

Niños furbito y niñas lo que sea

The gender gap in science and the consequences for the scientific knowledge that is created



Art by @oddrabid

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SUPERVISORS: ZAIDA CHINCHILLA-RODRÍGUEZ & NICOLÁS ROBINSON-GARCÍA

UNIVERSIDAD DE GRANADA

1. Who am I?
2. What do I do?
3. What am I doing here?

Who am I?

International Relations

History of Politics and Society Master

RRII + Scientometrics fellowship

PhD Scientometrics



Who am I?

International Relations

History of Politics and Society Master

RRII + Scientometrics fellowship

PhD Scientometrics



Not a
mathematician



What do I do?

PhD Scientometrics



He did his postdoc
with Tina!



Nicolás Robinson-García (Universidad de Granada)

Diversity in careers



Zaida Chinchilla-Rodríguez (IPP-CSIC)

Responsible metrics

What do I

Predicting the age of researchers using bibliometric data *

Gabriela F. Nane^a  , Vincent Larivière^b, Rodrigo Costas^c

Nicolas Robinson-Garcia  , Rodrigo Costas, Cassidy M. Lee^d, Thed N. van Leeuwen^e, Daniel G. Smith^f,                  

Meta-Research: Task specialization across research careers

Valuation regimes in academia: Researchers' attitudes towards their diversity of activities and academic performance [Get access >](#)

Nicolas Robinson-Garcia  , Rodrigo Costas, Gabriela F Nane, Thed N van Leeuwen



What do I do?

Scientometrics (bibliometrics, informetrics): Study of science literature

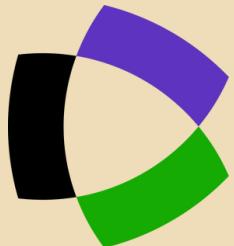
Mostly **quantitative**, but lately incorporating **qualitative** methodology



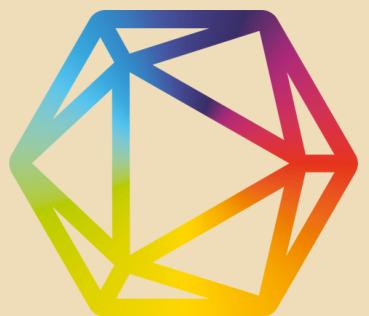
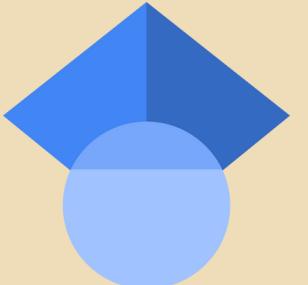
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Henk Moed



Scopus®



OpenAlex

arXiv

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2021, VOL. 12, NO. 4, 332–344
<https://doi.org/10.1080/19419899.2020.1729844>

Routledge
Taylor & Francis Group

OPEN ACCESS

What is gender, anyway: a review of the options for operationalising gender

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^aDepartment of Psychology, Lund University, Lund, Sweden; ^bDepartment of Psychology, Stockholm University, Stockholm, Sweden; ^cDepartment of Psychology, University of Gothenburg, Gothenburg, Sweden

ABSTRACT
In the social sciences, many quantitative research findings as well as presentations of demographics are related to participants' gender. Most often, gender is represented by a dichotomous variable with the possible responses of woman/man or female/male, although gender is not a binary variable. It is, however, rarely defined what is meant by gender. In this article, we deconstruct the concept 'gender' as consisting of several facets, and argue that the researcher needs to identify relevant aspects of gender in relation to their research question. We make a thorough exposition of considerations that the researcher should bear in mind when formulating questions about each facet, in order to exemplify how complex this construct is. We also remind the researcher that gender is not a binary category and discuss challenges in the balance between taking existing gender diversity into account and yet sorting participants into gender categorisations that function in statistical analyzes. To aid in this process, we provide an empirical example on how gender identity may be categorised when using a free-text response. Lastly, we suggest that other measurements than participants' gender might be better predictors of the outcome variable.

