The confidence gap predicts the gender pay gap among STEM graduates

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Women make less than men in some science, technology, engineering, and math (STEM) fields. While explanations for this gender pay gap vary, they have tended to focus on differences that arise for women and men after they have worked for a period of time. In this study we argue that the gender pay gap begins when women and men with earned degrees enter the workforce. Further, we contend the gender pay gap may arise due to cultural beliefs about the appropriateness of women and men for STEM professions that shape individuals’ self-beliefs in the form of self-efficacy. Using a three-wave NSF-funded longitudinal survey of 559 engineering and computer science students that graduated from over two dozen institutions in the United States between 2015 and 2017, we find women earn less than men, net of human capital factors like engineering degree and grade point average, and that the influence of gender on starting salaries is associated with self-efficacy. We find no support for a competing hypothesis that the importance placed on pay explains the pay gap; there is no gender difference in the importance placed on starting salaries is associated with self-efficacy. We find no support for a competing hypothesis that the importance placed on pay explains the pay gap; there is no gender difference in the importance placed on pay. We also find no support for the idea that women earn less because they place more importance on workplace culture; women do value workplace culture more, but those who hold such values earn more rather than less. Overall, the results suggest that addressing cultural beliefs as manifested in self-beliefs—that is, the confidence gap—commands attention to reduce the gender pay gap.

gender | pay gaps | cultural beliefs | STEM

A gender pay gap exists in science, technology, engineering, and mathematics (STEM), in part because within those majoring in STEM fields women choose majors that pay less, on average, than the majors men choose (1). Yet, even after accounting for college major, women make less than men with the same earned degrees in some STEM fields (2). In 2013, for instance, women with computing, mathematical, and engineering degrees earned between 82 and 87% of what men earned, or $65,000 on average annually compared to men’s $79,000 (3). The gender pay gap is consequential for a number of reasons. It affects who stays in the STEM workforce since lower compensation leads to less job satisfaction and greater exit from STEM jobs (4–6). Compensation also affects which individuals, after taking a break at critical junctures (e.g., for parental leave), are likely to return to the workforce (7).

One of the most striking aspects of the gender pay gap in STEM is that increasing women’s entry into STEM has been a primary policy aim for addressing the gender pay gap broadly. Specifically, women are directed into STEM-related majors and college degree programs as a part of initiatives designed to help women reach pay parity with men (8, 9). The main reason for this is that recipients of STEM degrees earn more, especially in lucrative, high-demand fields like engineering and computer science (CS), two fields with the largest representation in the STEM-employed workforce and some of the highest growth projections (10). Yet, of all the STEM fields, engineering and CS have some of the lowest levels of occupational entry by women alongside some of the highest gender pay differentials (11). In assessing why pay gaps in these fields exist, scholars have pointed to women’s competing domestic priorities due to family formation (12, 13) and employer discrimination (14), but such explanations have been found to be applicable across STEM fields and across professions more generally (1, 15–17) and are not specific to engineering and CS (5, 18).

In this paper we focus explicitly on engineering and CS and why cultural beliefs about the “fitness” of women and men for these fields may correspond to pay. We undertake an investigation of initial salaries at college exit for undergraduates that have recently earned engineering and CS degrees. A study of initial salaries at workforce entry is appealing not only because human capital differences between men and women at college graduation are largely negligible, but also because later-stage factors that influence careers (e.g., childcare responsibilities) are less apt to be present among individuals at college exit. Cultural beliefs about the appropriateness of women and men for engineering and CS that influence pay, meanwhile, may exist. Sociologists have argued that cultural beliefs about the appropriateness of women and men in engineering and CS professions operate at the societal level in ways that affect women’s and men’s beliefs (19, 20). Cultural archetypes such as “programmers” highlight men as more competent— and ultimately more fit—than women in CS and engineering fields. As a result, microlevel personal beliefs may come to be based on societal

**Significance**

Is there a gender pay gap among graduates in some science, technology, engineering, and math (STEM) fields? Women and men have near-identical human capital at college exit, but cultural beliefs about men as more fit for STEM professions than women may lead to self-beliefs that affect pay. We hypothesized that women and men would be paid differently upon college exit, and that gender gaps in self-beliefs about one’s abilities, or self-efficacy, would correspond to this difference. Using data from a three-wave longitudinal study of graduates of engineering programs from 2015–2017, we find a confidence gap that aligns with a gender pay gap. Overall, these findings point to the role that cultural beliefs play in creating gender disparities among STEM degree-holders.

