

# **Women are underrepresented and receive differential outcomes at ASM journals: A six-year retrospective analysis**

Running title: A six-year retrospective analysis of ASM journal outcomes

Ada K. Hagan<sup>\*1</sup>, Begüm D. Topçuoğlu<sup>1</sup>, Mia E. Gregory<sup>1</sup>, Hazel A. Barton<sup>2</sup>, Patrick D. Schloss<sup>1†</sup>

† To whom correspondence should be addressed: pschloss@umich.edu

\*Present address: Alliance SciComm & Consulting, LLC, Linden, MI

1. Department of Microbiology and Immunology, University of Michigan, Ann Arbor, MI
2. Department of Biology, University of Akron, Akron, OH

**1 Abstract**

2 Despite 50% of biology Ph.D. graduates being women, the number of women that advance in  
3 academia decreases at each level (e.g. from graduate to post-doctorate to tenure-track). Recently,  
4 scientific societies and publishers have begun examining internal submissions data to evaluate  
5 representation and evaluation of women in their peer review processes; however, representation  
6 and attitudes differ by scientific field and no studies to-date have investigated academic publishing  
7 in the field of microbiology. Using manuscripts submitted between January 2012 and August 2018  
8 to the 15 journals published by the American Society for Microbiology (ASM), we describe the  
9 representation of women at ASM journals and the outcomes of their manuscripts. Senior women  
10 authors at ASM journals were underrepresented compared to global and society estimates of  
11 microbiology researchers. Additionally, manuscripts submitted by corresponding authors that were  
12 women received more negative outcomes than those submitted by men. These negative outcomes  
13 were somewhat mediated by whether or not the corresponding author was based in the US, and by  
14 the type of institution for US-based authors. Nonetheless, the pattern for women corresponding  
15 authors to receive more negative outcomes on their submitted manuscripts held. We conclude with  
16 suggestions to improve the representation of and decrease structural penalties against women.

**17 Importance**

18 Barriers in science and academia have prevented women from becoming researchers and experts  
19 that are viewed as equivalent to their colleagues who are men. We evaluated the participation and  
20 success of women researchers at ASM journals to better understand their success in the field of  
21 microbiology. We found that women are underrepresented as expert scientists at ASM journals.  
22 This is, in part, due to a combination of both low submissions from senior women authors and more  
23 negative outcomes on submitted manuscripts for women compared to men.

**24 Introduction**

25 Evidence has accumulated over the decades that academic research has a representation problem.  
26 While at least 50% of biology Ph.D. graduates are women, the number of women in postdoctoral  
27 positions and tenure-track positions are less than 40 and 30%, respectively (1). There have been

28 many proposed reasons for these disparities, which include biases in training and hiring, the impact  
29 of children on career trajectories, a lack of support for primary caregivers, a lack of recognition, lower  
30 perceived competency, and less productivity as measured by research publications (1–8). These  
31 issues do not act independent of one another, instead they accumulate for both individuals and the  
32 community, much as advantages do (9–11). Accordingly, addressing these issues necessitates  
33 multi-level approaches from all institutions and members of the scientific community.

34 Scientific societies play an integral role in the formation and maintenance of scientific  
35 communities—they host conferences that provide forums for knowledge exchange, networking,  
36 and opportunities for increased visibility as a researcher. Scientific societies also frequently  
37 publish the most reputable journals in their field, facilitating the peer review process to vet new  
38 research submissions (12). Recently, scientific societies and publishers have begun examining  
39 internal submissions data to evaluate representation of and bias against women in their peer  
40 review processes. The American Geological Union found that while the acceptance rate of  
41 women-authored publications was greater than publications authored by men, women submitted  
42 fewer manuscripts than men and were used as reviewers only 20% of the time (13), a factor that  
43 is reported to be influenced by the gender of the editor (14). Several studies have concluded  
44 that there is no significant bias against papers authored by women (14–19). Recent reports of  
45 manuscript outcomes at publishers for ecology and evolution, physics, and chemistry journals have  
46 found that women-authored papers are less likely to have positive peer reviews and outcomes  
47 (20–23).

48 The representation of women scientists and gender attitudes differ by scientific field and no studies  
49 to-date have investigated academic publishing in the field of microbiology. The American Society  
50 for Microbiology (ASM) is one of the largest life science societies, with an average membership  
51 of 41,000 since 1990. A recent statement notes that “A diverse ASM enhances the microbial  
52 sciences, increases innovation, strengthens the community and sustains the profession” and  
53 pledges to “address all members’ needs through development and assessment of programs  
54 and services” that aims to ensure “equitable access and accountability through transparent  
55 procedures and communication” (24). One of the ASM’s services is the publication of microbiology  
56 research through a suite of research and review journals. Between January 2012 and August

57 2018, ASM published 25,818 original research papers across 15 different journals: *Antimicrobial*  
58 *Agents and Chemotherapy* (AAC), *Applied and Environmental Microbiology* (AEM), *Clinical*  
59 *and Vaccine Immunology* (CVI), *Clinical Microbiology Reviews* (CMR), *Eukaryotic Cell* (EC),  
60 *Infection and Immunity* (IAI), *Journal of Bacteriology* (JB), *Journal of Clinical Microbiology* (JCM),  
61 *Journal of Virology* (JVI), *mBio*, *Microbiology and Molecular Biology Reviews* (MMBR), *Genome*  
62 *Announcements* (GA, now *Microbiology Resource Announcements*), *Molecular and Cellular Biology*  
63 (MCB), *mSphere*, and *mSystems*. Two journals, EC and CVI, were retired during the period under  
64 study and three journals, GA/MRA, MMBR, and CMR, were excluded from the analysis due to  
65 their relatively low number of submissions. The goal of our research study was to describe the  
66 population of the ASM journals both through the gender-based representation of authors, reviewers,  
67 and editors and the associated peer review outcomes.

## 68 Results

69 Over 100,000 manuscript records were obtained for the period between January 2012 and August  
70 2018 (Fig. 1). Each of these were evaluated by editors and some by reviewers, leading to multiple  
71 possible outcomes. At the ASM journals, manuscripts may be immediately rejected by editors  
72 instead of being sent to peer review, often due to issues of scope or quality. These were defined  
73 as editorial rejections and identified as manuscripts rejected without review. Alternately, editors  
74 send a majority of manuscripts out for review by two or more experts in the field selected from  
75 a list of potential reviewers suggested by the authors and/or editors. Reviewers give feedback  
76 to the authors and editor, who decides whether the manuscript in question should be accepted,  
77 rejected, or sent back for revision. Manuscripts with suggested revisions that are expected to take  
78 more than 30 days to address are rejected, but generally encouraged to resubmit. If resubmitted,  
79 the authors are asked to note the previous manuscript and the re-submission is assigned a new  
80 manuscript number. Multiple related manuscripts were tracked together by generating a unique  
81 grouped manuscript number based on the recorded related manuscript numbers. This grouped  
82 manuscript number served dual purposes of tracking a single manuscript through multiple rejections  
83 and avoiding duplicate counts of authors for a single manuscript. After eliminating non-primary  
84 research manuscripts and linking records for resubmitted manuscripts, we identified 79,189 unique

85 manuscripts (Fig. 1).

86 We inferred gender for both the peer review participants (e.g., editor-in-chief, editors, reviewers)  
87 and authors on the manuscripts evaluated during this time period using a social media-informed  
88 classification algorithm with stringent criteria and validation process (Supp Text, Fig. S1). We  
89 recognize that biological sex (male/female) is not always equivalent to the gender that an individual  
90 presents as (man/woman), which is also distinct from the gender(s) that an individual may  
91 self-identify as. For the purposes of this manuscript, we choose to focus on the presenting  
92 gender based on first names (and appearance for editors), as this information is what reviewers  
93 and editors also have available. The sensitivity, specificity, and accuracy of our method were  
94 0.97 (maximum of 1.0) when validated against a curated set of authors (Table S1). The accuracy  
95 was 0.99 when applied to the list of editors, whose inferred genders were validated by hand  
96 using Google (Supp Text). In addition to identifying journal participants as men or women, this  
97 method of gender inference resulted in a category of individuals whose gender could not be reliably  
98 inferred (i.e., unknown). We included those individuals whose names did not allow a high degree of  
99 confidence for gender inference in the “unknown” category of our analysis, which is shown in many  
100 of the plots depicting representation of the population. These individuals were not included in the  
101 comparison of manuscript outcomes. Finally, we refer to editors and peer reviewers collectively  
102 as gatekeepers, which describes and recognizes their essential role in maintaining the scientific  
103 quality of manuscripts accepted (or rejected) at peer reviewed journals (25, 26).

104 **Men dominated as gatekeepers and senior authors.** We first evaluated the representation  
105 of men and women who were gatekeepers during the study period. Each journal is led by an  
106 editor-in-chief (EIC) who manages journal scope and quality standards through a board of editors  
107 with field expertise that, in turn, handle the peer review process. There were 17 EICs, 17.6% of  
108 which were women. Four years before retirement, the EIC of CVI transferred from a man to a  
109 woman, while JVI has had a woman as EIC since 2012. The total number of editors at all ASM  
110 journals combined over the duration of our study (senior editors and editors pooled) was 1015,  
111 28.8% of which were women.

112 Over 40% of both men and women editors were from US-based R1 institutions, defined as

113 doctoral-granting universities with very high research activity (27). Non-US institutions and U.S.  
114 medical schools or research institutions supplied the next largest proportions of editors (Fig. 2A)(27).  
115 Since 2012, there was a slow trend toward equivalent gender representation among editors (Fig.  
116 2B). Individual journal trends varied considerably, though most had slow trends toward parity (Fig.  
117 2C). CVI and *mSphere* were the only ASM journals to have accomplished equivalent representation  
118 of men and women, with CVI having a greater proportion of women editors than men before it was  
119 retired. EC was the only journal with an increasing parity gap.

120 Altogether, 30439 reviewers submitted reviews and 24.6% were inferred to be women (using our  
121 algorithm with 0.97-0.99 accuracy). The greatest proportion of reviewers (over 50% of all groups)  
122 came from non-US institutions, while R1 institutions supplied the next largest cohort of reviewers  
123 (Fig. 2D). The proportions of each gender group were consistent over time among reviewers at  
124 the ASM journals (Fig. 2E) and were representative of both the suggested reviewers at all journals  
125 combined, and the actual reviewer proportions at most journals (Fig. S2).

126 **Editorial workloads were not proportionate.** To evaluate the editorial workload for each gender,  
127 we calculated the proportion of manuscripts handled by editors of each gender (excluding editorial  
128 rejections), relative to their representation. If the workload is proportionate, then the workload for  
129 each gender will be equivalent to the gender's representation at that journal. Across all of the  
130 journals combined, men handled a slightly greater proportion of manuscripts than women, relative  
131 to their respective editorial representations (Fig. 3A). This trend was present at most journals with  
132 varying degrees of difference between workload and representation (Fig. 2C). For instance, at  
133 *mSphere*, both workload and representation were identical; however, CVI, *mBio*, and JVI each  
134 had periods at which the workload for women editors was much higher than their representation,  
135 with corresponding decreases in the workload of men. In the years preceding its retirement, the  
136 representation of women at CVI increased, decreasing the gap in editorial workload. However,  
137 representation and relative workloads for men and women editors at JVI held steady over time,  
138 while the proportional workload for women at *mBio* has increased.

139 The median number of manuscripts reviewed by men, women, and unknown gendered individuals  
140 was 2, for each group. Half of those in the men, women, or unknown gender groups reviewed

141 between one and 5, 4, or 3 manuscripts each, respectively (Fig. 3B). Conversely, 44.6% of men,  
142 40.1% of women, and 48.6% of unknown gendered reviewers reviewed only one manuscript,  
143 suggesting that women were more likely than other groups to review multiple manuscripts.  
144 Reviewers of all gender groups accepted fewer requests to review from women editors (average of  
145 47.8%) than from men (average of 53.3%; Fig. 3C). Reviewers were also less likely to respond to  
146 women editors than men (no response rate averages of 25.1 and 19.9%, respectively). Both men  
147 and women editors contacted reviewers from all three gender groups in similar proportions, with  
148 women editors contacting 76.4% of suggested reviewers and men contacting 74.1% (median of the  
149 percent contacted from each gender group).

150 **Women were underrepresented as authors.** Globally, microbiology researchers are 60% men  
151 and 40% women (28). In September 2018, 38.4% of ASM members who reported their gender  
152 were women. We wanted to determine if these proportions were similar for senior authors at the  
153 journals and to understand the distribution of each gender group among submitted manuscripts and  
154 published papers. We began by describing senior author (last/corresponding author) institutions  
155 by gender group. Over 60% of submitting senior authors were from non-US institutions, followed  
156 by about 20% from R1 institutions. The proportion of manuscripts submitted from US institutions  
157 by women senior authors was 31% versus 36% from women who were senior authors at non-US  
158 institutions. Women senior authors were more highly represented at low research universities and  
159 federal research institutions than at any other US-based institution (Fig. 4A). The proportions of all  
160 men and women (senior and co-) authors at the ASM journals decreased over time at equivalent  
161 rates, while the proportion of unknown gendered authors increased; the ratio of men to women  
162 authors was 4 to 3 (i.e., 57% men; Fig. 4B).

163 In the field of microbiology, order of authorship on a manuscript signals the type and magnitude  
164 of contributions to the finished product. First and last authorship are the most prestigious. First  
165 authors are generally trainees (e.g., students or post-docs) or early career researchers responsible  
166 for performing the bulk of the project, while last authors are generally lead investigators that  
167 supplied conceptual guidance and resources to complete the project. Middle authors are generally  
168 responsible for technical analyses and methods. Any author can also be a corresponding author,  
169 which we identified as the individual responsible for communicating with publishing staff during peer

170 review (as opposed to an author to whom readers direct questions), of which there can be multiple.

171 The proportion of manuscripts submitted with men or women as first authors remained constant at  
172 29.1 and 30.7%, respectively (Fig. 4C, dashed). The proportions of first author published papers  
173 were nearly identical at 33.1% for men and 33.8% for women (Fig. 4C, solid). The proportion of  
174 submitted manuscripts with men corresponding authors remained steady at an average of 41.6%  
175 and the proportion with women corresponding authors was 23.4% (Fig. 4D, dashed); the proportion  
176 of published unknown gender authors declined. Both men and women corresponding authors had  
177 a greater proportion of papers published than manuscripts submitted. Accordingly, manuscripts  
178 with corresponding authors of unknown gender were rejected at a higher rate than their submission.  
179 The difference between the percent of submitted manuscripts and published papers was 8.2%  
180 when men were corresponding authors, but only 0.9% when women were corresponding authors,  
181 making the submitted to published difference near equal (Fig. 4C, solid). This trend was similar for  
182 middle and last authors (Fig. S3).