ARTICLE HISTORY
Received 1 March 2019
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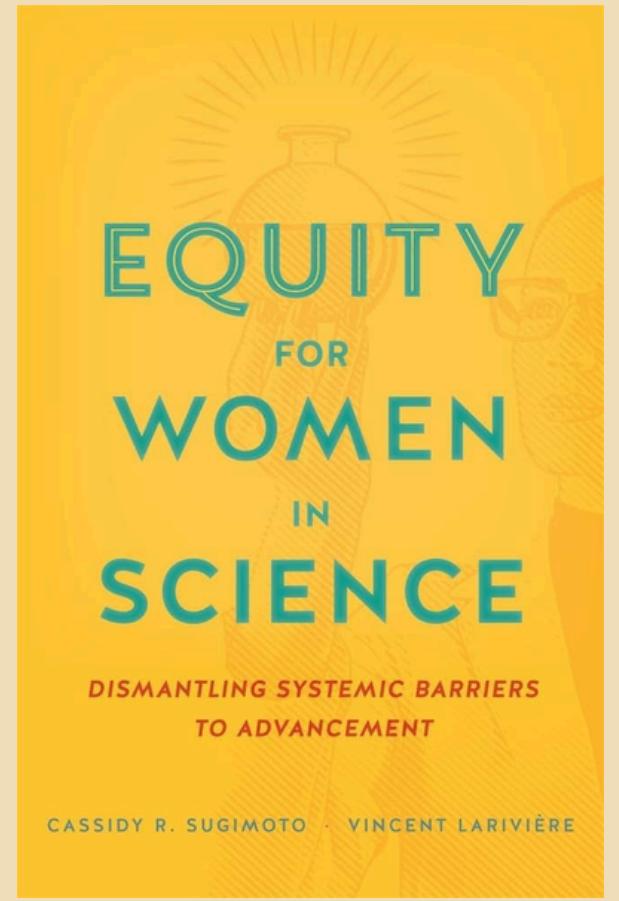
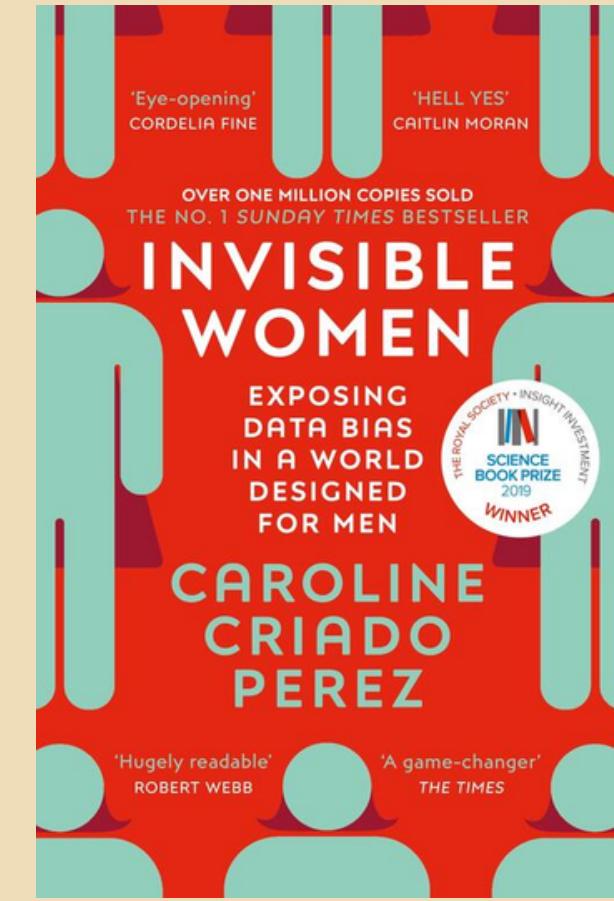
KEYWORDS
Gender; gender identity; transgender; research methods; cisnormativity

