

[advances.sciencemag.org/cgi/content/full/7/3/eabe0933/DC1](https://advances.sciencemag.org/cgi/content/full/7/3/eabe0933/DC1)

Supplementary Materials for  
**Systemic inequalities for LGBTQ professionals in STEM**

E. A. Cech\* and T. J. Waidzunas

\*Corresponding author. Email: [ecech@umich.edu](mailto:ecech@umich.edu)

Published 15 January 2021, *Sci. Adv.* 7, eabe0933 (2021)

DOI: [10.1126/sciadv.abe0933](https://doi.org/10.1126/sciadv.abe0933)

**This PDF file includes:**

Additional Methods and Materials Information

Tables S1 to S10

References

## **Additional Methods and Materials Information**

*STEM Inclusion Study Survey:* Between winter 2017 and spring 2019, we surveyed representative samples of the US-based membership of 21 STEM professional societies and organizations. In collaboration with the leadership and membership directors of each professional society, we fielded the survey electronically via email to either a random sample of the US-based members of each society (for societies over 10,000 members) or to the full US-based membership (for societies under 10,000 members). The survey was open at each society for a period of six weeks. Non-responders received up to two reminder emails. The average response rate was 20.1%, which is typical of surveys of members of voluntary organizations (56). Respondents could end the survey at any time and participation was anonymous. The study was approved by the human subjects board at each author's institution.

The survey asked respondents a variety of questions about their experiences with the climate in their workplaces, their interactions with colleagues, and their future plans. At the end of the survey, respondents were invited to enter a raffle for a \$100 gift card (one per professional society) to help offset the annual cost of their society membership. The survey included skip logics that could accommodate students, retired members, and members who worked in non-STEM jobs; we use here only data from respondents who were employed full-time in a non-social science STEM job in the United States at the time of survey participation (N=25,324). To avoid duplicate individual entries, the first question in the survey asked respondents whether they had taken the survey previously; those who answered affirmatively (2%) were thanked and skipped out of the survey.

### ***Survey Reliability and Validity***

The measures we used in our analysis were either replications of existing, validated survey items or items written and pretested for this study. Measures for education level, STEM field, supervisory status, employment sector, race/ethnicity, and age came from the National Science Foundation's National Survey of College Graduates (<https://www.census.gov/programs-surveys/nscg.html>). The health and wellness variables and the social exclusion measures were replications of questions from the National Survey of the Changing Workforce (<https://www.familiesandwork.org>). Questions on career opportunities and sufficient resources and comfort whistleblowing came from the US Office of Personnel Management's biennial Merit Principles Survey (<https://www.mspb.gov/studies/MPS2016.htm>).

The five measures that make up the professional devaluation scale were specifically designed for this survey. These measures assessed whether respondents' professional expertise was recognized, valued, and given proper credit by their colleagues. We pretested these items in a survey of STEM professionals at two NASA space flight

centers, and found a strong Cronbach's alpha (a measure of how well topically-related items from a question set vary together and form a coherent and reliable scale ) of .769 and high predictive validity. Indicating high test-retest reliability and internal consistency of the professional devaluation scale, we found consistently strong Cronbach's alphas (less than 10% variability) when we ran factor analyses of the professional devaluation measures for each of the 21 professional societies separately.

We assessed the validity of the survey instrument in its entirety through a number of steps. First, we established content validity by workshopping the survey with a panel of seven social scientists who are experts on workplace inequality and LGBTQ issues. Second, we checked the face validity of the survey with in-person talk-through sessions (i.e. cognitive interviews) with eight STEM professionals (three who identified as LGBTQ). In these sessions, informants took the survey while explaining aloud their interpretations of each survey question and answer option. These talk-throughs provided insight into the clarity and consistency of the survey questions and answer options (57). Third, we established construct validity through analysis of the convergence of similar concepts and divergence of dissimilar concepts in correlations and factor loadings (57-58). Evincing convergence validity, the questions used in the three scale measures (career opportunities, professional devaluation, and social exclusion) each loaded onto their respective factors; discriminant validity tests indicated that each measure in a given scale was more highly correlated with measures in its own scale than with measures in the other two scales. Tests for discriminant validity showed that within-index correlations ranged between .638-.760 (strongly correlated) while cross-index correlations were less than .400 (weakly correlated).

Finally, we calculated Cohen's  $d$  effect sizes ( $d$  = difference in means/pooled standard deviation) for the difference in means between LGBTQ and non-LGBTQ respondents (see Table S10). Suggesting that LGBTQ status is a particularly consequential axis of disadvantage in the context of STEM compared to the labor force overall, these effect sizes are 1.5 to 2 times greater on average than the LGBTQ status effect sizes Cech and Rothwell (23) found in their analysis of workplace inequalities for LGBTQ employees in the US federal workforce.

### ***Descriptive Statistics***

Table S1 presents means for all respondents in the sample and for LGBTQ and non-LGBTQ respondents separately. The p-value column indicates statistical significance of two-tailed bivariate difference in means tests (i.e., t-tests) between LGBTQ and non-LGBTQ respondents. Here, LGBTQ respondents were more likely to identify as women and as Latinx and Native American or Pacific Islander, and less likely to identify as white. LGBTQ persons were less likely than non-LGBTQ respondents to work in engineering but more likely to work in life sciences, computer science and mathematics, and other STEM occupations. Finally, compared to non-LGBTQ sample members, LGBTQ professionals were less likely to work in for-profit industries and more likely to work in government and university sectors. LGBTQ respondents were also younger (~3 years on average) than their non-LGBTQ peers. Given this variation in LGBTQ persons' representation across sectors and STEM fields, it was important to control for this variation via predicted means and multivariate regressions when assessing possible LGBTQ differences in the figures and regression models above.

