# Gender and Regional Differences in Internal and International Migration of Scholars Worldwide

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The literature on international scholarly migration shows that women are less mobile and move over shorter distances. We extend this literature by using data from Scopus on over 29 million publications by over 18 million scholars worldwide, and simultaneously focus on internal and international movements. We find that women are also less mobile in internal migration. However, the gender gap is smaller and less prevalent gender gap than in international migration. We decompose the gender differences in migration between the contribution of the specific migration rates specific by field of science and composition. The differences in rates contribute to most of the gap in both internal and international migration in all periods and regions, except for internal migration in Latin America and the Caribbean. The field of science composition has different contributions for internal and international migration. In internal migration, the gap would be bigger in some regions if men and women had the same field of science composition. Our results have implications for promoting more equitable science by showing that there is a cumulative gender gap in migration of scientists, with men being more mobile both internationally and internally.

Migration of scholars | Gender differences | Internal migration | International migration | Bibliometric data

igration is more common among some sub-populations. For instance, the 28 share of international migrants among scholars, about 9% in 2014, is more 29 than three times that of the general population, roughly 3% since the 1960s (1). 30 31 While good quality data about migration of the general population is rare (2), this lack is also present in the case of high-skilled population and scholars (3). This 32 33 lack makes it difficult to advance our understanding of the migration patterns of this specific subgroup and how they differ from the pattern observed for the overall 34 population. Previous studies have addressed this lack by re-purposing bibliometric 35 data to estimate migration based on changes in scholars' affiliation addresses (3-5). 36 There is a promising future in demographic research using bibliometric data due 37 to new services accumulating metadata of scientific publications and innovative 38 methods to prepare cleaner data which is openly accessible for research. It aids 39 in addressing crucial scientific and policy-relevant questions that would have been 40 41 impossible to answer otherwise (6).

42 Gender inequalities are particularly pronounced in academic migration (5). Women in academia often face unique challenges, including but not limited to 43 gender bias, less access to networking opportunities (7), and a disproportionate 44 45 parenting penalty due to unequal distribution of household responsibilities (8-10). 46 These factors contribute to different international migration patterns among women 47 compared to men (5). The intersection of gender with other social and professional 48 factors —such as regional differences between the Global North and Global South creates a complex landscape of inequality, influencing the decision to enter, remain, 49 migrate, or exit academia (7, 11). 50

51 Inequalities in science are also evident when analyzed from a global perspective (12). Scholars from different regions experience varied productivity levels and 52 53 exposure to international collaborations (13). However, the idea that some countries 54 form the core of the global science and others constitute its peripheries has been criticized in recent studies (14). One of the criticisms of the center-periphery 55 56 model is the increase in collaboration between scholars from the global periphery 57 compared to center-center and center-periphery collaborations (15). In addition, recent decades have seen an increase in China's share of global scientific production 58 and the emergence of strong scientific systems in countries such as India, Iran, 59 South Korea, and Brazil (15). 60

The inequalities between regions can also be identified in the internal migration of scholars. Some countries with large science systems are able to hire and promote

### Significance Statement

A scientist's mobility can advance their career by providing access to resources, collaborators, and research opportunities. Previous studies have shown that women tend to migrate less frequently and over shorter distances. However, little is known about internal migration. Using bibliometric data, we estimated the gender gap in both types of migration simultaneously. We found that women are also less mobile at the subnational level, though the gender difference is smaller. Additionally, we demonstrated that the distribution of women and men across fields of science accounts for only a small part of the gender gap in international migration. In internal migration, in almost all regions, the gender gap would be even smaller if men and women were equally distributed across fields.

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scholars for permanent positions, making internal migration 125 more frequent and relevant than international migration 126 (16). This relationship between internal and international 127 migration aligns with the integrated methodology suggested 128 by (17). Throughout their academic careers, scientists are 129 likely to move between institutions for training and take up 130 research positions that are scattered across different locations. 131 To some extent, it can be argued that all researchers are 132 mobile, but what varies is the scale (7). Since women migrate 133 less internationally, the difference may be smaller internally 134 or even in favor of women. Consequently, examining internal 135 and international migration simultaneously is essential to 136 gaining a comprehensive understanding of the gender gap in 137 the migration of scholars. 138