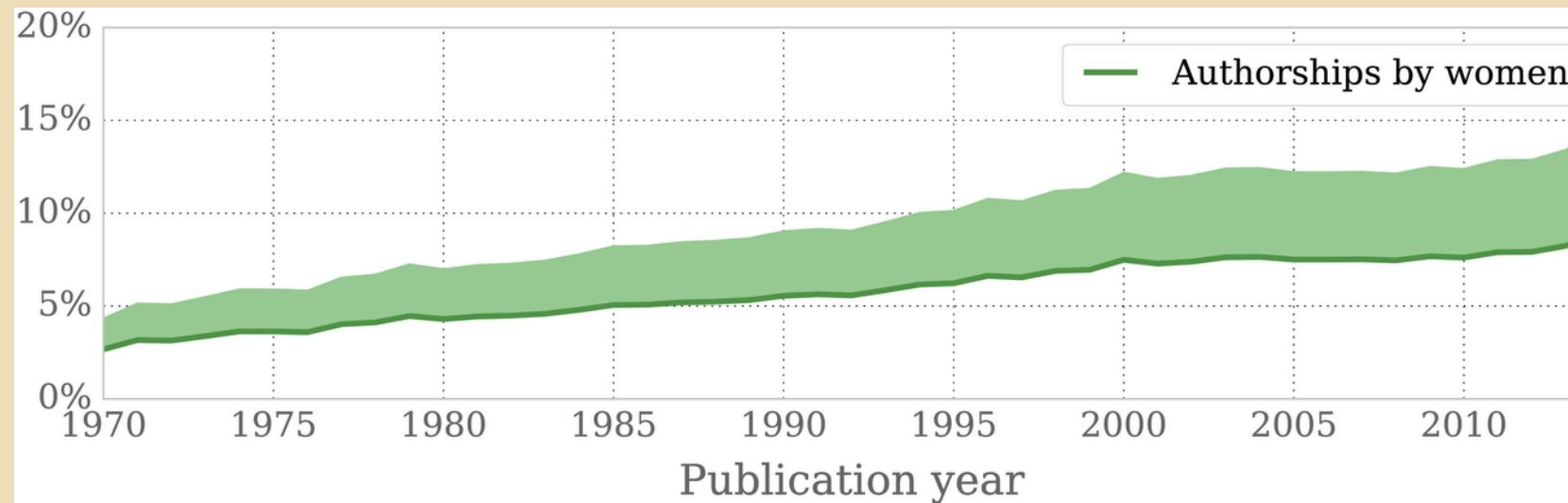
My interests

Interested in gender dynamics in science

It matters who gets to do the science
(*Situated knowledge*, Donna Haraway)