The rightmost column in Table S1 presents means for the STEM population nationally from 2017 National Science Foundation data (<https://nces.nsf.gov/pubs/nsf19304/data>). Compared to the STEM population overall,

our data over-represents those who are white, who work in engineering and physical sciences, and those who work in university and government sectors. As noted above, we conducted supplemental analyses where we weighted our sample to match the distribution in the NSF data by demographics, sector, and STEM field; the LGBTQ results patterns did not change when we used this weighting.

### *Analytic Strategy for Supplemental Tables*

The multivariate analyses presented in the tables below used OLS and logistic regression models, as appropriate, to predict outcome measures. We opted to use OLS and logistic regression models with dichotomous controls for each professional society rather than multilevel models because the former are easier to interpret for those without advanced quantitative training and because 21 level-two categories is just on the threshold of appropriateness for hierarchical models. We also used structural equation modeling (SEM) to test for mediation effects (Tables S4 and S5), and interaction term analysis (e.g., LGBTQ status X age, Tables S6 and S7) to test for intersectional patterns. Each model included the demographic and employment controls listed above and dichotomous controls for professional societies. Engineering is the comparison category for STEM field in the regression models because it is the largest subfield category in the sample. We use multiple imputation (MI chained technique in Stata 15 with 20 imputations) to handle missing data.

**Table S1.** Univariate and Bivariate Statistics for Demographics and Employment Controls for All Respondents and for LGBTQ and non-LGBTQ Respondents, and Descriptive Statistics from National Science Foundation Data on US STEM Professionals.

	ALL N=25,324	LGBTQ N=1,006	Non-LGBTQ N=24,318	P (LGBTQ vs Non-LGBTQ)	2017 NSF Data
LGBTQ	4.51%	---	---		---
Women (cisgender & transgender)	30.17%	32.63%	29.81%	***	29.00%
Men (cisgender & transgender)	69.81%	58.13%	69.57%	***	71.00%
Transgender & Gender Non-binary	0.85%	4.06%	0.09%	***	---
Black	2.19%	2.72%	2.16%		5.73%
Hispanic/Latinx	5.91%	5.97%	5.79%	**	7.58%
Asian	10.20%	13.28%	10.30%		20.09%
NAAPI	0.93%	1.45%	0.91%	*	0.57%
White	78.89%	65.80%	78.83%	*	66.02%
Engineering	38.58%	29.74%	39.11%	***	20.18%
Life Sciences	11.72%	11.71%	11.64%	*	7.12%
Physical Sciences	21.92%	23.54%	21.81%		4.30%
Computer Science & Mathematics	15.52%	20.17%	15.33%	***	39.90%
Other STEM Occupation	12.26%	14.84%	12.12%	**	28.49%
For-Profit Sector	33.71%	28.75%	33.95%	***	63.77%
University Sector	39.91%	42.49%	39.73%	*	13.01%
Government Sector	13.71%	15.25%	13.68%	*	11.60%
Nonprofit Sector	5.28%	5.97%	5.23%		5.31%
K-12	3.81%	4.46%	3.78%		3.19%
Other Sector	3.57%	3.07%	3.62%		3.12%
Age	49.84	47.24	50.12	***	---
Core Technical Work Indicator	39.15%	39.48%	39.10%		---
Employer Size	5.60	5.66	5.59		---
Parents' Highest Degree	4.30	4.30	4.29		---

Notes: \* p<.05; \*\* p<.01; \*\*\* p<.001; two-tailed test, comparing LGBTQ and non-LGBTQ respondents via t-tests. Gender categories for women and men include both cisgender and transgender persons who identify as women and men, respectively. Transgender and gender non-binary status is combined above to protect confidentiality. NAAPI=Native American and Asian Pacific Islander. Unlike the NSF survey, the SIS survey allowed respondents to indicate more than one racial/ethnic category.

**Table S2.** OLS and Logistic Regression Models Predicting Career Opportunities, Professional Devaluation, and Social Exclusion Measures with LGBTQ Status and Controls.

	Career Opportunities		Career Resources		Whistleblowing comfort		Professional Devaluation		Social Marginalization		Experienced Harassment	
LGBTQ	-.106	***	-.089	***	-.151	***	.137	***	.171	***	.050	***
Women (cis and trans)	-.125	***	-.132	***	-.253	***	.257	***	.105	***	.686	***
Black	-.091	*	.040		-.134	**	.249	***	.194	***	.268	*
Hispanic/Latinx	-.084	**	.034		-.027		.094	***	.005		.231	**
Asian	-.220	***	.108	***	-.162	***	.148	***	.029		.025	
NAAPI	-.059		-.191	**	-.083		.128	**	.184	***	.363	*
Life Sciences	-.048		.044		.031		-.007		-.036		-.114	
Physical Sciences	-.030		.015		-.017		.004		-.023		-.155	*
Computer Sci & Math	.008		.014		.008		-.046		-.022		-.268	*
Other STEM Occupation	-.043	*	.027		.011		-.001		-.013		-.049	
University sector	-.039	*	-.127	***	-.197	***	.148	***	.108	***	.279	***
Government sector	-.037		-.156	***	-.174	***	.080	***	.061	***	.293	***
Nonprofit sector	-.009		.058		-.044		-.002		.044		.121	
K-12	.044		-.140	***	-.224	***	-.043		.003		.203	
Other Sector	.133	***	.152	***	-.170	***	.162	***	.037		.192	
Age	.008	***	.002	***	.001	**	-.002	***	.012	***	-.008	***
Highest Degree	.016	**	-.032	***	-.001		.001		.000		.016	**
Core Tech work indicator	-.071	***	.144	***	-.058	***	.011		.012		.014	
Employer Size	.393	***	-.038	**	.085	***	-.033	**	-.013		.085	***
Supervisory Status	-.005		-.018	***	.008	*	.010	***	.012	***	-.018	
Parents' Highest Ed	.013	***	.003		.000		-.011	***	.003		-.173	***
Constant	3.400	***	4.008	***	2.779	***	1.976	***	1.949	***	-1.838	***

Notes: N=25,324; \* p<.05; \*\* p<.01; \*\*\* p<.001; two-tailed test. White is comparison category for race/ethnicity; men (cis and transgender) and gender non-binary respondents are comparison category for women; engineering is comparison category for STEM field; for-profit is comparison category for sector. Models also include controls for professional society. Logistic regression was used to predict the experiences of harassment (a dichotomous measure); OLS regression was used for all other outcome measures.