Author contributions: A.D.S., M.E.T., S.W., A.K., S.G., and S.S. designed research; A.D.S., M.E.T., S.W., and A.K. analyzed data; and A.D.S. and S.G. wrote the paper.

The authors declare no competing interest.

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2This article contains supporting information online at https://www.pnas.org/lookup/suppl/doi:10.1073/pnas.2010269117/-/DCSupplemental.


*In the United States, women graduating with STEM degrees have lower probabilities of first births early on after graduation and relatively higher probabilities in their mid to late 30s compared to women graduating in other fields (46, 47).
expectations that influence women’s and men’s own perceptions of their skills. Prior studies have investigated such personal beliefs, but they have concentrated on how these affect individuals earlier, such as when they are deciding a college major, or whether or not to continue in a STEM program (19, 20). By pivoting toward degree holders at workforce entry and hypothesizing about what occurs among those at college exit, we move a step closer to understanding how beliefs may influence an initial gender pay gap, if one exists.

The early-career stage is a particularly potent time for gendered personal beliefs to come to the fore. We specify personal beliefs here as one’s self-efficacy, a psychological concept defined as a judgment about one’s ability to perform a necessary course of action to achieve a goal (21). Prior work indicates that girls and women have lower confidence in their math and science ability than boys and men, net of actual ability in these subjects—that is, there is a confidence gap by gender in STEM (20, 22). During STEM matriculation in the secondary and postsecondary years, the lack of female STEM role models and faculty in classes can reinforce beliefs about the lesser abilities of women compared to men (19, 23–25). As women and men are first entering the workforce, personal beliefs in one’s engineering or technical ability may influence initial salaries.

Personal beliefs might have gendered effects on salaries in a number of ways. Women who are less confident about their technical abilities despite earning an engineering or CS degree may pursue less-competitive, lower-paying jobs than men who have recently earned those same degrees (9). This self-sorting might occur because less-confident women pursue lower-paying engineering or CS jobs, or because they opt to pursue jobs that are outside of engineering and CS which may pay less (5). Additionally, after receiving a job offer, less-confident women might be less likely to negotiate for higher salaries, or if they do attempt to negotiate those with less confidence might be less successful in receiving more compensation than more confident men (26). From the demand side of the labor market, even if less-confident women apply to lucrative jobs, studies indicate there are a number of ways such women might be reallocated or “steered” into applicant pools for jobs that pay less (27–29). Finally, less-confident individuals that may be considered for lucrative jobs might simply be rejected for those jobs, leaving them to pursue lower-paying options after one or more failed attempts at greater levels of remuneration. Each of these potential pathways for the effects of self-efficacy on the gender pay gap suggest the importance of personal beliefs. We expect the following:

H1) Women who graduate with engineering and CS degrees make less than men who graduate with engineering and CS degrees.

H2) Self-efficacy mediates the relationship between gender and salary among engineering and CS degree holders.

Empirical Considerations and Alternative Hypotheses

We compare our hypothesized relationships about gender, self-efficacy, and pay alongside two competing hypotheses that prior studies suggest might operate at the individual level to influence salaries: 1) socialized pay preferences and 2) female-friendly work environments. The first competing hypothesis, socialized pay preference, refers to girls and boys as well as women and men being socialized to expect different responsibilities across work and family as they get older. Scholars argue that girls and young women may be socialized to believe that money should matter little to them as they are not expected to be the primary breadwinners, so they should care less about salaries than men (30, 31). A second competing hypothesis, female-friendly work environments, refers to women placing greater emphasis on work culture than men. Women might prize environments that prioritize inclusion over competition, the latter being environments that might drive salaries higher (32). A gender pay gap might emerge to the extent that women place greater value on work cultures that pay less, not because they are less confident but because they prefer these environments more (33).

