The extent and drivers of gender and race/ethnicity imbalance in infectious disease dynamics research

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Abstract

Publication practices accumulate to affect credibility and career advancement. Understanding authorship and citation practices is critical to addressing inequities. While citation bias has been demonstrated in several fields, it remains uncharacterized in infectious disease dynamics (IDD), a quantitative, interdisciplinary domain highly visible during the COVID-19 pandemic. We analyze IDD articles and their bibliographies from 2000-2019 using machine-learning algorithms to infer the gender and race/ethnicity of each article's lead and senior author. We examine authorship and citation patterns by gender and racial group across geographic scales, including characterizing the author composition of each article's bibliography relative to the field. Our analysis reveals persistent gender and race imbalances in IDD research. Man-authored and White-authored publications dominate the field, with little progress in racial diversification of US and UK publications over the last two decades. Woman-authored articles have the most representative citation practices but are undercited, especially when women are senior authors. In the US and UK, most citations feature White lead and senior authors, even when citing articles with lead or senior authors of color. These findings underscore the urgent need for more inclusive IDD research practices. We discuss possible mechanisms and solutions to create opportunities for researchers from underrepresented groups.

Introduction

Diverse teams do better science [1,2] yet numerous barriers prevent underrepresented individuals from entering, flourishing, and staying in scientific research (e.g., [3–5]). As it stands, White men generally dominate scientific fields: while women compose 51% of the United States (US) population, for example, they only make up 35% of the US STEM workforce [6]. The gender disparity in STEM fields follows a scissor-shaped curve: women slightly outnumber men at the undergraduate level, but in later career stages (Ph.D. and beyond), men dominate, widening the gap between genders at a particularly visible professional level [7]. There are similar disparities in racial and ethnic representation of STEM fields [6, 8]. Identifying these inequities within each academic field is necessary to address field-specific equity issues and, ultimately, to improve national and international scientific output and individual research experiences.

Publication and citation practices are two critical markers of scientific success that are particularly vulnerable to bias and could be contributing to these disparities. These metrics are integral to visibility, credibility, career advancement, and attaining leadership positions but are not always objectively constructed and may be

amplifying existing inequities [9]. Characterizing how publication rates and citation practices vary within the scientific ecosystem is one way to evaluate the diversity and equity within a scientific field. This work has been conducted across disciplines [10–12] and in several specific fields including political science [13], international relations [14,15], neuroscience [16–18], physics [19], astronomy [20], and medicine [21]. For example, Bertelero and colleagues' found that White neuroscience authors preferentially cite other White authors and that neuroscience authors of color are being increasingly undercited despite authoring an increasing proportion of articles [17]. Likewise, Caplar and colleagues' found that women are lead-authoring an increasing proportion of astronomy publications, but their work is undercited compared to men [20]. Within infectious disease research, a recent study on editors and authorship in 40 infectious disease journals found that women are underrepresented in senior author and editor positions and suggested that recruiting more women editors could help increase the publication rate for women [22]. The field of infectious disease dynamics is distinct from most infectious disease research due to its highly quantitative nature and is highly visible to the public due to its ability to provide actionable insights during health crises. These distinctions make the field both prone to gender and race disparities and a priority for ensuring equitable research methods. However, there have been no comprehensive analyses of publication and citation practices in infectious disease dynamics to date.

Here, we address this gap by examining and quantifying the author composition and citation practices by gender and race/ethnicity within the interdisciplinary and impactful field of infectious disease dynamics. We compile a dataset of over 12,000 articles in infectious disease dynamics and infer the gender and race/ethnicity of lead (first) and senior (last) authors of these articles using validated machine-learning packages. We test three hypotheses: (1) men and White (lead/senior) authors are over-represented in articles in the field (especially in the Global North); (2) even once accounting for the composition of the field, articles with women (lead/senior) authors are undercited compared to articles authored by men authors; and (3) articles by women and authors of color exhibit more representative citation practices than articles with men and White (lead/senior) authors. Our findings characterize the degree of inequity in infectious disease dynamics publication and citation practices and suggest potential ways to remediate these issues.

Methods

Data collection

Infectious disease dynamics (IDD) is an interdisciplinary field that publishes in a broad range of disciplinary and general scientific journals. Therefore, it is challenging to define the boundaries of the published literature in infectious disease dynamics by selecting journals or conducting keyword searches. Similarly, because no IDD international professional organization exists, membership in the field is not feasible to identify. For our analysis, we thus identified authors in the IDD field by identifying the articles that cite a set of influential primary research and review IDD articles before 2020. To identify this set of influential research, we searched Web of Science (WoS) and selected articles that developed or discussed dynamical models of population-scale disease transmission and had accrued an average of 50 or more citations per year since publication. This resulted in 23 research articles published between 1990 and 2015. (Additional details and the the complete list of articles can be found in the Supplement.) To characterize authorship patterns in IDD, we define the *authorship dataset* as the set of articles published in 2000-2019 in the Web of Science Core Collection that cite these influential articles as of November 25, 2024 and are cited by another authorship dataset article by November 25, 2024.