**Table S2, Cont.** OLS Regression Models Predicting Health and Wellness Difficulties and Turnover Intentions with LGBTQ Status and Controls.

	Minor Health Problems		Insomnia		Stressed from Work		Depressive Symptoms		Thought about Leaving Job		Intentions to Leave Prof.	
LGBTQ	.206	***	.246	***	.259	***	.178	***	.163	***	.173	***
Women (cis and trans)	.339	***	.150	***	.327	***	.034	**	.114	***	.102	***
Black	-.130	**	-.095	*	-.206	***	-.060	*	.159	***	.266	***
Hispanic/Latinx	-.001		-.010		-.054		-.041	*	-.004		.057	
Asian	-.133	***	-.047	*	-.271	*	.049	**	-.125	***	.278	***
NAAPI	.135	*	.075		.058		.060		.020		.067	
Life Sciences	.014		.029		.067		.034		.078		.103	
Physical Sciences	-.019		-.009		.047		.036		-.027		-.099	**
Computer Sci & Math	.024		-.011		.067	*	.060	*	.048		.058	
Other STEM Occupation	.013		-.007		.012		.019		.136	***	.206	***
University sector	.090	***	.106	***	.120	***	.096	***	-.161	***	-.135	***
Government sector	.069	**	.090	***	.031		.054	***	-.078	*	-.004	
Nonprofit sector	.000		.072	**	.023		.023		-.112	**	-.068	
K-12	-.013		-.027		-.004		.038		-.193	***	.072	
Other Sector	-.046		.012		-.080	*	-.060	*	-.060		-.291	***
Age	-.013	***	-.009	*	-.024	***	-.011	***	-.013	***	.008	
Core Tech work indicator	.001		.002		-.027	*	.002	*	-.093	**	-.087	***
Supervisory Status	-.005	**	.023		.066	***	-.022		-.059	**	-.173	***
Employer Size	-.010		-.002		.003		-.003		.009		-.005	
Parents' Highest education	.003	***	-.003		.006		-.007	**	-.003		-.015	**
Constant	2.934	***	2.231	***	3.781	***	2.557	***	2.709	***	1.899	***

Notes: N=25,324; \* p<.05; \*\* p<.01; \*\*\* p<.001; two-tailed test. White is comparison category for race/ethnicity; men (cis and trans) and gender non-binary respondents are comparison category for women; engineering is comparison category for STEM field; for-profit is comparison category for sector. Models also include controls for professional society.

**Table S3.** Direct Effects of LGBTQ Status and Inequality Measures, and Indirect Effects of LGBTQ Status through Inequality Measures as Mediators, Predicting Health and Wellness Outcomes

	<b>Direct Effect:</b> LGBTQ→ Health Outcome (Coeff/signif)	<b>Direct Effect:</b> Career Opps→ Health Outcome (Coeff/signif)	<b>Direct Effect:</b> LGBTQ→ Career Opps (Coeff/signif)	<b>Indirect Effect:</b> LGBTQ→ Career Opps → Health Outcome(Coeff/Signif)
<b>Mediator: Career Opportunities</b>				
Outcome: Minor Health Problems	.266***	-.066***	-.192***	.013***
Outcome: Insomnia	.276***	-.086***	-.193***	.017***
Outcome: Stressed from Work	.372***	-.096***	-.193***	.019***
Outcome: Depressive Symptoms	.207***	-.112***	-.194***	.022***
	<b>Direct Effect:</b> LGBTQ→ Health Outcome (Coeff/signif)	<b>Direct Effect:</b> Career Resorc→ Health Outcome (Coeff/signif)	<b>Direct Effect:</b> LGBTQ→ Career Resorcs (Coeff/signif)	<b>Indirect Effect:</b> LGBTQ→ Career Resources → Health Outcome (Coeff/Signif)
<b>Mediator: Career Resources</b>				
Outcome: Minor Health Problems	.267***	-.093***	-.130***	.012***
Outcome: Insomnia	.276***	-.108***	-.132***	.014***
Outcome: Stressed from Work	.374***	-.146***	-.132***	.019***
Outcome: Depressive Symptoms	.211***	-.117***	-.132***	.015***
	<b>Direct Effect:</b> LGBTQ→ Health Outcome (Coeff/signif)	<b>Direct Effect:</b> Prof. Deval→ Health Outcome (Coeff/signif)	<b>Direct Effect:</b> LGBTQ→ Prof. Deval (Coeff/signif)	<b>Indirect Effect:</b> LGBTQ→ Prof. Deval → Health Outcome (Coeff/Signif)
<b>Mediator: Professional Devaluation</b>				
Outcome: Minor Health Problems	.260***	.135***	.179***	.024***
Outcome: Insomnia	.265***	.209***	.182***	.038***
Outcome: Stressed from Work	.361***	.219***	.181***	.034***
Outcome: Depressive Symptoms	.196***	.214***	.181***	.039***

Notes: N=25,324; \* p<.05; \*\* p<.01; \*\*\* p<.001; two-tailed test. Direct and indirect effects produced from generalized structural equation models (GSEM) in Stata 14. All models include controls for the demographic, STEM field, employment sector, and job controls listed in Table S2, as well as controls for professional society.



