1 2	Assessing Science Training Programs: Structured Undergraduate Research Programs Make a Difference
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55 **Abstract:**

56 Training in science, technology, engineering, and mathematics (STEM) is a top priority 57 for driving economic growth and maintaining technological competitiveness. We 58 propose that exposure to a rigorous research program as an undergraduate leads to 59 success in a research STEM career. We compared the scientific outcomes of 88 60 participants from five National Science Foundation Research Experiences for 61 Undergraduates (REU) Site programs with demographically-similar applicants to assess 62 the impact that formal, organized, and funded undergraduate summer research 63 experiences have on participants. Our study demonstrates that REU participants are 64 more likely to pursue a PhD program and generate significantly more valued products, 65 including presentations, publications, and awards, relative to applicants. We believe 66 that key components of the program include funding for personal and professional 67 needs; access to diverse intellectual, analytical, and field resources; and the presence 68 of other undergraduate researchers who support each other and share their goals and 69 interests.

70

71 Introduction:

Scientific, technological, and economic competitiveness is motivating greater interest and investment in science, technology, engineering, and mathematics (STEM) training around the world (Gentile et al. 2017), with an emphasis on addressing the current shortage of STEM PhDs (Brewer et al. 2011, U.S. National Academy of Sciences et al. 2010). With annual spending on STEM training well over US\$14 billion in the United States (U.S. Department of Labor 2007), guiding future investments in STEM training 78 demands a good understanding of effective approaches (U.S. National Science Board 79 2015, Hanauer et al. 2017, Lopatto 2004, Wei and Woodin 2011). For example, 80 undergraduate research experience is often credited with preparing students for 81 success in STEM careers (Graham et al. 2013, Hernandez et al. 2018, Kolb and Kolb 82 2005, U.S. Department of Labor 2007, U.S. National Academy of Sciences et al. 2010). 83 However, quantitative assessments of STEM training are rare (Hanauer et al. 2017, 84 Linn et al. 2015) due to a variety of problems, including the difficulty of tracking long-85 term scientific outcomes in a controlled fashion.

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87 Considering the need to identify effective models for STEM training (Barney 2017), we 88 guantitatively analyzed data from demographically-matched students who participated 89 (hereafter "participants") or applied but did not participate ("applicants") to the same 90 United States National Science Foundation (NSF) Research Experiences for 91 Undergraduates (REU) Site summer program held from 2009 to 2011. These 92 independent and geographically dispersed training programs fully support REU 93 participants to conduct independent research projects. Participants are awarded an 94 NSF-defined "take-home" stipend and travel and housing support. During fiscal years 95 2015-2017 NSF REU Site programs across the entire Foundation spent more than 96 \$185M on more than 500 grants, and trained over 150,000 REU participants (grant data 97 available at https://www.nsf.gov/funding/pgm_summ.jsp?pims_id=5517). In this study, we used a matched pair sampling design (Faresjö and Faresjö 2010) of 88 98 99 participant-applicant pairs of undergraduate students associated from five field-biology

and ecology-based REU Sites supported by NSF to determine the impact of structured

101 research experience on future STEM productivity (measured as the number of scientific 102 presentations, publications, and merit-based academic awards, as well as the highest 103 degree pursued at the time of tracking: see data available in *Supplementary* 104 *Information*). Given our interest to best match REU participants and applicants prior to 105 outcome data collection, we used as much provided demographic information as 106 possible. Prior research experience was not considered when matching REU 107 participants and applicants. Some applicants (32%) participated in other research 108 experiences, including other REU programs (8%). Thus, our results may be viewed as 109 conservative and actually underestimate the impact and value of undergraduate 110 research programs for participants.

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112 Methods:

113 **Participant and applicant information**

114 Participant and applicant data from five field-based NSF REU Site programs held during 115 summer 2009 (three Sites, 23 student pairs), 2010 (three Sites, 22 student pairs), and 116 2011 (five Sites, 43 student pairs) were collected from REU grant principal investigators 117 (see Supplementary Information). A "participant" is defined as a student who was 118 admitted and who successfully completed the program. An "applicant" is defined as a 119 qualified undergraduate student who applied for admission to one of the participating 120 REU Site programs, but was not admitted. Each student (participant and applicant) was 121 tracked between 5 and 7 years post-REU experience. To account for this variation in 122 time, paired (participant and applicant) data were treated as a random factor in our 123 statistical analyses. For each REU Site, a demographically-similar (e.g., gender, race,

ethnicity, age, home institution type (private or public) and size, major, focus area, and grade point average) applicant was paired with an REU participant (see *Supplementary dataset A*).

