to support URM students in STEM (Gándara
71 and Maxwell-Jolly 1999; Cota-Robles and Gordan 1999; Matsui, Liu, and Kane 2003;
72 Summers and Hrabowski 2006; Buchwitz et al. 2012; Barlow and Villarejo 2004). The overall
73 lack of quantitative studies on diversity support programs in STEM could be due to low
74 numbers of participating students, unfavorable results, or the inability to disseminate their
75 data with the wider scientific community. Regardless, in order to clarify positive strategies,

76 institutional programs should rigorously and regularly self-assess student performance in a
77 manner consistent with the way STEM researchers address their own scientific questions.

78 Here, we analyze a longitudinal data set of students enrolled in Biological Sciences at
79 Cornell University between fall 2008 and fall 2015. We compare performance metrics among
80 non-URM and URM students that either participated in an institutional support initiative or
81 did not. The Biology Scholars Program is an undergraduate program based out of Cornell
82 University's Office of Undergraduate Biology in collaboration with the College of
83 Agriculture and Life Sciences and the College of Arts and Sciences. The program's mission
84 is to increase the satisfaction, retention, and graduation of historically underrepresented
85 students in the Biological Sciences, and to promote the value of educating a diverse
86 population of students in the sciences.

87 To assess the impact of the BSP at Cornell University, we evaluated the preparedness
88 and performance of students that varied in their URM status and whether they participated
89 in the BSP program with three metrics: 1) SAT scores; 2) cumulative GPA; 3) graduation
90 rates. Thus, we use a quantitative approach that is modified from Matsui et al. (2003) to
91 examine variation in both preparedness and performance among biology students at Cornell.

93 **Research Participants**

94 We gathered a longitudinal data set that spans 15 semesters from fall 2008 through fall 2015.
95 We compared the academic performance of 3159 students distributed among four groups: 1)
96 non-URM non-BSP (n = 2221; 'the majority'); 2) non-URM BSP (n = 51; including low-
97 socioeconomic status or first-generation college students); 3) URM non-BSP (n = 706);
98 4) URM BSP (n = 181). We considered participants of the BSP students who are either
99 currently active members or those who remained in the program for at least four semesters;
100 we removed fifty students because they did not fit these criteria. Through follow up surveys
101 with students who left the BSP, we found two emerging reasons students leave the program:
102 because they decided to pursue a non-science career path, or the BSP was too large of a time
103 commitment. All students who were included in the analysis were intended biology majors,
104 or those who stated in their admissions application that they intend to study the
105 fundamentals of biology and declare a concentration in one of the following: animal
106 physiology; biochemistry; computational biology; ecology and evolutionary biology; genetics,
107 genomics, and development; insect biology; marine biology; microbiology; molecular and cell

108 biology; neurobiology and behavior; human nutrition; plant biology; systematics and biotic
109 diversity. Of our entire student population who graduated, 41% of all students who entered
110 biological sciences graduated with a bachelor of science (N = 615), and 59% graduated with
111 a Bachelor of Arts (N = 877).

112 All experimental procedures on participants were approved by Cornell's Institutional
113 Research Board for human participants (protocol 1410005010). Anonymized data is
114 accessible through the DRYAD digital repository.

115

116 **Program description**

117 Between 2008 and 2015, 925 of 3,199 students enrolled in Biological Sciences
118 described themselves as URM (29% of students). Of those URM students, 24% participated
119 in the BSP, representing 7% of all biology majors. From data available between 2009-2015,
120 244 students were accepted to participate in the BSP out of a pool of 599 applicants (41%)
121 entering Biological sciences. Prior to the program's conception, the university did not
122 provide any unique support to historically underserved students. In response to a national
123 call to action led by the Howard Hughes Medical Institute in 2005, teams of administrators
124 from a number of universities across the country met to discuss the state of the nation's
125 historically underserved student populations, and to generate new ideas on how to better
126 support them. Following that meeting, the group that attended from Cornell met regularly in
127 order to develop what became the BSP, including an on-site visit to Meyerhoff Scholars
128 Program at University of Maryland, Baltimore County. Therefore, the BSP was conceived in
129 an effort to promote and retain URM students within the sciences, with the ultimate goal of
130 diversifying the STEM workforce. The program is institutionally funded out of provost
131 office at Cornell.