In this study, we use bibliometric data from more than 139 29 million publications by more than 18 million scholars 140 worldwide to extend the literature on gender differences and 141 study it in internal and international scholarly migration. We 142 present gender differences in migration rates and show how 143 these differences vary by global region. We also decompose 144 these differences to determine the extent to which they are due 145 to differences in the composition of fields of science between 146 men and women. 147

## <sup>149</sup> Results

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At the country level, the gender gap in international migration 151 is larger and more prevalent than internal migration. We used 152 the female-to-male ratio of migration rates to assess the 153 gender gap in migration. We referred to the female-to-male 154 ratio as the Migration Gender Parity Index (mGPI), since 155 a value of one represents parity between men and women in 156 migration, controlling for differences in parity in the overall 157 population of scholars. An mGPI greater than one indicates 158 that women migrate proportionally more than men, while 159 an mGPI lower than one indicates that men migrate more 160 than women. We estimated the mGPI using 2,000 random 161 posterior samples, and the uncertainty intervals correspond 162 to the respectives percentiles of the results. The mGPI was 163 estimated at the country level by Field of Science (FOS) 164 and aggregated by country using the mean weighted by the 165 proportion of scholars in each field. In Figure 1, the maps 166 show, for the latest period (2013-2017), the size of the gender 167 gap in international and internal migration, along with the 168 level of uncertainty used to assess the presence of a difference. 169 The sankeys plots in the same figure show how countries 170 in each gender gap category for international migration are 171 classified for internal migration. The sankey plot for the 172 latest period helps interpret and compare the international 173 and internal migration maps. 174

Gender differences favoring men in international migration 175 were prevalent at the country level, which is consistent with 176 previous studies (5, 7). We found a gender gap favoring men 177 in international migration in more than 80% of countries 178 in all periods. Although the proportion of countries with 179 a gender gap has remained relatively stable over time, the 180 proportion of countries with a migration rate for men that 181 is more than 25% higher than for women has increased from 182 27% in the first period (1998-2002) to 42% in the latest period 183 (2013-2017).184

<sup>185</sup> The gender gap in internal migration was less prevalent than in international migration, and, when present, it was usually smaller. While a gender gap in international migration was identified in more than 80% of countries in every period, this proportion varied from 59% in the first period to 47% in the most recent period for internal migration. In addition to fewer countries having gender differences in migration, the number of countries with a difference greater than 25% in internal migration decreased from 27% to 6%. The decrease in countries with higher differences is the opposite of what was observed with international migration.

In most countries, migration rates for men were higher than for women, but there were a few exceptions where women's rates were higher. In the most recent period, only Uganda and Sri Lanka, women's migration rates were higher than men's migration rates. However, we did not find a difference in the international migration rates of women in two or more periods for the same country. When a consistent difference in international migration over time was identified, it was always in favor of men. Countries with higher rates of internal migration for women were more common than those with higher rates of international migration for women, but they were still exceptions. In the latest period, five countries had higher women's internal migration rates than men's internal migration: Spain, the Netherlands, China, Jordan, and Qatar. However, only the differences in the Netherlands and China were consistent over time. Since 2003-2007, women in the Netherlands have been more internally mobile than men, and since 2008-2012, women in China have been more internally mobile than men.

Our results indicate that when a difference is present in international migration, it does not necessarily extend to internal migration. However, the opposite is true. In other words, a gender gap in internal migration is generally accompanied by a gender gap in international migration. This pattern is evident in Fig. 1's Sankey plot, where the left bars are frequently linked to the yellow bars on the right, but the green bars on the left are almost exclusively linked to other green bars on the right. For example, in 2013-2017, Japan and Kenya were the only two exceptions. In these two countries, there was no difference in international migration rates, but internal migration rates for men were higher than for women. In the case of Japan, although no difference was identified for international migration, the internal migration rate for men was more than 25% higher than the rate for women.

Over time, the gender gap in internal migration either decreased or remained stable in nearly all regions. This resulted in a greater difference between internal and international gaps. To assess trends over time and differences between regions, we aggregated the migration rates for men and women using UN-M49 regions, and estimated the mGPI for each region. The regional rate was aggregated using the mean of the countries rates weighted by the population of scholars. Due to the number of countries with data, we combined some of the UN-M49 regions. Figure 2 shows the internal and international mGPI by period for each UN-49 region. The gender gap in migration is equal to the difference between the mGPI and one (represented in red line). The internal migration gender gap was smaller than the international one in all periods for regions in the Global North and in most periods for regions in the Global South. Regions in the Global North showed a consistent trend of decreasing