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128 Measured outcomes

129 Five individual-specific outcomes were considered in this study, including (#1) general 130 field of study (STEM or non-STEM; Figure 1A) and (#2) highest degree pursued 131 (doctorate (PhD), healthcare (e.g., Doctor of Medicine, Doctor of Veterinary Medicine, 132 Doctor of Pharmacy), masters of science (MS), masters of business administration 133 (MBA), bachelor of science (BS), associate of arts (AA), and high school (HS); Figure 134 1B)) and number of (#3) scientific conference presentations and (#4) publications and 135 (#5) academic awards (Figure 2; see Supplementary data). Information for outcomes 136 was collected using a combination of REU Site PI tracking data, social media (i.e., 137 LinkedIn, Facebook), scientific databases (Google Scholar, PubMed), and internet 138 searches. Identities of each student were confirmed by name, undergraduate 139 institution, and year of graduation. Pairs were included in our analyses only if all data 140 were available for both members of the pair. Publications were carefully matched with 141 REU participants or applicants and counted only if they were (1) published, (2) scientific, 142 and (3) available in searched databases, social media profiles, or PI reports. In most 143 cases, publications were peer-reviewed journal articles. Undergraduate research-based 144 honors theses were also included. Non-scientific publications, such as fashion blog 145 articles, were not counted. Scientific awards (including grants) associated with merit 146 and related to scientific contributions were counted. GPA related awards, such as

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147 Dean's List, were not included. Finally, presentations (including oral and poster 148 formats) were counted if they were scientific in nature but did not include formal 149 presentations directly associated with the REU program (such as a final poster 150 symposium).

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152 Statistical analyses

153 Chi-square and Fisher Exact tests were used to compare the number of REU 154 participants and applicants according to their discipline (STEM or non-STEM, Figure 155 1A) and highest degree pursued when tracked (Figure 1B). A generalized linear mixed 156 model fit by maximum likelihood (Laplace approximation) with Poisson distribution and 157 "Pair" as a random factor was used to compare the scientific outcomes (i.e., 158 presentations, publications, or awards) of de-identified, demographically paired NSF 159 REU participants and applicants (main effect = REU experience). The REU experience 160 effect is interpreted as the multiplicative increase in scientific productivity an REU 161 participant exhibits relative to the demographically-similar applicant. Thus, an "REU 162 effect" where the 95% confidence intervals include 1 indicates that paired students had 163 statistically similar outcomes. An REU effect >1 (lower 95% confidence interval >1) 164 indicates that an REU applicant was more productive than a paired applicant. Chi-165 square and generalized linear mixed model statistics were conducted using the base 166 package of R and the *glmer* function in the *lme4* package of R (Bates et al. 2015), 167 respectively.

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De-identified data associated with REU participant and applicant pairs are available in the *Supplementary information*. Requests for additional information about this study can be made directly to the corresponding author, Alan Wilson, at wilson@auburn.edu.

172

173 **Results:**

174 As a group, applicants to, and participants in, NSF REU Site programs are similarly 175 biased toward selecting STEM field careers (Figure 1A; chi-square P = 0.214). When 176 considering all degrees, no statistical difference was observed between REU 177 participants' and applicants' highest degree types pursued at the time of tracking 178 (Figure 1B, Fisher's Exact Test P = 0.10). Given our interest in determining if REU 179 experiences encourage greater interest and pursuit of advanced STEM degrees, we 180 conducted a chi-square test using only PhD and MS degree data. In this analysis, we 181 found that REU participants pursued significantly more PhD (+48%) and less MS (-45%) 182 degrees than applicants within six years after completing their baccalaureate degree 183 (Figure 1B, chi-square P = 0.018). The matched paired design means that the positive 184 effect of the REU experience on the pursuit of a PhD is not a function of self-selecting 185 populations since REU participants were matched with demographically-similar 186 applicants to the same REU Site. This result alone supports the hypothesis that 187 structured independent undergraduate research experience is an important stepping 188 stone to a STEM terminal degree.