To test the main and alternative hypotheses, we use data collected from the Engineering Majors Survey (EMS), a three-wave longitudinal survey administered at 27 institutions in the United States (Materials and Methods). The first wave of surveys (EMS 1.0) was administered in winter/spring 2015 when students were still enrolled in college; the second wave (EMS 2.0) was administered in April 2016 and the third wave (EMS 3.0) in October 2017. We use data from EMS 1.0 and 3.0, in which we measured students’ engineering self-efficacy while still in school (EMS 1.0) and their workforce experiences to examine initial salaries among recent graduates (EMS 3.0). This study importantly measures self-efficacy prior to workforce entry and its correspondence to salary.

Results

At workforce entry, we find a significant difference between women and men in starting salaries (P < 0.01; Table 1). On average, women in our sample that graduate with engineering degrees earn less than $61,000 annually, while men earn above $65,000 annually. We model the relationship between sex and starting salaries of engineering and CS degree holders both without and with covariates (Table 2, models 1 and 2, respectively). Covariates include fixed effects for respondents’ home institutions, industry of employment, major field of study, and grade point average (GPA), as well as variables for degree type (B.A. or B.S.), year of graduation, whether they received a second degree, and whether they previously had an internship with the current employer (see SI Appendix, Table S1 for the correlation matrix for main variables and SI Appendix, Table S2 for a description of the covariates). We regress the logarithm of annual salary on female and find it is associated with a 9% decrease in salary overall without the covariates included (P < 0.01) and 7% decrease with covariates included (P < 0.05). Our first hypothesis, that women make less than men upon college exit when holding engineering and CS degrees, is supported.

Next, we examine self-efficacy by sex. Women have lower levels of self-efficacy than men on average prior to graduation (P < 0.001; Table 1), and women have lower self-efficacy than men across the percentile distribution (Fig. 1). For example, men at their 50th percentile of engineering self-efficacy have the same level of engineering self-efficacy as women in the 67th to 75th percentile.

We therefore investigate hypothesis 2 regarding whether or not self-efficacy mediates the relationship between gender and salary for engineering and CS degree holders. Our equations for mediation are as follows (34):

\[
Y = \beta_{10} + \beta_1 X + e_1 \tag{1}
\]

\[
Me = \beta_{20} + \beta_2 X + e_2 \tag{2}
\]

\[
Y = \beta_{30} + \beta_3 X + \beta_3 Me + e_3 \tag{3}
\]

where Y is salary, the \( \beta \)s are the coefficients, Me is the mediator of interest, and e is the error term. We examine the relationship between engineering self-efficacy (ESE) and salary and find there is a positive and statistically significant relationship, net other factors (\( \beta = 0.07, P < 0.01; SI Appendix, Table S3 \)). To assess the simple mediation model, we carry out the other two mediation steps and find ESE is negatively associated with being female (\( \beta = -0.37, P < 0.01; SI Appendix, Table S4 \)). Once ESE is included in the full model, the negative effect of being female is
Why Do Personal Beliefs Affect the Salaries of Women and Men? To build on possible explanations for why the confidence gap may impact the gender pay gap, we consider a number of explanations. We began by considering the possibility of performance differences across women and men. There is no evidence that women at the same level of self-efficacy as men perform worse

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**P < 0.01; ***P < 0.001. Difference values are calculated with more significant digits than what is shown above, and statistics are from two-sample t tests.