We also define an alternative definition for the IDD authorship dataset by identifying articles that have cited three seminal IDD books: *Infectious Diseases of Humans: Dynamics and Control* [23], *Modeling Infectious Diseases in Humans and Animals* [24], and *Mathematical Tools for Understanding Infectious Disease Dynamics* [25]. These articles were extracted on April 13, 2023 from the Web of Science Core Collection. We present our findings based on this alternative definition in the Supplementary Information.

To characterize citation patterns in IDD, we define the *citation dataset* as the set of articles cited in the articles of the authorship dataset. To exclude irrelevant citations, citation dataset articles must either be in the authorship dataset or have been cited sufficiently based on their date of publication (from more than 2 times for 2019, to more than 5 times for 2000, determined using the 90th percentile of the number of citations for other articles published the same year). We perform sensitivity analyses on this threshold for the citation dataset (Figures S16, S17).

Inferring identities

We inferred the gender and race/ethnicity of the lead and senior authors of each article in the authorship and citation datasets using the machine-learning algorithms implemented in `genderize.io` [26, 27] and `ethnicolr` [28], respectively. Only identities identified with over a 70% confidence were included; we performed sensitivity analyses on this threshold (Figures S12, S13, S14, S15). We also assessed the sensitivity of our results to the choice of algorithm by additionally using `namsor` [29] for classification (Figures S10, S11). Given the limitations of these methods, we had to reduce gender to a binary (man or woman), and race/ethnicity to four levels (White, Black, Asian, and Hispanic). We additionally use the affiliation for the senior author to assign a home country to each article, allowing us to specify our analysis to certain countries (e.g., US, UK) or the Global North (which is broadly comprised of Northern America, Europe, Israel, Japan, South Korea, Australia, and New Zealand). All geographically restricted analyses exclude articles that are missing senior author country affiliations. For articles with missing author first names or non-delineated first and last names, we pulled additional author name information from the CrossRef API, if available. To disambiguate authors who are recognized by different names across articles or with only initials for first names (i.e., E. Halloran vs Elizabeth Halloran), we used the algorithm in [16] to identify authors in the dataset with first and last names matching the initials and last names, and replaced initials with full name. If there were different full author names that match the initials, then we did not replace the name (i.e., Elizabeth Halloran and Edward Halloran).

Analyses

To understand authorship patterns, we measured the proportion of articles in the authorship dataset authored by each gender and race/ethnicity group across time and geography. When analyzing author collaborations we removed any articles with only a single author. We performed bootstrapping (1000 draws) to address small sample sizes. These proportions are compared relative to the gender and race/ethnicity representation of the general population in the respective geography (e.g., global, Global North, US, UK).

To characterize gendered citation practices, we calculated the cumulative fraction of articles of each lead and senior authorship gender type (man-man (MM), woman-woman (WW), woman-man (WM), man-woman (MW)) across time; this distribution represents the pool of articles available to cite in a given bibliography. We then compared the makeup of each bibliography, excluding self-citations of the lead or senior authors, with that of the field one month prior to publication to assess which author types were being over- or undercited. We included citations of articles written before 2000 by assuming that the composition of the field had not changed before 2000. To understand whether the identity of the citing authors affected their citation practices, we calculated the proportion of each gender pairing in each bibliography for articles with men lead and senior authors versus articles with a woman lead and/or senior author. We bootstrapped both of these estimates by randomly selecting articles from the authorship dataset 1000 times and analyzing their bibliographies. We repeat the same analyses for race/ethnicity with a focus on White-White and lead or senior person of color(POC)-authored articles due to low sample size of mixed race authorship.

We further analyzed citation practices using linear regression to understand how lead and senior author identity predict citation rate (defined as the ratio of the number of total citations of an article and the number of years since publication). We exclude self-citations and control for 2023 Journal Impact Factor (JIF) and whether an article was published by authors in the Global North. (See Supplement for sensitivity

analyses and diagnostics; articles missing senior author affiliation country are excluded). The model was structured as:

$$
\begin{aligned}
\log_{10}(\text{citation rate})_i \sim {} & \beta_0 + \beta_1 \text{ManWoman}_i + \beta_2 \text{WomanMan}_i + \beta_3 \text{WomanWoman}_i + \\
& \beta_4 \text{WhitePoC}_i + \beta_5 \text{PoCWhite}_i + \beta_6 \text{PoCPoC}_i + \beta_7 \sqrt{\text{JIF}}_i + \beta_8 \mathbb{1}_{\text{Global North}_i} + \\
& \beta_9 \text{ManWoman} \sqrt{\text{JIF}}_i + \beta_{10} \text{WomanMan} \sqrt{\text{JIF}}_i + \beta_{11} \text{WomanWoman} \sqrt{\text{JIF}}_i + \\
& \beta_{12} \text{WhitePoC} \sqrt{\text{JIF}}_i + \beta_{13} \text{PoCWhite} \sqrt{\text{JIF}}_i + \beta_{14} \text{PoCPoC} \sqrt{\text{JIF}}_i + \epsilon_i.
\end{aligned}
$$

Results

We identify 10,660 articles in the authorship dataset and 16,195 unique articles for the citation dataset cited in 220,570 citations across the bibliographies of the authorship dataset. The published literature of the IDD field grew rapidly and consistently across the last two decades (Figure 1A).The authorship dataset also spans 101 countries of affiliation, with the majority of articles (65%) published between 2000 and 2019 written by senior authors from the Global North (Figure 1B). The two countries with the most published articles per population size were the US with 2,855 articles (8.70 per million population) and the UK with 938 articles (14 per million population) (Figure S2). We were able to infer senior author gender for 85% of authorship articles and 86% of unique citation articles and senior author race for 63% of authorship articles and 53% of unique citation articles.

