

Scientists & Women Scientists: Exploring Gender Biases in Institutional Category Systems



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Introduction

- Typicality is influenced by social biases and stereotypes.
- People = men bias: men are the more generic or typical example of a human
- There is often an asymmetry in how typical versus atypical members of a category are labelled.
- We can measure bias towards women in a category system as an asymmetry in the explicit labelling of categories for women versus men.

We explore how typicality manifests in two real-world settings: The Library of Congress Subject Headings (LCSH) and Wikipedia.

Goal 1: Determine if these systems mirror, mitigate, or exaggerate biases found in language use and if this has changed with time.

Goal 2: Use nationality and ethnicity categories (identities) to understand how the people = men bias manifests in a system and job categories to show how frequency influences typicality.

Category Systems

The LCSH

Postmodern theology

[BT83.597] BT Theology, Doctrinal

Postmodernism

UF Post-modernism

Postmodernism (Philosophy)

- Controlled vocabulary of terms that represent book subjects.
- First published in 1909.

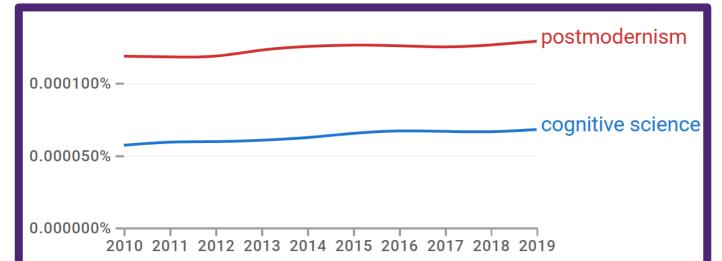
Wikipedia

Categories: Evil clowns | Black comedy | Fictional clowns | Halloween costume | Horror villains Postmodernism | Stock characters | Urban legends

<u>Categories</u>: Cognition | Cognitive Science | Branches of linguistics | Applied psychology

- User-created terms for grouping related articles.
- Introduced in 2004.

Google Ngram



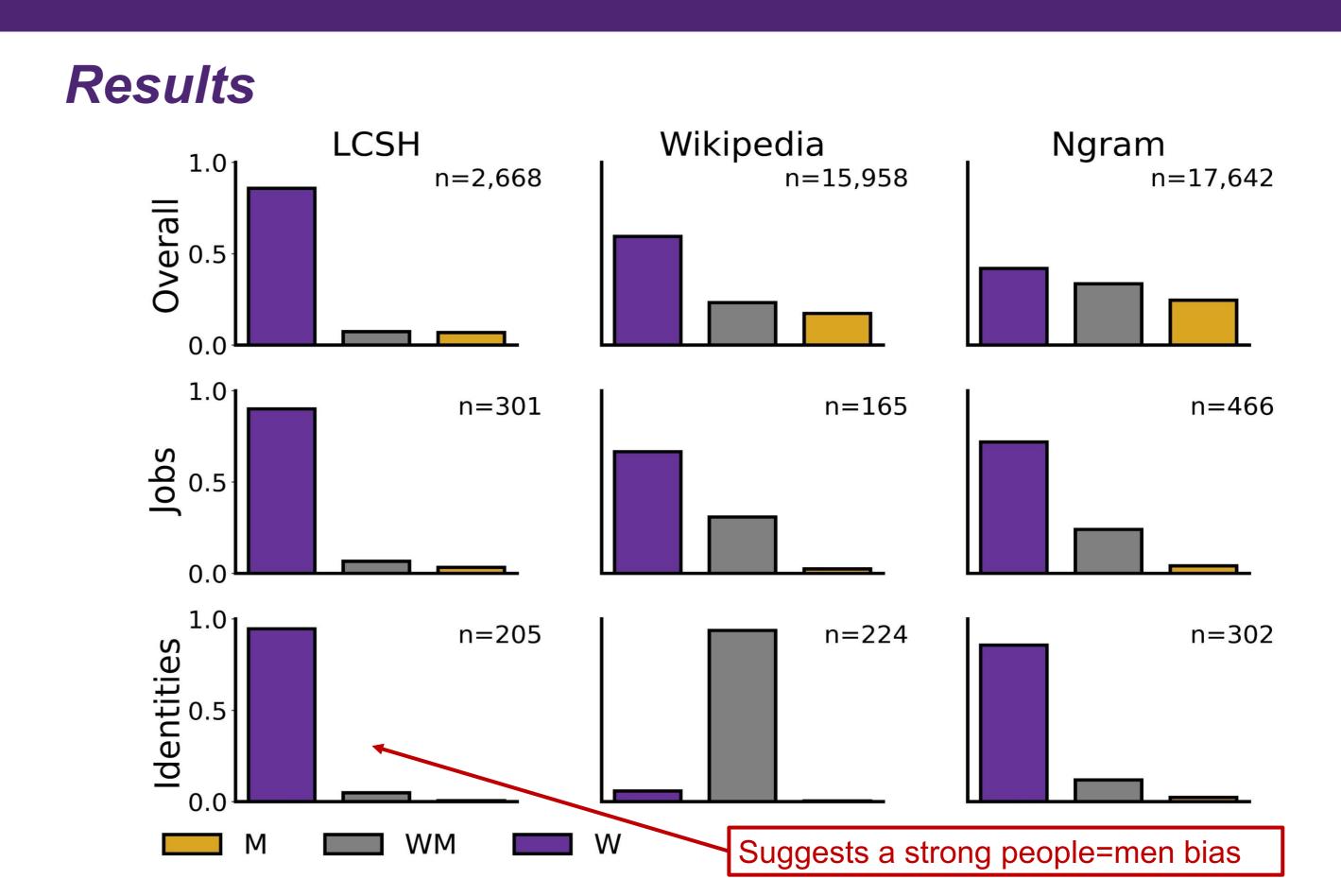
- Term frequencies from millions of published books.
- A proxy for (written) language use.