(2013-2017). A country is classified with a gender gap for women if the migration Gender Parity Index (mGPI) is bigger than one considering the uncertainty interval, for men if

the mGPI is smaller than one and not identified if the uncertainty interval (UI) includes one (see methods section for details). A gender gap in favor of women is represented

by purple, while a gap in favor of men is represented by green. The brightness of the color indicates the size of the gap and the level of uncertainty used to identify it. For

example, dark green indicates that men's migration rates are more than 25% higher than women's, and a 95% uncertainty interval does not include one. Second darkest green

indicates that the mGPI is smaller than one when using a 95% uncertainty interval, but men's rates are up to 25% higher than women's. The lightest green indicates that the

male migration rate is higher than the female migration rate, but since the uncertainty interval used is high, we do not represent the size of the difference. mGPIs that are not

different from one when an 80% interval is used are represented by light yellow. To assess the relationship between internal and international gender gap, C presents the

cross-classification of gender gap by type of migration of each period. For example, a flow from a green color to a purple color in the right indicates that a country with a gender

gap in favor of men in international migration has a gender gap in favor of women in internal migration in the same period.

internal gender gaps and relatively stable international gender
 gaps. In contrast, regions in the Global South experienced
 heterogeneous trends.

The five regions in the Global South presented different 376 patterns. Central and Southern Asia experienced an increase 377 in the international gender gap, followed by a decrease in the 378 internal gender gap. This pattern is similar to that observed 379 in Eastern and Southern Asia, except for the first period. 380 Despite this similar trend, the gender gap was smaller in 381 Eastern and Southeastern Asia than in Central and Southern 382 Asia. In Latin America and the Caribbean, both internal 383 and international gender gaps remained stable, with smaller 384 disparities in internal migration. The North Africa and 385 Western Asia region had a stable internal gender gap and 386 an increasing international gender gap. The largest gender 387 gap was observed in this region for international migration, 388 with a rate that was 64% [57%, 70%] higher for men than for 389 women. Sub-Saharan Africa was the only region with a clear 390 increase in the internal gender gap, and, during the 2008-2013 391 period, it was also the only region in which no difference 392 was identified between the gender gaps in international and 393 internal migration. In all other regions, the gender gap was 394 larger for international migration than for internal migration. 395



Fig. 2. Migration Gender Parity Index (mGPI) for international (yellow diamonds) and internal (blue squares) out-migration by 5-year period (only the beginning year of the interval is printed on the x-axis) for (A) the UN M49 regions classified as the Global North and (B) the UN M49 regions classified as the Global South. The markers are filled with white when the 95% uncertainty interval includes one.

425 Most of the gender gap is due to differences in field-specific 426 migration rates, except for internal migration in Latin America 427 and the Caribbean. Men and Women have different FOS 428 composition. From 2012 to 2017, 61% of men were in 429 agricultural, engineering, and natural sciences; 31% were 430 in medical and health sciences; and 7% were in humanities 431 and social sciences. During the same period, women were less 432 concentrated in agriculture, engineering, and natural sciences 433 (48%) than men, and, consequently, more concentrated in 434

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medical and health sciences (41%), and in humanities and 435 social sciences (11%). These percentages vary by time and 436 region, but the overall pattern remains consistent. Due to the 437 international nature of some fields, scholars are more likely 438 to migrate internationally in those fields than in others (7). 439 Therefore, part of the observed gender gap could be attributed 440 to the difference in FOS composition by gender. To assess the 441 extent to which the difference in FOS composition contributes 442 to the difference in migration rates between men and women, 443 we used a Kitagawa decomposition (18). We decomposed the 444 total gender difference in migration rate between FOS-specific 445 migration rates and FOS composition. 446

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The contribution of FOS composition to gender differences in international migration is higher in regions of the Global North than in the Global South. Considering only the points with a difference found using a 95% uncertainty interval, the contribution of the FOS composition ranged from 16% for Europe (1998–2003) to 27% for North America (2013–2017) in the Global North. In the Global South, the range was from -6% for Sub-Saharan Africa (2003–2007) to 14% for Northern Africa and Western Asia (1998–2002). The negative contribution indicates that the observed gender gap would be 6% larger if men and women had an FOS composition equal to the mean for both genders. In both the Global North and the Global South, FOS composition only contributes a small part to the observed gender gap. However, the highest contribution observed in Global South regions was smaller than the lowest value observed in Global North regions.