Figure 1. Infectious disease dynamics publications have rapidly increased in the last two decades, but most senior authors are concentrated in the Global North. (A) Number of infectious disease dynamics publications per year from 2000 to 2019. (B) Distribution of articles published per million citizens by country from 2000 to 2019, based on the affiliations of the senior authors.

IDD publication authorship is dominated by men while woman-authored articles are undercited

We find that men lead-authored 61% of articles and senior-authored 67% of the articles published in IDD from 2000 to 2019, despite men only composing 50.4% of the global population. The ratio of man-authored to woman-authored articles has remained relatively constant since the early 2000s for both lead and senior authors (Figures 2A, S5). Concerning collaboration, 43% of articles have men lead and senior authors while only 6% of articles have women lead and senior authors. These ratios are heavily skewed compared to the

global population and equal gender representation within the field, but do not indicate unequal collaboration practices after taking into account the field's disproportionate number of men authors (Figure 2B). These findings are consistent across geographic scales, gender-inference approaches, field definitions, and after accounting for uncertainty in our sample (Figures S7, S8, S10, S3).

On citation practices, women lead- and senior-authored articles are undercited and the woman's authorship position affects the degree of undercitation. WM articles are undercited by 3%, MW by 15%, and WW by 33% (Figure 2C). In 2019, 58% of references in articles authored by men lead and senior authors are publications authored by men lead and senior authors and this trend has been fairly consistent over time (Figure 2D). Articles with women lead and senior authors have had increasingly gender-diverse bibliographies with only 40% of their citations corresponding to articles authored by men lead and senior authors by 2019 (Figure 2E). These findings are robust to geography and gender inference package (Figures S7, S10) but are sensitive to the definition of the IDD field (Figure S8).

Figure 2. Globally, infectious disease dynamics publications are predominantly authored by men, but collaboration practices across genders do not appear biased. Women are undercited, particularly by man-man authored articles and when women are in more senior authorship positions. (A) Proportion of articles senior-authored by each gender from 2000 to 2019. Expected proportions based on (global) population (EP) for 2019 are shown in the color bar at right. (B) Proportion of articles authored by a given (lead & senior author) gender pair from 2000 to 2019. Expected proportions based on the field (EF) in 2019 are shown in the color bar at right; these expectations are based on the observed proportions in Figure S5.(C) Rates of over- and undercitation by article lead and senior author gender based on authorship gender composition prior to publication date. (D) Authorship gender breakdown of articles cited in articles by men lead and senior authors.(E) Authorship gender breakdown of articles cited in articles written by women lead and senior authors.

Race/ethnicity disparities are present in authorship and citation practices

Because race is a social construct that is experienced and treated differently across contexts, we focus on the race/ethnicity of authors in the Global North, where race is understood more similarly. We restrict our

analysis to four race/ethnicity categories in the US and UK (White, Black, Hispanic, and Asian) so that we can compare authorship composition with the general population of each country. We find that most US publications have White authors with growing Asian authorship, which is less diverse than the general population (Figure 3A, S6). Similarly, nearly all UK publications have White authors, but authorship more closely reflects the general population (Figure 3B, S6). These findings are consistent across definitions of the IDD field, race/ethnicity inference methods, and after accounting for uncertainty in our sample (Figures S9, S11, S4).

Likewise, citation practices tell a biased story. Articles with a lead and/or senior author of color are undercited, while articles with White lead and senior authors are overcited in the Global North (Figure 3C). The rates of over- and undercitation are more extreme when author's bibliographies are expected to reflect global race/ethnicity authorship (Figure 3C). Authors of color are no better at citing diverse authors with respect to race compared to White authors (Figure 3D, E).

Figure 3. White authored articles are overcited compared to articles that have a non-White lead and/or senior author, but author racial diversity isn't associated with increased diverse citation practices. (A) Proportion of articles senior-authored by each race/ethnicity group in the USA from 2000 to 2019. Expected values (EP) based on 2019 census data are shown in color bar to the right. (B) Proportion of articles senior-authored by each race/ethnicity group in the UK from 2000 to 2019. Expected values (EP) based on 2019 Office of National Statistics Data are shown in color bar to the right. (C) Rates of over- and undercitation by article lead and senior author race/ethnicity based on authorship race/ethnicity composition prior to publication date. PoC author denotes that the lead and/or senior author was a person of color; sample size was too small to disaggregate by the author of color's position. Filled boxes denote results for articles published by authors with affiliations within the Global North citing other work from the Global North. Dotted boxes denote results for articles published by authors with affiliations within the Global North citing articles with any affiliation location. (D) Racial/ethnic composition of articles cited in articles written by White lead and senior authors in the Global North, citing other papers authored in the Global North. (E) Racial/ethnic composition of articles cited in articles written by a lead and/or senior author of color

Author race & gender influence citation rate after controlling for publication venue

We found that author gender pairing and race pairing are significantly associated with article citation rate, even after accounting for the impact factor of the journal where the article was published (Figure 4). Woman-woman, PoC-White, and PoC-PoC papers had lower citation rates, while journal impact factor increased citation rate. However, woman-woman, PoC-White, and PoC-PoC publications in high-impact factor journals had higher citation rates.