183 Of the 38594 multi-author manuscripts submitted by men corresponding authors, 23.5% had  
184 zero authors inferred to be women. In contrast, 7253 (36.3%) of the manuscripts submitted by  
185 women corresponding authors had more than half of the authors inferred as women, exceeding  
186 those submitted by men corresponding authors in both the number (3247) and percent (8.4) of  
187 submissions. Additionally, the proportion of women authors decreased as the number of authors  
188 increased, such that when the number of authors exceeded 30 on a manuscript (N=59), the  
189 proportion of individuals inferred to be women was always below 51% (Fig. S4). Men submitted  
190 225 single-authored manuscripts while women submitted 69 single-authored manuscripts.

191 We hypothesized that we would be able to predict the inferred gender of the corresponding author  
192 using a logistic regression model trained on the following variables: whether the corresponding  
193 author's institution was in the U.S., the total number of authors, the proportion of authors that were  
194 women, whether the paper was published, the gender of senior editors and editors, the number  
195 of revisions, and whether the manuscript was editorially rejected at any point. We measured the  
196 model's performance using the area under the receiver operating characteristic curve (AUROC). The  
197 AUROC value is a predictive performance metric that ranges from 0.0, where the model's predictions

198 are completely wrong, to 1.0, where the model distinguishes perfectly between outcomes. A value  
199 of 0.5 indicates that the model did not perform better than a random assignment. The median  
200 AUROC value of our model to predict the corresponding author's inferred gender was 0.7 (Fig.  
201 S5A, panel A). The variable with the largest absolute weight (i.e., the most predictive value), in our  
202 model was the proportion of women authors (Fig. S5C). These results indicate that manuscript  
203 submission data was capable of predicting the inferred gender of the corresponding author, but that  
204 the prediction was primarily driven by the percentage of authors that were inferred to be women.

205 As described above, first authors were slightly more likely to be women (30.7%W vs 29.1%M), but  
206 corresponding authors were significantly more likely to be men (23.44%W vs 41.59%M). A concern  
207 is that if authors are not retained to transition from junior to senior status, they will be left out of  
208 the gatekeeping roles. Since authorship conventions indicate that last and corresponding authors  
209 are typically senior authors, we combined both first and middle authors into the "junior" author role  
210 and tracked individuals through the possible roles at the ASM journals. There were 75451 women  
211 who participated as junior authors (first/middle) at the ASM journals. Of those junior authors who  
212 were women, 8.2% also participated as senior authors (last/corresponding), 8.9% were potential  
213 reviewers and 5.4% participated as reviewers. 0.2% of women junior authors became editors  
214 at the ASM journals over the 6 year period studied. For men, there were a total of 83727 junior  
215 authors, where 13.6% also participated as senior authors, 16.7% were potential reviewers, and  
216 11.1% actually reviewed. 0.7% of men junior authors became editors at the ASM journals. Overall,  
217 women who participated at the ASM journals as junior authors were half as likely to move to senior  
218 author or reviewer roles, and 30% as likely to be an editor than men at the ASM journals.

219 **Manuscripts submitted by women have more negative outcomes than those submitted by**  
220 **men.** To further investigate the difference in percents of published and submitted proportions  
221 for men and women authors (Fig. 4CD, Fig. S3), we compared the rejection rates of men and  
222 women at each author stage (first, middle, corresponding, and last). To more easily visualize and  
223 understand the differences in outcomes according to author gender, we calculated the outcome  
224 rate for each gender then subtracted the rate for women from men to generate the percentage point  
225 difference. To correct for the disparity in participation by women compared to men, all percentage  
226 point comparisons were made relative to the gender and population in question. Where men

227 received an outcome more often than women, the value is positive (blue) while a negative value  
228 (orange) indicates that women outperformed men in the given metric. For the following analyses,  
229 only manuscripts authored by an individual inferred to be a man or woman were included. Finally,  
230 these analyses were conducted on all available manuscripts, not a statistical sampling. As a result,  
231 statistical tests were only required for correlative analyses.

232 Middle authors were rejected at equivalent rates for men and women (a 0.23 percentage point  
233 difference across all journals). However, manuscripts with senior women authors were rejected  
234 more frequently than those authored by men with -6.7 and -6.0 percentage point differences for  
235 corresponding and last authors, respectively (Fig. 5A, vertical lines). The overall trend of increased  
236 rejection for women was most pronounced at MCB, JB, IAI and AAC. The greatest differences  
237 were observed when comparing the outcome of corresponding authors by gender, so we used  
238 this sub-population to further examine the difference in manuscript acceptance and rejection rates  
239 between men and women.

240 We next compared the rejection rates for men and women corresponding authors after two review  
241 points, initial editor review and the first round of peer review. Manuscripts authored by women  
242 were editorially rejected by as much as 12 percentage points more often than those authored by  
243 men (Fig. 5B). The difference at all of the ASM journals combined was -3.8 percentage points  
244 (vertical line). MCB and *mBio* had the most extreme percentage point differences. Manuscripts  
245 authored by men and women were equally likely to be accepted after the first round of review (Fig.  
246 5C, right panel). However, women-authored papers were rejected (left panel) more often while  
247 men-authored papers were more often given revision (center panel) decisions. The differences for  
248 rejection and revision decisions after review were -5.6 and 5.6 percentage points, respectively (Fig.  
249 5C, vertical lines). JB, AAC, and MCB had the most extreme differences for rejection and revision  
250 decisions. Percentage point differences were not correlated with journal prestige as measured by  
251 2018 impact factors ( $R^2 = -0.022$ ,  $P = 0.787$ ).

252 In addition to manuscript decisions, other disparate outcomes may occur during the peer review  
253 process (29). To determine whether accepted women-authored manuscripts spent more time  
254 between being submitted and being ready for publication, we compared the number of revisions,

255 days spent in the ASM peer review system, and the number of days between submission and  
256 being ready for publication to those authored by men. Manuscripts authored by women took slightly  
257 longer (from submission to ready for publication) to complete than those by men at some journals  
258 (*mSphere*, *mBio*, *mSystems*, CVI, JB, JCM, AEM) despite spending similar amounts of time in the  
259 ASM journal peer review system (Fig. S6), and having the same median number of revisions prior  
260 to acceptance (Median = 2, IQR = 0).

261 To understand how a gatekeeper's (editor/reviewer) gender interacted with decision types (e.g.,  
262 Fig. 5C), we grouped editor decisions and reviewer suggestions according to the gatekeeper's  
263 inferred gender (unknowns excluded). Both men and women editors rejected proportionally  
264 more women-authored papers, however the percentage point difference in decisions were  
265 slightly larger for men-edited manuscripts (Fig. 6A). Reviewers were more likely to suggest  
266 rejection for women-authored manuscripts as compared to men and a minimal difference in  
267 revise recommendations was observed (Fig. 6B). Both men and women reviewers recommended  
268 rejection more often for women-authored manuscripts although men recommended acceptance  
269 and revision more frequently for men-authored manuscripts than women did (Fig. 6C).

270 To evaluate if inferred gender played a role in manuscript editorial decisions, we trained a logistic  
271 regression model to predict whether a manuscript was reviewed (i.e., editorially rejected or not).  
272 We used the inferred genders of the senior editor, editor, and corresponding author, as well as the  
273 proportion of authors that were women as variables to train the model (Fig. S5B). The median  
274 AUROC value was 0.61 (Fig. S5A, panel B), which indicated that editorial decisions were not  
275 random, however, the relatively low AUROC value indicated that there are factors not included in  
276 our model that influence editorial decisions.

277 **Multiple factors contribute to the overperformance of men.** The association between inferred  
278 gender and manuscript decision could be attributed to implicit gender bias by journal gatekeepers,  
279 however, there are other types of bias that may contribute to, or obscure, gender bias; for  
280 instance, a recent evaluation of peer-review outcomes at *eLife* found evidence of preference  
281 for research submitted by authors from a gatekeeper's own country or region (20). Other studies  
282 have documented prestige bias, where men are over-represented in more prestigious (i.e., more

283 respected and selective) programs (30). It is therefore possible, that what seems to be gender bias  
284 could be geographic or prestige bias interacting with the increased proportion of women submitting  
285 from outside the US or from lower prestige institutions (e.g., the highest rate of submissions from  
286 women were at low research institutions, 37%; Fig. 4A).

287 To quantify how these factors affected manuscript decisions, we next looked at the outcome of  
288 manuscripts submitted only by corresponding authors at US institutions, because these institutions  
289 represented the majority of manuscripts and could be classified by using the Carnegie Classification  
290 of Institutions of Higher Education (TM) (27). We used the same strategy as described above.  
291 When only considering US-based authors, the percentage point difference for editorial rejections  
292 increased from -3.8 to -1.4 percentage points (Fig. 7A). The trend of percentage point difference  
293 in decisions after review for US-based authors mirrored those seen for all corresponding authors  
294 at the journal level (Fig. 7B). The over-representation of women in rejection decisions increased  
295 from -5.6 to -4.4 percentage points, and the over-representation of men in revise only decisions  
296 decreased from 5.6 to 4.2 (Fig. 7B). The difference in the rate of accept decisions changed from  
297 -1.4 to 0.2 percentage points after restricting the analysis to US-based authors. These results  
298 suggest that the country of origin (i.e., US versus not) accounted for some of the differences in  
299 outcomes by inferred gender, particularly for editorial rejections.

300 To address institution-based prestige bias, we split the US-based corresponding authors according  
301 to the type of institution they were affiliated with (based on the Carnegie classification) and  
302 re-evaluated the differences for men and women (27). Editorial rejections occurred most often for  
303 women from medical schools or institutes, followed by those from R2 institutions: 32% and 28%  
304 of manuscripts from each institution were submitted by women, respectively (Fig. 7C, Fig. S7A).  
305 This percentage point difference in the editorial rejections of corresponding authors from medical  
306 schools or institutes was spread across most of the ASM journals, while the editorial rejection of  
307 papers submitted from women at R2 institutions was driven primarily by submissions to JCM (Fig.  
308 S7A). Evaluating the percentage point difference in acceptance rates by institution and inferred  
309 gender mirrored that of editorial rejections for some journals, where submissions from men received  
310 better outcomes than submissions from women (Fig. 7CD and S7BC). For instance, manuscripts  
311 submitted by men from medical schools or institutes were accepted up to 10 percentage points

312 more often than those submitted by women (Fig. 7C).

313 To evaluate if these factors affect manuscript decisions, we trained a logistic regression model  
314 to predict whether a manuscript was editorially rejected using the variables: origin (US vs non),  
315 institution (US institution type), number of authors, proportion of authors that were women, and the  
316 inferred genders of both gatekeepers and corresponding authors. The model had a median AUROC  
317 value of 0.67 (Fig. S5A, panel C), which indicated a non-random interaction between these factors  
318 and editorial decisions. Manuscripts from authors at U.S. “other” institutions, men EICs, men that  
319 were corresponding authors from “other” U.S. institutions, and women from medical schools and  
320 institutes were all more associated with editorial rejections (Fig. S7D). Conversely, manuscripts  
321 from R1 institutions, authors from the U.S., EICs that were women, and the number of authors were  
322 all more likely to be associated with review (Fig. S7D). These results confirm that the country of  
323 origin and class of institution impact decisions in a non-random manner, though not as much as  
324 gender.

325 A final factor we considered was whether the type of research pursued by men as opposed  
326 to women may impact manuscript outcomes. Black women philosophers and physicists have  
327 described the devaluation of non-traditional sub-disciplines in their fields (31–33). This concept  
328 originally described bias against Black women—the intersection of two historically marginalized  
329 identities. However, the idea that researchers in an established core field might be skeptical of less  
330 established, or non-traditional, sub-field research likely applies elsewhere. The disparate outcomes  
331 of sub-fields in a gendered context has recently been observed in the biomedical sciences, where  
332 NIH proposals focusing on womens’ reproductive health were the least likely to be funded (34).  
333 To explore this phenomenon in the ASM journals, we looked at the editorial rejection rates of  
334 manuscripts (regardless of origin or institution) for each research category at the five largest ASM  
335 journals: AAC, AEM, IAI, JVI, and JCM. Together, these journals account for 47% of the manuscripts  
336 analyzed in this study and comprise 55 categories.

337 The number of submissions in each category ranged from 1 (“FDA Approval” at AAC) to 2952  
338 (“Bacteriology” at JCM) while the acceptance rates varied from 29.4% (“Chemistry:Biosynthesis”  
339 at AAC) to 71.3% (“Structure and Assembly” at JVI) (Table 1). We argued that the number of

340 submissions to each category could help indicate core versus periphery subfields, (i.e., core  
341 subfields would have more submissions than periphery subfields) and based on the literature  
342 to-date, we expected that periphery subfields might have a higher participation of women (31–33).  
343 Women submitted on average 35.3% of the manuscripts to each category, ranging from 20% to  
344 86% (Table 1). There was not a correlation between the proportion of women authors and the  
345 number of submissions ( $R^2 = -0.0177$ ,  $P = 0.779$ ) to each category. Nor was there a correlation  
346 between the proportion of women authors and the category acceptance rate ( $R^2 = 0.041$ ,  $P =$   
347 0.078). These data suggest that there was not a relationship between the participation of women  
348 and either the number of submissions or the acceptance rate of categories in our dataset.

349 We next looked at the percentage point differences in performance for men and women in each  
350 category at two decision points: editorial rejection and rejection after the first review. Each journal  
351 focuses on a different facet of microbiology or immunology, making the results difficult to compare  
352 directly. However, the pattern of increased rejection rates for women was maintained across  
353 most categories with some displaying major differences in gendered performance (Fig. S8). For  
354 instance, the “Biologic Response Modifier” (e.g., immunotherapy) sub-category at AAC, had extreme  
355 differences for both editorial rejections and rejections after review, about -30 and -40 percentage  
356 points, respectively. While that category had a relatively low number of submissions ( $N = 44$ ), 43%  
357 were from women (Fig. S8A). “Mycology”, was a category at two journals, AEM and JCM. At both  
358 journals, men received favorable outcomes relative to women in this category. At AEM, there were  
359 73 “Mycology” submissions, 44% from women authors with a percentage point difference of almost  
360 -20 for editorial rejection outcomes and -10 for rejections after review (Fig. S8B). JCM had 587  
361 “Mycology” submissions with a submission rate of 39% from women authors (Fig. S8D). Differences  
362 between outcomes were almost -10 for editorial rejections and -12 for rejections after review at  
363 JCM.