With metadata on authors names, we can infer gender and do cool things

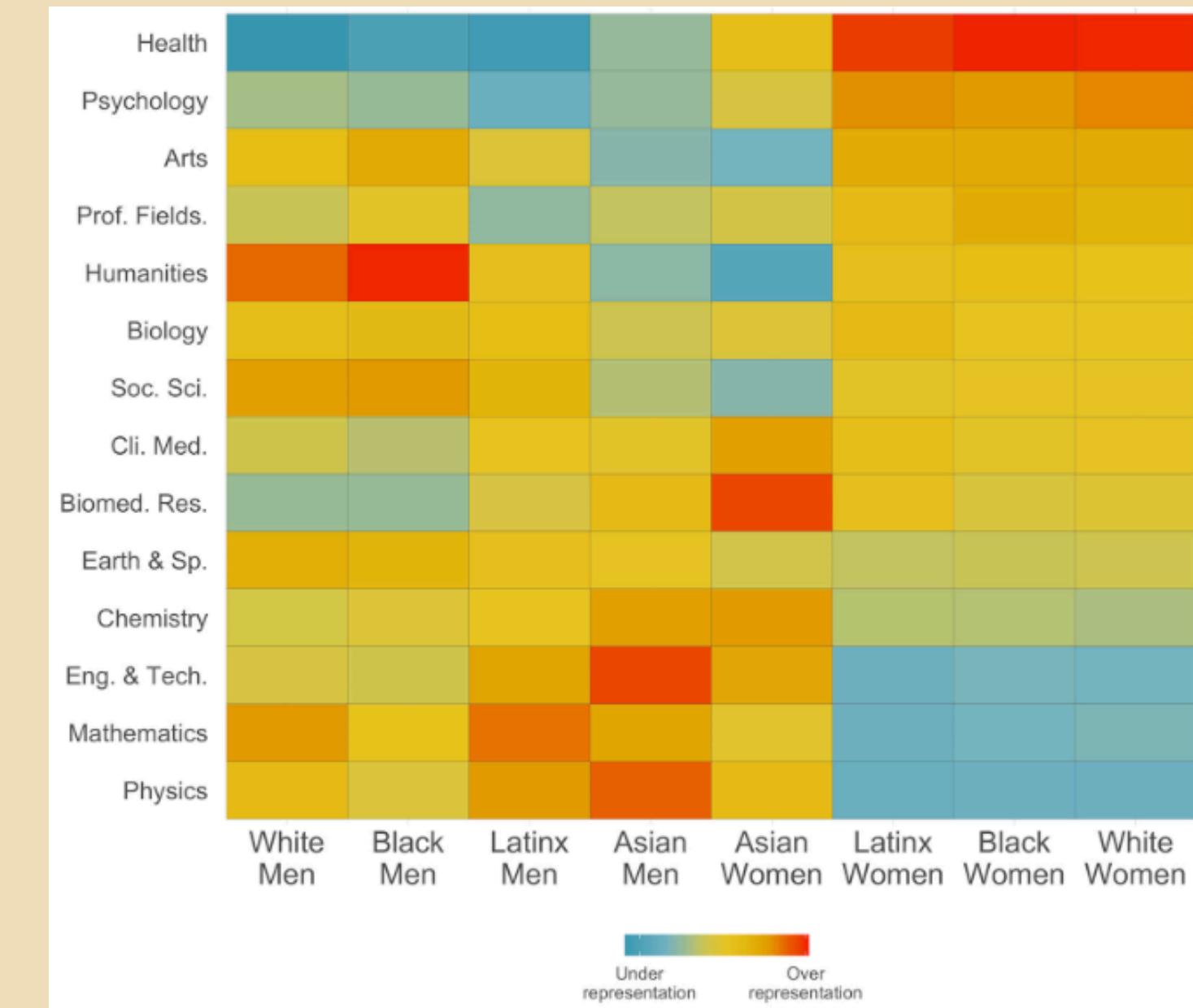




Mihaljević-Brandt, H., Santamaría, L., & Tullney, M. (2016). The Effect of Gender in the Publication Patterns in Mathematics. *PLOS ONE*, 11(10), e0165367.
<https://doi.org/10.1371/journal.pone.0165367>

Strumia case!

among both *M* and *F* authors. This suggests extending my considerations from possible sociological issues to possible biological issues.



Kozlowski, D., Larivière, V., Sugimoto, C. R., & Monroe-White, T. (2022). Intersectional inequalities in science. *Proceedings of the National Academy of Sciences*, 119(2), e2113067119.
<https://doi.org/10.1073/pnas.2113067119>

Differences

Authorship

Collaboration

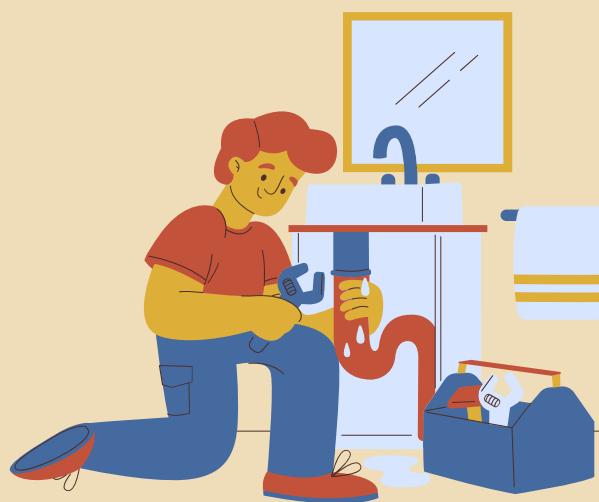
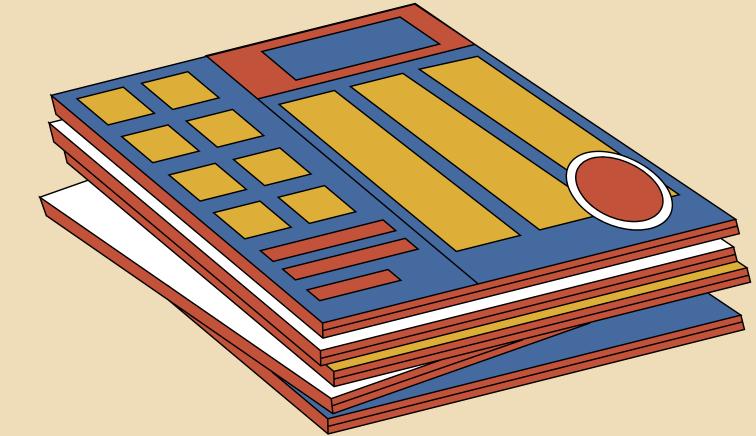
Citations

Number of publications

Academic status

Journals

Conferences



“ ”