132 According to personal communication with program directors (November 12, 2016),
133 incoming URM students who intend to major in biological sciences are eligible for the BSP,
134 which serves students primarily from economic, gender, ethnic, or historically
135 underrepresented cultural groups and first-generation college students. In the summer prior
136 to matriculation, all incoming freshmen biological sciences majors are notified about the BSP
137 and must apply by the end of August. The program strongly encourages applicants from
138 Cornell's pre-freshman Summer Program. The online application consists of questions and
139 essays that help the BSP selection committee choose students who will be a good fit with

140 BSP. Students begin the program after university matriculation in mid-September. The BSP
141 selection committee is comprised of staff from the Office of Undergraduate Biology and
142 from each of the two Cornell colleges that support biological sciences majors. The
143 committee looks not only for applicants who may need academic support, but those who are
144 able to demonstrate a commitment to diversity in science are also eligible for the program
145 (personal communication). Approximately 35 freshmen are accepted into the BSP each year
146 and they remain members as long as they meet the program's expectations and continue on a
147 science-related career path.

148 Activities that characterize the BSP take place in the first four semesters, and
149 participation is voluntary. These include: 1) **academic monitoring and support** through
150 participation in study groups. Biology Scholars are required to attend a weekly, two-hour
151 study group for biology, chemistry, physics, or math courses through their sophomore year.
152 Final grades in science and math courses are monitored by program coordinators and
153 support is provided to any struggling students. Program coordinators are notified if
154 struggling students perform poorly on an exam. Following the notification, they contact
155 students to check on them, make sure they know what resources are available to them (e.g.
156 the Learning Strategies Center, tutoring through the College of Arts and Life Sciences), and
157 develop a plan to improve study habits. The BSP study group leaders are also notified so
158 they can focus their efforts on struggling students; 2) **leadership development**. Leaders of
159 the study groups described above are BSP members who received high grades in the courses
160 they tutor. They are required to attend weekly training sessions where they discuss that
161 week's study group experience and mentor more junior study group leaders, or are provided
162 guidance by more senior study group leaders. BSP members may also serve on the BSP
163 Executive Board. Approximately half of the students continue to serve in some leadership
164 capacity within the BSP after the first two years of support, but this varies dramatically
165 depending upon the cohort.; 3) **interaction with faculty**. Biology Scholars are required to
166 participate in two one-credit seminar courses in their first and second years, in which they
167 meet and work with faculty to learn how to interpret and articulate scientific literature.
168 During this time, students may tour labs and are encouraged to pursue undergraduate
169 research. Because we do not know the number of students who engaged in an authentic
170 research experience as part of the BSP we do not know the impact of this experience on
171 GPA or probability of graduating; 4) **career and professional development**. The seminars

172 also provide information about medical and graduate school and advice for pursuing medical
173 degrees. For example, the BSP offers trips to visit graduate and medical schools as well as
174 financial support to attend off-campus science-related conferences; 5) **sense of community**.
175 Students are required to participate in community service and social events each semester,
176 and have access to a study space housed within the Office of Undergraduate Biology.
177 Required activities for BSP participants only go through the first 4 semesters, and
178 continuation in the BSP is voluntary.

179

180 **Statistical analyses**

181 *Effect of URM Status and BSP Participation on Academic Performance*

182 We used generalized linear models to quantify differences among the four previously
183 described student groups with respect to three metrics: combined math and verbal SAT
184 score, cumulative GPA upon graduation or at the time of data collection, graduation rates.
185 We constructed a generalized linear model with a Gaussian distribution to quantify main and
186 interactive effects of URM status and participation in the BSP program on combined SAT
187 scores and cumulative GPA. We calculated a marginal R^2 value for our generalized linear
188 models as an indicator of model fit and variance explained (Nakagawa and Schielzeth 2012)
189 using functions in the MuMIn package (Bartoń 2009). We urge readers to exercise caution in
190 comparing R^2 values for generalized linear models across studies, however, because
191 substantial variation in methods for generating these summary statistics and their underlying
192 assumptions preclude widespread generalizations (Nakagawa and Schielzeth 2012; Johnson
193 2014). We also computed the least-square means among all four groups to determine the
194 statistical significance of each pairwise comparison. To examine variation in graduation rates
195 among groups, we determined whether each student graduated from the university, and
196 modeled this binary response variable with a logistic regression using the same predictor
197 variables: URM status, BSP participation, the interaction effect between these two factors,
198 and combined SAT scores. We also calculated a marginal R^2 value for this logistic regression
199 of graduation rates (Nakagawa and Shielzeth 2012).