364 Because of these extreme percentage point differences in categories with high women authorship,  
365 we next asked if the number of women participating in a particular category was related to  
366 manuscript outcomes. There was no correlation between the difference in editorial rejection  
367 by category and the percent of women that were either authors ( $R^2 = -0.003$ ,  $P = 0.363$ ) or  
368 editors ( $R^2 = -0.018$ ,  $P = 0.765$ ). The percent of women authors and percent of women editors

369 in journal categories did not correlate either ( $R^2 = -0.007$ ,  $P = 0.682$ ), which is likely related  
370 to the underrepresentation of women editors in categories dominated by women authors (e.g.,  
371 "Epidemiology"). These data suggest the possibility of persistent negative outcomes against women  
372 in particular fields (e.g., "Mycology"), though it does not seem to relate to either the number of  
373 submissions or participation of women in those subfields.

374 **Discussion**

375 We described the representation of inferred men and women participating in the submission  
376 and peer review process at the ASM journals between January 2012 and August 2018 and  
377 compared editorial outcomes according to the authors' inferred gender. Women were consistently  
378 under-represented (30% or less in all levels of the peer review process) excluding first authors,  
379 where women represented about 50% of authors where we could infer a gender (Figs. 2 and 4).  
380 Women and men editors had proportionate workloads across all of the ASM journals combined, but  
381 those workloads were disproportionate at the journal level and the overburdened gender varied  
382 by journal (Figs. 2 and 3). Additionally, manuscripts submitted by women corresponding authors  
383 received more negative outcomes (e.g., editorial rejections) than those submitted by men (Figs. 5  
384 and 6). These negative outcomes were somewhat mediated by whether the corresponding author  
385 was based in the US, the type of institution for US-based authors, and the research category (Figs.  
386 7 and S8). However, the trend for women corresponding authors to receive more negative outcomes  
387 held across all analyses, indicating a pattern of gender-influenced editorial decisions regardless of  
388 journal prestige (as determined by impact factor). Together, these data indicate a persistent penalty  
389 for senior women microbiologists who participate at the ASM journals.

390 The proportion of women as first authors was higher than data obtained globally and from  
391 self-reported ASM membership data, which was higher than the proportion of senior women  
392 authors at the ASM journals. Only half as many women who were junior authors at the ASM  
393 journals were also senior authors when compared to men. The representation of women decreased  
394 as the prestige (e.g., reviewer, editor) increased. These trends are consistent with the representation  
395 of senior women in academic biological sciences, the observation that women are more likely to  
396 leave academia during the transition from postdoc (junior) to investigator (senior), and mentorship

397 bias (35, 36). These data indicate that microbiology (as represented by the ASM journals) is not  
398 exempt from the issues that limit the retention of women through academic ranks.

399 How to define representation and determine what the leadership should look like are recurring  
400 questions in STEM. Ideally, the representation for men and women corresponding authors,  
401 reviewers, and editors would reflect the number of Ph.D.s awarded (about 50% each, when  
402 considered on a binary spectrum). We argue that the goal should depend on the workload and  
403 visibility of the position. Since high visibility positions (e.g., editor, EIC) are filled by a smaller  
404 number of individuals that are responsible for recruiting more individuals into leadership, filling  
405 these positions should be done aspirationally (i.e., 50% should be women if the goal were an  
406 aspirational leadership). This allows greater visibility for women as experts, expansion of the  
407 potential reviewer network, and recruitment into those positions (37–39). Conversely, lower visibility  
408 positions (e.g., reviewers) require effort from a greater number of individuals and should thus  
409 be representational of the field to avoid overburdening the minority population (i.e., since 23.5%  
410 of corresponding authors at the ASM journals are women, then 20–25% of reviewers should be  
411 women). Balancing the workload is particularly important given the literature indicating that women  
412 faculty have higher institutional service loads than their counterparts who are men (40).

413 In contrast to institutional service, the editing workload at the ASM journals seems to be  
414 predominantly borne by men (Fig. S1). A possible explanation for the difference in gatekeeper  
415 representation and editor workloads is that women are more likely to study non-traditional  
416 sub-disciplines (31–33). Their separation from the traditional center of a field decreases their  
417 perceived competency, which could result in research typecasting and lower manuscript handling  
418 responsibilities (6, 31–33). However, our data could not confirm this phenomenon at the ASM  
419 journals. Another possibility is the increased proportion of potential reviewers that either did not  
420 accept, or did not respond to, requests to review from women editors (Fig. 3C). This increases the  
421 proportion of reviewers that women editors must contact, adding additional time and work to their  
422 editorial burdens, thus making them seem less efficient (i.e., less capable) than men editors. Three  
423 journals, *mBio*, CVI, and JVI were exceptions with regards to editorial workloads (Fig. S1). At these  
424 journals, the editorial workloads of women exceeded their representation. A possible explanation  
425 for CVI and JVI is that both of these journals have been led by women EICs. The tendency for

426 reviewers to reject requests to review from women editors, may also extend to editors; men editors  
427 may be more likely to reject requests to handle manuscripts from women EICs. Our data differ from  
428 those of Fox, Burns, and Meyer who found that the gender of the editor influenced the gender of  
429 the contacted reviewers (14), but supports findings that women editors contact more reviewers  
430 than men (41).

431 Our data also revealed some disturbing patterns in gendered authorship that have implications for  
432 the retention of women microbiologists. Previous research suggests that women who collaborate  
433 with other women receive less credit for these publications than when they collaborate with men  
434 (42), and that women are more likely to yield corresponding authorship to colleagues that are men  
435 (21). In our linear regression models, the number of authors on a manuscript was the largest  
436 contributor to avoiding editorial rejections, suggesting that highly collaborative research is preferred  
437 by editors (43). This observation was supported by the positive correlation between citations  
438 and author count (Fig. S7). Thus it concerns us that when the number of authors exceeded 30  
439 on a manuscript (N=59), the proportion of individuals inferred to be women was always below  
440 51%, despite equivalent numbers of trainees in the biological sciences (Fig. S4). And while  
441 women corresponding authors submitted fewer manuscripts, more of them (both numerically and  
442 proportionally), had a majority of co-authors inferred to be women, compared to those submitted by  
443 men corresponding authors. These data support previous findings that women are more likely to  
444 collaborate with other women (23, 44–46). Additionally, the proportion of women authors was the  
445 greatest predictor of corresponding author gender. This gender-based segregation of collaborations  
446 at the ASM journals likely has had consequences in pay and promotion for women microbiologists  
447 and could be a factor in the decreased retention of senior women. We predict that the low retention  
448 is aggravated by the under representation of women as corresponding authors, which also has  
449 negative consequences for both their careers and microbiology. Since senior authorships impact  
450 status, visibility, and salary, the under representation of women as senior authors and reviewers  
451 likely hampers their career progression and desire to progress (18, 47). The retention of women  
452 (and other marginalized groups) is important to the progress of microbiology since less diversity in  
453 science limits the diversity of perspectives and approaches, thus stunting the search for knowledge.

454 Whether academic research journals support women has been the topic of many papers, which

455 note the lack of women authors publishing relative to men in high impact journals (48–54). However,  
456 submissions data are required to determine if the lack of representation is due to low submissions  
457 or bias during peer review. Using such data, we have shown that there is a disparity in submissions  
458 from senior women in microbiology compared to men, but this does not fully account for the  
459 difference in publications by men and women corresponding authors at the ASM journals (Fig.  
460 4). When examining manuscript outcomes, we found a consistent trend of positive outcomes for  
461 manuscripts submitted by men corresponding authors (Fig. 5). Manuscripts submitted by women  
462 corresponding authors were editorially rejected at greater rates and recommended for rejection  
463 more often by gatekeepers of both genders. Neither geographic (i.e., US or not), institution type,  
464 nor sub-discipline could fully account for the observed gender-based outcomes (Fig. 6, 7, S7, and  
465 S8). Instead, the presence of outcomes that favor men over women from U.S. R1 institutions and  
466 medical schools and institutes suggests that the penalty for women persists, even in environments  
467 with generally excellent resources and infrastructure for research. Science and the peer review  
468 system select for decisions based on the assumption that scientists are objective, impartial experts.  
469 But scientists who believe themselves immune to bias are making decisions that inherently rely on  
470 cognitive biases to speed the process, some of which have negative impacts (55–57). For instance,  
471 previous studies show that a greater burden of proof is required for women to achieve similar  
472 competency as men (6, 58, 59). This and similar implicit biases might train women to be more  
473 conservative in their manuscript submissions, which makes our observed difference in outcomes  
474 even more concerning.

475 Even if a gatekeeper does not know the corresponding author or their gender, there remain ample  
476 avenues for implicit bias during peer review. The stricter standard of competency has led women  
477 to adopt different writing styles from men, resulting in manuscripts with increased explanations,  
478 detail, and readability than those authored by men (29, 60). Additionally, women are often at a  
479 disadvantage for the resources required for highly competitive fields due to cumulative penalties  
480 (9–11). As a result, corresponding authors that are women may be spending their resources  
481 in research fields where competition impacts are mitigated and/or on topics that are historically  
482 understudied, thus these are cues to gender and perceived competency (31–33). Alternatively,  
483 non-traditional research may be seen as less impactful, leading to poorer peer review outcomes

484 (34). These possibilities are reflected in our data, since while the number of revisions before  
485 publication is identical for both men and women, manuscripts authored by women have increased  
486 rejection rates and time spent on revision. This suggests that manuscripts submitted by women  
487 receive more involved critiques (i.e., work) from reviewers and/or their competency to complete  
488 revisions within the prescribed 30 days is doubted, compared to men. Women may also feel that  
489 they need to do more to meet reviewer expectations, thus leading to longer periods between a  
490 decision and resubmission. Finally, our data show a penalty for women researching mycology (Fig.  
491 S8). Despite being among the most deadly infectious diseases in 2016 (along with tuberculosis  
492 and diarrheal diseases), mycology is an underserved, and underfunded, field in microbiology that  
493 has historically been considered unimportant (61–64). Microbiology would benefit from a more  
494 nuanced evaluation of sub-fields to better understand how they interact with gender and peer review  
495 outcomes.

496 A limitation to our methodology is the use of an algorithm to infer gender from first names. While we  
497 report a high accuracy (0.97–0.99) where gender was inferred, this method left us with a category  
498 of unknown gendered individuals. Additionally, the gender of an individual may be interpreted  
499 differently according to the reader (e.g., Kim is predominately a woman's name in the U.S., but likely  
500 a man's name in other cultures). The increase in unknown gendered authors corresponds to an  
501 increase in submissions to the ASM journals from Asian countries, particularly China. Anecdotally,  
502 most editorial rejections are poor quality papers from Asia, and our method had low performance on  
503 non-gendered languages from this region (Supp Text, Fig. S1), thus excluding many Asia-submitted  
504 manuscripts and increasing our confidence that the trends observed were gender-based. For  
505 corresponding authors, manuscript submissions are the end product of several other prior decisions  
506 such as a mentor's implicit bias(es), postdoctoral fellowships, faculty applications, start-up funding  
507 negotiations, and grant proposals. These prior factors, which cannot be accounted for in our  
508 analysis, along with the small effect size observed in some analyses, limit quantifying the role of  
509 gatekeeper decisions in the disparate gender-influenced outcomes. However, the consistency of  
510 decisions to benefit men corresponding authors over women across all of the journals included  
511 in this study, in addition to accumulated literature to-date, confirms that this descriptive study is  
512 highly relevant for the ASM as a society. Our findings offer opportunities to address gendered

513 representation in microbiology and systemic barriers in peer review at our journals.

514 All parties have an opportunity and obligation to advance marginalized groups in science (65). We  
515 suggest that journals develop a visible mission, vision, or other statement that commits to equity,  
516 justice, and inclusion and includes a non-discrimination clause regarding decisions made by editors  
517 and editors-in-chief. This non-discrimination clause should be backed by a specific protocol for the  
518 reporting of, and response to, instances of discrimination and harassment. Second, society journals  
519 should begin collecting additional data from authors and gatekeepers such as race, ethnicity, gender  
520 identity, and disabilities. These data should not be available to journal gatekeepers but instead kept  
521 in a dis-aggregated manner that allows for public presentation, tracking the success of inclusive  
522 measures, and to maintain accountability. Third, society journals can implement mechanisms to  
523 explicitly provide support for women and other marginalized groups, reward inclusive behavior  
524 by gatekeepers, nominate more women to leadership positions, and recruit manuscripts from  
525 sub-fields that are more likely to attract women and other marginalized groups (34). We can all help  
526 advance women (and other marginalized groups) within the peer review system by changing how  
527 we select experts in our field. For instance, authors can suggest more women as reviewers using  
528 “Diversify” resources (66), while reviewers can agree to review for women editors more often. Editors  
529 can rely more on manuscript reference lists and data base searches than personal knowledge to  
530 recruit reviewers (67), and journals can improve the interactivity and functionality of the reviewer  
531 selection software. Given the propensity for journals to recruit editors and EiCs from within their  
532 already skewed reviewer pools, opening searches to include more scientists in their reviewer pool  
533 and/or editors from outside the journal while enacting more transparent processes could be a major  
534 component of improving representation. Growing evidence suggests that representation problems  
535 in STEM are due to retention rather than recruitment. We need to align journal practices to foster  
536 the retention of women and other marginalized groups.

537 Most approaches to disparate outcomes focus on choices made by individuals, such as  
538 double-blinded reviews and implicit bias training. These cannot fully remedy the effects of implicit  
539 bias and may even worsen outcomes (68, 69). Since disparate outcomes (by gender, geographic,  
540 prestige, or otherwise) are partially the result of accumulated disadvantages and actions resulting  
541 from implicit biases and systemic “-isms”, a structural, system-wide approach is required. Broadly,

542 peer review is a nebulous process with expectations and outcomes that vary considerably, even  
543 within a single journal. Academic writing courses suffer similar issues and have sought to remedy  
544 them with rubrics. When implemented correctly, rubrics can reduce implicit bias during evaluation  
545 and enhance the evaluation process for both the evaluator and the evaluatee (70–73). We argue  
546 that rubrics could be implemented in the peer review process to focus reviewer comments, clarify  
547 editorial decisions, and improve the author experience. Such rubrics should increase the emphasis  
548 on solid research, as opposed to novel or “impactful” research, the latter of which is a highly  
549 subjective measure (74, 75). This might also change the overall negative attitude toward replicative  
550 research and negative results, thus bolstering the field through reproducibility. We also argue that  
551 reconsidering journal scope and the membership of honorary editorial boards might help address  
552 structural penalties resulting from implicit bias against women (and other marginalized groups)  
553 in peer review. Expanding journal scope and adding more handling editors would improve the  
554 breadth of research published, thus providing a home for more non-traditional and underserved  
555 research fields (the case at *mSphere* with an increased pool of editors). Implementing these  
556 steps to decrease implicit bias and structural penalties—review rubrics, increased focus on solid  
557 research, expansion of journal scopes and editorial boards—will also standardize competency  
558 principles for researchers at the ASM journals and improve microbiology as a whole.