In the case of internal migration, the FOS composition contributed negatively or nearly zero to the gender gap in all regions of the Global North and most regions of the Global South. While scholars from the fields of agriculture, engineering, and natural sciences had higher international migration rates than scholars from other fields, scholars from medical and health sciences migrated with similar or greater intensity in internal migration. These differences in migration rates by field led to different contributions to the observed gender gap in internal and international migration. The Latin America and the Caribbean region was an exception with the composition contribution varying from 73% of the total gender difference in 1998-2002 to 50% in 2013-2018. Unlike in all other regions, scholars from the Medical and Health Sciences did not show an equal or higher rate than scholars from the Agricultural, Engineering, and Natural Sciences. The latter had a rate twice as high as the former.

#### Discussion

Our aim was to analyze differences between genders in the migration of scholars by focusing simultaneously on internal and international migration. We demonstrated that a gender gap also exists in the internal migration of scholars, though it is smaller than the gap observed in international migration. Women cannot make up for lower international mobility by moving more internally, which increases inequality in access to opportunities for career advancement. Decomposing the differences between rates and composition contributions revealed that the effects of differences in FOS composition on internal and international migration were different. Therefore, field-specific interventions to promote mobility may have different effects on internal and international gender gaps.



Fig. 3. Contribution of the differences in field-specific out-migration rates (dark orange) and field composition (light orange) to the total difference between female and male out-migration rates for international and internal migration by 5-year period for (A) the UN M49 regions classified as the Global North and (B) the UN M49 regions classified as the Global South.

Using bibliometric data has allowed us to study the internal migration of researchers worldwide simultaneously, but these data also have limitations present in our study. A limitation is related to the gender information. The gender of authors was inferred by an algorithm with different accuracy between countries and reduced to a binary gender classification without taking into account more diverse gender identities. Another limitation is that Elsevier's 2020 Scopus snapshot data mainly included English-language publications that undercover countries and regions, mostly from the Global South. 

Analyzing the mechanisms that explain the gender differ-ences in scholars' migration was beyond the scope of our work. The data used allows for longitudinal tracking of each scholar over the years, and future studies could use this feature to investigate how internal and international migration interact at the micro level, whether they complement, substitute, or precede each other. Another challenge for future studies would be to combine these data with other databases that allow us to study the mechanisms at work in the inequalities described in this study. 

## Materials and Methods

We use 29+ million "articles" and "reviews" by 18+ million authors indexed in Elsevier's 2020 Scopus snapshot provided to us by the German Competence Network for Bibliometrics (19) through the Max Planck Digital Library. Since bibliometric data have not been collected to study the migration of scholars, some pre-processing steps and assumptions are necessary to re-purpose the data (3, 6).

**Pre-processing and variables.** The first necessary step is to do the author name disambiguation, i.e, identify and link all papers of each unique author. For this step, we use the identification numbers provided by Scopus. The Scopus author name disambiguation algorithm identifies all publications of a single author in 94.4% of the cases and incorrectly merges by mistake the publications of two different authors in the same author id in only 1.9% of the cases (20).

The second necessary step is to disambiguate and geolocate affiliations at the subnational level. We used the data after disambiguation and geolocation of the authors' affiliation addresses at the subnational level, as described by Akbaritabar (21). In the analysis, we only considered authors whose affiliation location was identified at a subnational level and excluded those affiliation addresses that could not be geocoded.

After author and affiliation disambiguation, it is possible to identify the migration events. We use the change in the modal region of affiliation for each year to identify migration events. This strategy was found to be most suitable for analyzing scholars' migration with bibliometric data. A more detailed description of 

the migration identification process can be found in Akbaritabar 621 et al. (3). 622

We use the United Nations' "Standard Country or Area Codes 623 for Statistical Use" (UN-M49) to group countries. Additionally, we grouped the UN-M49 regions "Europe," "North America," and 624 "Oceania" into the "Global North" and the remaining regions 625 into the "Global South." This division is similar to the "More 626 developed regions" category in the UN Human Development Index 627 (excluding Japan) and is also consistent with the WEIRD\*/non-628 WEIRD categories. However, the Global North and Global South 629 classifications are more appropriate for studying the migration of scholars because these categories reflect country characteristics that 630 have influenced the historical development of the social sciences as 631 a whole (22).