**Table S3, Cont.** Direct Effects of LGBTQ Status and Inequality Measures, and Indirect Effects of LGBTQ Status through Inequality Measures as Mediators, Predicting Health and Wellness Outcomes

	<b>Direct Effect:</b> LGBTQ→ Health Outcome (Coeff/signif)	<b>Direct Effect:</b> Whistleblowing → Health Outcome (Coeff/signif)	<b>Direct Effect:</b> LGBTQ→ Whistleblowing (Coeff/signif)	<b>Indirect Effect:</b> LGBTQ→ Whistleblowing → Health Outcome (Coeff/Signif)
<b>Mediator: Comfort Whistleblowing</b>				
Outcome: Minor Health Problems	.261***	-.098***	-.248***	.024***
Outcome: Insomnia	.266***	-.097***	-.238***	.023***
Outcome: Stressed from Work	.366***	-.106***	-.237***	.025***
Outcome: Depressive Symptoms	.202***	-.100**	-.238***	.024***
	<b>Direct Effect:</b> LGBTQ→ Health Outcome (Coeff/signif)	<b>Direct Effect:</b> Social Excludn → Health Outcome (Coeff/signif)	<b>Direct Effect:</b> LGBTQ→ Social Excludn (Coeff/signif)	<b>Indirect Effect:</b> LGBTQ→ Social Exclusion → Health Outcome (Coeff/Signif)
<b>Mediator: Social Exclusion</b>				
Outcome: Minor Health Problems	.258***	.094***	.189***	.018***
Outcome: Insomnia	.266***	.121***	.189***	.023***
Outcome: Stressed from Work	.362***	.125***	.190***	.024***
Outcome: Depressive Symptoms	.200***	.130***	.189***	.025***
	<b>Direct Effect:</b> LGBTQ→ Health Outcome (Coeff/signif)	<b>Direct Effect:</b> Harassment → Health Outcome (Coeff/signif)	<b>Direct Effect:</b> LGBTQ→ Harassment (Coeff/signif)	<b>Indirect Effect:</b> LGBTQ→ Harassment → Health Outcome (Coeff/Signif)
<b>Mediator: Harassment</b>				
Outcome: Minor Health Problems	.271***	.116***	.067***	.008***
Outcome: Insomnia	.271***	.173***	.067***	.012***
Outcome: Stressed from Work	.371***	.184***	.067***	.012***
Outcome: Depressive Symptoms	.206***	.116***	.067***	.008***

Notes: N=25,324; \* p<.05; \*\* p<.01; \*\*\* p<.001; two-tailed test. Direct and indirect effects produced from generalized structural equation models (GSEM) in Stata 14. All models include controls for the demographic, STEM field, employment sector, and job controls listed in Table S2, as well as controls for professional society.

**Table S4.** Direct Effects of LGBTQ Status and Inequality Measures, and Indirect Effects of LGBTQ Status through Inequality Measures as Mediators, Predicting Intentions to leave STEM Job and STEM Profession

	<b>Direct Effect:</b> LGBTQ→ Leaving Outcome (Coeff/signif)	<b>Direct Effect:</b> Career Opps → Leaving Outcome (Coeff/signif)	<b>Direct Effect:</b> LGBTQ→ Career Opps (Coeff/signif)	<b>Indirect Effect:</b> LGBTQ→ Career Opps → Leaving Outcome (Coeff/Signif)
<b>Mediator: Career Opportunities</b>				
Outcome: Thought abt Leaving Job	.106**	-.317***	-.198***	.063***
Outcome: Intend to Leave Prof	.077	-.114***	-.202***	.023***
<b>Mediator: Career Resources</b>				
Outcome: Thought abt Leaving Job	.123**	-.244***	-.133***	.032***
Outcome: Intend to Leave Prof	.092*	-.023***	-.123***	.004*
<b>Mediator: Professional Devaluation</b>				
Outcome: Thought abt Leaving Job	.067**	.394***	.183***	.072***
Outcome: Intend to Leave Prof	.079	.096***	.202***	.019***
<b>Mediator: Comfort Whistleblowing</b>				
Outcome: Thought abt Leaving Job	.135***	-.245***	.231***	.058***
Outcome: Intend to Leave Prof	.084	-.040***	.245***	.010***
<b>Mediator: Social Exclusion</b>				
Outcome: Thought abt Leaving Job	.100**	.262***	.193***	.051***
Outcome: Intend to Leave Prof	.076	.748***	.206***	.015***
<b>Mediator: Harassment</b>				
Outcome: Thought abt Leaving Job	.129***	(Coeff/signif)	.067***	.031***
Outcome: Intend to Leave Prof	.085	.054**	.066***	.004*

Notes: N=25,324; \* p<.05; \*\* p<.01; \*\*\* p<.001; two-tailed test. Direct and indirect effects produced from GSEM in Stata 14. All models include controls for the demographic, STEM field, employment sector, and job controls listed in Table S2, and controls for professional society.

**Table S5.** Focal Coefficients and Significance Levels from Regression Models Predicting Outcome Measures with Interaction Terms between LGBTQ Status and Gender Identity, Race/Ethnicity, and Age (included in models separately)

	<b>LGBTQ x Women (cis &amp; trans)</b>	<b>LGBTQ x Transgender &amp; Gender Non-binary</b>	<b>LGBTQ x Black</b>	<b>LGBTQ x Latinx</b>	<b>LGBTQ x Asian</b>	<b>LGBTQ x NAAPI</b>	<b>LGBTQ x Age</b>
Career Opportunities	.063	-.182	-.205	-.186 *	-.017	-.177	-.001
Career Resources	.009	.004	-.088	.079	-.185	-.286	-.005
Whistleblowing comfort	-.052	.094	-.280	.072	-.145	.471 *	-.003
Professional Devaluation	-.068	.111	.179	-.004	.098	.415 *	-.001
Social Exclusion	-.020	-.133	.178	.055	.068	-.023	-.002
Harassment	.027	.047	.091	.032	.129 *	-.012	-.001
Minor Health Problems	-.113	.327 *	.182	-.120	-.147	-.287	-.003
Insomnia	.060	.089	.287	-.076	.012	-.022	-.005 *
Stressed from Work	-.024	.482 **	.300	.041	-.019	.286	-.003
Depressive Symptoms	.056	.388 ***	.245 +	-.017	-.053	.278	-.011 **
Thought about Leaving	.029	.423 *	.008	-.069	.235	.083	-.001
Intentions to Leave Prof.	.031	.083	.107	.151	.060	.213	.011 *

Notes: N=25,324; \* p<.05; \*\* p<.01; \*\*\* p<.001; two-tailed test. NAAPI=Native American Asian Pacific Islander. Interaction term coefficients produced from OLS or logit models predicting each outcome (as appropriate). Interaction terms were included in the models separately. To protect confidentiality of gender non-binary respondents, we created a combined category of transgender and gender non-binary individuals for the interaction analysis in the second column above. All models include controls for the demographic, STEM field, employment sector, and job controls listed in Table S2, as well as controls for professional society.