559 Although the level of bias at many of the ASM journals is small, it is present. Peer review at the ASM  
560 journals is not immune to the accumulated disadvantages against women in microbiology. However,  
561 the adaptation of women and other marginalized groups to implicit bias (e.g., area of research and  
562 communication styles), make it impossible to address at the individual level. Instead, we must  
563 commit to changing the fundamental structure and goals of peer review to minimize the impact  
564 of such bias. We encourage the ASM journals, as well as other societies, to institute more fair  
565 and transparent procedures and approaches of peer review. The self-correcting nature of science  
566 is a badge that scientists wear proudly, but no single report or action can correct the inertia of a  
567 centuries-old institution. Instead, it requires the long-standing and steady actions of many. Our  
568 findings reflect many similar reports, and suggest concrete actions to correct the inertia of peer  
569 review at all levels. The next step is commitment and implementation.

570 **Data and Methods**

571 **Data.** All manuscripts handled by the ASM journals (e.g., *mBio*, *Journal of Virology*) that received  
572 an editorial decision between January 1, 2012 and August 31, 2018 were supplied as XML files  
573 by the ASM's publishing platform, eJP. Data were extracted from the XML documents provided,  
574 manipulated, and visualized using R statistical software (version 3.4.4) and relevant packages.  
575 Variables of interest included: the manuscript number assigned to each submission, manuscript  
576 type (e.g., full length research, erratum, editorial), category (e.g., microbial ecology), related (i.e.,  
577 previously submitted) manuscripts, number of versions submitted, dates (e.g., submission, decision),  
578 author data (e.g., first, last, and corresponding authorship, total number of authors), reviewer data  
579 (e.g., recommendation, editor decision), and personal data (names, institutions, country) of the  
580 editors, authors, and reviewers. Since reviews and commentaries are often commissioned, only  
581 original, research-based manuscripts were included in this analysis, e.g., long- and short-form  
582 research articles, New-Data Letters, Observations, Opinion/Hypothesis articles, and Fast-Track  
583 Communications. To help protect the confidentiality of peer review, names were removed from  
584 all records, and identifying data (e.g., manuscript numbers, days of date) were replaced with  
585 randomized values.

586 **Institution classification.** To identify the communities represented, we used the Carnegie  
587 classifications to group US-based academic institutions into R1 research (very high research  
588 activity), R2 research (high research activity), four-year medical schools, or low research (i.e., not  
589 R1, R2, or medical school) (27). Research institutes (e.g., Mayo Clinic, Cold Springs Harbor),  
590 industry (e.g., pharmaceutical), and federal (e.g., NIH, CDC) research groups were identified using  
591 the internet. Four-year medical schools and research institutions were grouped together since these  
592 typically share research prestige and have considerable resources to support research. Industry  
593 and federal research were their own groups. The "Other" category represents uncategorized US  
594 institutions. Non-US institutions were their own category.

595 **Gender inference.** The genderize.io API was used to infer an individual's gender based on their  
596 given name and country where possible. The genderize.io platform uses data gathered from social  
597 media to infer gender based on given names with the option to include an associated language or

598 country to enhance the probability of successful inference. Since all manuscripts were submitted  
599 in English, which precludes language association for names with special characters, names  
600 were standardized to ASCII coding (e.g., “José” to “Jose”). We next matched each individuals’  
601 country against the list of 242 country names accepted by genderize.io. Using the GenderGuesser  
602 package for R, all unique given names associated with an accepted country were submitted to  
603 the genderize.io API and any names returned without an inferred value of either male or female  
604 were resubmitted without an associated country. The data returned include the name, inferred  
605 gender (as “male”, “female”, “unknown”), the probability of correct gender inference (ranging from  
606 0.5 to 1.0), and the number of instances the name and gender were associated together (1 or  
607 greater). The inferred genders of all given names (with and without an associated country) whose  
608 probabilities were greater or equal to a modified probability (pmod) of 0.85 were used to infer  
609 genders (man/woman) of the individuals in our data set (Supp Text). The presenting gender  
610 (man/woman) of editors and senior editors in our data set was inferred by hand using Google where  
611 possible, and the algorithm was validated using both editor and published data (Supp Text)(5).

612 **Manuscript outcome analysis.** To better visualize and understand the differences in outcomes  
613 according to author gender, we calculated the difference in percentage points between the  
614 proportion of that outcome for men and women. To correct for the disparity in the participation  
615 of women relative to men at the ASM journals, all percentage point comparisons were made  
616 relative to the gender and population in question. For instance, the percentage point difference in  
617 acceptance rates was the acceptance rate for men minus the acceptance rate for women. A positive  
618 value indicated that men received the outcome more often than women, whereas a negative value  
619 indicated that women outperformed men in the given metric.

620 **Logistic regression models.** For the L2-regularized logistic regression models, we established  
621 modeling pipelines for a binary prediction task (76). First, we randomly split the data into training  
622 and test sets so that the training set consisted of 80% of the full data set while the test set was  
623 composed of the remaining 20% of the data. To maintain the distribution of the two model outcomes  
624 found with the full data set, we performed stratified splits. The training data was used to build the  
625 models and the test set was used for evaluating predictive performance. To build the models, we  
626 performed an internal five-fold cross-validation where we tuned the cost hyper-parameter, which

627 determines the regularization strength where smaller values specify stronger regularization. This  
628 internal cross-validation was repeated 100 times. Then, we trained the full training data set with  
629 the selected hyper-parameter values and applied the model to the held-out data to evaluate the  
630 testing predictive performance of each model. The data-split, hyper-parameter selection, training  
631 and testing steps were repeated 25 times to get a reliable and robust reading of model performance.  
632 Models were trained using the machine learning wrapper caret package (v.6.0.81) in R (v.3.5.0).

633 **Code and data availability.** Data and code for all analysis steps, logistic regression pipeline, and  
634 an Rmarkdown version of this manuscript, are available at [https://github.com/SchlossLab/Hagan\\_Gender\\_mBio\\_2019/](https://github.com/SchlossLab/Hagan_Gender_mBio_2019/)

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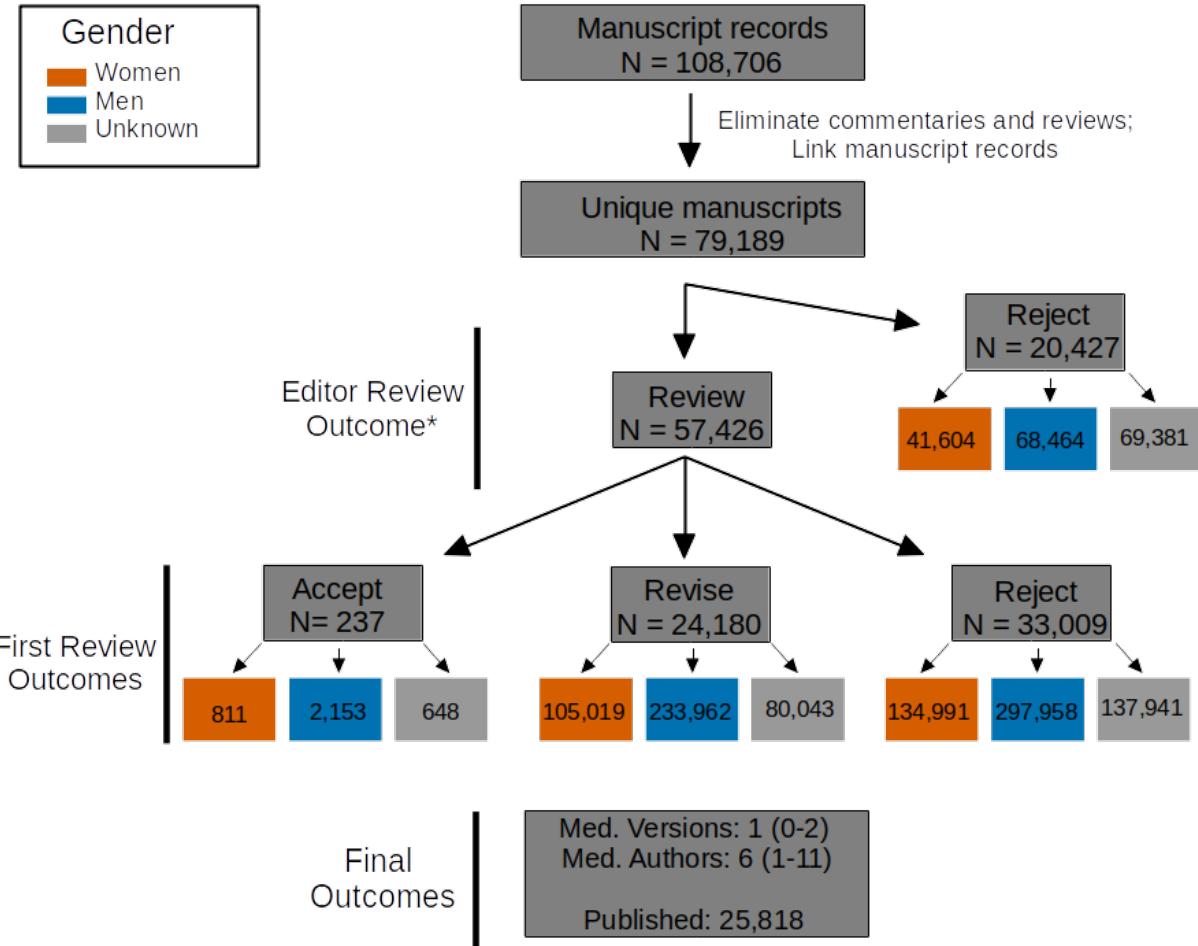
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829 Table 1. Analysis of sub-discipline participation by women corresponding authors at five ASM  
 830 journals.

Journal	Category	N	% Accepted	% Editors	% Women Authors
AAC	Analytical Procedures	135	43.0	14	29
AAC	Antiviral Agents	836	56.5	6	33
AAC	Biologic Response Modifiers	44	40.9	12	43
AAC	Chemistry; Biosynthesis	109	29.4	10	32
AAC	Clinical Therapeutics	1060	48.9	13	31
AAC	Epidemiology and Surveillance	765	52.3	14	40
AAC	Experimental Therapeutics	1329	57.4	13	28
AAC	FDA Approvals	1	NA	NA	NA
AAC	Mechanisms of Action: Physiological Effects	597	51.8	14	30
AAC	Mechanisms of Resistance	1783	60.0	14	36
AAC	Pharmacology	878	66.6	13	29
AAC	Susceptibility	1051	46.8	12	39
AEM	Biodegradation	302	38.4	35	26
AEM	Biotechnology	802	37.9	30	27
AEM	Environmental Microbiology	2395	30.3	35	42
AEM	Enzymology and Protein Engineering	340	46.5	28	24
AEM	Evolutionary and Genomic Microbiology	279	48.4	32	30
AEM	Food Microbiology	1216	38.2	33	39
AEM	Genetics and Molecular Biology	587	51.8	32	36
AEM	Geomicrobiology	151	44.4	34	37
AEM	Invertebrate Microbiology	317	45.7	29	37
AEM	Methods	529	39.7	30	29
AEM	Microbial Ecology	1121	35.8	29	37
AEM	Mycology	73	47.9	33	44
AEM	Physiology	356	50.3	32	31