200

201 **Results**

202 *Effect of URM Status and BSP Participation on Academic Preparedness and Performance*

203 Comparing incoming SAT scores of URM BSP and URM non-BSP students by
 204 computing the least-squares means revealed a significant difference between each student
 205 group (Figure 1A; Table 1). Non-URM non-BSP students had the highest combined SAT
 206 score (mean = 2167; SE = 3.32), followed by non-URM BSP students (mean = 2033; SE =
 207 22.04), URM non-BSP students (mean = 1967; SE = 5.94), and URM BSP students (mean =
 208 1885; SE = 11.25). Computing the least-squares means for each of the four groups revealed
 209 statistically significant differences in mean cumulative GPA (Figure 1B; Table 1). We found
 210 that the cumulative GPA of non-URM non-BSP students (mean = 3.49; SE = 0.01) and
 211 non-URM BSP students (mean = 3.44; SE = 0.07) were significantly higher than both URM
 212 non-BSP students (mean = 3.04, SE = 0.02) and URM BSP students (mean = 3.10, SE =
 213 0.03). There was no statistically significant difference in least-squares means between non-
 214 URM non-BSP students and non-URM BSP students nor between URM non-BSP students
 215 and URM BSP students.

216 Within our generalized linear models, we found a significant effect of URM status
 217 ($\beta_{\text{URM}} = -0.45$, $t = -21.99$; $P = 8.57 \times 10^{-100}$) on cumulative GPA. BSP participation ($\beta_{\text{BSP}} = -$
 218 0.05 , $t = -0.75$; $P = 0.45$) and the interaction effect between BSP participation and URM
 219 status were not significant ($\beta_{\text{URM} \times \text{BSP}} = 0.11$; $t = 1.42$; $P = 0.16$). The marginal R^2 value for
 220 this model was 0.15. We found a significant effect of URM status ($\beta_{\text{URM}} = -0.87$; $\chi = -3.67$; P
 221 $= 2 \times 10^4$) on the probability of graduation with a degree. BSP participation ($\beta_{\text{BSP}} = 13.65$; $\chi =$
 222 0.025 ; $P = 0.980$) and the interaction effect between these two predictor variables ($\beta_{\text{URM} \times \text{BSP}}$
 223 $= -13.65$; $\chi = -0.024$; $P = 0.981$) were not significant. The marginal R^2 value for this model
 224 was 0.49. These results suggest that there is a decrease in the probability of graduation for
 225 URM students, but that there is no statistically significant difference among URM students
 226 that participate in the BSP program and those that do not.

227

228 *Effect of Incoming Preparedness on Academic Performance*

229 When we included SAT scores as an index of incoming preparedness in our
 230 generalized linear models, we found a significant effect of SAT score ($\beta_{\text{SAT}} = 1.02 \times 10^{-3}$; $t =$
 231 17.658 ; $P = 3.37 \times 10^{-66}$) and URM status ($\beta_{\text{URM}} = -0.24 \times 10^{-3}$; $t = -10.20$; $P = 5.53 \times 10^{-24}$) on
 232 cumulative GPA (Figure 2). BSP participation ($\beta_{\text{BSP}} = -0.24 \times 10^{-3}$; $t = 1.42$; $P = 0.156$) and
 233 the interaction effect between URM status and BSP participation ($\beta_{\text{BSP} \times \text{URM}} = -0.03$; $t =$

234 0.426, $P = 0.670$) were not significant predictors of cumulative GPA when SAT scores were
235 included in the model. The marginal R^2 value for this model was 0.23.

236 When we included SAT scores in our logistic regression of graduation probability,
237 we found that SAT score was the sole statistically significant predictor of graduation
238 probability ($\beta_{SAT} = 2.30 \times 10^{-3}$; $\chi = 3.11$; $P = 0.002$). The remaining predictor variables,
239 including URM status ($\beta_{URM} = -0.484$; $\chi = -1.64$; $P = 0.10$), BSP participation ($\beta_{BSP} = 0.14$; χ
240 $= 13.93$; $P = 0.981$), and the interaction effect between these two predictors ($\beta_{BSP \times URM} = -$
241 13.02 ; $\chi = -0.02$; $P = 0.982$) were not statistically significant. The marginal R^2 value for this
242 model was 0.50.