Gender Bias in Institutional Category Systems

Methods

- Find all categories and Ngrams that contain a gendered term ('men', 'women', 'male', or 'female').
- 2. Group terms in three ways:
 - a) W = terms for women, but not men (e.g., Women and war)
 - **b) M** = terms for men, but not women (e.g., Male beauty)
 - c) WM = terms for both (e.g., Young women; Young men)
- 3. Create subsets for job categories (e.g., Male high school teachers) and identities (e.g., Canadian women).

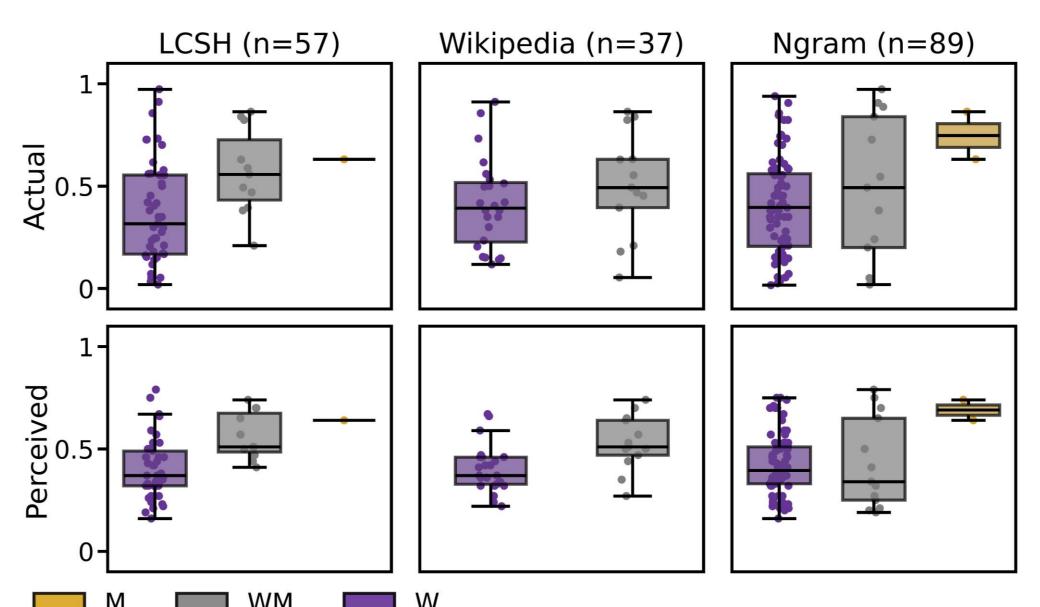
RQ: To what extent do gender biases exist in each system? How do they compare?