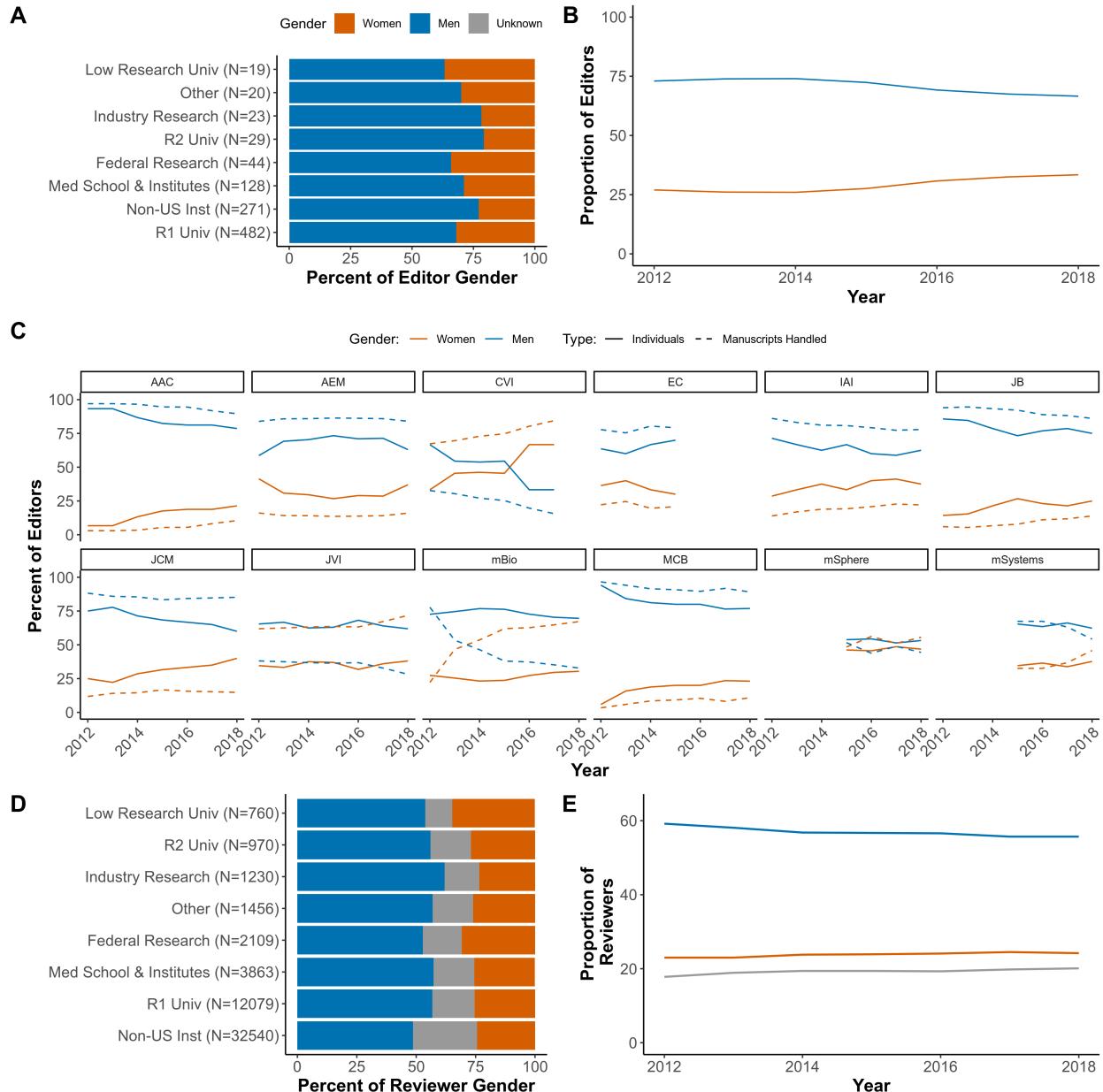
Journal	Category	N	% Accepted	Editors	% Women
				Authors	
AEM	Plant Microbiology	346	36.4	29	39
AEM	Public and Environmental Health	893	34.0	32	45
	Microbiology				
IAI	Bacterial Infections	716	58.4	35	36
IAI	Cellular Microbiology: Pathogen-Host Cell	685	55.2	35	37
	Molecular Interactions				
IAI	Fungal and Parasitic Infections	353	59.5	33	33
IAI	Host Response and Inflammation	763	50.2	35	40
IAI	Host-Associated Microbial Communities	7	57.1	43	86
IAI	Microbial Immunity and Vaccines	342	56.4	35	32
IAI	Molecular Genomics	33	60.6	37	33
IAI	Molecular Pathogenesis	617	68.4	35	31
JCM	Bacteriology	2952	33.2	27	41
JCM	Chlamydiology and Rickettsiology	80	32.5	25	41
JCM	Clinical Veterinary Microbiology	364	32.7	29	40
JCM	Epidemiology	854	29.7	30	45
JCM	Fast-Track Communications	5	40.0	33	40
JCM	Immunoassays	139	36.0	31	41
JCM	Mycobacteriology and Aerobic	510	42.9	32	41
	Actinomycetes				
JCM	Mycology	587	37.3	19	39
JCM	Parasitology	337	33.2	27	34
JCM	Virology	1140	37.5	29	41
JVI	Cellular Response to Infection	604	51.2	36	32
JVI	Gene Delivery	98	41.8	32	20
JVI	Genetic Diversity and Evolution	883	51.1	39	27

Journal	Category	N	% Accepted	Editors	% Women	% Women Authors
JVI	Genome Replication and Regulation of Viral Gene Expression	813	64.6	39	23	
JVI	Pathogenesis and Immunity	1622	60.4	35	33	
JVI	Prions	92	69.6	56	22	
JVI	Structure and Assembly	725	71.3	39	29	
JVI	Transformation and Oncogenesis	154	59.1	39	36	
JVI	Vaccines and Antiviral Agents	1149	59.2	36	28	
JVI	Virus-Cell Interactions	2414	63.6	40	30	



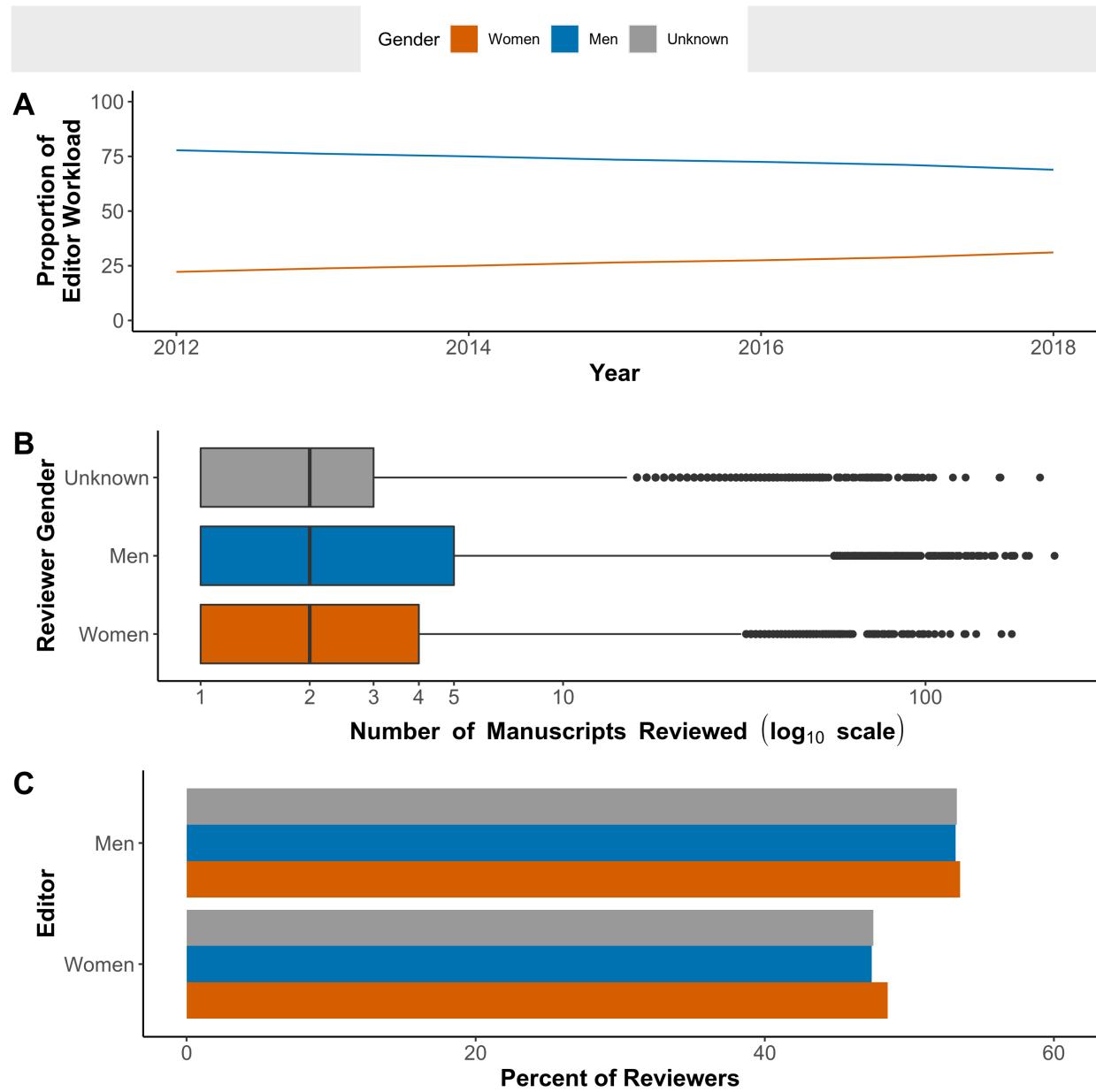
831

832 **Figure 1. Overview of manuscript outcomes.** 108,706 manuscript records were obtained for  
 833 the period between January 2012 and August 2018. After eliminating non-primary research  
 834 manuscripts and linking records for resubmitted manuscripts, we processed 79,189 unique  
 835 manuscripts. The median number of versions was 1 (IQR=0-2) with a median of 6 (IQR=1-11)  
 836 authors per manuscript. As of August 2018, 34,196 of these were published at the ASM journals.  
 837 Revisions were requested for 24,016 manuscripts and 53,436 manuscripts were rejected at their  
 838 first submission. The number of individuals (e.g., author, editor, reviewer) involved in each category  
 839 of manuscript decision are indicated in the colored boxes: women (orange), men (blue), and  
 840 unknown (gray). \*A small number were given revise (242) or acceptance (1094) decisions without  
 841 review.



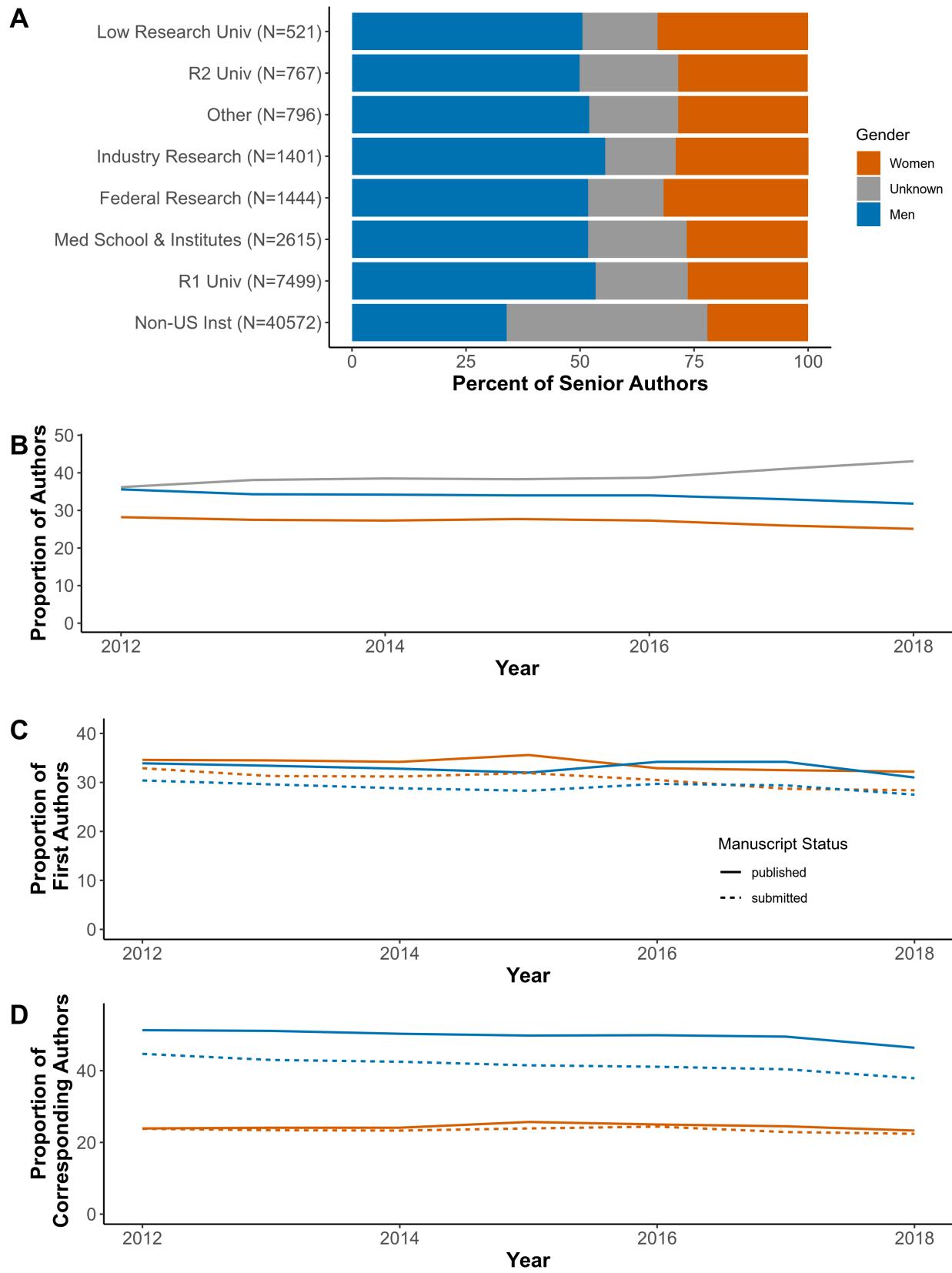
842

843 **Figure 2. Gendered representation among gatekeepers.** Proportion of editors from (A)  
 844 institution types and (B) over time. Editors and senior editors are pooled together. (C) The  
 845 proportion of editors (solid line) and their workloads (dashed line) at each of the ASM journals from  
 846 2012 to 2018. Proportion of reviewers from (D) institution types and (E) over time. (A,D) Each  
 847 gender equals 100% when all institutions are summed.(B,E) Each individual was counted once per  
 848 calendar year, proportions of each gender add to 100% per year.



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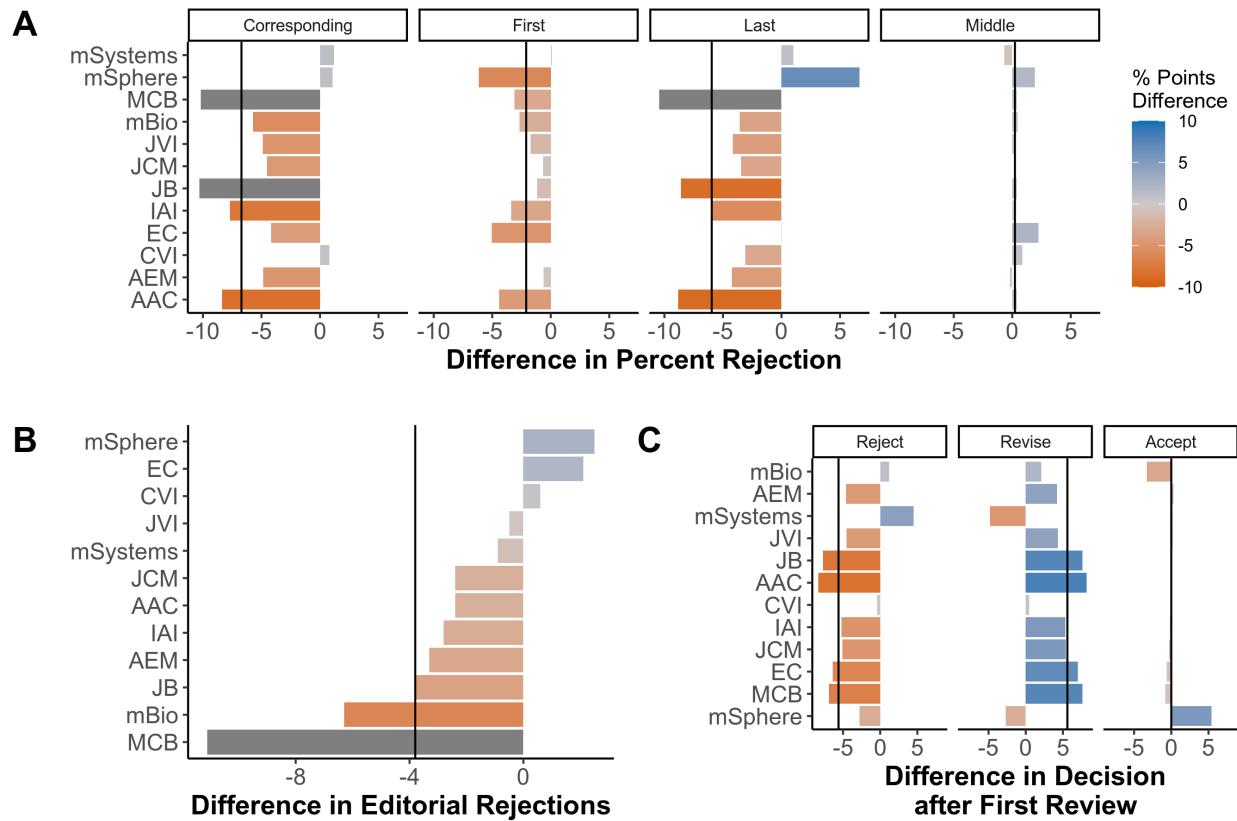
850 **Figure 3. Gatekeeper workload and response to requests to review.** (A) Proportion of  
 851 manuscripts handled by men and women editors, editorial rejections excluded. (B) Box plot  
 852 comparison of all manuscripts by reviewer gender on a  $\log_{10}$  scale. (C) The percent of reviewers by  
 853 gender that accepted the opportunity to review, split according to the editor's gender.