Factors



Networking

Discrimination

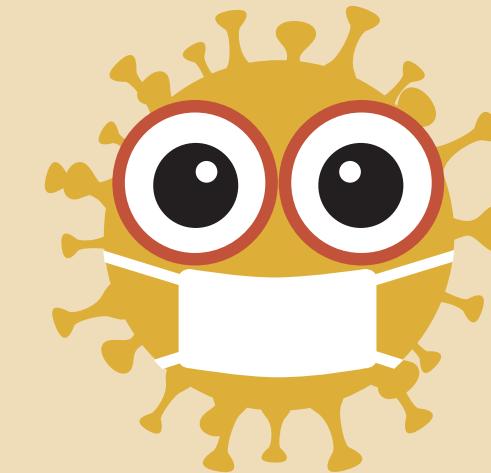
Stereotypes

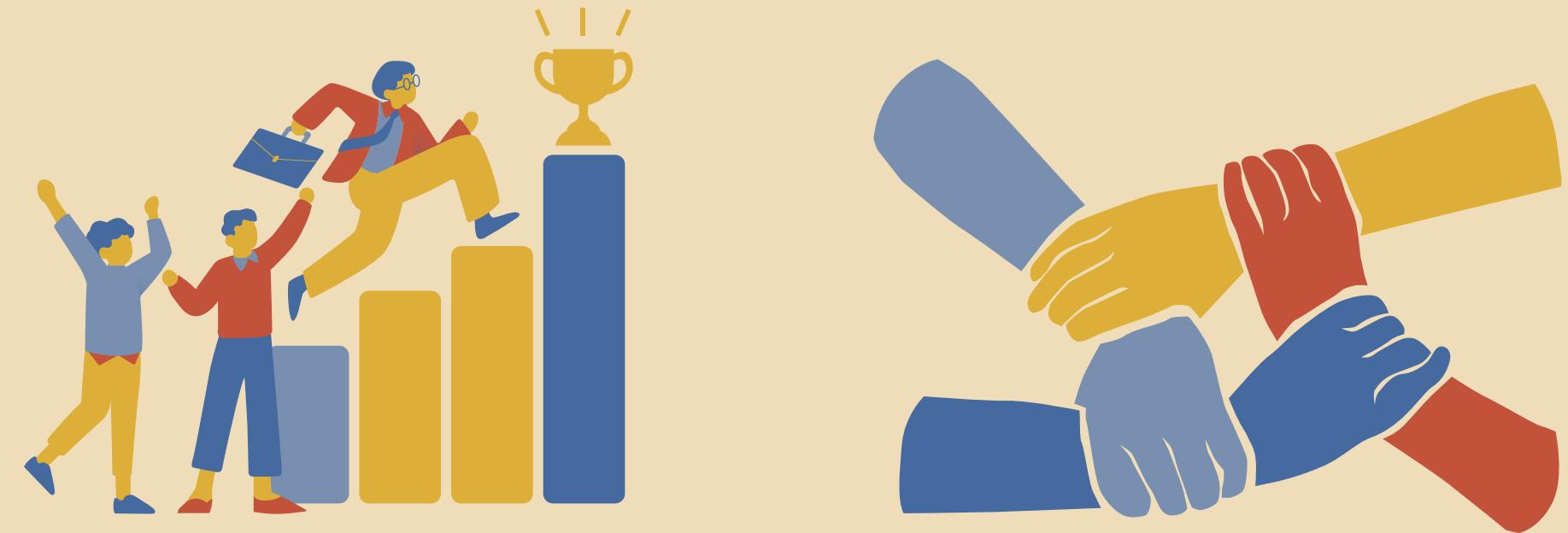
Evaluations

Families

Mobility

Coronavirus





Consequences

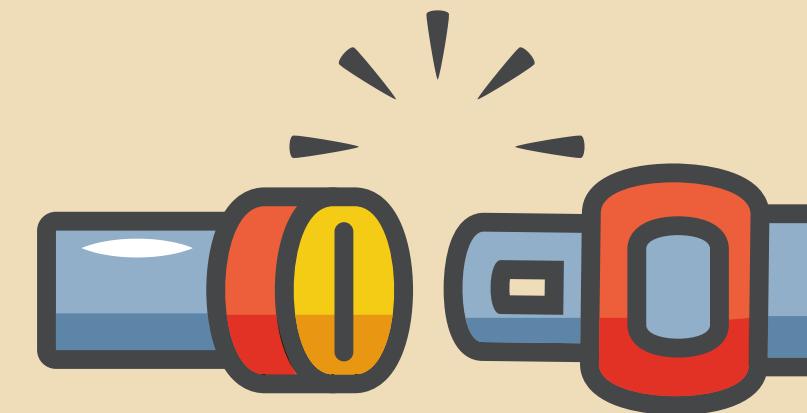
Women's careers

Diversity

Research topics/knowledge
that is generated

Research question

How do gender differences in science impact the kind of knowledge that is generated?