189

Additionally, of the REU participants included in our analyses who provided

demographic information (gender 72%; race 75%), females and under-represented

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192 minorities (including African-Americans, Hispanic Americans, and Native Americans) 193 accounted for 59% and 42%, respectively. These demographics were similar to the 194 matched applicant pool based on available data (female 64% and/or under-represented 195 minorities 44%). However, across the five REU Sites, the most common applicants, 196 including REU participants, were female (68%) and/or Caucasian (84%) based on 197 available data. Thus, further diversification is generated from NSF's expectation that 198 REU participants be selected from a broad range of schools, especially minority serving 199 institutions and institutions with limited research opportunities. Typically, PhD students 200 come from research intensive public universities or private liberal arts colleges 201 (Fiegener and Proudfoot 2013). Thus, our results suggest an important broader impact 202 of REU programs; namely that they serve as a powerful tool for supporting the 203 economic and cultural diversification of PhD-level scientists.

204

205 Potentially as a result of an increase in advanced degrees pursued by REU participants, 206 we found that participation in REU Site programs was also effective at boosting 207 research productivity (Figure 2). For example, REU participants produced 2.14 times 208 and 1.58 times as many scientific presentations and publications, respectively, and, 209 earned 1.37 times as many academic awards than applicants (Figure 2D; generalized 210 linear mixed model all $P \le 0.012$). Considering that these outcomes are central forms of 211 intellectual currency and indicators of future success in STEM (Laurance et al. 2013, 212 Morales et al. 2017), our findings suggest that there are both short-term (products) and 213 long-term (careers) benefits to participating in NSF REU Site programs. We observed a 214 greater range of products for applicants than REU participants (see Supplementary

dataset B), however, variance did not scale with the observed data ranges because
many applicants had 0 products for a specific scientific outcome.

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218 Although we were not interested in evaluating differences across the five REU Sites 219 included in this study, we conducted additional generalized linear mixed model analyses 220 including REU experience, Site, and REU experience x Site interaction for the three 221 primary outcomes (presentations, publications, and awards) for thoroughness (see 222 Supplementary dataset C). In general, our findings from these additional analyses were 223 consistent with our primary analyses (presented in Figure 1B). For example, all three of 224 the analyses showed significant REU experience effects (all $P \le 0.024$). Moreover, we 225 did find a significant interaction (all $P \le 0.0074$) between the REU experience and REU 226 Site program location (any of the five participating programs) for each scientific 227 outcome. However, for all Sites and outcomes except one (publications at Site C; within 228 Site comparisons results not shown), the estimated effect of the REU experience was 229 positive (albeit not always significantly so). Thus, the effect of REU experience was 230 estimated to be positive across almost all REU Sites and outcomes, suggesting that the 231 REU experience, in general, drives the patterns we observed despite variation in 232 program location, management, and implementation.

233

234 **Discussion:**

Our quantitative results show the potential effectiveness of undergraduate research
experiences (Figure 2) are consistent with earlier qualitative (Hernandez et al. 2018,
Linn et al. 2015, Lopatto 2004) and quantitative findings (Hanauer et al. 2017) –

structured independent research training is effective at cultivating future scientists.

However, an important question remains – Why do these experiences work? (Gentile etal. 2017).

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242 In general, NSF REU Sites provide structured and fully funded research experiences for 243 student cohorts (~8-10 students) for several weeks (~8-10 weeks per summer) where 244 REU participants collaboratively work with a senior scientist with an active research 245 program and peer scientists while participating in a variety of professional development 246 enrichment activities, such as learning to read research articles, presenting oral or 247 poster presentations, preparing applications for graduate school, and networking with 248 other scientists. Although all or some of these training characteristics could explain our 249 findings that compared demographically-matched participants and applicants 250 (Abudayyeh 2003, Auchincloss et al. 2014, Fox et al. 2017, Linn et al. 2015, Morales et 251 al. 2017, Rocchi et al. 2016, Shanahan et al. 2015, Taraban and Logue 2012), it is 252 impossible to completely eliminate potential confounding factors without a controlled, 253 replicated experiment. Thus, alternative factors beyond the REU experience itself may 254 explain our findings. For example, the participant selection process could bias towards 255 students who are more likely to be successful in science. While demographic matching 256 does not completely eliminate the possibility that the selection process introduced 257 confounding factors that explained student outcomes, rather than the outcomes being 258 generated by the program, that outcome is highly unlikely. REU PIs independently use 259 a variety of data to choose REU participants, including essays, transcripts, letters of 260 recommendation, fit for program and mentors, future aspirations, interview experience,