Figure 4. Coefficients from linear regression predicting citation rate (num. citations/years since publication) based on article and author information. Journal impact factor is square-root transformed to ensure normality. Teal point-ranges show coefficient estimate and 95% confidence interval for the predictor alone, while gold point-ranges denote the same values for the interaction between the predictor and journal impact factor. PoC stands for person of color.

Discussion

Infectious disease dynamics is a rapidly growing area of interdisciplinary research. As the field reflects on lessons learned from the early stages of the COVID-19 pandemic and identifies areas of future work, it is paramount to consider which voices are underrepresented and underrecognized in the field. This introspection is essential for promoting inclusivity, but also because publications and citations are key ways that academic success, credibility, and potential are measured, and public dissemination is achieved. We address this pressing issue by comprehensively analyzing the authorship and bibliographic landscape of infectious disease dynamics publications. We overcome challenges to define the field by curating a list of influential articles and

analyze the relevant articles citing these influential works. We use validated machine-learning packages [30] to infer the gender and race/ethnicity identity of the authors of each article and conduct sensitivity analyses on this identity inference, our definition of the field, and using varying spatial scales to demonstrate that our results are robust.

We find that publication and citation practices in the field of infectious disease dynamics are notably biased. Most publications are authored by men or White authors which doesn't reflect the composition of the general population (see intersectional results in Figure S28). Work authored by men or White scientists is overcited even after accounting for their disproportionate composition of the field. In contrast, work authored by women or people of color is heavily undercited, and this amplifies with gender or race assortativity. Author position also affects the rate of undercitation, with greater bias when women are in more senior authorship positions. The same trend may apply for authors of color, but we did not have sufficient sample size to detect it. Our regression analysis shows that these issues persist even after accounting for Journal Impact Factor. However, we do learn that citation rates of the articles by women or people of color in high-impact journals are higher relative to man-man and White-White authored articles, respectively. We speculate that this might reflect the phenomenon of overperformance for underrecognized individuals, in which systemically disadvantaged individuals are driven to and held to a higher standard for reaching the top of their field [31].

These gender results are consistent across spatial scales, but our race/ethnicity findings show greater geographic variability. The United States and the United Kingdom produced the most articles in the Global North, but authorship within these countries is not representative of the racial/ethnic makeup of the general population. Citation practices by race/ethnicity are similarly biased. Articles authored by White lead and senior authors are overcited while articles with at least one author of color are undercited by more than 30%. This difference cannot solely be explained by White lead and senior authors citing fewer authors of color (Figure 3D,E).

We recognize that our interpretation of gender as binary and race/ethnicity fitting into four classifications is limiting and prevents analysis of some of the most vulnerable scientists, including, but not limited to, non-binary individuals and Indigenous populations. Furthermore, race/ethnicity categorization can meaningfully differ across countries. Using names to infer gender and race can be inaccurate and ignores physical forms of identity expression [32]. However, names and affiliations are the key identifying information that editors, reviewers, and authors have when choosing to read, accept, review, or cite a given article. Thus, inaccuracies in gender or race/ethnicity inference may well reflect how an unknown author is perceived by others. We focus on authors listed first and last, ignoring shared authorship positions and the identity makeup of the full authorship list. Likewise, we acknowledge that our sample may be missing some IDD publications or include non-IDD work, but we must start to address these issues of inequity even in the absence of perfect data. Despite these limitations, our findings align with those from other disciplines, such as neuroscience, physics, astronomy, and medicine [16, 19–21].

As the field of infectious disease dynamics moves forward with additional urgency and scrutiny due to the COVID-19 pandemic, it is imperative to take action to increase the diversity and equity in the field [33]. We must also recognize how the pandemic has likely only exacerbated many of the disparities we have documented here [34]. For example, women had lower research productivity than men [35], including submitting fewer manuscripts during the early months of the pandemic [36]. We advocate for increased journal editorial board and reviewer diversity [22, 37–39] to leverage the effect of homophily. Additionally bibliographic transparency [19], in which editors and authors ensure that the composition of submitted bibliographies reflects the gender and race/ethnicity composition of the field, is critical. To facilitate this process, we've created a tool so that authors can check their own bibliographies relative to expected compositions for the field of IDD [40]. Additionally, individuals should consider citing both older initial work and recent applications by more junior and diverse authors in their manuscripts, a practice that can be aided by the removal of reference length limits by journals. Finally, academia should consider broadening the definitions of scientific contributions to reflect the variety of work that advances the field but may not be traditionally valued. To move forward, the field must assess and examine our current composition and practices so that we can measure future progress; we begin such an effort here and call on others to join us in acknowledging and correcting the inequities within our field.

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Data availability

This project utilizes a commercial dataset provided by Clarivate and the raw data cannot be republished. For further information, please contact Clarivate. The data and code used in our analysis is provided for reproducibility at `https://github.com/bansallab/IDD_imbalance`.