854

855 **Figure 4. Author representation by gender.** The proportion of (A) men, women, and unknown

856 gender senior authors from each institution type, (B) men, women, and unknown (senior and co-)  
857 authors from 2012 - 2018. Each individual was counted once per calendar year. The proportion  
858 of (C) first authors and (D) corresponding authors from 2012 - 2018 on submitted manuscripts  
859 (dashed line) and published papers (solid line).



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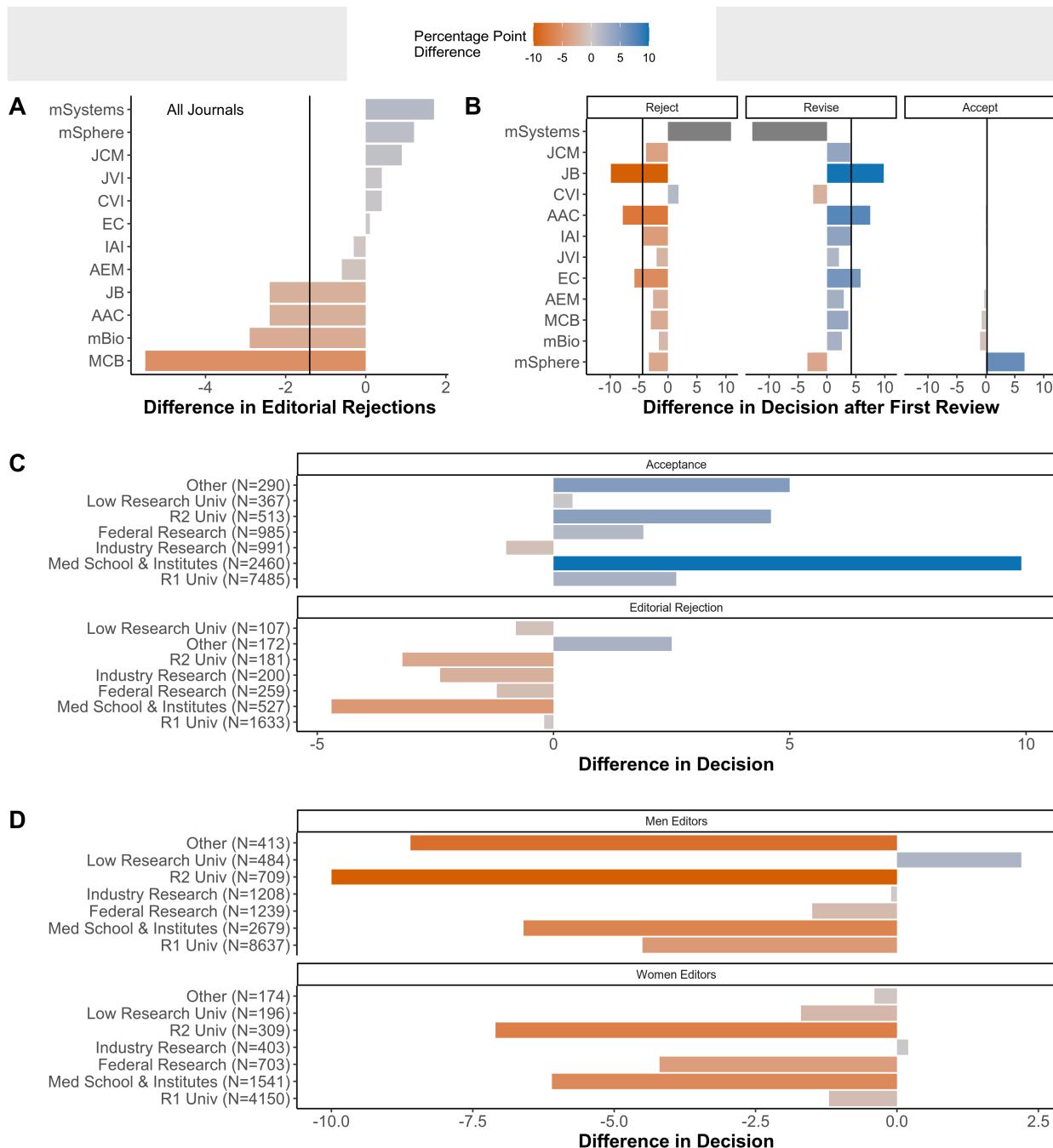
861 **Figure 5. Difference in manuscript outcomes by author gender.** The difference in the percent  
 862 of manuscript outcomes was calculated by subtracting the percent of women who received the  
 863 outcome, from the percent of men. A negative value (orange) indicates women received the  
 864 outcome more often, 0 (or no bar) indicates equal rates of the outcome, and positive (blue)  
 865 indicates that men received the outcome more frequently. Vertical lines indicate the difference value  
 866 for all journals combined. (A) The difference in percent rejections by author gender and type (e.g.,  
 867 corresponding, first, last, middle) at any stage across all journals. (B) The difference in percent  
 868 editorial rejection rates for corresponding authors at each journal. (C) The difference in percentage  
 869 points between each decision type for corresponding authors following the first peer review.



870

871 **Figure 6. Difference in decisions or recommendations according to the gatekeeper gender.**

872 The difference in the percent of manuscript outcomes was calculated by subtracting the percent of  
 873 women who received the outcome, from the percent of men. A negative value (orange) indicates  
 874 women received the outcome more often, 0 (or no bar) indicates equal rates of the outcome, and  
 875 positive (blue) indicates that men received the outcome more frequently. (A) Effect of editor gender  
 876 on the difference in decisions following review. (B) Difference in percentage points for review  
 877 recommendations and (C) how that is affected by reviewer gender. (A-C) All journals combined.



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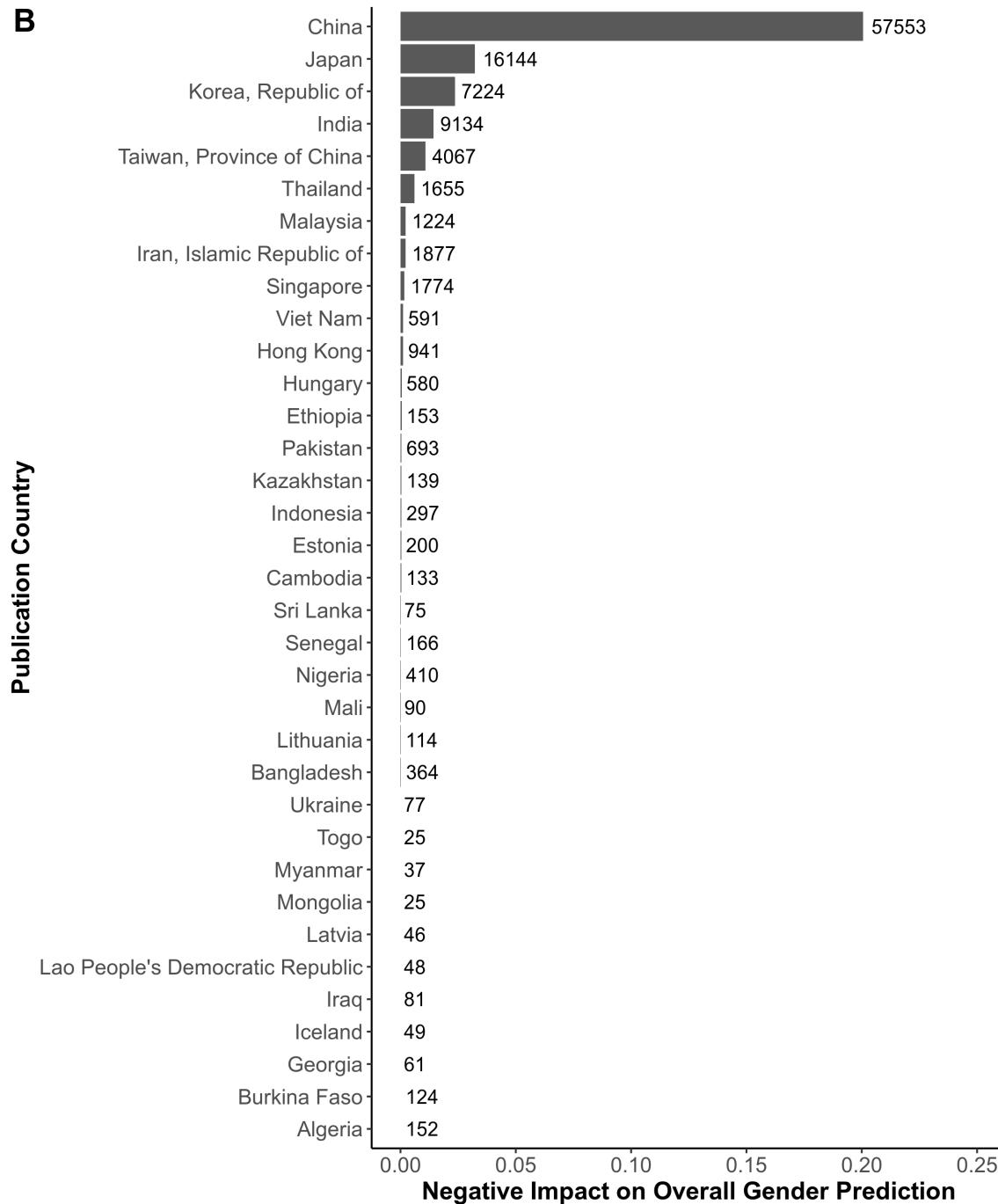
879 **Figure 7. Impact of origin and U.S. institution type on manuscript decisions by gender.** The  
 880 difference in the percent of manuscript outcomes was calculated by subtracting the percent of  
 881 women who received the outcome, from the percent of men. A negative value (orange) indicates  
 882 women received the outcome more often, 0 (or no bar) indicates equal rates of the outcome, and  
 883 positive (blue) indicates that men received the outcome more frequently. Vertical lines indicate  
 884 the difference value for all of the ASM journals combined. Difference in percentage points for

885 (A) editorial rejections and (B) following first review of manuscripts submitted by US-based  
886 corresponding authors. Vertical line indicates value for all of the ASM journals combined. (C)  
887 Difference in percentage points for acceptance and editorial rejections according to institution types  
888 and (D) acceptance decisions by editor gender and institution type.

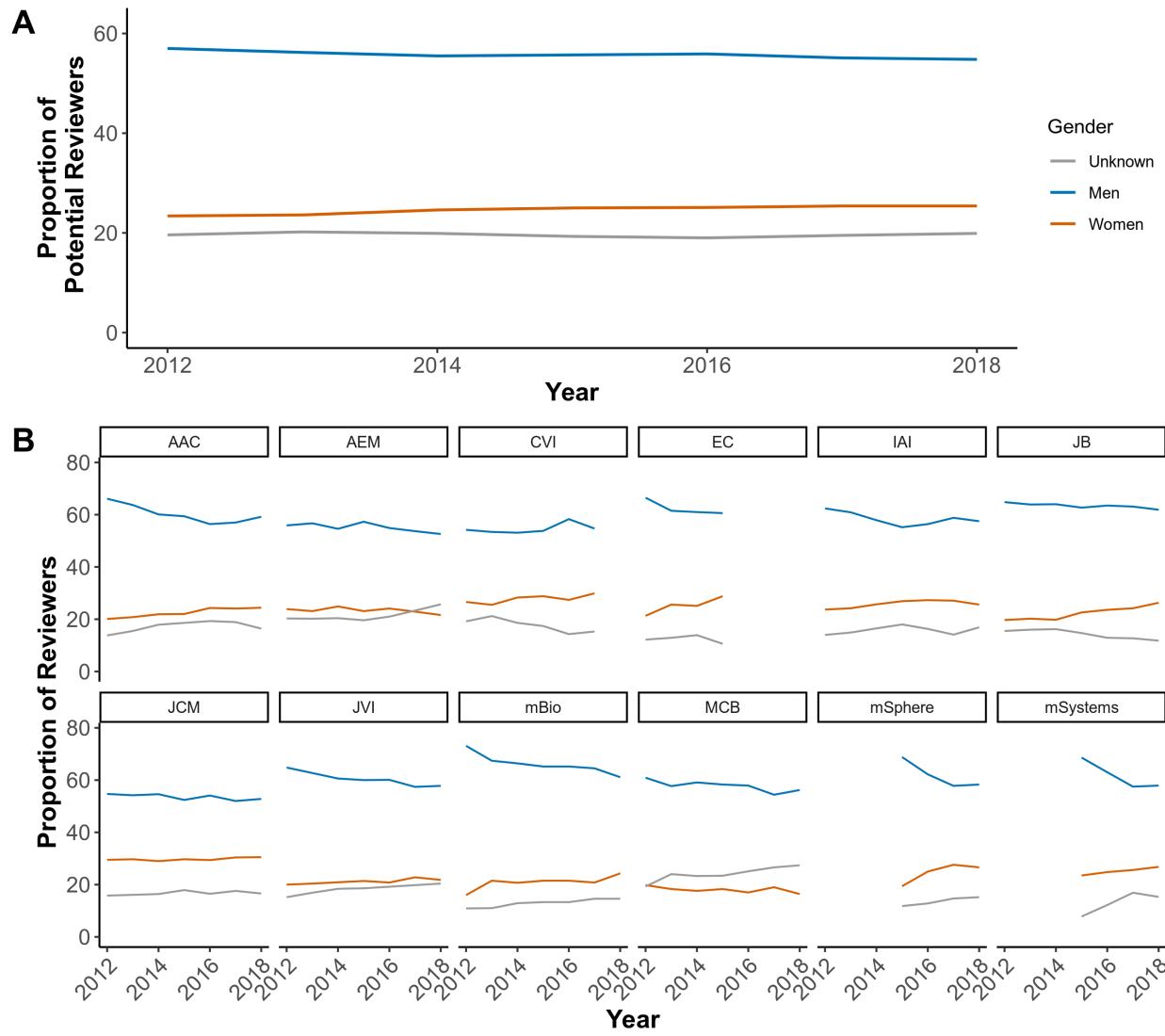
**A**

$$Impact_C = \left| \frac{(\% Unpredicted_C - \% Unpredicted_{Total}) \times \left( \frac{Observations_C}{Observations_{Total}} \right)}{\% Unpredicted_{Total}} \right|$$