Table 2. OLS regressions predicting annual salary upon workforce entry

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|                | (1)     | (2)     | (3)     | (4)     | (5)     | (6)     |
|                | Female  | ESE     | Importance of compensation | Importance of workplace culture | Industry FEs | Institution FEs | Field FEs | GPA FEs | Degree FEs | Year of degree FEs | Second degree dummy | Internship employer dummy | Adjusted R² | Residual SE (df = 557) | Residual SE (df = 495) | Residual SE (df = 494) | Residual SE (df = 494) | Residual SE (df = 494) |
|                | −0.091** | (0.031) | 0.067** | (0.019) | 0.081** | (0.015) | 0.042** | (0.015) | 0.028* | (0.014) | 0.346 | 0.304 | 0.306 | 0.296 |

*P < 0.05; **P < 0.01; two-tailed hypothesis testing. Robust SEs in parentheses. The abbreviation FEs indicates model is run with the indicated fixed effects.

df, degrees of freedom.
We also suggested personal beliefs stemming from broader cultural beliefs about the appropriateness of women and men for some STEM professions—that is, the confidence gap—would explain this pay difference. Using data from a longitudinal study of 559 degree earners in engineering and CS undergraduate programs, we find support for our hypotheses. Women earn less than men in their initial jobs. We find no evidence that this gap in pay is due to men’s having stronger interests in high-paying jobs than women, or because women have preferences for the culture of workplaces that might correspond with lower pay. Rather, we find that women have lower levels of self-efficacy than men, and that self-efficacy mediates the relationship between gender and initial salaries.

In focusing on what occurs at a critical transition—that is, college to work—we offer key insights into the gender pay gap. Prior work has established a number of factors that influence the gender pay gap at the midpoint career stage, including women facing competing priorities on their time (36–39), more challenges in accumulating human capital (40), and discrimination in being promoted into higher and/or more technical positions in organizations, all of which have been shown to affect pay (41, 42). Less clear has been what transpires as women and men are making their initial transitions into the workplace, at a juncture that may affect earnings over the course of a career (43). We offer evidence that gendered beliefs may not only influence how women and men rate their own abilities when deciding which jobs to take but also on employer rejection, whereby less-confident individuals who majored and graduated in engineering or CS do not end up working in engineering or CS jobs. We posited various reasons including self-sorting and employer rejection, whereby less-confident individuals who majored and graduated in engineering or CS do not end up working in engineering or CS jobs. We look at pay for men and women that take initial jobs in engineering and CS and find men earn on average $1,996 more yearly than women, a difference that is not statistically significant (t statistic = 0.87).¹ Meanwhile degree holders that took engineering and CS jobs upon graduation were paid more than those that did not take engineering and CS jobs (t statistic = 4.01), suggesting one potential way self-efficacy may be influencing salaries is through entry into engineering and CS jobs. We find that self-efficacy influences the intent to enter engineering and CS jobs, and this in turn partially influences individuals’ actual entry into engineering and CS jobs (see SI Appendix, Fig. S4 for the positive relationship between ESE and engineering and CS job entry).² Performing a path analysis, we calculate that 33% of the influence that self-efficacy has on salary is explained by the effect of entering an engineering or CS job.³

Discussion

In this paper we hypothesized that among engineering and CS degree holders women earn less than men at workforce entry.

¹We have information regarding the salaries of the first jobs the full sample had after graduation. For a subset (n = 460) we also know the type of initial job they entered (for those whose first job was their current job).

²On the EMS 1.0 survey prior to graduating, students were asked about their intent to work in an engineering/CS job (see SI Appendix, section S1). Women had lower intentions of entering engineering jobs than men (t statistic = 2.55), but we found no statistically significant difference by sex once ESE was taken into account. This suggests the effect of self-efficacy on salaries is partly explained by its effects on intentions to enter and then actual entry into engineering and CS jobs.

³Iterative path analyses were performed using seemingly unrelated regression, sureg models in Stata 14. All pathways with a theoretical basis were tested, and support for the following pathways were found: ESE → Engineering Intentions → CS/Engineering job → Salary; ESE → Engineering Intentions → Salary; ESE → Salary; Female → ESE.