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Supplementary Information

Appendix A: Defining the field

To define the authorship dataset, we identify a set of influential articles in the field of infectious disease dynamics. To to this, we searched Web of Science for articles matching the search terms in Table S1. This resulted in 69,654 articles. We then excluded all articles that had an average annual citation rate below 50 citations per year, which resulted in 275 articles. We then excluded articles that were not relevant to the field (excluded journal names and title terms can be found in Table S1). The resulting list has 23 articles published from 1990 to 2015 and can be found in Table S2.

Search terms	'infectious disease dynamics' OR 'mathematical epidemiology' OR 'compartmental model' OR 'infectious disease model' OR 'SIR model' OR 'disease modeling' OR 'epidemic model' OR 'disease modelling' OR 'mathematical modeling of diseases' or 'mathematical modeling of infectious disease' OR 'infectious disease mathematical model' OR 'infectious disease mathematical modeling' OR ('mathematical model' AND 'disease') OR 'Mathematical modelling of diseases' or 'Mathematical modelling of infectious disease' OR 'compartmental disease transmission model' OR 'model of disease transmission' OR 'reproductive number' OR 'reproduction number' OR 'reproduction ratio' OR ('contact network' AND 'disease') OR ('epidemics' AND 'dynamics') OR ('networks' AND 'epidemic')
Excluded journals	'Cancer Research', 'European Respiratory Journal', 'Human Brain Mapping', 'Cell Stem Cell', 'Hepatology', 'British Journal Of Cancer', 'Lancet Oncology', 'Nature Reviews Drug Discovery', 'Journal Of Experimental Medicine', 'International Journal Of Radiation Oncology Biology Physics', 'American Journal Of Physiology-Regulatory Integrative And Comparative Physiology', 'Journal Of The American College Of Cardiology', 'Circulation Research', 'Nature Neuroscience', 'Stem Cells', 'Nature Reviews Gastroenterology & Hepatology', 'Pediatric Research', 'Nature Cell Biology', 'IEE-ACM Transactions On Networking'
Excluded title terms	'cells', 'fibroblasts', 'metagenome', 'diagnostic', 'logistic', 'envelopes', 'lung', 'mice', 'linkage', 'pressure', 'heartbeat', 'practical', 'mixture', 'cancer', 'mouse', 'neurons', 'macromolecular', 'tissue', 'simplex', 'tumor', 'anticancer', 'kidney', 'mini-guts', 'longitudinal', 'survey', 'influence maximization', 'scaling laws', 'cellular', 'men', 'fast', 'mapdamage2.0', 'mobile', 'eating'

Table S1. Web of Science search terms and exclusion criteria to identify set of 23 influential articles in infectious disease dynamics.

Authors	Article Title	Journal	Publication Year	Citation Rate
Diekmann, O. et al	On The Definition And The Computation Of The Basic Reproduction Ratio R0 In Models For Infectious-Diseases In Heterogeneous Populations	Journal Of Mathematical Biology	1990	97.50
Hethcote, H.W. et al	The Mathematics Of Infectious Diseases	Siam Review	2000	171.17
Pastor-Satorras, R. et al	Epidemic Spreading In Scale-Free Networks	Physical Review Letters	2001	182.04
Van Den Driessche, P. et al	Reproduction Numbers And Sub-Threshold Endemic Equilibria For Compartmental Models Of Disease Transmission	Mathematical Biosciences	2002	276.59
Newman, M.E.J. et al	Spread Of Epidemic Disease On Networks	Physical Review E	2002	112.45
Eubank, S. et al	Modelling Disease Outbreaks In Realistic Urban Social Networks	Nature	2004	67.05
Castillo-Chavez, C. et al	Dynamical Models Of Tuberculosis And Their Applications	Mathematical Biosciences And Engineering	2004	59.40
Lloyd-Smith, J.O. et al	Superspreading And The Effect Of Individual Variation On Disease Emergence	Nature	2005	87.21
Ferguson, N.M. et al	Strategies For Containing An Emerging Influenza Pandemic In Southeast Asia	Nature	2005	70.32
Keeling, M.J. et al	Networks And Epidemic Models	Journal Of The Royal Society Interface	2005	63.47
Ferguson, N.M. et al	Strategies For Mitigating An Influenza Pandemic	Nature	2006	84.28
Altizer, S. et al	Seasonality And The Dynamics Of Infectious Diseases	Ecology Letters	2006	60.33
Wallinga, J. et al	How Generation Intervals Shape The Relationship Between Growth Rates And Reproductive Numbers	Proceedings Of The Royal Society B-Biological Sciences	2007	57.00
Chitnis, N. et al	Determining Important Parameters In The Spread Of Malaria Through The Sensitivity Analysis Of A Mathematical Model	Bulletin Of Mathematical Biology	2008	57.75
Granich, R.M. et al	Universal Voluntary Hiv Testing With Immediate Antiretroviral Therapy As A Strategy For Elimination Of Hiv Transmission: A Mathematical Model	Lancet	2009	98.33
Fraser, C. et al	Pandemic Potential Of A Strain Of Influenza A (H1N1): Early Findings	Science	2009	95.40
Balcan, D. et al	Multiscale Mobility Networks And The Spatial Spreading Of Infectious Diseases	Proceedings Of The National Academy Of Sciences Of The United States Of America	2009	59.73
Diekmann, O. et al	The Construction Of Next-Generation Matrices For Compartmental Epidemic Models	Journal Of The Royal Society Interface	2010	78.14
Funk, S. et al	Modelling The Influence Of Human Behaviour On The Spread Of Infectious Diseases: A Review	Journal Of The Royal Society Interface	2010	58.21
Brockmann, D. et al	The Hidden Geometry Of Complex, Network-Driven Contagion Phenomena	Science	2013	76.55
Granell, C. et al	Dynamical Interplay Between Awareness And Epidemic Spreading In Multiplex Networks	Physical Review Letters	2013	67.91
Cori, A. et al	A New Framework And Software To Estimate Time-Varying Reproduction Numbers During Epidemics	American Journal Of Epidemiology	2013	56.91
Pastor-Satorras, R. et al	Epidemic Processes In Complex Networks	Reviews Of Modern Physics	2015	265.11

Table S2. List of influential seed papers for defining the authorship dataset.