**B**

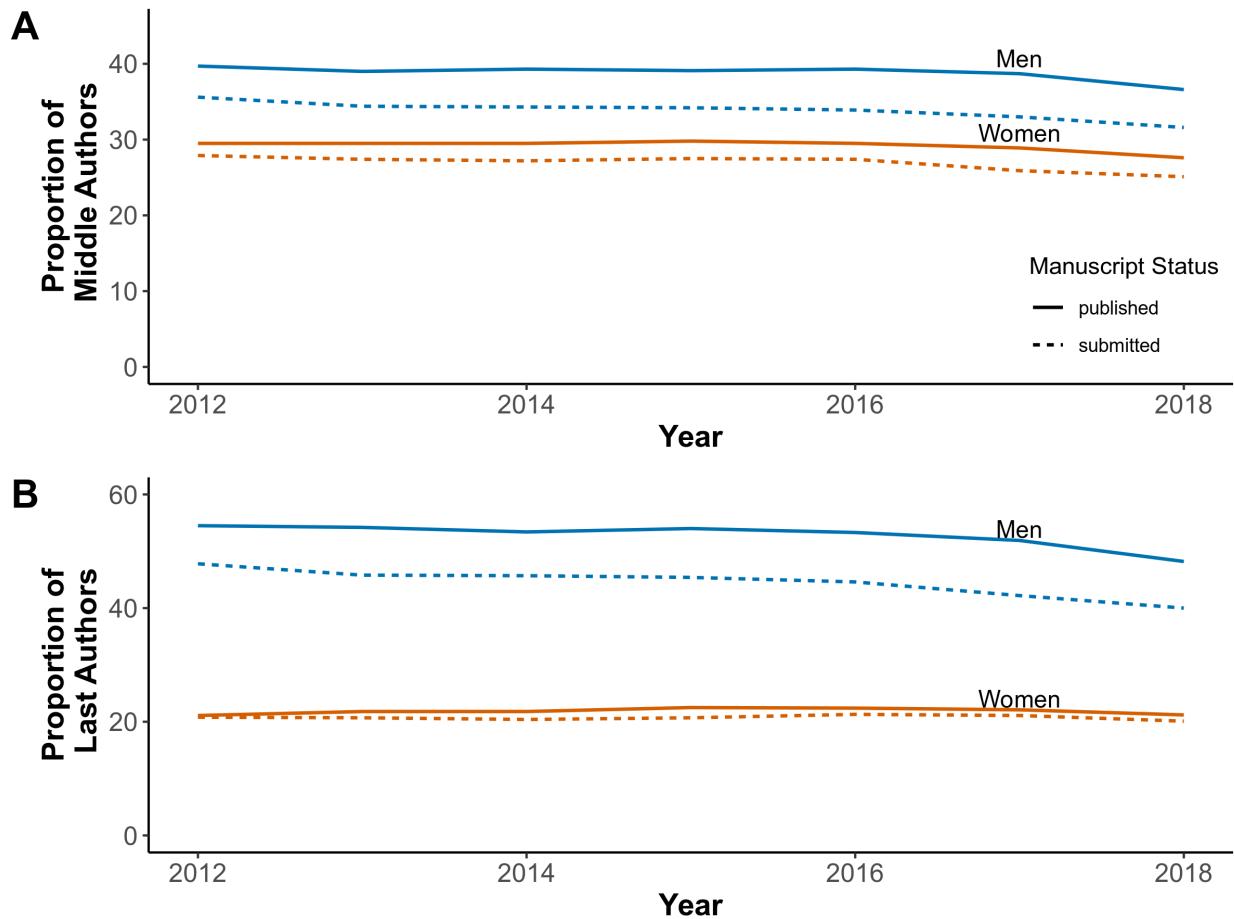


890 S1. (A) Equation for calculating negative bias by genderize algorithm. C indicates a country. (B)  
891 The negative impact of each country on the overall gender inference of the full data-set. Number to  
892 the right of each column is the total number of names associated with that country.



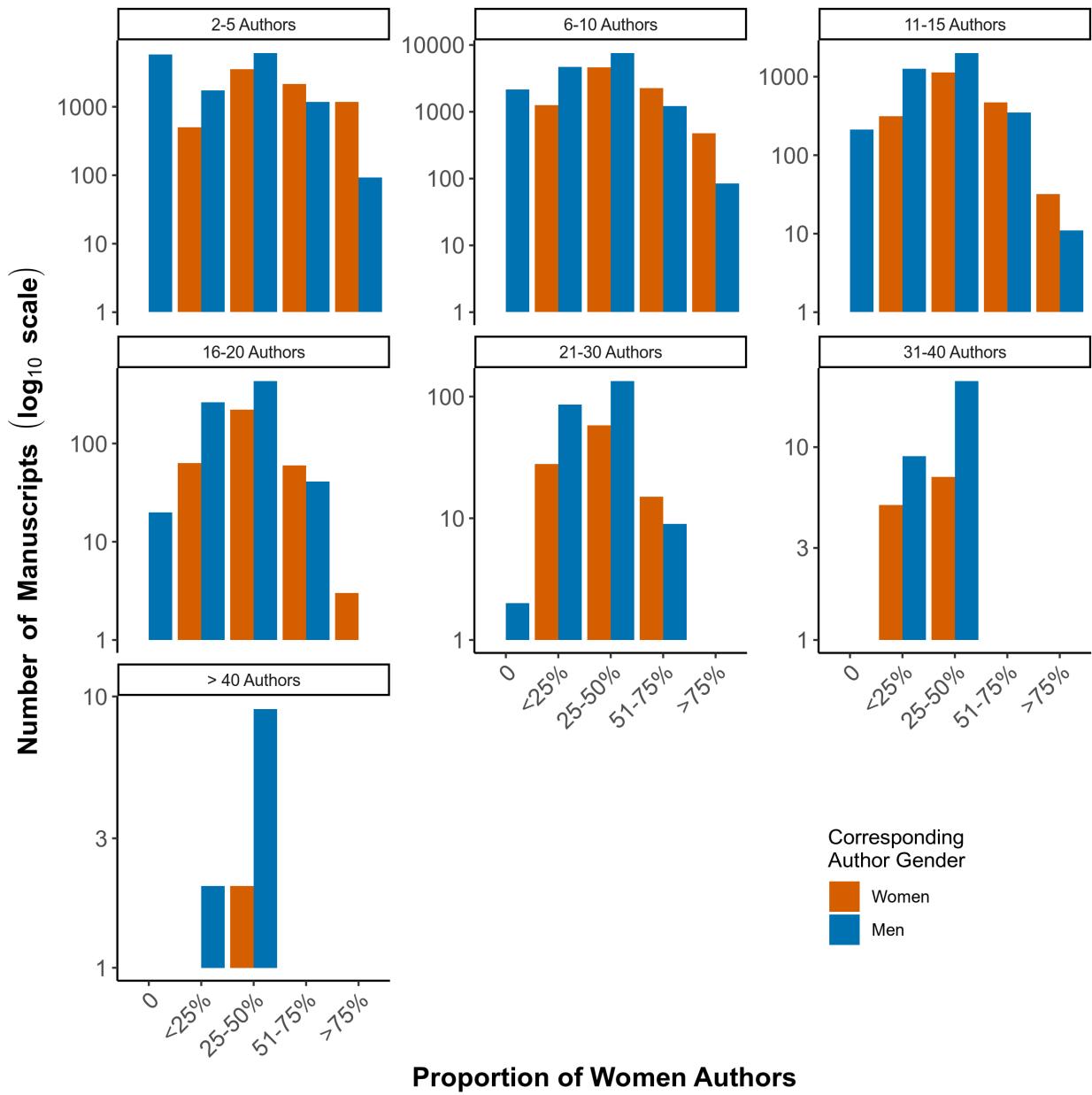
893

894 Figure S2. The proportion of (A) potential reviewers at all of the ASM journals combined, (B)  
895 reviewers at each ASM journal.



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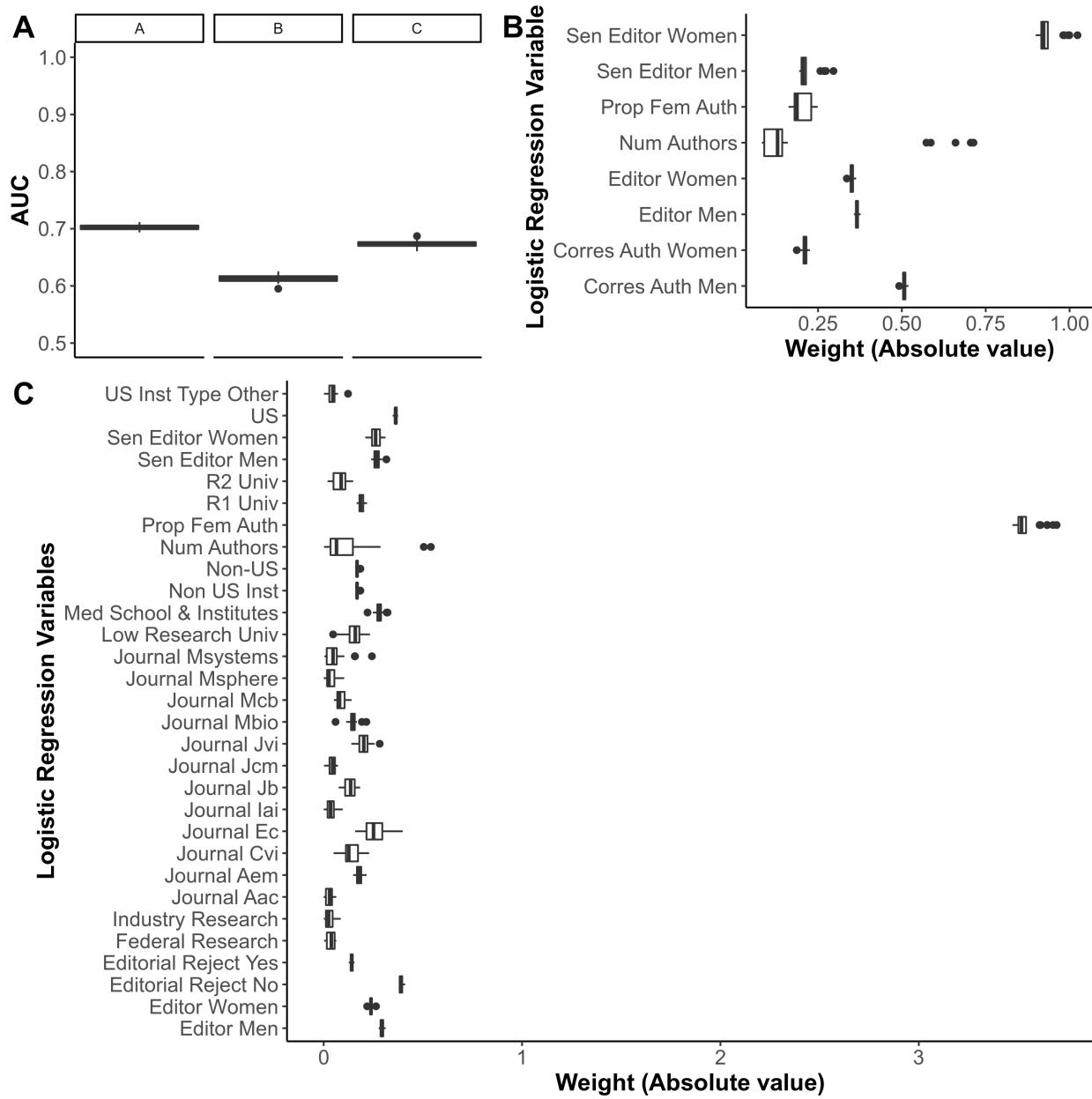
897 Figure S3. The proportion of all submitted (dashed line) and published (solid line) (A) middle and  
 898 (B) last authors by gender at each ASM journal.



