1	Knowledge is em(power)ing: Historically underrepresented groups benefit from exposure to information
2	about advanced degrees in psychology.
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ABSTRACT

21 Despite the growing interest in Science, Technology, Engineering, and Mathematics (STEM) education 22 and careers, the "leaky pipeline" continues to disproportionally exclude underrepresented minorities from 23 STEM fields. One feature of the leaky pipeline is unequal access to knowledge about graduate education 24 and careers in STEM fields. Disparities in representation can be reduced by making this knowledge more 25 accessible, which may be particularly beneficial at the transition from undergraduate to post-graduate 26 research experience and education. Here, we sought to investigate the value of large-scale, open-27 access programming as a mechanism for achieving this objective, by using survey data from the 2020 28 Yale University Research Psychology Bootcamp as a case study. We found that this programming 29 increased confidence and understanding of how to navigate the field of research psychology in all survey 30 respondents, with a targeted benefit for Black and First Generation/Low-Income participants. Based on 31 these findings, we argue that large-scale, open-access programming offers a unique and practical utility 32 as an intervention for eliminating disparities in STEM higher education and STEM fields. 33 34 Keywords: diversity, higher education, educational opportunities, doctoral education, open-access, 35 online programming

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INTRODUCTION

The fields of Science, Technology, Engineering, and Mathematics (STEM) have seen rapid professional growth in recent years. In 2018, the Pew Research Center reported a remarkable 79% increase in STEM employment in the US since 1990, and the average STEM professional now earns about two-thirds more than the average non-STEM professional [1]. Despite the lucrativeness and popularity of these fields, STEM participation by underrepresented minorities (URMs) is rife with inequality across both racial and socioeconomic lines, particularly at the level of post-graduate research and higher education.

46 Although 11% of the US workforce is Black and another 17% is Hispanic, Black and Hispanic 47 individuals only make up 9% and 8% of STEM professionals, respectively [2]. Black, Native American, 48 and Latinx students acquire STEM-related degrees at slower and lower rates when compared to their 49 White and Asian American counterparts, despite matched interest in majoring in STEM disciplines [3]. 50 This disconnect between interest and subsequent success in STEM participation is mirrored in first 51 generation and low income (FG/LI) individuals, where first-generation students enroll in [4] and persist 52 with [5] STEM-field related degrees at lower rates than their non-first generation peers. These extant 53 disparities stand at odds with the demonstrated interest in pursuing STEM-related education and careers, 54 with one longitudinal study showing that, despite matched aspiration to continue education and acquire 55 an advanced degree, FG/LI students were less likely than their peers to subsequently earn graduate 56 degrees [6]. These findings illustrate the "leaky pipeline" problem, in which structural and systemic 57 barriers disproportionately filter out URM students at every stage of STEM education and professional 58 advancement [7].

Aside from the formal requirements that gate-keep STEM pathways (e.g., high GPA and high standardized test scores, evident research productivity, and strong letters of recommendation), students may benefit from direct access to insider knowledge of the rules of academia that dictate less obvious expectations and knowledge of cultural norms [8]. Underrepresented minority (URM) students may especially benefit, as they are less likely to have access to support systems that can provide full information about academia [9]. Access to this "insider perspective" has tangible value – in one national

65 survey of Black professors in STEM fields, respondents indicated that advising and mentorship were

66 critical to their academic and professional success [10].