**Table S6.** Focal Coefficients from Regression Models Predicting Inequality Measures with Interaction Terms between LGBTQ Status and STEM Field and Employment Sector (included in models separately)

	LGBTQ x Life Sci	LGBTQ x Physical Sci	LGBTQ x Comp Sci & Math	LGBTQ x Other STEM	LGBTQ x University	LGBTQ x Gvmt	LGBTQ x K-12	LGBTQ x Nonprofit	LGBTQ x Other Sector
Career Opportunities	.022	.030	-.016	.025	-.098	-.080	.163	.178	.062
Career Resources	-.080	.047	.034	.096	-.178 *	-.130	.212	.214	-.241
Whistleblowing comfort	-.150	-.120	-.161	-.175	.145 *	-.239 *	-.262	.074	-.005
Professional Devaluation	-.135 *	-.024	.035	-.076	.087	.111	.022	-.056	-.133
Social Exclusion	-.063	.031	.044	-.172 *	.049	.105	-.141	-.141	-.075
Harassment	-.003	-.061	-.012	.091	-.001	.028	-.063	-.052	-.071
Minor Health Problems	-.016	-.087	.141	-.014	.056	.111	-.184	.079	-.050
Insomnia	.097	-.086	.017	-.014	-.041	-.143	-.275	.063	-.001
Stressed from Work	.136	.009	.146	.139	.064	-.037	-.038	-.102	.205
Depressive Symptoms	.111	-.069	.029	-.006	.056	-.032	-.004	-.027	-.005
Thought about Leaving	-.082	-.039	-.015	.034	.108	.149	-.090	-.078	-.111
Intentions to Leave Prof.	-.138	-.070	-.086	-.107	-.191 *	-.153	-.382	-.199	-.096

Notes: N=25,324; \* p<.05; \*\* p<.01; \*\*\* p<.001; two-tailed test. Interaction term coefficients produced from OLS or logit models predicting each outcome (as appropriate). All models include controls for the demographic, STEM field, employment sector, and job controls listed in Table S2, as well as controls for professional society.

**Table S7.** OLS and Logistic Regression Models Predicting Career Outcomes, Devaluation, Marginalization, Health and Wellness Difficulties and Intentions to Leave among LGBTQ STEM Professionals Only, with Demographic Measures and Controls

	Career Opportunities	Career Resources	Whistleblowing comfort	Professional Devaluation	Social Exclusion	Harassment
Transgender & Gender Nonbinary	-.263	-.299	-.139	.278 *	.229	.170
Women (cis & trans)	-.089	-.124	-.207 **	.189 ***	.083	.141 **
Black	-.374	-.072	-.609 **	.445 **	.447 *	.074
Hispanic/Latinx	-.189	.031	-.050	.022	.048	.072
Asian	-.257 *	-.187	-.257 *	.258 **	.107	.162 *
NAAPI	-.335	-.548	.251	.599 **	.176	.047
Life Sciences	-.026	-.034	-.145	-.102	.065	-.114
Physical Sciences	-.006	.098	-.014	.036	.086	-.081
Computer Sci & Math	-.026	-.133	-.163	.102	.099	-.007
Other STEM Occupation	-.135	.062	-.071	.075	-.031	.137
University sector	-.115	-.353 **	-.273 *	.204 *	.149	.071
Government sector	-.062	-.331 **	-.440 ***	.172	.182	.147 *
Nonprofit sector	.128	.333 *	-.033	.072	-.001 ***	.033
K-12	.041	.125	-.362	.067	-.063	-.091
Other Sector	.211	.098	-.187	.177	.036	.077
Age	.007 *	-.005	-.003	-.001	-.002	-.003
Core Tech work indicator	-.035	.270 **	-.080	-.003	.026	-.114 ***
Employer Size	-.021	-.011	.008	.054 **	.019	.021
Parents' Highest Ed	.010	-.045 *	-.032	-.009	.002	-.005
Constant	3.873 ***	4.33 ***	2.529 ***	1.673 ***	1.891 ***	1.132 ***

Notes: N=1,006; \* p<.05; \*\* p<.01; \*\*\* p<.001; two-tailed test. Transgender and gender non-binary status is combined in a single category to protect confidentiality. Logistic regression was used to predict the dichotomous harassment measure; OLS regression was used to predict all other outcomes. All models include controls for professional society.