Figure S1. Number of articles by year based on alternative definition of the field. Here, the field is defined as articles citing [23–25]. (A) Number of articles published per year globally. (B) Number of articles published per million residents in 2019.

Figure S2. Total number of articles published in the field of Infectious Disease Dynamics. Number of articles published in the field of Infectious Disease Dynamics from 2000-2019.

Appendix B: Uncertainty in authorship and citation practices

Figure S3. Gendered authorship and citation practices of the field with 95% confidence intervals. (A) Proportion of articles senior-authored by each gender from 2000 to 2019, with shaded region representing 95% confidence interval. (B) Proportion of articles authored by a given (lead & senior author) gender pair from 2000 to 2019, with shaded region representing 95% confidence interval. (C) Authorship gender breakdown of articles cited by men lead and senior authors (MM), with shaded region representing 95% confidence interval. (D) Authorship gender breakdown of articles cited by women lead and senior authors (WW), with shaded region representing 95% confidence interval.

Figure S4. Race/ethnicity authorship and citation practices of the field with 95% confidence intervals. (A) Proportion of articles senior-authored by each race/ethnicity group in the USA from 2000 to 2019, with shaded region representing 95% confidence interval. (B) Proportion of articles senior-authored by each race/ethnicity group in the UK from 2000 to 2019, with shaded region representing 95% confidence interval. (C) Racial/ethnic composition of articles cited by White lead and senior authors in the Global North, citing other papers authored in the Global North, with shaded region representing 95% confidence interval. (D) Racial/ethnic composition of articles published in the Global North cited by a lead and/or senior author of color, with shaded region representing 95% confidence interval.

Appendix C: Demographic comparison of lead to senior author

Figure S5. Globally, lead authors are more representative of gender demographics than senior authors. (A) Proportion of articles lead-authored by each gender from 2000 to 2019. Expected values (EP) for 2019 based on the global population are shown in the color bar at right. (B) Proportion of articles senior-authored by each gender from 2000 to 2019.

Figure S6. In both the USA and UK, lead authors are more diverse than their senior counterparts. (A) Proportion of articles lead-authored by each race/ethnicity group in the USA from 2000 to 2019. Expected values (EP) based on 2019 census data are shown in color bar to the right. (B) Proportion of articles lead-authored by each race/ethnicity group in the UK from 2000 to 2019. Expected values (EP) based on 2019 Office of National Statistics Data are shown in color bar to the right. (C) Proportion of articles senior-authored by each race/ethnicity group in the USA from 2000 to 2019. (D) Proportion of articles senior-authored by each race/ethnicity group in the UK from 2000 to 2019.

Appendix D: Gender results for the Global North

Figure S7. Authorship and citation practices in the Global North mirror global patterns with respect to gender. (A) Proportion of articles senior-authored by each gender from 2000 to 2019. Expected proportions based on (global north) population (EP) for 2019 are shown in the color bar at right. (B) Proportion of articles authored by a given (lead & senior author) gender pair from 2000 to 2019. Expected proportions based on the field (EF) in 2019 are shown in the color bar at right. (C) Rates of over- and undercitation by article lead and senior author gender based on authorship gender composition prior to publication date. (D) Authorship gender breakdown of articles cited in articles by men lead and senior authors.(E) Authorship gender breakdown of articles cited in articles written by women lead and senior authors.

Appendix E: Results based on different definition of the field

Figure S8. Authorship patterns remain largely unchanged, with some deviations in citation patterns as a result of changes in definition of the field. Here, the field is defined as articles citing [23–25]. (A) Proportion of articles senior-authored by each gender from 2000 to 2019. Expected values (EP) for 2019 based on global proportions are shown in the color bar at right. (B) Proportion of articles authored by a given (lead & senior author) gender pair from 2000 to 2019. Expected values (EF) produced based on the gender composition of the field in 2019 are shown in the color bar at right.(C) Rates of over- and undercitation by article lead and senior author gender based on authorship gender composition prior to publication date. (D) Authorship gender breakdown of articles cited in articles by men lead and senior authors.(E) Authorship gender breakdown of articles cited in articles written by women lead and senior authors.