67 More broadly, the *pathways intervention model* of increasing URM participation in higher 68 education has yielded promising results, by aiming to elevate students' perceptions of the tangibility of 69 their desired futures and aid them in progressing towards these aspirational selves [11]. For example, 70 incoming college students that received programming that aligned with and highlighted their diverse 71 backgrounds in the context of academic success had higher end of year GPAs than those that received 72 programming that didn't highlight these differences [12], and were more equipped to handle the stresses 73 of college over time [13]. Further highlighting the importance of identity-aligned programming, a meta-74 analysis of 24 field studies on the presence of URM "ingroup role models" showed a significantly positive 75 impact on underrepresented students aged 10-21 years pursuing STEM education [14]. 76 Critically, much of this work so far has been conducted at the high school and college level. 77 Consequently, much less is known of the potential value of interventions at the post-graduate level. 78 Nonetheless, several institutions have taken approaches to provide this programming via application-79 based diversity events, such as the University of Virginia's "Diversifying Psychology Visit Day" 80 (psychology.as.virginia.edu/diversifying-psychology-visit-day) and Brown University's "Diversity Preview 81 Day" (https://www.brown.edu/academics/gradschool/news/2019-08/diversity-preview-day-phd-students). 82 Others have implemented longitudinal STEM mentorship programs, such as the University of Maryland -83 Baltimore County Meyerhoff Scholars Program [15] and Central State University's Benjamin Banneker 84 Scholars Program (BBSP) [16]. While initiatives such as these aim to fill a demonstrated need, they are 85 limited in two major ways. First, they are limited in scope. Participants apply for a limited number of slots 86 or are assigned a mentor from a limited number of available academics and professionals. While this 87 does not take away from the importance of these types of programs, given the demonstrated interest from 88 URMs in pursuing STEM careers [3, 6], the amount of need for these programs may surpass the extent of 89 resources available. Second, data on the success of these programs are largely absent. Some initiatives, 90 like BBSP and the Meyerhoff Scholars program, have published data on the positive impact of their 91 programming [15, 16], but more data on the variety of programs currently in place are needed to illustrate 92 the fuller picture. Future efforts to diversify the STEM pipeline would greatly benefit from evidence-based

93 recommendations about what offerings make the most difference for participants, and particularly those

94 from underrepresented backgrounds.

95 Here, we evaluate the benefits and efficacy of large scale, open-access programming on the topic 96 of higher education and careers in STEM, using data from a research psychology "Bootcamp" event as a 97 case study. Participants of this virtual programming completed a pre-bootcamp and post-bootcamp 98 survey to explore changes in their knowledge of, and confidence in, pursuing careers in psychological 99 research. Using an intervention model of analysis, we assessed the impact of the bootcamp on these 100 response metrics. 101 102 **METHODS** 103 **Participants** 104 A total of 1,378 students registered to either attend the bootcamp or receive the bootcamp 105 materials. Of those registered, 659 started the pre-bootcamp survey, 544 of whom completed it. 318 106 participants started the post-Bootcamp survey, 224 of whom completed it. Following a data filtering 107 procedure (see Data Filtering and Handling), the subsequent dataset consisted of pre- and post-108 bootcamp responses from 148 participants. Participant age was not requested in the original survey, but 109 participant academic level revealed that most of the participants were college seniors (n=53), followed by 110 post-baccalaureates (n=52), college juniors (n=33), college sophomores (n=8), and pre-college students 111 (n=2). Full demographic breakdown is provided in Table 1. Exemption determination was made by the 112 Human Research Protection Program Institutional Review Board at Yale University. 113 114 115

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		Pre-College Student		College Sophomore			College Junior			College Senior			Post-Baccalaureate			Grand	
		FG/LI	Total	FG/LI	Not FG/LI	Total	FG/LI	Not FG/LI	Total	FG/LI	Not FG/LI	Total	FG/LI	Not FG/LI	Total	Total	
Asian	Female				1	1	3	1	4	4	2	6	3	13	16	27	
	Gender Non- Binary/Prefer not to Say							1	1	1		1	1	1	2	4	
	Male	1	1					1	1		1	1		3	3	6	
	Total	1	1		1	1	3	3	6	5	3	8	4	17	21	37	
Black/ African	Female			3		3	3	7	10	2 7 9		9		5	5	27	
Amcan	Gender Non- Binary/Prefer not to Say				1	1				1	2	3				4	
	Male	1	1					2	2		1	1	1	2	3	7	
	Total	1	1	3	1	4	3	9	12	3	10	13	1	7	8	38	
White	Female				1	1	1	1	2	3	4	7	3	3	6	16	
	Gender Non- Binary/Prefer not to Say										1	1				1	
	Male						1	1	2				1	1	2	4	
	Total				1	1	2	2	4	3	5	8	4	4	8	21	
Hispanic/ Latinx	Female			1		1	3	2	5	5	6	11	6		6	23	
Launx	Male						2		2	2	1	3	2		2	7	
	Total			1		1	5	2	7	7	7	14	8		8	30	
Multiracial	Female				1	1	1	2	3	4	2	6	3	2	5	15	
	Male						1		1				1		1	2	
	Total				1	1	2	2	4	4	2	6	4	2	6	17	
Did Not Say	Female									2	1	3				3	
	Total									2	1	3				3	
Other	Female									1		1		1	1	2	
	Total									1		1		1	1	2	
Grand Total		2	2	4	4	8	15	18	33	25	28	53	21	31	52	148	