**Table S7, Cont.** OLS Regression Models Predicting Career Outcomes, Devaluation, Marginalization, Health and Wellness Difficulties and Intentions to Leave among LGBTQ STEM Professionals, with Demographic Measures and Controls

	Minor Health Problems	Insomnia	Stressed from Work	Depressive Symptoms	Thought about Leaving Job	Intentions to Leave Prof.
Transgender & Gender Nonbinary	.216	.154	.359 *	.452 ***	.149 *	.071
Women (cis & trans)	.185	.190 ***	.281 **	.032 ***	.083	.141 **
Black	-.084	.125	.057	.139 **	.447 *	.074
Hispanic/Latinx	-.297 *	-.153	-.170	-.149	.048	.072
Asian	-.408 ***	-.102	-.366 **	-.044 **	.107	.162 *
NAAPI	-.166	-.066	.316	.326 **	.176	.047
Life Sciences	.204	.123	.221	.148	.065	-.114
Physical Sciences	-.065	-.059	.150	.051	.086	.081
Computer Sci & Math	.136	.053	.030	.077	.099	-.007
Other STEM Occupation	.177	.081	.246	.069 *	-.031	.137
University sector	.151	-.004	.108	.159 *	.149	.071
Government sector	.166	.040	-.031	-.012	.182	.147 *
Nonprofit sector	.180	.150	.014	.117	-.001 ***	.033
K-12	-.084	-.288	-.126	.135	-.063	-.091
Other Sector	.104	.070	.147	.058	.036	.077
Age	-.017 ***	-.014 ***	-.029 ***	-.015 ***	-.002	-.003
Core Tech work indicator	.109	.033	-.047	.014	.026	-.114
Employer Size	-.004	.001	.037	.033	.019 *	.021 *
Parents' Highest Ed	-.021	-.013	.013	-.006	.002	-.005
Constant	3.482 ***	2.639	4.493 ***	3.335 ***	1.891	1.132 ***

Notes: N=1,006; \* p<.05; \*\* p<.01; \*\*\* p<.001; two-tailed test. Transgender and gender non-binary status is combined in a single category to protect confidentiality. Logistic regression was used to predict the dichotomous harassment measure; OLS regression was used to predict all other outcomes. All models include controls for professional society.

**Table S8.** OLS and Logistic Regression Models Predicting Education Level, Hours Worked, Core Technical Work, and Work Dedication Measures, by LGBTQ Status and Controls.

	Highest Degree			Hours Worked per Week			Does Core Technical Work as Primary Work Responsibility			Puts in a Great Deal of Effort Beyond What's Required of Job			Work is an Important Part of Personal Identity		
	Coeff	p	SE	Coeff	p	SE	Coeff	P	SE	Coeff	p	SE	Coeff	p	SE
LGBTQ	-.060		.036	.213		.379	.002		.069	-.050		.027	.018		.028
Women (cis & trans)	-.057	**	.017	-1.091	***	.180	-.336	***	.069	.003		.013	.082	***	.013
Black	.168	**	.048	1.142	*	.514	-.198	*	.033	.229	***	.038	-.193	***	.038
Hispanic/Latinx	.049		.034	.254		.365	.074		.097	.125	***	.026	.055	*	.027
Asian	.370	***	.024	.620	*	.257	.546	***	.065	.087	***	.018	.079	***	.019
NAAPI	-.173	*	.074	1.117		.785	-.396	*	.045	.071		.057	-.068		.058
Life Sciences	.283	***	.044	.430		.470	.317	***	.156	.122	***	.034	.072	*	.034
Physical Sciences	.349	***	.030	-.614		.323	.140	*	.084	.057	*	.023	.050	*	.024
Comp Sci & Math	-.007		.039	-.513		.418	-.058		.059	.049		.030	-.056	*	.031
Other STEM Occupation	-.264	***	.027	.125		.295	-1.145	***	.076	.043	*	.021	-.054	***	.021
University sector	1.221	***	.022	2.402	***	.244	-.256	***	.062	.011		.017	.293	***	.017
Government sector	.534	***	.025	-.721	**	.265	-.222	***	.040	.033		.019	.096	***	.019
Nonprofit sector	.692	***	.035	-.128		.375	-.115		.046	.008		.027	.188	***	.027
K-12	-.188	***	.046	6.274	***	.491	-4.121	***	.067	.264	***	.035	.403	***	.036
Other Sector	.334	***	.043	-7.102	***	.457	.472	***	.505	.166	***	.033	.212	***	.034
Age	.011	***	.001	-.082	***	.006	-.020	***	.080	.000	***	.005	.013	***	.000
Core Tech work indicator	.333	***	.016	-1.027	***	.171	---		---	.005	***	.000	.001	***	.004
Supervisory Status	.175	***	.015	5.796	***	.162	-.420	***	.001	-.062	***	.012	.010	***	.003
Employer Size	.016	***	.004	.447	****	.048	.075	***	.029	.213	***	.012	.018		.028
Parents' Highest Edu	.057	***	.004	.051		.040	.027	***	.009	-.013	***	.003	.082	***	.013
Constant	4.930	***	.064	42.848	***	.736	-1.155	***	.151	-.014	***	.003	3.324	***	.048

Notes: N=25,324; \* p<.05; \*\* p<.01; \*\*\* p<.001; two-tailed test. White is comparison category for race/ethnicity; men and gender non-binary respondents are comparison categories for women; engineering is comparison category for STEM field; for-profit is comparison category for sector. Models also include controls for professional society. Model predicting whether respondent does core technical work as their primary work activity uses logistic regression; all other outcome measures are predicted with OLS regression.

**Table S9.** Coefficients and Standard Errors for LGBTQ Coefficient in Regression Models With and Without Control for Job Satisfaction, Predicting Each Focal Outcome.

	LGBTQ Coefficient <u>without</u> Job Satisfaction control			LGBTQ Coefficient <u>with</u> Job Satisfaction control		
	Coeff		SE	Coeff		SE
Career Opportunities	-.105	***	.027	-.057	*	.025
Sufficient Resources	-.089	**	.033	-.035		.034
Whistleblowing Comfort	-.151	***	.031	-.104	***	.029
Professional Devaluation	.137	***	.023	.093	***	.020
Social Marginalization	.171	***	.025	.132	***	.024
Harassment	.049	**	.018	.042	*	.017
Minor Health Problems	.206	***	.030	.190	***	.029
Insomnia	.246	***	.028	.226	***	.028
Stressed	.260	***	.031	.233	***	.030
Unable to Control	.242	***	.030	.210	***	.029
Difficulties Piling Up	.265	***	.029	.234	***	.028
Thought about Leaving Job	.163	***	.041	.087	*	.036
Plans to leave Occupation	.164	***	.041	.137	**	.044

Notes: N=25,324; \* p<.05; \*\* p<.01; \*\*\* p<.001; two-tailed test. All models include controls for the demographic, STEM field, employment sector, and job controls listed in Table S2, as well as controls for professional society. Logistic regression used to predict dichotomous harassment measure; all other models use OLS regression.