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that overemphasize confidence in one practitioners and industry experts to reconsider hiring practices is $63,709, SD women, compared with 21% women nationally). The average starting salary stratified scheme. Compared to the national population of engineering population from which the 27 EMS institutions were sampled using a (Stanford IRB #31803 and Stanford IRB #35539). All participants provided This study was approved by the Stanford Institutional Review Board (IRB) flect technical ability and may correspond to gender, doing so likely job success. Since self-perceptions may not accurately re-

In closing, there are several important practical implications from this study. Our results highlight the importance of career guidance as well as possibly internships and co-ops to strengthen students’ self-assessments and provide stronger bridges to engineering and CS jobs with higher pay (45). Our results also ask practitioners and industry experts to reconsider hiring practices that overemphasize confidence in one’s ability as an indication of likely job success. Since self-perceptions may not accurately reflect technical ability and may correspond to gender, doing so may only stand to widen the gender pay gap.

Materials and Methods
This study was approved by the Stanford Institutional Review Board (IRB) (Stanford IRB #31803 and Stanford IRB #35539). All participants provided informed consent. The 350 US engineering schools in 2011 composed the population from which the 27 EMS institutions were sampled using a stratified scheme. Compared to the national population of engineering students, women are overrepresented (in the analytical sample 35% are women, compared with 21% women nationally). The average starting salary is $63,709, SD = $21,370 (Table 1). There are n = 559 respondents. All of the CS graduates in the study were awarded degrees from engineering schools.4

Fig. 2. Parallel multiple mediation model depicting the effect of being female on salary as mediated by engineering self-efficacy, importance of compensation, and importance of workplace culture. The specific indirect effects (a, b, c) for the mediators and direct effect for sex (c') are shown with their SEs and 95% CIs. n = 559.

Figures

Measures. Following prior literature, the dependent variable is the log of self-reported annual salary (see SI Appendix, Fig. S1 for salary distributions). We asked respondents to self-identify by sex and report self-identified sex in the results. To assess individual beliefs about one’s technical ability we measure ESE, a five-item validated measure on a five-point scale (0 = “not confident” to 4 = “extremely confident,” alpha = 0.87; SI Appendix, section S1). Participants were asked, “How confident are you in your ability to do each of the following at this time?”: 1) design a new product or project to meet specified requirements; 2) conduct experiments, build prototypes, or construct mathematical models to develop or evaluate a design; 3) develop and integrate component subsystems to build a complete system or product; 4) analyze the operation or functional performance of a complete system; and 5) troubleshoot a failure of a technical component or system.

Data Analyses. To test hypothesis 1 we ran fixed effect ordinary least squares (OLS) models with robust SEs to address heteroscedasticity in Table 2. To test hypothesis 2 we ran a simple mediation model to assess the indirect effect of self-efficacy on the relationship between sex and salary (34) as described in Eqs. 1–3. To inspect the specific indirect effect of self-efficacy controlling for the two alternative mediators, sex, and the other covariates we ran the parallel multiple mediation model (35) as shown in Fig. 2. We used 1,000 bootstrapped iterations to compute a bias corrected 95% CI for the specific indirect effects.

Data Availability. A subset of anonymized data and code used to test hypotheses 1 and 2 are available on Open Science Framework (https://osf.io/yzqy9).

ACKNOWLEDGMENTS. We thank Tom Byers, Helen Chen, Kai-Jun Chew, Qu Jin, and Mark Schar for their contributions to the launch of the EMS study, as well as our project liaisons at the 27 EMS schools. We also thank the anonymous reviewers for their critical comments that led to an improved manuscript, Francis (Frank) Flynn for his consultancy on methods, and Anita Rollins for her editing support. The EMS study was conducted with funding from NSF Grant/Award 1636442 and DUE-1125457. The research reported here was supported in part by the Institute of Education Sciences, US Department of Education, through Grant R305B140009 to the Board of Trustees of the Leland Stanford Junior University. The opinions expressed are those of the authors and do not represent views of the Institute, the US Department of Education, or the Board of Trustees of the Leland Stanford Junior University.

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