243

244 Discussion

245 Recent calls to action urge educators and institutions to increase the retention and
246 performance of all students in STEM fields (eg. Brewer and Smith 2011). Our longitudinal
247 study adds to a growing body of literature that highlights the need for national efforts to
248 quantitatively assess diversity support programs and institute effective practices. After URM
249 students participated in the BSP program at Cornell, we found that the statistically significant
250 gaps in academic preparedness among URM students closed in terms of actual academic
251 achievement. However, BSP participation does not improve the GPA of URM students
252 beyond non-participants. Future research should identify which strategies among diversity
253 support programs contribute most to URM student success.

254 We acknowledge one limitation to this study could be the self-selection of high-
255 performing students to the BSP, since more motivated students may be more likely to apply
256 to such a program. However, participation in the BSP program did not affect graduation
257 rates. Furthermore, when we included SAT scores as a measure of incoming preparedness in
258 our model, we found SAT scores and URM status strongly predict GPA upon graduation.
259 SAT score was also the sole positive predictor of student graduation rates.

260

261 *Changing strategies*

262 The persistent performance gap between URM and non-URM students highlights
263 the importance of implementing specific strategies that promote URM demographics.
264 Measuring a range of programs in thoughtful and deliberate ways will allow us to identify the
265 most effective approaches. Institutional support programs that have quantitatively assessed

266 student performance offer points of comparison, but also differ widely in their approaches
267 to supporting students. To further support URM students, Cornell University's BSP plans to
268 experimentally implement a number of new evidence-based strategies. In addition to an
269 annual quantitative assessment using the data presented here as a baseline comparison, the
270 BSP will implement multiple approaches that are describe below. These actions were chosen
271 based on their success in other programs that improved the academic performance or other
272 relevant metrics for URM participants.

273 One experimental strategy will be to increase student engagement with research
274 opportunities for undergraduates. While the BSP currently encourages students to conduct
275 research, students may be more willing to pursue these opportunities if they are financially
276 supported to do so or are given directed research credits. Research experiences place
277 students in the middle of ongoing research in active laboratories on campus (Hernandez et
278 al. 2013, Maton et al. 2012, Matsui et al. 2003, Olson and Riordan 2012, Villarejo et al. 2008).
279 Through research opportunities, students are exposed to the process of discovery through
280 an authentic project, and engage with professors and graduate students. This strategy may
281 lead to publications, presentation opportunities, and other activities that serve as important
282 steps in building a CV and academic confidence for students. This also places students in
283 close proximity to faculty, who serve as important role models and collaborators. Hernandez
284 (2013) showed in a longitudinal analysis of interventions across 38 institutions that the single
285 most effective strategy that significantly contributes to positive academic motivation of
286 minority students was engagement in undergraduate research.

287 Another common strategy employed by successful programs, and one that Cornell's
288 BSP will implement starting fall 2016, is student guidance through mentorship by graduate
289 students and faculty. Mentors can be people with whom students develop supportive
290 relationships and from whom they receive professional advice throughout their
291 undergraduate career. Positive role-models and regular contact with faculty are considered
292 key experiences in higher education associated with student retention and development
293 (Epstein et al. 2015, Wilson et al. 2012), including degree aspirations (Kim and Sax 2009),
294 potential for degree completion (Newman 2011), and academic performance (Kim and Sax
295 2009).

296 Finally, the BSP will incorporate established learning theory into practice. For
297 example, the BSP plans to extend programming to include juniors and seniors, with a focus

298 on continued use of collaborative learning with peer groups (Toven-Lindsey et al. 2015). In
299 this scenario, we expect increased motivation and persistence through elements of social
300 constructivism, in which learning happens through social interactions with others (Au 1998).
301 Beyond quantifying persistence and performance, an appropriate assessment tool for the
302 BSP to quantify the effects of extended programming on motivation is the Motivated
303 Strategies for Learning Questionnaire (Pintrich et al. 1993). Other established learning
304 theories that the BSP will employ is the growth mindset and lay theory approaches to
305 learning (Yeager David Scott and Dweck 2012, Yeager David S et al. 2016). In practice,
306 teaching students about growth mindset is to stress that intelligence and performance are
307 malleable; lay theory stresses the high prevalence of emotional challenges experienced by
308 other students as they entered college. These types of interventions reduce the susceptibility
309 of stereotype threat and narrow institutional achievement gaps (Levy et al. 1998, Yeager
310 David S et al. 2016).