632 We use gender data inferred from authors' names by a character-633 based neural network model. The model was trained using a Wikidata-dump containing approximately 4.3 million names, each 634 with an assigned gender. Of those names, 870,000 were female. 635 The model was trained using 80% of randomly chosen 870,000 male 636 and 870,000 female names and validated using the remaining 20%. 637 The model's overall accuracy was 93.5%. However, we acknowledge 638 that this algorithmic gender labeling, despite being scalable to large-scale data such as ours, is binary and neglects more diverse 639 gender identities. 640

We assign each author in the database a Field of Science (FOS). 641 We define an author's FOS as a time-invariant variable based 642 on their entire career. The assignment of an FOS consists of 643 two steps. First, we assign an FOS to each publication. Then, we select the FOS with the highest proportion for each author. 644 We use the OECD FOS classification (23), which has six major 645 fields. We aggregate the six major fields into three groups: 1) 646 Agricultural, Engineering, and Natural Sciences; 2) Humanities 647 and Social Sciences; and 3) Medical and Health Sciences.

648 Statistical model. We use a Bayesian hierarchical model to estimate 649 the gender and FOS-specific internal and international in- and 650 out-migration rates at the country level. Due to the size of the exposure in some countries, we grouped the data into four periods 651 of five years: 1998-2002, 2003-2007, 2008-2012, and 2013-2017. 652 Additionally, we grouped countries with fewer than ten thousand 653 person-years in a given period. We fit the same model twice: one 654 for out-migration and one for in-migration. 655

We assume that  $y_{it}^g$  – the number of observed in- and out-migrants for a gender, country, and field in a 5-year period – follows a Poisson distribution with an expected unobserved value  $\mu_{it}^{g}$ . In this notation, g represents the gender, i represents the combination of country and field, and t represents the five-year period.

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- 1. M Czaika, S Orazbayev, The globalisation of scientific mobility, 1970-2014. Appl. Geogr. 96, 1-10 (2018).
- 2. F Willekens, D Massey, J Raymer, C Beauchemin, International migration under the microscope. Science 352, 897-899 (2016).
- 3. A Akbaritabar, T Theile, E Zagheni, Bilateral flows and rates of international migration of 665 scholars for 210 countries for the period 1998-2020. Sci. Data 11, 816 (2024).
  - 4. A Miranda-González, S Aref, T Theile, E Zagheni, Scholarly migration within mexico
- 666 Analyzing internal migration among researchers using scopus longitudinal bibliometric data. 667 EPJ Data Sci. 9, 34 (2020).
- 668 5. X Zhao, A Akbaritabar, R Kashyap, E Zagheni, A gender perspective on the global migration of scholars. Proc. Natl. Acad. Sci. 120 (2023-03-07). 669
- 6. R Kashyap, et al., Digital and computational demography in Research Handbook on Digital 670 Sociology. (Edward Elgar Publishing), pp. 48-86 (2023).
- 671 7. CR Sugimoto, V Larivière, Equity for Women in Science: Dismantling Systemic Barriers to Advancement. (Harvard University Press), (2023-03-21). 672
- 8. AC Morgan, et al., The unequal impact of parenthood in academia. Sci. Adv. 7, eabd1996 673 (2021-02-01).
- 674 9. KA Shauman, Y Xie, Geographic mobility of scientists: Sex differences and family constraints. Demography 33, 455-468 (1996-11-01). 675
- 10. GE Derrick, PY Chen, T van Leeuwen, V Larivière, CR Sugimoto, The relationship between 676 parenting engagement and academic performance. Sci. Reports 12, 22300 (2022).
- 11. D Kaminski, C Geisler, Survival Analysis of Faculty Retention in Science and Engineering 677 by Gender. Science 335, 864-866 (2012). 678
- 12. A Akbaritabar, AF Castro Torres, V Larivière, A global perspective on social stratification in 679 science. Humanit. Soc. Sci. Commun. 11, 1-10 (2024).
- 13. JJ Lee, JP Haupt, Scientific globalism during a global crisis; Research collaboration and 680 open access publications on COVID-19. High. Educ. 81, 949-966 (2021-05-01). 681

$$\mathbf{y}_{it}^g \sim \mathrm{Poisson}(\mu_{it}^g)$$
 [1] <sup>684</sup>

We model the gender and FOS-specific log-migration rates  $m_{it}$ hierarchically. We reparameterize our hierarchical model using non-centered priors to avoid convergence issues (24).

$$\ln(\mu_{it}^g) = m_{it}^g + \ln(P_{it}^g)$$
<sup>[2]</sup>

$$m_{it}^g = \overline{m}_{r_{it}^g} + \sigma_{r_{it}^g} \cdot z_{it}^g \tag{3}$$

Where,  $P_{it}^g$  is the number of person-years,  $\overline{m}_{r_{it}}^g$  is the gender-and FOS-specific mean log-rate of the UN-M49 region of the country,  $\sigma_{r_{it}}^{g}$  is the standard deviation of the UN-M49 region.