Objectives



Objective (1)
Map the differences



Objective (2)
Topic selection

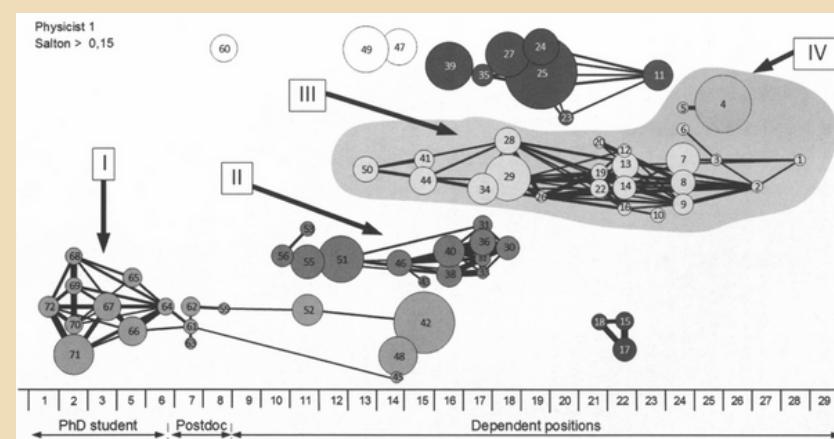


Objective (3)
Funding

Gender parity:
By fields & topics
By countries

Case study: How does gender affect the
selection of a research topic?

Are there differences in funding obtained
for male-dominated and female-
dominated topics? Do funding agencies
perpetuate gender inequalities when
funding male-dominated topics?



What am I doing here?

As part of objective 1

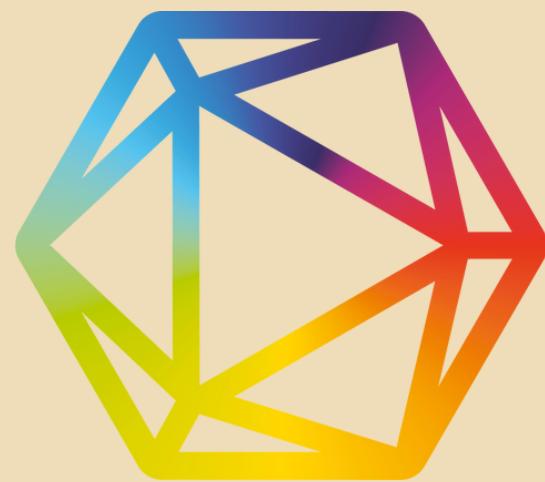
What leads to gender parity in a country?

Objective: **predicting parity** in science worldwide and by country and **identify national factors** influencing the increase or decrease of gender parity



What am I doing here?

Data 1990-2020



+



+



6m researchers

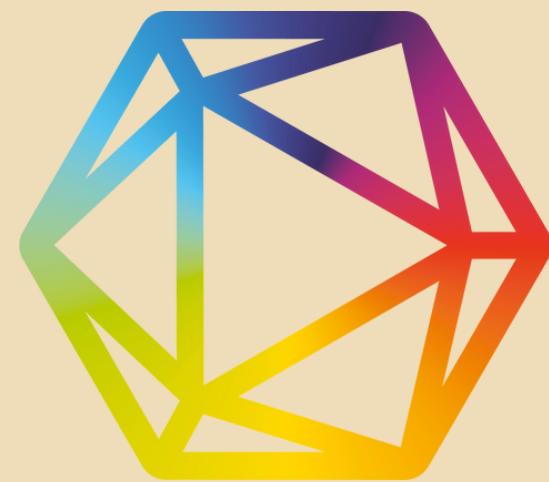
Educational variables
Social variables
Economic related variables

What am I doing here?