Figure S9. Authorship and citation practices remained biased towards White authors even with a different definition of the field. Here, the field is defined as articles citing [23–25]. (A) Proportion of articles senior-authored by each race/ethnicity group in the USA from 2000 to 2019. Expected values (EP) based on 2019 census data are shown in color bar to the right. (B) Proportion of articles senior-authored by each race/ethnicity group in the UK from 2000 to 2019. Expected values (EP) based on 2019 Office of National Statistics Data are shown in color bar to the right. (C) Rates of over- and undercitation by article lead and senior author race/ethnicity based on authorship race/ethnicity composition prior to publication date. PoC author denotes that the lead and/or senior author was a person of color; sample size was too small to disaggregate by the author of color's position. Filled boxes denote results for articles published by authors with affiliations within the Global North citing other work from the Global North. Dotted boxes denote results for articles published by authors with affiliations within the Global North citing articles with any affiliation location. (D) Racial/ethnic composition of articles cited in articles written by White lead and senior authors in the Global North, citing other papers authored in the Global North.(E) Racial/ethnic composition of articles cited in articles written by a lead and/or senior author of color.

Appendix F: Results based on different gender/race inference tools

Figure S10. Authorship and citation results remain biased towards men authors even with a different gender inference model. (A) Proportion of articles senior-authored by each gender from 2000 to 2019. Expected values (EP) for 2019 based on global proportions are shown in the color bar at right. (B) Proportion of articles authored by a given (lead & senior author) gender pair from 2000 to 2019. Expected values (EF) produced based on the gender composition of the field in 2019 are shown in the color bar at right.(C) Rates of over- and undercitation by article lead and senior author gender based on authorship gender composition prior to publication date. (D) Authorship gender breakdown of articles cited in articles by men lead and senior authors.(E) Authorship gender breakdown of articles cited in articles written by women lead and senior authors.

Figure S11. Authorship and citation practices remain biased towards White authors even with a different race/ ethnicity inference model. (A) Proportion of articles senior-authored by each race/ethnicity group in the USA from 2000 to 2019. Expected values (EP) based on 2019 census data are shown in color bar to the right. (B) Proportion of articles senior-authored by each race/ethnicity group in the UK from 2000 to 2019. Expected values (EP) based on 2019 Office of National Statistics Data are shown in color bar to the right. (C) Rates of over- and undercitation by article lead and senior author race/ethnicity based on authorship race/ethnicity composition prior to publication date. PoC author denotes that the lead and/or senior author was a person of color; sample size was too small to disaggregate by the author of color's position. Filled boxes denote results for articles published by authors with affiliations within the Global North citing other work from the Global North. Dotted boxes denote results for articles published by authors with affiliations within the Global North citing articles with any affiliation location. (D) Racial/ethnic composition of articles cited in articles written by White lead and senior authors in the Global North, citing other papers authored in the Global North.(E) Racial/ethnic composition of articles cited in articles written by a lead and/or senior author of color.

Appendix G: Results based on different gender/race threshold probabilities

Figure S12. Authorship and citation practices remained biased towards men authors even with a lower, 60% probability threshold, for the gender inference model. (A) Proportion of articles senior-authored by each gender from 2000 to 2019. Expected values (EP) for 2019 based on global proportions are shown in the color bar at right. (B) Proportion of articles authored by a given (lead & senior author) gender pair from 2000 to 2019. Expected values (EF) produced based on the gender composition of the field in 2019 are shown in the color bar at right.(C) Rates of over- and undercitation by article lead and senior author gender based on authorship gender composition prior to publication date. (D) Authorship gender breakdown of articles cited in articles by men lead and senior authors.(E) Authorship gender breakdown of articles cited in articles written by women lead and senior authors.

Figure S13. Authorship and citation practices remained biased towards White authors even with a lower, 60% probability threshold, for the race/ethnicity inference model. (A) Proportion of articles senior-authored by each race/ethnicity group in the USA from 2000 to 2019. Expected values (EP) based on 2019 census data are shown in color bar to the right. (B) Proportion of articles senior-authored by each race/ethnicity group in the UK from 2000 to 2019. Expected values (EP) based on 2019 Office of National Statistics Data are shown in color bar to the right. (C) Rates of over- and undercitation by article lead and senior author race/ethnicity based on authorship race/ethnicity composition prior to publication date. PoC author denotes that the lead and/or senior author was a person of color; sample size was too small to disaggregate by the author of color's position. Filled boxes denote results for articles published by authors with affiliations within the Global North citing other work from the Global North. Dotted boxes denote results for articles published by authors with affiliations within the Global North citing articles with any affiliation location. (D) Racial/ethnic composition of articles cited in articles written by White lead and senior authors in the Global North, citing other papers authored in the Global North.(E) Racial/ethnic composition of articles cited in articles written by a lead and/or senior author of color.

Figure S14. Authorship and citation practices remained biased towards men authors even with a higher, 80% probability threshold, for the gender inference model. (A) Proportion of articles senior-authored by each gender from 2000 to 2019. Expected values (EP) for 2019 based on global proportions are shown in the color bar at right. (B) Proportion of articles authored by a given (lead & senior author) gender pair from 2000 to 2019. Expected values (EF) produced based on the gender composition of the field in 2019 are shown in the color bar at right.(C) Rates of over- and undercitation by article lead and senior author gender based on authorship gender composition prior to publication date. (D) Authorship gender breakdown of articles cited in articles by men lead and senior authors.(E) Authorship gender breakdown of articles cited in articles written by women lead and senior authors.