**Table S10.** Effect Sizes of Mean Differences between LGBTQ and Non-LGBTQ Respondents on Outcome Measures

<b>Effect Sizes</b>	
Career Opportunities	.218
Career Resources	.124
Whistleblowing Comfort	.228
Professional Devaluation	.255
Social Marginalization	.250
Harassment	.129
Minor Health Problems	.365
Insomnia	.381
Stressed	.462
Depressive Symptoms	.371
Thought about Leaving	.205
Intentions to Leave Profession	.110

Note: Columns above represent Cohen's *d* effect sizes [ $d$ =difference in means / pooled standard deviation] on differences in means on each inequality measure between LGBTQ and non-LGBTQ respondents.

## REFERENCES AND NOTES

1. NAESM, *Promising Practices for Addressing the Underrepresentation of Women in Science, Engineering, and Medicine: Opening Doors* (National Academies of Sciences Engineering and Medicine, 2020).
2. NAESM, *Sexual Harassment of Women: Climate, Culture, and Consequences in Academic Sciences, Engineering, and Medicine* (National Academies of Sciences Engineering and Medicine, 2018).
3. Y. Tao, Earnings of academic scientists and engineers: intersectionality of gender and race/ethnicity effects. *Am. Behav. Sci.* **62**, 625–644 (2018).
4. S. R. Sommers, On racial diversity and group decision making: Identifying multiple effects of racial composition on jury deliberations. *J. Pers. Soc. Psychol.* **90**, 597–612 (2006).
5. S. S. Levine, E. P. Apfelbaum, M. Bernard, V. L. Bartelt, E. J. Zajac, D. Stark, Ethnic diversity deflates price bubbles. *Proc. Natl. Acad. Sci. U.S.A.* **111**, 18524–18529 (2014).
6. C. Díaz-García, A. González-Moreno, F. Jose Sáez-Martínez, Gender diversity within R&D teams: Its impact on radicalness of innovation. *Innovation* **15**, 149–160 (2013).
7. M. Nathan, N. Lee, Cultural diversity, innovation, and entrepreneurship: Firm-level evidence from London. *Econ. Geogr* **89**, 367–394 (2013).
8. S. E. Page, *The Diversity Bonus: How Great Teams Pay Off in the Knowledge Economy* (Princeton University Press, 2017).
9. B. Hofstra, V. V. Kulkarni, S. Munoz-Najar Galvez, B. He, D. Jurafsky, D. A. McFarland, The diversity-innovation paradox in science. *Proc. Natl. Acad. Sci. U.S.A.* **117**, 9284–9291 (2020).
10. Y. Xie, K. A. Shauman, *Women in Science: Career Processes and Outcomes* (Harvard University Press, 2005).
11. J. S. McIlwee, J. G. Robinson, *Women in Engineering: Gender, Power, and Workplace Culture* (SUNY Press, 1992).
12. L. Smith-Doerr, *Women's Work: Gender Equality vs. Hierarchy in the Life Sciences* (Lynne Reinner Publishers, 2004).
13. R. Eglash, Race, sex, and nerds: From black geeks to asian American hipsters. *Social Text* **20**, 49–64 (2002).
14. J. R. Oberst, in *Science* (American Association for the Advancement of Science, 2010).

15. J. Freeman, in *Nature* (Springer Nature Limited, 2018), vol. 559, pp. 27–28.
16. M. V. L. Badgett, B. Sears, H. Lau, D. Ho, Bias in the workplace: Consistent evidence of sexual orientation and gender identity discrimination 1998–2008. *Chicago-Kent Law Review* **84**, 559–595 (2009).
17. A. Tilcsik, Pride and prejudice: Employment discrimination against openly gay men in the United States. *Am. J. Sociol.* **117**, 586–626 (2011).
18. G. R. Arabsheibani, A. Marin, J. Wadsworth, Variations in gay pay in the USA and UK, in *Sexual orientation discrimination: An international perspective*, L. Badgett, J. Frank, Eds. (Routledge, 2007).
19. J. M. Grant, L. A. Mottet, J. Tanis, J. Harrison, J. L. Herman, M. Keisling, *Injustice at every turn: A report of the national transgender discrimination survey* (National Center for Transgender Equality and National Gay and Lesbian Taskforce, 2011).
20. Supreme Court of the United States, *Bostock v Clayton County Georgia* (590), no. 17–1618 (2020).
21. S. Davidson, Gender inequality: Nonbinary transgender people in the workplace. *Cogent. Soc. Sci.* **2**, 1236511 (2016).
22. E. A. Cech, M. V. Pham, Queer in STEM organizations: Workplace disadvantages for LGBT employees in STEM related federal agencies. *Soc. Sci.* **6**, 12 (2017).
23. E. A. Cech, W. R. Rothwell, LGBT workplace inequality in the federal workforce: Intersectional processes, organizational contexts, and turnover considerations. *ILR Rev.* **73**, 25–60 (2020).
24. B. Sears, C. Mallory, *Documented Evidence of Employment Discrimination & Its Effects on LGBT People* (Williams Institute, 2011).
25. L. Doan, A. Loehr, L. R. Miller, Formal rights and informal privileges for same-sex Couples: Evidence from a National survey experiment. *Am. Sociol. Rev.* **79**, 1172–1195 (2014).
26. T. W. Smith, J. Son, *Trends in Public Attitudes about Sexual Morality* (NORC, 2013).
27. H. Dryburgh, Work Hard, Play hard: Women and professionalization in engineering—Adapting to the culture. *Gender Soc* **13**, 664–682 (1999).
28. A. Abbott, *The System of Professions: An Essay on the Division of Expert Labor* (University of Chicago Press, 1988).
29. Y. Xie, M. Fang, K. Shauman, STEM Education. *Annu. Rev. Sociol.* **41**, 331–357 (2015).