Table 1: Demographic breakdown of respondents that had valid data in both the pre-bootcamp and post-

bootcamp surveys.

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- 126 Procedure 127 **Psychology Bootcamp Event** 128 The Yale Psychology Committee on Diversity and Inclusion (YPCDI) organized a virtual 129 Psychology "Bootcamp" in the summer of 2020, in collaboration with the Yale Office of Graduate Student 130 Development and Diversity (OGSDD). Delivered via Zoom Video Webinar 131 (https://explore.zoom.us/en/products/webinar/), this three-hour event was free of cost and open to all. 132 Advertising was targeted towards underrepresented minority students and post-baccalaureates interested 133 in pursuing careers in research psychology. The bootcamp was divided into three, 1-hour sections, each 134 with a distinct theme. The first hour was dedicated to "Careers in Research Psychology", the second 135 section to "Getting Research Experience", and the third to "How to Apply to Graduate School". The video 136 recordings from each of these sections are freely available online 137 (https://psychology.yale.edu/graduate/admissions/psychology-graduate-school-bootcamp). 138 Importantly, the format of this event included presentation and panel discussion and Question 139 and Answer from URM graduate students, faculty, and industry professionals. Presenters and panelists 140 spoke not only on the insider tips for succeeding in academia, but also on the specific challenges and 141 opportunities they experienced as underrepresented scientists and professionals. 142 143 Survey Data Collection Mechanism 144 The pre-bootcamp and post-bootcamp feedback surveys were administered via the Qualtrics 145 survey platform (qualtrics.com). The link to the pre-bootcamp survey was sent via email to registered 146 participants leading up to the day of the event. In order to encourage participation, participants were 147 informed that completion of both the pre-bootcamp and post-bootcamp surveys would enter them in a 148 raffle to win one of ten \$50 Amazon gift cards. No information regarding the purpose of this study was 149 known to the participants. 150 The link to the post-survey was displayed on the last slide of each section's presentation and 151 shared over Zoom. This was done in addition to sharing the link after the event via email to registered
- 152 participants to ensure that people who only attended a subset of the sections would still receive it

153 immediately after viewing the content. In order to link participants' pre- and post-bootcamp surveys, each 154 participant was given a randomly generated code in the pre-bootcamp survey that they were required to 155 enter in the post- survey. Questions consisted of Likert scale and fill-in-the-blank questions, as well as 156 inputs for demographic information. Likert scales used for participants' responses ranged from "1 -157 Strongly disagree" to "5 - Strongly agree." (see Supplementary Table S1 for a complete list of Likert scale 158 survey questions). The analyses below were conducted using the Likert scale questions and 159 demographic information. 160 161 **Data Analysis** 162 Data Filtering and Handling 163 We wanted to limit our analyses to individuals who had completed the pre-bootcamp survey 164 before viewing the bootcamp, and the post-bootcamp survey after viewing the bootcamp. Because there 165 was no way to ensure that people who watched the bootcamp videos asynchronously completed the 166 surveys in the correct order, we limited inclusion to pre-bootcamp surveys completed before the start of 167 the live bootcamp, i.e. before the post-survey was available. We excluded repeat entries, pre-bootcamp 168 surveys with no accompanying post-surveys, and vice versa. Prior to analysis, we removed a subset of 169 Likert Scale survey questions the content of which was not a focus of the bootcamp event. A full list of 170 included and excluded questions is provided in Supplementary Table S1. 171 First Generation participants were categorized by self-endorsement of first generation status as 172 defined by the US Department of Education ("if neither parent completed a baccalaureate degree; 173 https://nces.ed.gov/pubs2018/2018009.pdf"). Low Income participants were categorized by self-174 endorsement of low-income status, as defined by the US Department of Health and Human Services 175 2020 Poverty Guidelines (https://aspe.hhs.gov/2020-poverty-guidelines). 176 To achieve a more balanced dataset prior to our analyses, we first re-coded some demographic 177 responses. First, we coded the individuals who selected multiple of the listed racial identities as 178 "Multiracial." We then coded all individuals who responded to the question of racial identity with either 179 "Prefer not to Answer" or by only inputting a custom answer as "Other." Because only one respondent 180 identified as Pacific Islander, we included them in the "Other" category. Next, we re-coded specific custom