261 prior research experience, current institution type, and/or demographics. Since each 262 program has its own selection process, and it is unlikely that the broad range of 263 selection processes used across Sites would generate a consistent effect of the REU 264 program on participants. Furthermore, given that PIs typically lack any formal human 265 resources training, we would not expect that REU PIs would be more effective at picking 266 more successful participants than groups with formally trained human resources 267 staff. In fact, REU PIs are required to review their REU Site each summer and part of 268 this process includes reflecting on the students selected to participate in the program. 269 Despite REU PI selection efforts, participants always range widely in performance (see 270 Supplementary datasets A and B).

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272 Although we cannot completely discount the influence of any confounding factors 273 associated with the participant selection process for an REU Site, NSF REU Sites are 274 not prescribed. Instead, REU PIs have significant flexibility in leveraging available 275 laboratory, analytical, field, and human infrastructure to create the most impactful 276 experiential learning for their REU participants. Each REU PI approaches the selection 277 of their REU participants independently, considering the nature of their REU program, 278 and with NSF's guidance in mind regarding creating opportunities for under-represented 279 students, students with disabilities, and students from limited research opportunities. 280 Considering the latter (and its influence on future funding for an REU Site), many REU 281 Pls tend to recruit students with limited to no prior research experience. Therefore, 282 despite variation across NSF REU Sites in their research, professional development, 283 and networking activities, we found strong effects of NSF REU experiences (Figure 2D).

284 These findings are even more impressive considering that our comparison of paired 285 participants and applicants to the same REU Site did not exclude applicants who 286 conducted other undergraduate research, including participating in other NSF REU 287 Sites or similar programs. Thus, our analyses may actually underestimate the impact of 288 participating in undergraduate research programs, in general. Given the positive 289 impacts of undergraduate research, we argue for continued investment in such 290 programs, including making certain they are available to all demographic groups 291 (Economy et al. 2014, Hernandez et al. 2018, Linn et al. 2015, MacLachlan 2012, U.S. 292 National Academy of Sciences et al. 2010), as a way of maintaining a strong, global 293 STEM workforce.

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377 Figure legends

Figure 1. Interests and pursued degrees of REU participants and applicants. Number of students (black bar = participant; white bar = applicant) included in the REU assessment associated with their (a) current discipline (chi-square P = 0.214) and (b) highest degree pursued at the time of being tracked for this study (all degrees Fisher's Exact Test P = 0.10; PhD and MS only chi-square P = 0.018). Degrees included Doctor

383 of Philosophy (PhD), Master of Science (MS), and Bachelor of Science (BS).

384 Healthcare includes all health related advanced degrees, such as medical doctor.

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386 **Figure 2.** Scientific outcome comparisons of Research Experiences for Undergraduates 387 (REU) participants vs. demographically-similar applicants. Box plots showing paired 388 response differences (n=88 pairs; calculated as REU participant - applicant) for three 389 assessed scientific outcomes, including (a) presentations, (b) publications, and (c) 390 awards. A value of 0 would mean that the paired students have the same outcome 391 (e.g., 0-0, 1-1, 2-2, etc.). Box plots show 10% and 90% of the paired difference data 392 (black whisker caps), 25% to 75% of data (gray box), mean (black line in gray box; the 393 median for each outcome was 0), and outliers (black circles outside whisker caps). (d) 394 REU effects (estimate ± 95% CI) for the three scientific outcomes is the multiplicative 395 increase in scientific productivity by an REU participant relative to a demographically-396 similar applicant (presentations P = 0.000000068, publications P = 0.0002, and awards 397 P = 0.012). P-values are from a generalized linear mixed model fit by maximum

- 398 likelihood (Laplace approximation) with Poisson distribution and "Pair" as a random
- 399 factor.
- 400