899

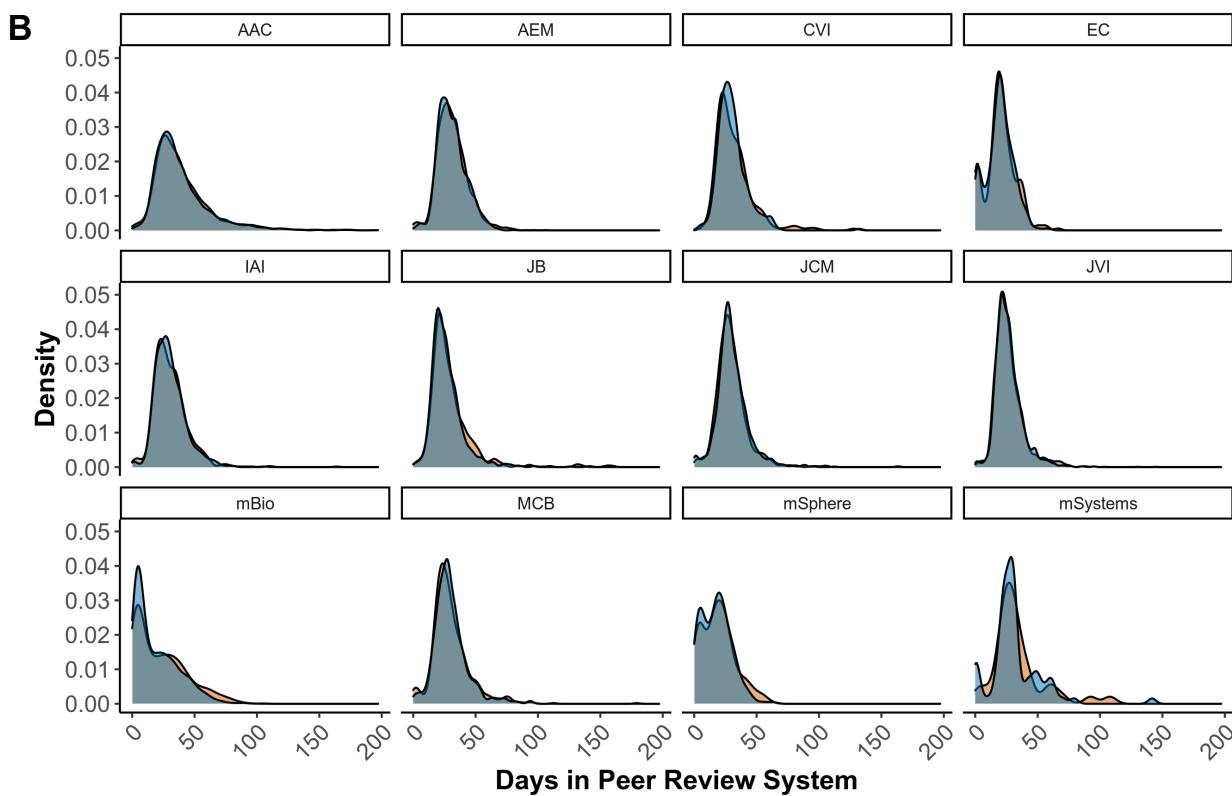
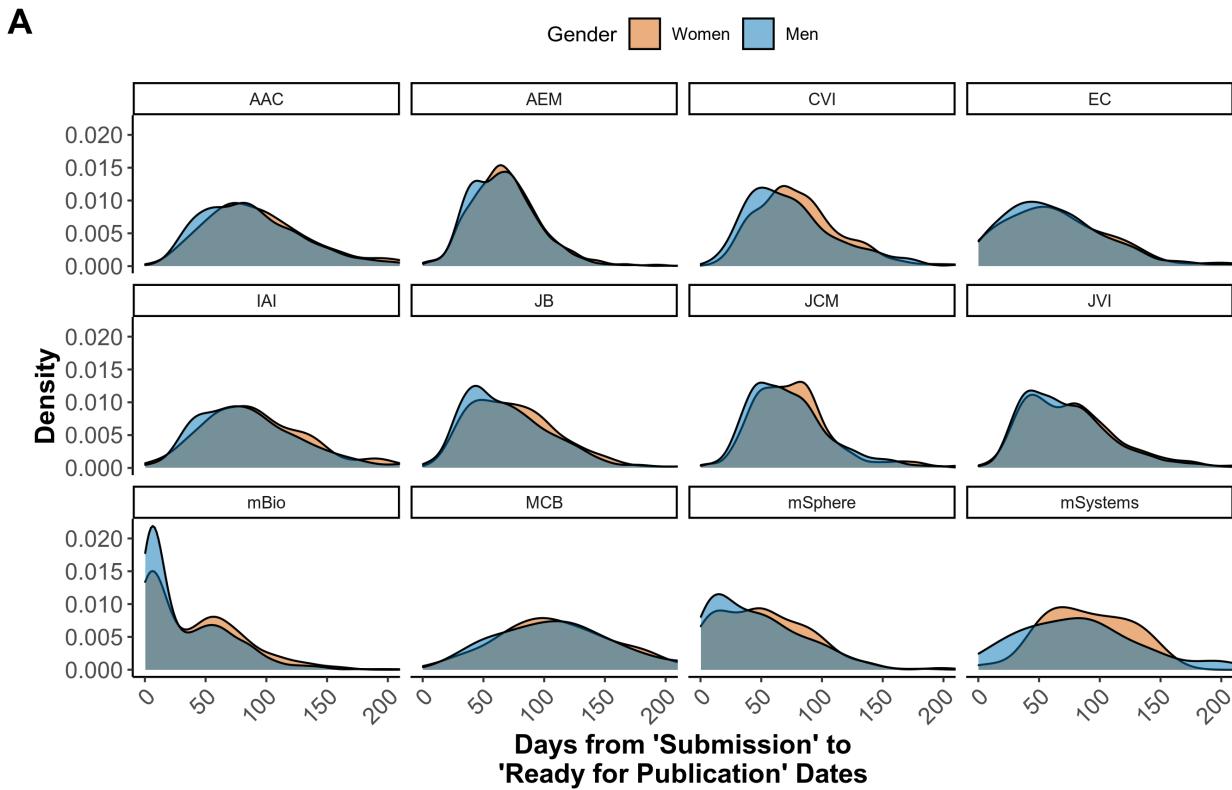
900 Figure S4. The proportion of women authors (x-axis) on submitted manuscripts (y axis,  $\log_{10}$  scale)  
 901 according to the number of co-authors (individual plot) and the gender of the corresponding author  
 902 (orange/blue). Single author papers were eliminated and the manuscripts grouped according to the  
 903 number of co-authors: groups of 5 authors up to 20, groups of 10 authors up to 40, and a single  
 904 group of manuscripts with 40+ authors. The manuscripts were then split according to the proportion  
 905 of co-authors that were inferred to be women: 0, up to 24%, 25-50%, 51-75%, and more than 75%.  
 906 The manuscripts were then further split according to the inferred gender of the corresponding  
 907 author. Regardless of the number of co-authors, women corresponding authors (orange) submitted

908 more manuscripts with a majority (>50%) of women co-authors than men corresponding authors.  
909 Men corresponding authors submitted more manuscripts with less than 50% women co-authors  
910 than women corresponding authors did, and the trend of this gap increased with the number of  
911 co-authors.



912

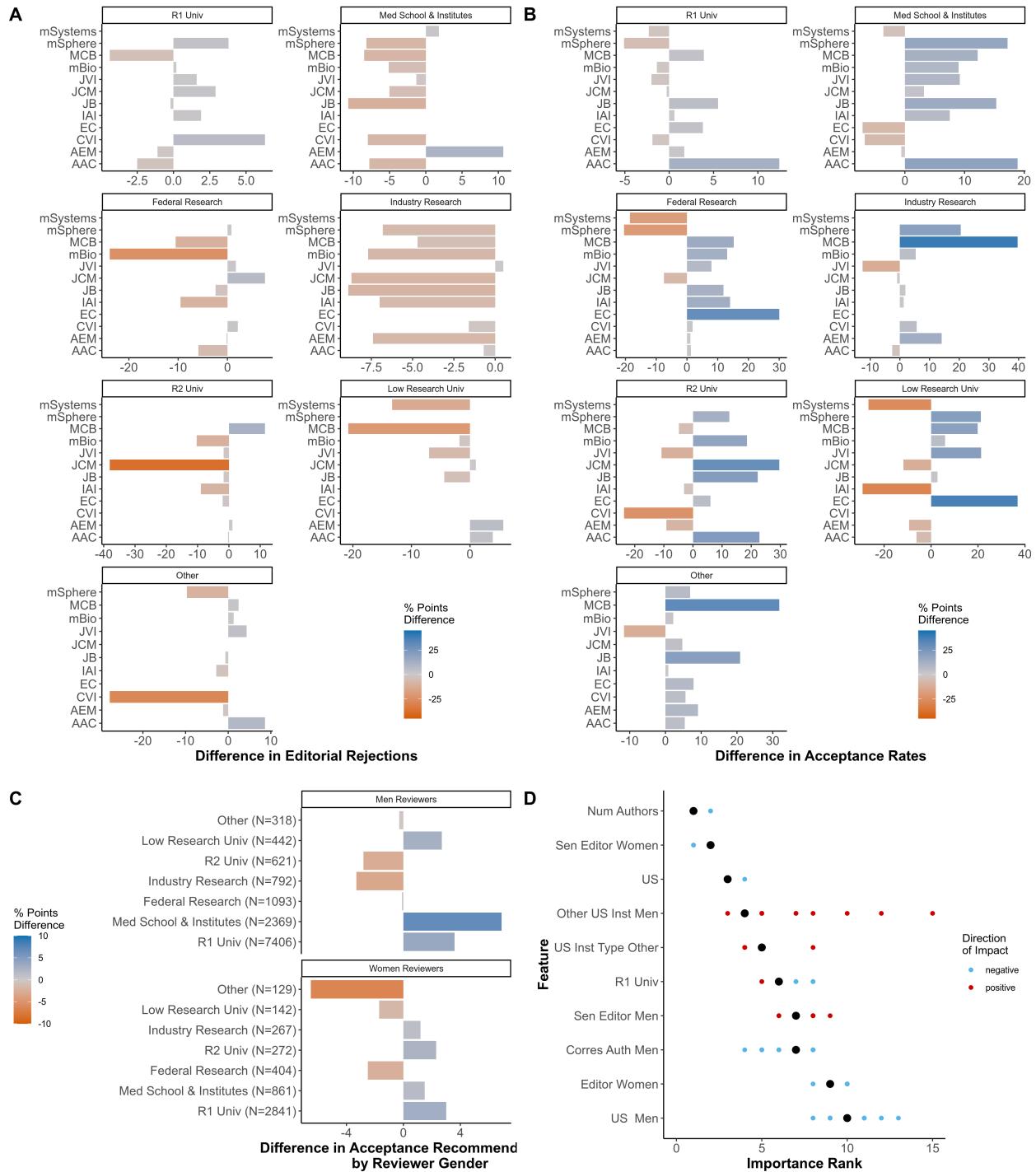
913 Figure S5. Boxplots of linear regression results from 25 data-splits. A) AUC values, each panel  
 914 represents a different prediction model: A - Corresponding author's gender; B - Author gender on  
 915 editorial decisions; C - Institution on editorial decisions. B,C) Effect of variables on the logistic  
 916 regression model represented as the absolute values of the variable weight. B) Author gender from  
 917 editorial decisions. C) Corresponding author's gender.



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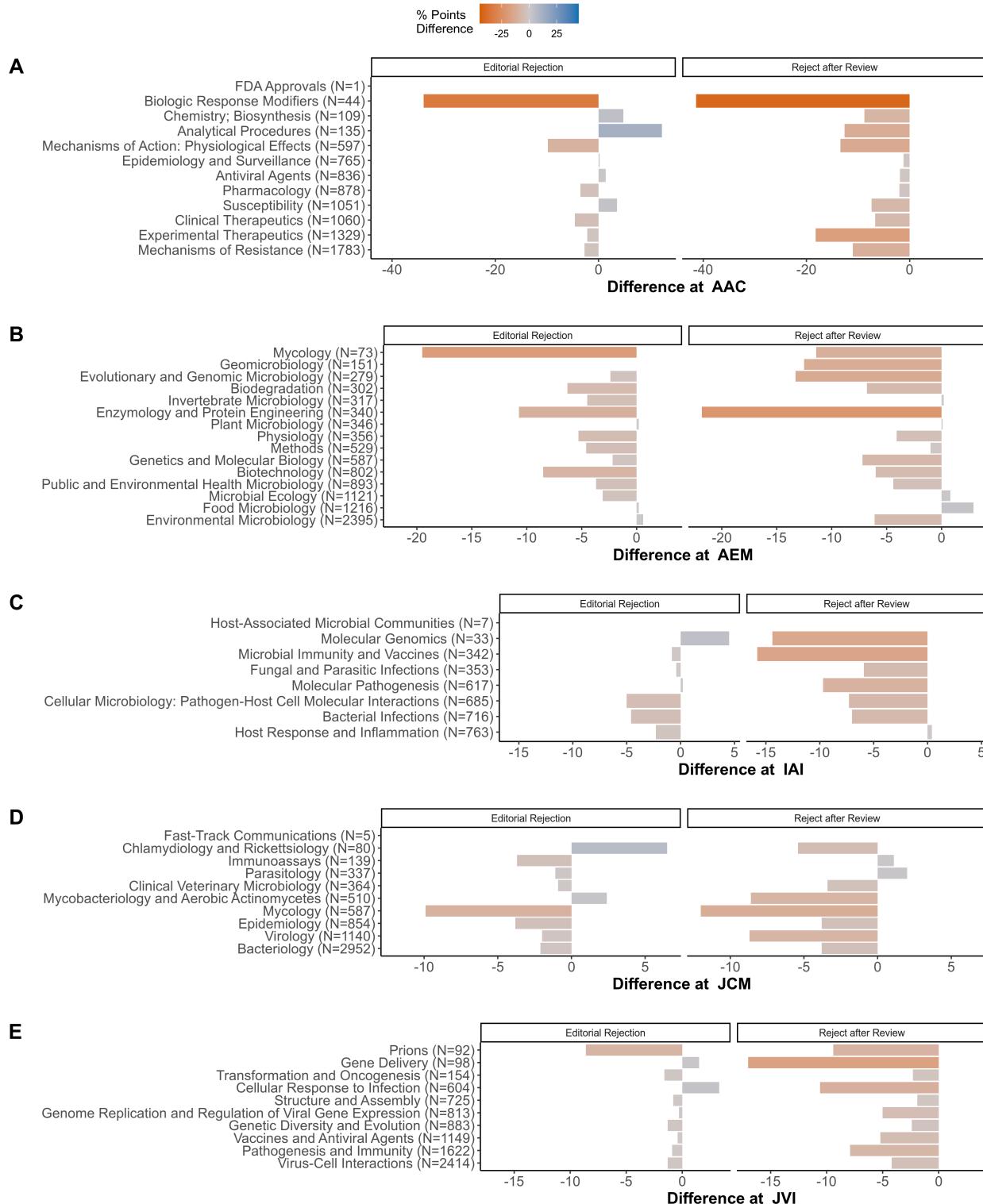
919 Figure S6. Comparison of time to final decision and impact by gender. The number of days (A)

920 between when a manuscript is initially submitted and finally published or (B) that a manuscript  
921 spends in the ASM peer review system.



922

923 **Figure S7.** Difference in A) editorial rejection and B) acceptance rates by journal and institution  
 924 type. C) Difference in review recommendations by reviewer gender and author institution type. D)  
 925 Median importance (black dot) of features affecting editorial rejections, and their range. Color of  
 926 smaller dots (N=25) indicate the direction of the impact.



927

928 **Figure S8. Difference in editorial rejections and rejections after review by corresponding author**  
929 **gender and manuscript category at (A) AAC, (B) AEM, (C) IAI, (D) JCM, and (E) JVI. In parentheses:**

<sup>930</sup> N = the number of manuscripts submitted.