Data 1990-2020

Missing data



+



+



6m researchers

Educational variables
Social variables
Economic related variables

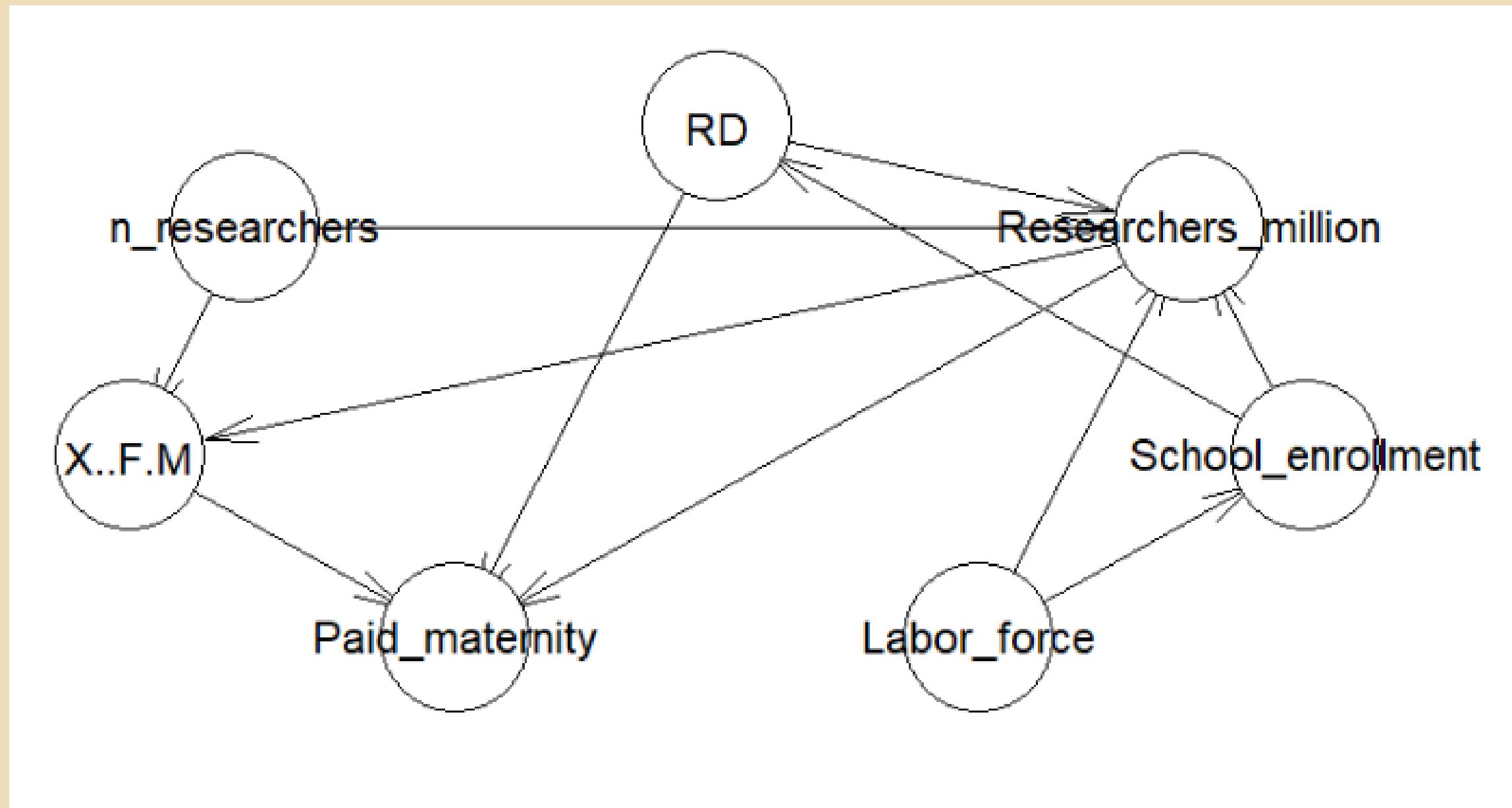
Country	Year	n_researchers	% F/M	Region	Arts and Hum.	Business, Adm.	Agriculture, Forestry	share of grads	Education progr.	Engineering, Tech.	Health and V.	Information Sci.	Natural Scienc.	other fields	Services proj.	Social Scienc.	unknown or	tion of time	of paid mate	Adolescents	Labor force p.	Labor force p.	School enrol.	Researchers	Research and devel.				
NL	2021	11913	35,454,231	Europe & Central Asia	112	0,2300100029	70,482	60,569
NL	2020	12158	34,385,443	Europe & Central Asia	112	0,2217800008	70,119	59,887	0,9980400205	2,304,500,103	..
NL	2019	11084	33,164,094	Europe & Central Asia	112	0,1816799939	70,571	59,812	1,000079989	2,184,350,014	..
NL	2018	10507	32,082,708	Europe & Central Asia	112	0,263310999	70,042	59,098	0,9990500212	2,138,799,06	..
NL	2017	9775	31,016,909	Europe & Central Asia	54,417,82	47,338,61	55,368,63	29,292,93	77,507,88	23,056,07	75,929,68	14,472,67	43,751,99	61,402,33	51,977,92	68,172,19	55,789,47	..	112	0,2495899945	69,979	58,797	0,9995899887	2,178,570,032	..
NL	2016	9544	29,798,113	Europe & Central Asia	112	0,1832399964	69,498	58,882	1,000069978	2,150,810,003	..
NL	2015	8897	29,601,228	Europe & Central Asia	112	0,1675599962	69,709	58,803	0,9995800257	4548,1363	2,14805999	..
NL	2014	8688	28,489,119	Europe & Central Asia	57,552,13	46,720,29	54,585,37	25,316,58	78,824,05	24,210,43	74,266,69	16,572,12	40,910,36	62,262,48	55,428,22	68,546,09	44,811,46	..	112	..	67,992	58,481	..	4519,15376	2,173,300,028	..
NL	2013	8432	27,348,699	Europe & Central Asia	58,122,86	46,453,54	54,439,13	24,608,95	79,622,83	22,476,98	74,464,96	15,407,04	43,805,31	62,596,4	55,250,04	69,137	44,834,77	..	112	..	70,341	59,255	..	4561,23136	2,156,05998	..
NL	2012	7870	25,914,058	Europe & Central Asia	58,806,55	..	54,478,46	..	78,098,63	22,121,96	75,072,66	52,970,89	..	58,191,13	14,722,22	112	0,08731099993	71,084	59,298	0,999810994	4372,42304	1,918,270,018	..	
NL	2011	7385	23,532,174	Europe & Central Asia	58,741,79	..	54,946,3	..	79,930,29	21,083,34	74,757,13	53,639,27	..	57,449,21	..	112	0,1846099985	69,578	58,772	0,9992399812	3874,97708	1,881,30998	..	