Both the LCSH and Wikipedia exhibit stronger gender biases than language use overall, with LCSH being the most biased.

Base Rates and Gender Bias

RQ: Can gender base rates (frequency) predict if jobs are grouped into W, WM, or M?



Actual:

Gender breakdown of US jobs.

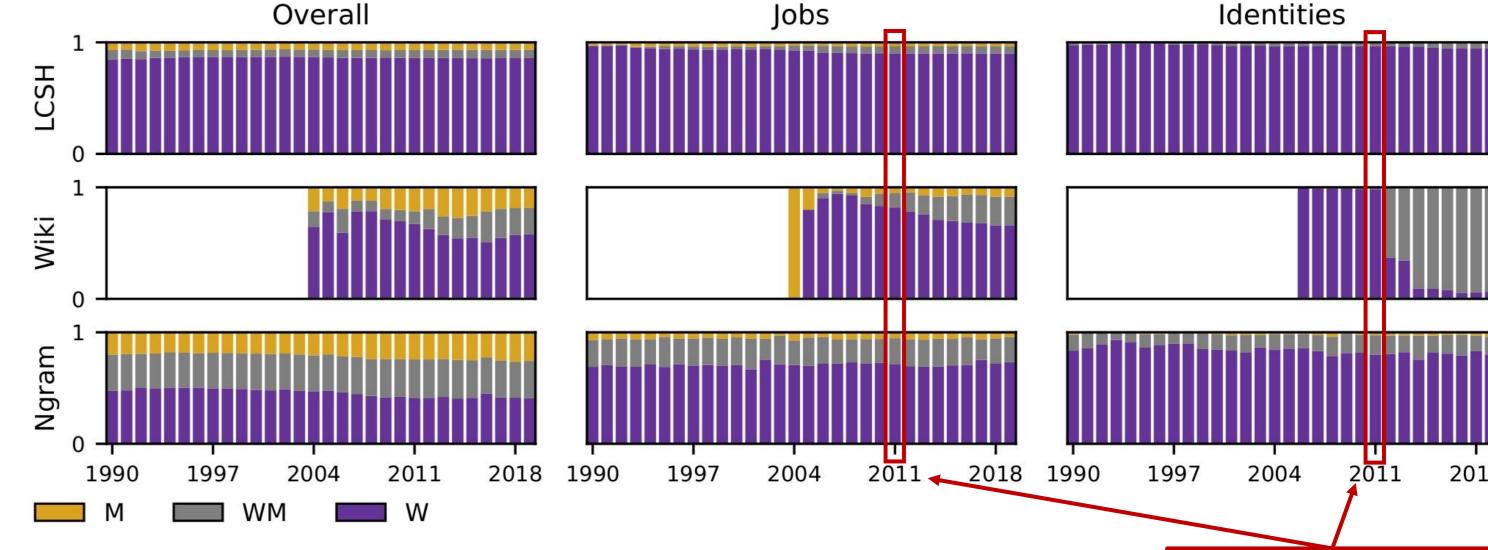
Perceived:

People's beliefs about the gender breakdown of jobs.

People's beliefs about gender base rates predict asymmetries in Wikipedia and the LCSH better than actual base rates.

Gender Bias Across Time

RQ: Have asymmetries in category labelling decreased with time?



Excluding Wikipedia, gender biases have remained unchanged over the past 30 years. Papers highlighting gender bias in Wikipedia emerge

Takeaways

- The LCSH tends to exaggerate gender bias found in language whereas Wikipedia mirrors it (jobs) or mitigates it (identities).
- People's beliefs about the gender breakdown of jobs (perceived) frequency) can explain some of the asymmetries found.
- Although unequal labelling can address historical exclusion, it can also reinforce whose membership is assumed and whose is not.
- Wikipedia has done a better job than the LCSH at mitigating gender bias despite criticisms of the LCSH existing since the 70s.