We use weakly informative priors chosen by a prior predictive check as suggested by Gelman et al. (25).

$$z_{it}^g \sim \mathcal{N}(0, 1)$$

$$\overline{m}_{r_{i,i}^g} \sim \mathcal{N}(-3, 0.1)$$

$$\sigma_{r^g} \sim \mathcal{N}^+(0, 0.3)$$

Here,  $\mathcal{N}^+(0,\sigma)$  denotes a half-normal distribution, i.e., a normal distribution truncated at zero to ensure non-negativity.

#### \*Western, Educated, Industrial, Rich, and Democratic (WEIRD)

We implement the model in Stan v2.36.0 (26) and R v4.4.3 (27). We fit the model with 4 chains for 4,000 iterations and 2,000 warm-up iterations. We evaluate the convergence of the model by checking the estimated R-hat criterion and visually checking the rank plots (28).

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- 14. M Kwiek, W Roszka, Academic vs. biological age in research on academic careers: A large-scale study with implications for scientifically developing systems. Scientometrics 127, 3543-3575 (2022).
- 15. S Marginson, What drives global science? The four competing narratives. Stud. High. Educ. 47, 1566-1584 (2022-08-03).
- 16. A Akbaritabar, MJ Dańko, X Zhao, E Zagheni, Global subnational estimates of migration of scientists reveal large disparities in internal and international flows. Proc. Natl. Acad. Sci. 122 (year?) Publisher: Proceedings of the National Academy of Sciences
- 17. R King, R Skeldon, 'Mind the Gap!' Integrating Approaches to Internal and International Migration. J. Ethn. Migr. Stud. 36, 1619-1646 (2010).
- 18. SH Preston, P Heuveline, M Guillot, Demography: measuring and modeling population processes. (Blackwell Publishers), (year?).
- 19. M Schmidt, et al., The data infrastructure of the german kompetenznetzwerk bibliometrie: An enabling intermediary between raw data and analysis (2024).
- 20. J Baas, M Schotten, A Plume, G Côté, R Karimi, Scopus as a curated, high-quality bibliometric data source for academic research in quantitative science studies. Quant. Sci. Stud. 1, 377-386 (2020).
- 21. A Akbaritabar, A quantitative view of the structure of institutional scientific collaborations using the example of Berlin. Quant. Sci. Stud. 2, 753-777 (2021).
- 22. AF Castro Torres, D Alburez-Gutierrez, North and South: Naming practices and the hidden dimension of global disparities in knowledge production. Proc. Natl. Acad. Sci. 119. e2119373119 (2022).
- 23. OECD DSTI/EAS/STP/NESTI, Revised Field of Science and Technology (FOS) Classification in the Frascati Manual (OECD), (year?).
- R McElreath, Statistical Rethinking: A Bayesian Course with Examples in R and Stan, Texts 24 in Statistical Science. (Chapman & Hall/CRC, Boca Raton, FL), 1st edition, p. 487 (2015).
- 25. A Gelman, et al., Bayesian workflow (2020-11-03).

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687

745	26 Stan Development Team. Stan Modeling Language Likers Guide and Petersnes Manual	
745	20. Stan Development team, Stan Modelling Language Users Guide and Helerence Manual, (2020) Varcing 2 36	807
746	27. R Core Team. R: A Language and Environment for Statistical Computing (R Foundation for	808
747	Statistical Computing, Vienna, Austria), (2025).	809
748	28. A Vehtari, A Gelman, D Simpson, B Carpenter, PC Bürkner, Rank-normalization, folding,	810
740	and localization: An improved r <sup>^</sup> for assessing convergence of MCMC (with discussion).	010
749	Bayesian Analysis 16, 667–718 (2021) Publisher: International Society for Bayesian	811
750	Analysis.	812
751		813
752		814
753		815
755		010
754		816
755		817
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