311

312 *Challenges and opportunities for STEM*

313 The critical importance of effective diversity programs for minority students has
314 strong implications for the achievement of equity in STEM disciplines. In order to reveal
315 positive outcomes and efficient use of resources, more quantitative research is required. One
316 difficulty for many universities is selecting how to distribute funds for URM-support
317 programs. While most large universities have URM support programs, few studies have
318 explored the optimal allocation of limited resources to best serve students: does a university
319 invest a finite amount of resources across a large pool of students or into few individuals? If
320 they choose the latter option, is it better to invest in low-achieving students who most need
321 the intervention, or top-achieving students who are most likely to succeed? There may be a
322 critical financial threshold below which the amount of funding will not benefit students, or
323 above which programs should consider widening their pool of recipients.

324 Another area that would benefit from further study is the exploration of nuanced
325 quantitative metrics beyond GPA and retention rate that capture the positive effects of
326 URM-support programs. Such metrics may include measures of intellectual breadth,
327 extracurricular depth, self-efficacy and motivation, academic or extracurricular
328 accomplishments, and life-long impacts. The lack of studies on these other metrics means
329 that we cannot test how Cornell's BSP affects different facets of student success; yet the BSP

330 may affect URM students in ways we have not quantified. Rigorous research on alternative
331 metrics of performance is required if our field aims to evaluate the generality of different
332 program impacts.

333 This assessment presents the BSP with the unique opportunity to apply and monitor
334 evidence-based methodologies to close the majority-minority gap. In spite of inherent
335 challenges, the promotion of diversity in STEM fields will be made possible through
336 continued collaborative assessment and systemic change.

337

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343

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345 the data: NAM. Wrote the paper: CJB and NAM. Edited and improved the manuscript: CJB
346 and NAM.

347

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	URM		Non-URM	
	BSP	Non-BSP	BSP	Non-BSP
SAT Score (SE)	1885.33 (11.25)	1967.01 (5.94)	2032.95 (22.04)	2166.85 (3.32)
GPA (SE)	3.10 (0.03)	3.04 (0.01)	3.44 (0.07)	3.49 (0.01)
n	181	706	51	2221

349

350 Table 1: Results from generating least-squares means to compare incoming SAT scores and
 351 cumulative GPA of students who differ based on their racial minority status (URM or non-
 352 URM) and participation in Cornell's Biology Scholars Program (BSP or non-BSP). Standard
 353 errors are shown in parentheses.