responses that better fit one of the listed racial identities than the "Other" category. The three respondents who identified as "South Asian/Indian" and one who identified as "Uyghur" were re-coded as Asian. The two respondents who identified as "Jewish" and as "Armenian descent" were re-coded as White. Lastly, the two respondents who identified as "Biracial South Asian and Caucasian" and "Middle Eastern" were re-coded as Multiracial. We then combined the 46 respondents that identified as Low Income with the 51 who identified as First Generation to create one FG/LI category that encompassed both identities and their overlap.

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189 Overall Assessment

190 We first sought to assess the overall impact of the bootcamp event. We subset the data to only 191 include responses to questions from participants who had attended the relevant Section (i.e., excluding 192 responses to a question pertaining to Section 1 material from participants who did not attend Section 1). 193 We then calculated a Response Difference measure for each participant and each question by 194 subtracting their post-bootcamp Likert scale responses from their pre-bootcamp responses. Thus, positive 195 values indicate increased agreement with the statement being rated. We then conducted one-sample 196 Wilcoxon signed rank tests to determine whether this Response Difference was significantly above or 197 below zero for each session.

198

199 Participant Demographic Baseline

200 We then sought to identify differences in pre-bootcamp survey responses across participant racial 201 and socioeconomic groups. The objective of this analysis was to determine the extent to which baseline 202 knowledge of different aspects of research psychology was unevenly distributed across demographics. 203 For bootcamp sections 1 and 3, we analyzed the subset of responses from participants who indicated that 204 they had attended that section. To investigate differences across racial identity, we conducted a linear 205 mixed effects regression predicting response on a Likert scale, with survey question and racial identity as 206 predictors, plus an interaction term, as well as random intercepts for each participant. As there was only 207 one survey question related to Section 2, we performed the same data subsetting procedure and used a

- 208 linear instead of a linear mixed effects model, with only racial identity as a predictor. We conducted an
- 209 analogous set of analyses for differences across FG/LI status.
- 210 Effect of Event
- 211 For the subset of survey questions for which there were group differences in baseline response, 212 we then sought to evaluate the effect of the bootcamp in modulating those differences. Taking a data-213 driven approach, we conducted the next set of analyses only on those questions and demographics for 214 which the baseline differences were significant. The objective of this analysis was to determine whether, 215 beyond any overall effects, the bootcamp event additionally served to close knowledge gaps between 216 groups, as evidenced by differences in baseline. 217 To that end, we performed linear mixed effects regressions predicting responses to the questions 218 showing baseline variability. Racial identity, survey timepoint (pre- or post-bootcamp), and their 219 interaction served as predictors, with random intercepts per participant. We again conducted an 220 analogous set of analyses for differences across FG/LI status. 221 222 RESULTS 223 224 **Overall Positive Effect of Bootcamp** 225 One-sample Wilcoxon signed-rank tests with Bonferroni correction revealed that the positive 226 change in response was significant for all three sessions (all p's <0.001) (Fig 1). 227 228 Fig 1. Participants show positive effect of the bootcamp across all three sessions, as evidenced by 229 greater post- compared to pre-bootcamp responses. 230 231 Lower Baseline for Black and Low Income/First Generation Participants 232 We first investigated differences in baseline response as a function of racial identity. For Section 233 1 Pre-bootcamp survey responses across race, linear mixed effects modeling revealed a marginal 234 interaction between Question and Race (F(30,780)=1.144, p=0.063). This model was a better fit to the 235 data than a main effects model ($\chi^2(30)$ =43.85, p=0.049). Given this marginal effect and our a priori

236 predictions, we conducted follow-up pairwise Wilcoxon Signed-Rank tests of which survey responses

237 varied between which racial identities. However, none of these analyses survived correction for multiple

comparisons (p's >0.35).