NL	2010	6812	22,256,149	Europe & Central Asia	57,378,58	46,814,16	54,860,36	20,921,23	79,707,78	20,453,92	74,789,68	10,922,59	40,251,57	62,789,29	54,573,93	67,714,51	56,435,64	..	112	..	68,954	58,715	0,9965599775	3228,97939	1,704,040051	..
NL	2009	5096	20,957,628	Europe & Central Asia	58,708,01	..	51,684,72	..	81,118,11	19,799,24	75,249,52	54,721,13	112	..	69,926	58,805	0,9890390877	2833,03227	1,665,099959	..	
NL	2008	5265	18,800,260	Europe & Central Asia	58,544,75	..	51,805,04	..	81,352,39	19,200,88	75,462,77	56,909,80	..	58,363,84	..	112	..	69,53	58,351	0,9852700233	3070,6673	1,622,689982	..	
NL	2007	4874	16,836,131	Europe & Central Asia	58,095,15	..	50,107,07	..	80,853,89	19,104,72	75,569,36	57,369,02	..	57,112,07	..	112	..	68,901	57,412	0,9824799895	3101,31227	1,670,30007	..	
NL	2006	4593	16,345,907	Europe & Central Asia	58,011,85	..	50,722,22	..	80,077,24	18,446,79	75,328,69	58,351,46	..	56,5	16,944,44	112	..	66,651	56,157	0,9790599942	3240,63531	1,740,689986	..	
NL	2005	4104	14,412,353	Europe & Central Asia	60,805,08	..	46,584,12	21,012,21	79,559,92	17,445,05	75,523,23	9,395,48	42,132,51	..	62,019,91	..	48	..	112	1,552639961	66,849	55,643	0,9772109988	2930,13944	1,773,880,005	..
NL	2004	3513	13,222,297	Europe & Central Asia	59,276,89	48,058,57	48,444,07	20,157,34	79,211,98	17,208,14	78,385,4	8,937	40,623,11	63,344,77	58,907,89	59,189,47	112	..	72,212	56,965	0,9747200012	2977,96208	1,789,010,048	..
NL	2003	2941	11,049,429	Europe & Central Asia	58,693,79	49,111,46	44,957,77	..	78,326,68	14,011,54	76,208,38	63,462,06	55,129,01	55,992,41	..	15,138,89	112	..	73,536	56,51	0,9764800072	2709,46121	1,783,900,023	..
NL	2002	2743	10,808,398	Europe & Central Asia	59,513,04	47,523,87	45,856,63	18,490,97	78,104,38	14,010,93	74,667,74	13,130,7	35,893,1	62,678,26	59,257,14	55,429,4	18,918,92	..	112	1,049430013	73,287	55,919	0,9788299799	2729,1512	1,745,429993	..
NL	2001	2523	9,777,347	Europe & Central Asia	58,668,67	47,709,12	40,257,81	18,049	75,995,32	13,639,43	74,667,73	13,755,16	34,407,08	61,755,59	58,572,26	53,299,26	41,178,47	15	112</td										

Bayesian Networks

BNs are graphical models which **capture dependencies between multiple variables**. The dependencies are first-hand modelled through **arcs** from **nodes** (which represent random variables) and the structure of the BNs can be **learned completely from data**.

(Scutari & Denis, 2021)





Dynamic Bayesian Networks?

Thank you! Any questions?