354

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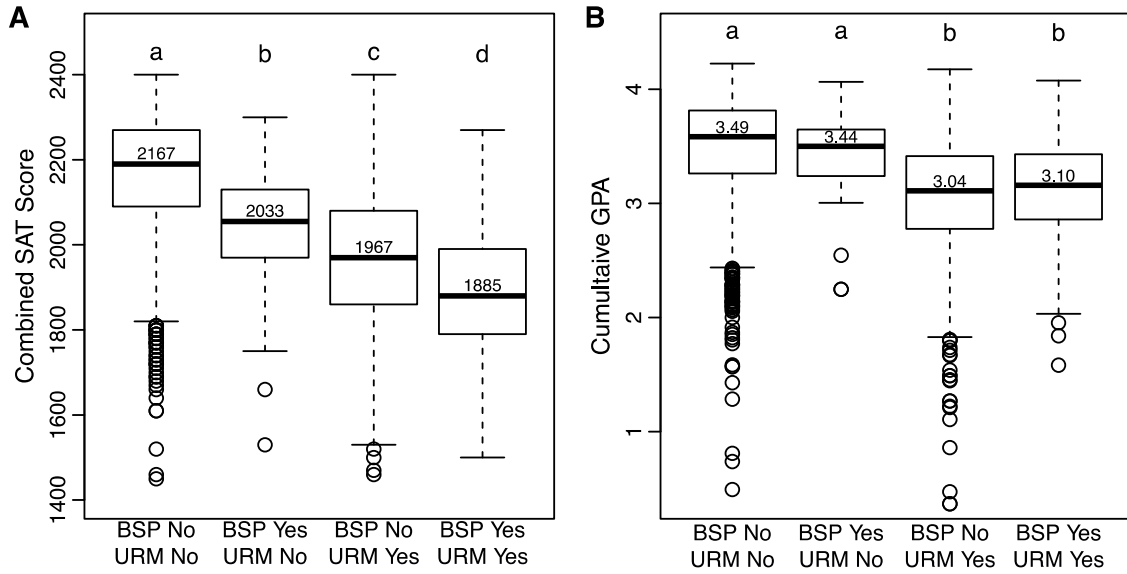
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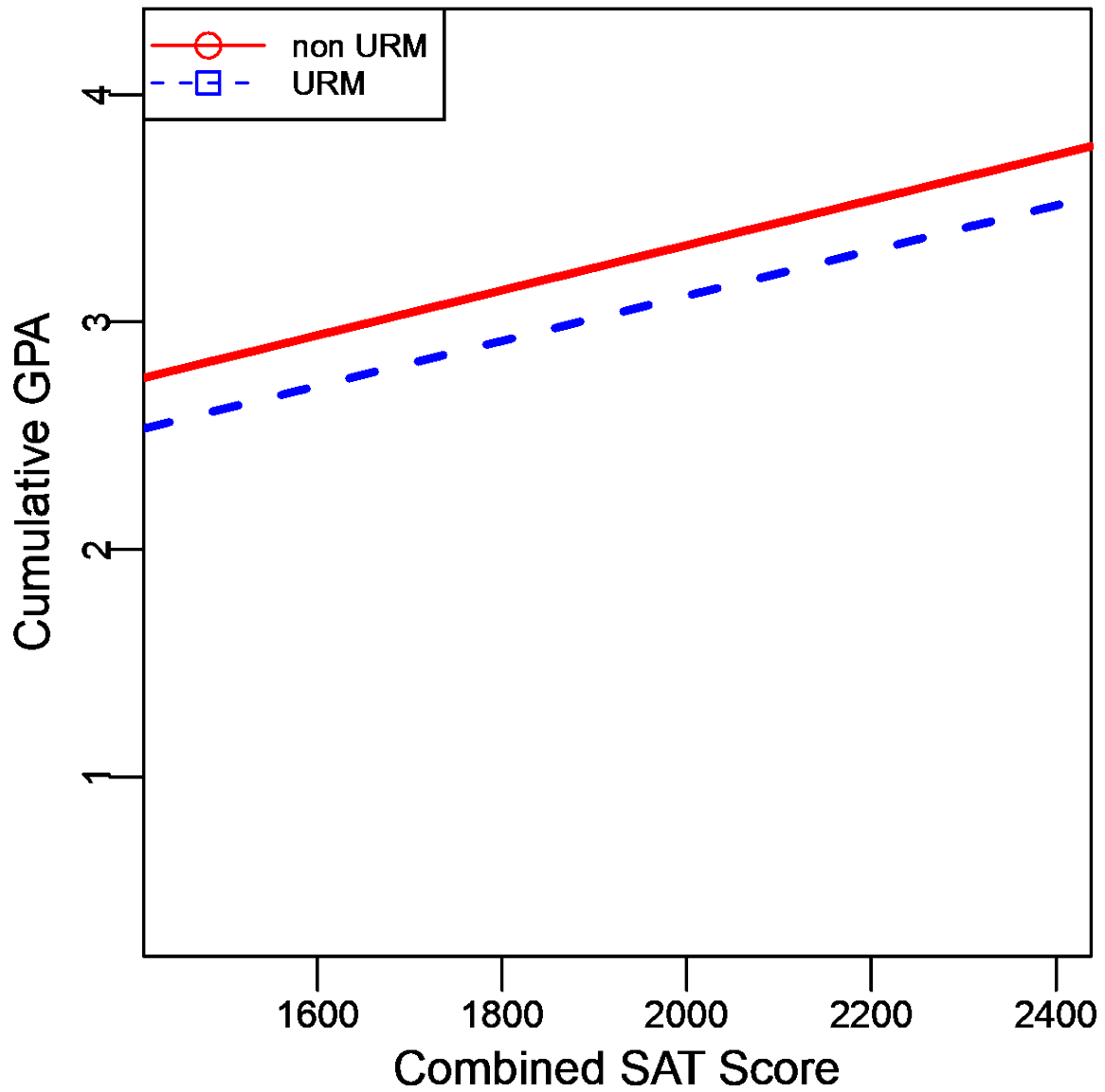
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438 **Figure Legends**



439

440 Figure 1: Differences in preparedness as estimated with (A) combined SAT scores and (B)
 441 cumulative GPA among students that varied in minority status and participation in the BSP
 442 program. The mean values are shown for each group above the median bar in the bar plot.
 443 Outliers are shown as circles. Significant differences in pairwise comparisons of least squares
 444 means estimates are shown above each boxplot.



445

446 Figure 2: Scatterplot of combined SAT score and cumulative GPA, which shows a positive

447 correlation between these two metrics. The solid line and circles represent non-URM

448 students, while the dotted line and squares represent URM students.