239 Section 2 linear modeling revealed no main effect of Race (F(5,125)=1.25, p=0.291).

240 Section 3 linear mixed effects modeling revealed a marginal main effect of Race (F(5,127)=2.07,

p=0.074) and no interaction F(20,508)=1.00, p=0.464). However, this model did not produce a better fit to

the data than a main effects model ($\chi^2(20)=20.48$, p=0.428). The main effects model revealed an effect of

243 Question (F(4,528)=70.24,p<0.001), as well as a marginal effect of Race (F(5,127)=2.07,p=0.074).

244 Pairwise Wilcoxon Signed-Rank tests with Bonferroni correction revealed that, overall, participants

endorsed the statements "I know what graduate school interviews look like and how to prepare for them"

and "I'm aware of the unwritten rules and informal expectations of the graduate school application

process" to a lesser extent than other Section 3 statements (p's <0.001), with no difference in

endorsement between the two statements (p's>0.79). The remaining three statements also did not differ

in endorsement (p's>0.79).

Given the marginal effect of Race and our a priori predictions, we additionally conducted followup pairwise Wilcoxon Signed-Rank tests with Bonferroni correction to investigate overall differences in response across race. We found that Black/African participants responded lower overall in comparison to Asian (p=0.004) and White (p=0.011) participants (Fig 2).

254

Fig 2. Black participants, compared to White and Asian participants, report less understanding of the graduate school application process.

257

258 We then investigated baseline differences as a function of FG/LI status. Linear mixed effects 259 modeling of responses in Section 1 revealed an interaction between Survey Question and FG/LI status 260 (F(6,822)=2.33, p=0.031). This model fit better than a main effects model ($\chi^2(6)$ =14.07, p=0.029).

261 Because the intercept for this model was a meaningful baseline (i.e., not Low Income/First Gen), unlike 262 that of the linear model predicting effects of Race, we then investigated the beta weights of the specific

263 contrasts for which there was an effect. FG/LI participants specifically endorsed the statements "I know

what a career in research psychology at a university looks like" (ß=-0.56, t=-2.70, p=0.007) and "I

understand what it takes to earn a PhD in psychology" (ß=-0.66, t=-3.20, p=0.001) to a lesser extent than

those who did not identify as FG/LI. We conducted follow-up Wilcoxon Signed-Rank tests with Bonferroni

267 correction on these specific contrasts. Only the difference in responses to the statement "I understand

what it takes to earn a PhD in psychology" showed an effect via this test (p=0.022) (Fig 3).

269

270 Fig 3. FG/LI participants report less understanding of what it takes to earn a PhD in psychology,

271 compared to non-FG/LI participants.

272

273 <u>Bootcamp Reduces Disparities Revealed in Baseline Responses</u>

274 Given these established differences in baseline response across race and FG/LI status, we

sought to determine whether attendance to the bootcamp had the effect of reducing these disparities. For

the effect of race on Section 3 responses, we focused our analysis on Black/African, Asian, and White

277 participants given the demonstrated differences in baseline response. Linear mixed effects modeling of

278 response to Section 3 questions as a function of Time Point (pre- and post- bootcamp) and Race

revealed an effect of Time Point (F(1,2222)=608.87,p<0.001), and an interaction between Time Point and

280 Race (F(2,2222)=4.89,p=0.008). This model fit better than a main effects model (χ^2 (2)=9.76, p=0.008).

281 Follow-up Wilcoxon Signed-Rank tests with Bonferroni correction revealed an overall increase in

endorsement of Section 3 statements (p<0.001). Further, tests revealed that, while Black/African

participants initially endorsed statements less than White (p=0.007) and Asian participants (p=0.004),

these differences were not evident in the post-bootcamp responses (p's>0.22) (Fig. 4).