Figure S15. Authorship and citation practices remained biased towards White authors even with higher, 80% probability threshold, for race/ethnicity inference model. (A) Proportion of articles senior-authored by each race/ethnicity group in the USA from 2000 to 2019. Expected values (EP) based on 2019 census data are shown in color bar to the right. (B) Proportion of articles senior-authored by each race/ethnicity group in the UK from 2000 to 2019. Expected values (EP) based on 2019 Office of National Statistics Data are shown in color bar to the right. (C) Rates of over- and undercitation by article lead and senior author race/ethnicity based on authorship race/ethnicity composition prior to publication date. PoC author denotes that the lead and/or senior author was a person of color; sample size was too small to disaggregate by the author of color's position. Filled boxes denote results for articles published by authors with affiliations within the Global North citing other work from the Global North. Dotted boxes denote results for articles published by authors with affiliations within the Global North citing articles with any affiliation location. (D) Racial/ethnic composition of articles cited in articles written by White lead and senior authors in the Global North, citing other papers authored in the Global North.(E) Racial/ethnic composition of articles cited in articles written by a lead and/or senior author of color.

Appendix H: Results based on lower threshold to be considered part of IDD field

Figure S16. Citation practices remain biased toward articles with men lead and senior authors, despite lower, 75th percentile consensus threshold for a cited article to be considered in the field. (A) Rates of over- and undercitation by article lead and senior author gender based on authorship gender composition prior to publication date. (B) Authorship gender breakdown of articles cited in articles by men lead and senior authors.(C) Authorship gender breakdown of articles cited in articles written by women lead and senior authors.

Figure S17. Citation practices remain biased to papers with White lead and senior authors, even with a lower, 75th percentile consensus threshold for a cited article to be considered in the field. (A) Rates of over- and undercitation by article lead and senior author race/ethnicity based on authorship race/ethnicity composition prior to publication date. PoC author denotes that the lead and/or senior author was a person of color; sample size was too small to disaggregate by the author of color's position. Filled boxes denote results for articles published by authors with affiliations within the Global North citing other work from the Global North. Dotted boxes denote results for articles published by authors with affiliations within the Global North citing articles with any affiliation location. (B) Racial/ethnic composition of articles cited in articles written by White lead and senior authors in the Global North, citing other papers authored in the Global North.(C) Racial/ethnic composition of articles cited in articles written by a lead and/or senior author of color.

Appendix I: Regression diagnostics and sensitivity analyses

Using a chi-squared test, we determined that neither first and last author gender ($\chi^2 = 84.8$, $p < 0.0001$) nor first and last author race are independent ($\chi^2 = 1122.8$, $p < 0.0001$). This result justified the use of author pairings in our regression analysis. By regressing square-root transformed impact factor on paired author identity, we determined that both gender and JIF (Table S18) and race and JIF are collinear (Table S19). These findings justify the inclusion of a interactive effect between gender and JIF, and race and JIF but warrant caution in interpreting our results. By regressing $\log_{10}$citation rate on gender alone, and race alone, we found there was a weak association between author gender and article citation rate (Table S20) and a strong association between author race and article citation rate, if both authors were non-White (Table S21). Impact factor could be complicating these results and when included suggests that articles with both first and last authors women authors (Table S22) or authors of color (Table S23) have lower citation rates.

These results are not sensitive to the inclusion of self-citations (Table S25). When the analysis is restricted to articles published in the Global North or authorship articles that aren't sufficiently cited on their own are excluded, race does not affect citation rate.

Figure S18. Regression of square-root transformed Journal Impact Factor on author pairing gender suggests these predictors are slightly collinear.

Figure S19. Regression of square-root transformed Journal Impact Factor on author pairing race suggests these predictors are collinear.

Figure S20. Coefficients from linear regression predicting citation rate (num. citations/years since publication) based on article and author information. This analysis excludes 2023 Journal Impact Factor.

Figure S21. Coefficients from linear regression predicting citation rate (num. citations/years since publication) based on article and author information. PoC includes Asian, Hispanic, and Black authors. This analysis excludes 2023 Journal Impact Factor.

Figure S22. Coefficients from linear regression predicting citation rate (num. citations/years since publication) based on article and author information. This analysis ignores author race.

Figure S23. Coefficients from linear regression predicting citation rate (num. citations/years since publication) based on article and author information. PoC includes Asian, Hispanic, and Black authors. This analysis ignores author gender.

Figure S24. Coefficients from linear regression predicting citation rate (num. citations/years since publication) based on article and author information. PoC includes Asian, Hispanic, and Black authors. This is identical to the regression in the main text with the exclusion of a Global North indicator covariate.

Figure S25. Coefficients from linear regression predicting citation rate (num. citations/years since publication) based on article and author information. PoC includes Asian, Hispanic, and Black authors. This analysis permits self-citations.

Figure S26. Coefficients from linear regression predicting citation rate (num. citations/years since publication) based on article and author information. PoC included Asian, Hispanic, and Black authors. Limited to articles written by senior authors in the Global North.

Figure S27. Coefficients from linear regression predicting citation rate (num. citations/years since publication) based on article and author information. PoC includes Asian, Hispanic, and Black authors. Articles must be sufficiently cited as defined in the manuscript, with no exceptions for articles in the authorship dataset that have fewer citations.

Appendix J: Intersectional results

Figure S28. Intersectional authorship composition for lead and senior authors over time. This analysis uses the global authorship dataset from the main text.