30. S. Shapin, Cordelia's love: Credibility and the Social Studies of Science. *Perspect. Sci.* **3**, 255–275 (1995).
31. E. Cech, The (Mis)Framing of Social Justice: Why Ideologies of Depoliticization and Meritocracy Hinder Engineers' Ability to Think About Social Injustices, in *Engineering Education for Social Justice*, J. C. Lucena, Ed. (Springer, 2013), vol. 10, chap. 4, pp. 67–84.
32. J. Yoder, A. Mattheis, Queer in STEM: Workplace experiences reported in a national survey of LGBTQA individuals in science, technology, engineering, and mathematics careers. *J. Homosex.* **63**, 1–27 (2016).
33. D. Bilimoria, A. J. Stewart, "Don't Ask, Don't Tell": The academic climate for lesbian, gay, bisexual, and transgender faculty in engineering. *NSWA J.* **21**, 85–103 (2009).
34. E. V. Patridge, R. S. Barthelemy, S. R. Rankin, Factors impacting the academic climate for LGBTQ STEM faculty. *J. Women Minorities Sci. Eng.* **20**, 75–98 (2014).
35. B. E. Hughes, Coming out in STEM: Factors affecting retention of sexual minority STEM students. *Sci. Adv.* **4**, eaao6373 (2018).
36. E. A. Cech, T. J. Waidzunus, Navigating the heteronormativity of engineering: The experiences of lesbian, gay, and bisexual students. *Eng. Stud.* **3**, 1–24 (2011).
37. E. A. Cech, W. R. Rothwell, LGBTQ inequality in engineering education. *J. Engr. Educ.* **107**, 583–610 (2018).
38. A. Mattheis, D. C.-R. De Arellano, J. B. Yoder, A model of queer STEM identity in the workplace. *J. Homosex.* **67**, 1839–1863 (2019).
39. W. Faulkner, 'Nuts and Bolts and People': Gender-troubled engineering identities. *Soc. Stud. Sci.* **37**, 331–356 (2007).
40. E. A. Cech, H. M. Sherick, Depoliticization and the Structure of Engineering Education, in *International Perspectives on Engineering Education*, S. H. Christensen, C. Didier, A. Jamison, M. Meganck, C. Mitcham, B. Newberry, Eds. (Springer, 2015).
41. K. Schilt, L. Westbrook, Doing gender, doing heteronormativity: "Gender normals," transgender people, and the social maintenance of heterosexuality. *Gender Soc.* **23**, 440–464 (2009).
42. M. Collier, M. Daniel, The production of trans illegality: Cisnormativity in the U.S. immigration system. *Sociol. Compass* **13**, e12666 (2017).
43. M. P. Miceli, J. P. Near, M. T. Rehg, J. R. Van Scotter, Predicting employee reactions to perceived organizational wrongdoing: Demoralization, justice, proactive personality, and whistle-blowing. *Human Relat.* **65**, 923–954 (2012).

44. E. A. Cech, Ideological wage inequalities? The technical/social dualism and the gender wage gap in engineering. *Soc. Forces* **91**, 1147–1182 (2013).
45. B. K. Attell, K. K. Brown, L. A. Treiber, Workplace bullying, perceived job stressors, and psychological distress: Gender and race differences in the stress process. *Soc. Sci. Res.* **65**, 210–221 (2017).
46. B. F. Reskin, D. B. McBrier, Why not ascription? Organizations' employment of male and female managers. *Am. Sociol. Rev.* **65**, 210–233 (2000).
47. C. R. Waldo, Working in a majority context: A structural model of heterosexism as minority stress in the workplace. *J. Couns. Psychol.* **46**, 218–232 (1999).
48. A. Solazzo, T. N. Brown, B. K. Gorman, State-level climate, anti-discrimination law, and sexual minority health status: An ecological study. *Soc. Sci. Med.* **196**, 158–165 (2018).
49. L. Turk-Bicakci, A. Berger, *Leaving STEM: STEM Ph.D. Holders in Non-STEM Careers* (American Institutes for Research, 2014).
50. R. P. Steel, N. K. Ovalle, A review and meta-analysis of research on the relationship between behavioral intentions and employee turnover. *J. Appl. Psychol.* **69**, 673–686 (1984).
51. P. H. Collins, *Black Feminist Thought* (Unwin Hyman, 1990).
52. K. Crenshaw, Mapping the margins: Intersectionality, identity politics, and violence against women of color. *Stanford Law Rev.* **43**, 1241–1299 (1991).
53. M. J. Finkel, R. D. Storaasli, A. Bandele, V. Schaefer, Diversity training in graduate school: An exploratory evaluation of the Safe Zone project. *Prof. Psychol. Res. Pract.* **34**, 555–561 (2003).
54. M. A. Whooley, A. L. Avins, J. Miranda, W. S. Browner, Case-finding instruments for depression: Two questions are as good as many. *J. Gen. Intern. Med.* **12**, 439–445 (1997).
55. I. H. Settles, L. M. Cortina, N. T. Buchanan, K. N. Miner, Derogation, discrimination, and (dis)satisfaction with jobs in science: A gendered analysis. *Psychol. Women Q* **37**, 179–191 (2012).
56. R. M. Guion, *Assessment, Measurement, and Prediction for Personnel Decisions* (Routledge, 2011).
57. M. S. Litwin, *How to Measure Survey Reliability and Validity* (Sage, 1995), vol. 7.
58. C. Van Mol, Improving web survey efficiency: the impact of an extra reminder and reminder content on web survey response. *Int. J. Soc. Res. Method* **20**, 317–327 (2017).