285

Fig 4. Participation in the bootcamp eliminates differences between Black and Asian/White participants inunderstanding of the graduate school application process.

288

Based on our pre-bootcamp survey results, we then ran a linear mixed effects model on

290 responses to the Section 1 statement "I understand what it takes to earn a PhD in psychology" as a

291 function of FG/LI status and Time Point. This model revealed an effect of FG/LI status

Running Head: Historically underrepresented groups benefit from exposure to information about advanced degrees in psychology. 292 (F(1,137)=4.05,p=0.046), Time Point (F(1,137)=204.87,p<0.001), and an interaction 293 (F(1,137)=6.12,p=0.015). This model fit better than a main effects model ($\chi^2(1)$ =6.07, p=0.014). Follow-up 294 Wilcoxon Signed-Rank tests revealed an overall increase in endorsement of the statement (p<0.001), as 295 well as lower pre-bootcamp endorsement by FG/LI participants compared to their counterparts (p=0.032). 296 This difference in endorsement was no longer evident post-bootcamp (p>0.99) (Fig. 5). 297 298 Fig 5. Participation in the bootcamp eliminates differences between FG/LI and non-FG/LI participants in 299 understanding of graduate school and career options in research psychology. FG/LI=First generation/Low 300 income. 301 302 DISCUSSION 303 In the current investigation, we examined differences in the understanding of graduate programs and 304 careers in research psychology in a diverse sample of individuals. We demonstrated a disparity between 305 individuals from underrepresented groups and their peers in their baseline knowledge of these topics and found 306 that individuals from underrepresented groups showed a particular benefit from informational programming about 307 higher education in psychology. Specifically, before the bootcamp, FG/LI students, compared to non-FG/LI 308 students, reported less understanding of what earning a doctoral degree in psychology entails. However, 309 attending the bootcamp eliminated these differences between the two groups. In addition, we showed that, Black 310 students, compared to White and Asian students, reported less understanding of how to apply to graduate school 311 in psychology before the program. Crucially, differences in understanding among Black, Asian, and White 312 students equalized after attending the bootcamp. 313 314 Identifying Disparities in Knowledge of Graduate Education 315 Interest in STEM careers are at an all-time high [1]. Although aspirations to pursue careers in STEM are 316 similar across demographic groups [6], individuals from historically underrepresented groups pursue STEM 317 education at a lower rate than their White peers [3]. Here we identified one factor that may contribute to 318 underrepresentation in doctoral programs: unequal access to diverse and informative insider perspectives. The 319 level of knowledge individuals have about career opportunities in STEM is associated with their intentions on

320 pursuing STEM careers in the future [17]. Therefore, reducing disparities in knowledge about how to pursue 321 graduate studies among underrepresented groups is critical in efforts to increase their representation in doctoral 322 programs and STEM fields.

323 Our findings suggest that specific underrepresented groups vary in their overall understanding of 324 graduate programs in psychology and possible career opportunities in psychology with a doctoral degree, aligning 325 with the large body of literature illustrating resource disparities as a function of race and class [see 18,19, 20]. 326 Specifically, we observed differences across racial group membership and FG/LI income status in two distinct 327 domains: knowledge of the graduate school application process and knowledge of what graduate school entails, 328 respectively. Black individuals reported less understanding of the graduate school application process than their 329 Asian and White counterparts. FG/LI individuals reported less knowledge of what is required in PhD psychology 330 programs than their non-FG/LI peers. Identifying disparities in knowledge of doctoral programs is a key first step 331 in developing strategies to address these disparities and increase diverse representation in STEM graduate 332 studies.

333

334

Overall Efficacy of Large Scale, Open-Access, Diversity-Aligned Programming

335 Once disparities in knowledge and opportunities are identified, programs can be implemented to target 336 problem areas. At present, a majority of programs (e.g. Meyerhoff Scholars Program) take an intensive 337 mentorship approach. In such university-based programs, a small cohort of students are accepted to receive 338 regular mentorship, placement in research labs, and professional development opportunities designed to help 339 them thrive in STEM fields and become competitive applicants for STEM graduate school programs. While these 340 programs have an important place in increasing diversity in higher education, the time and financial resources 341 required limit the number of students that can be reached at any given time. Here we highlight that large scale, 342 open-access programming in a virtual setting offering diverse perspectives is effective at increasing knowledge of 343 career opportunities and graduate education options in psychology for large and diverse audiences. Across all 344 three programming sessions, participants as a whole reported more confidence in their knowledge of career 345 opportunities, how to gain research experience, and how to apply to graduate school in research psychology after 346 participating in the bootcamp compared to before. Programs with large, virtual formats have the advantage of 347 reaching individuals across the globe, and in various stages of their early career, from those simply considering

careers in psychology to others who are planning to apply to doctoral programs, while requiring minimal financialor time commitments.

350 Critically, we demonstrate that this programming structure is effective at eliminating disparities in 351 knowledge of the graduate school process in multiple underrepresented groups. The diverse group of presenters 352 and panelists provided identity-specific perspectives on the topics and questions raised by bootcamp participants. 353 Black and FG/LI students showed a disadvantage compared to their majority counterparts in prior 354 graduate school knowledge before attending the bootcamp. Both of these underrepresented groups showed 355 increases in knowledge after attending the bootcamp such that their self-reported levels of knowledge became 356 indistinguishable from non-minoritized counterparts. In other words, the bootcamp programming diminished the 357 disparities in confidence and knowledge about careers and graduate school in psychology between Black and 358 FG/LI individuals and their peers. This finding provides evidence that large-scale, open-source programming such 359 as the bootcamp are effective for general audiences, and work to eliminate inequalities in knowledge of graduate 360 opportunities in disadvantaged populations when diverse perspectives are provided.

361

362 Alignment with Principles of Open-Science

363 The philosophy of open-access science has the ultimate goal of advancing society by making science 364 transparent and available to all [21]. As STEM fields guickly evolve to embrace open-access values [22], it is 365 important to ensure that graduate education outreach programs evolve with the changing times. With over 1000 366 individuals registered to receive the program materials, the Yale Psychology Bootcamp is an example of virtual, 367 open-access programming that allows resources to reach an extremely large audience. Program materials will 368 exist in perpetuity and can be referenced by subsequent generations of aspirational research psychologists. 369 Furthermore, the multi-session nature of the Yale Psychology Bootcamp expands the relevance of the 370 programming to individuals with a wide array of experience and interest in pursuing careers that require doctoral 371 degrees. With three distinct, open-access sessions, participants can watch the material that is most relevant to 372 their level of experience, and refer back to the curated materials as they progress towards applying to graduate 373 school. Programs with the aforementioned features are specifically designed to ensure that the knowledge 374 required to pursue graduate education and careers in STEM are accessible to all.

376

Conclusion

377	There is little doubt that continued efforts are needed to combat the systemic barriers and climate of
378	inequity that contribute to the leaky pipeline in STEM. The Yale Psychology Bootcamp is a case study for the
379	ways in which higher education outreach programming can be designed to reach large and diverse populations
380	and keep them in the pipeline. Our results quantify that not all individuals are granted equal access to the
381	information necessary to understand and pursue a graduate degree and career in research psychology.
382	Furthermore, we demonstrate that large-scale, open-access programming is not only successful in increasing
383	awareness of graduate education and possible careers in STEM across the board, but also has a specific,
384	targeted benefit for individuals from historically disadvantaged populations. The program model described in this
385	study offers a low cost and effective strategy for increasing diversity in STEM graduate education and beyond.
386	

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389	
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393	Availability of data and material: Data and material is available on public repository [xx].
394	Code availability: Code is available on public repository [xx].
395	Authors' contributions: KNG and ERT developed the research question and analysis plan and analyzed
396	the data with input from SWCC. KNG and ERT drafted the manuscript with feedback from SWCC and
397	DAS.
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463	Supporting Information
464	S1 Table. Post-bootcamp survey questions divided into their respective sections. Questions that were
465	not addressed in any section were excluded from analysis and are listed under "Excluded."