



Measuring Fairness in Ranked Results: An Analytical and Empirical Comparison

Amifa Raj



People and
Information
Research
Team



Grant No. IIS 17-51278



2022

Raj, A., & Ekstrand, M. D. (2022, July). Measuring Fairness in Ranked Results: An Analytical and Empirical Comparison. In Proceedings of the 45th International ACM SIGIR Conference on
[DOI 10.1145/3477495.3532018](https://doi.org/10.1145/3477495.3532018)



information access systems



query

All Images News Shopping Videos More Tools

About 2,850,000,000 results (0.53 seconds)

https://www.technologyinthearts.org › what-is-an-infor...

What Is An Information Access System? - Technology In The Arts

Nov 2, 2021 — An Information Access System (IAS) provides a secure way to manage who has access to which information your organization holds.

People also ask

What are the 4 types of access control?

What are the 3 types of access control?

What is an example of information access?

What are access systems?

Feedback

goodreads

Home My Books Browse Community

Search books

Recommendations > Fiction Genre

Here are some books we recommend based on the books you've added in this genre. Other readers with similar interests have enjoyed them. How to improve your recommendations...

updated: NOV 10, 2022 11:38AM

View covers | list



Want to Read buttons and star ratings for the first row of books



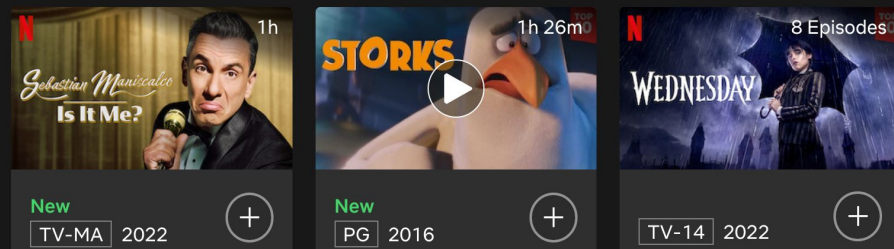
NETFLIX

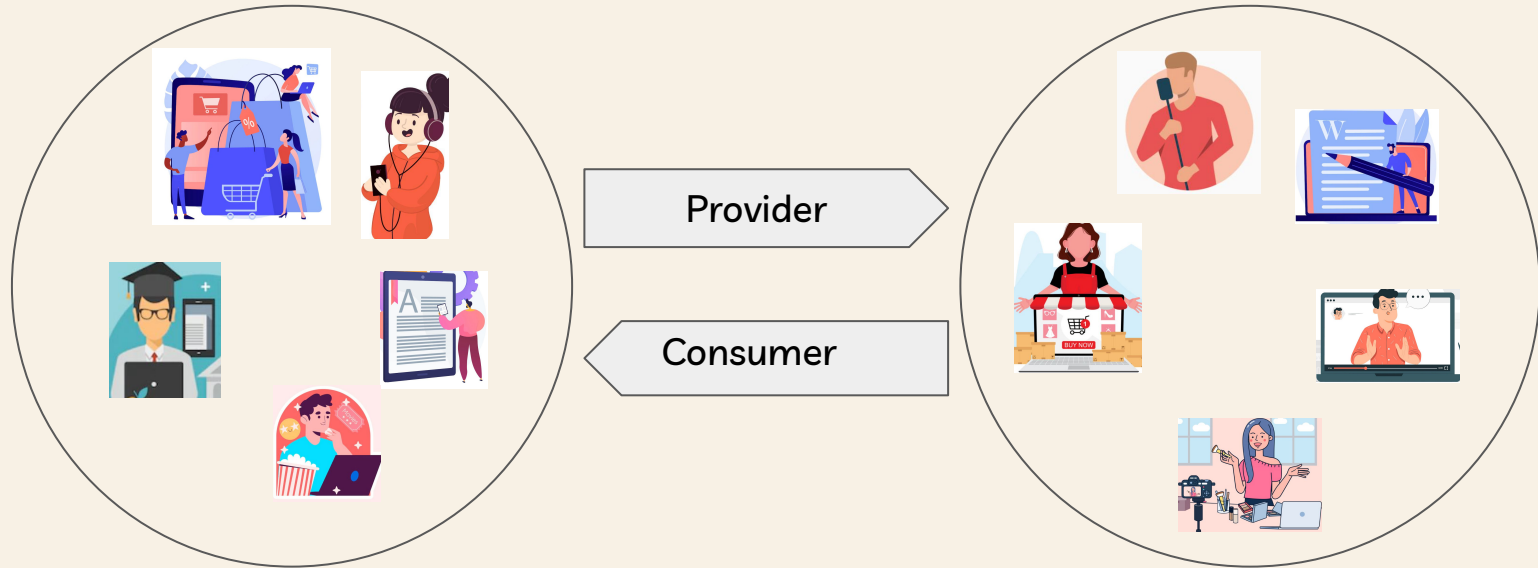
Home TV Shows Movies New & Popular My List Browse by Languages

Trending Now











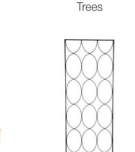

More Like This





- WUGU1
- Soils
- Seeds & Accessories
- Outdoor Gardening
- Organic Gardening
- Lawn Maintenance**
- Landscape Supplies
- Grass Seeds
- Wheelbarrows
- Landscape Edging
- Landscape Rocks
- Retaining Wall Blocks
- Lawn Mowers
- Riding Lawn Mowers
- Sheds & Outdoor Storage
- Mulch Calculator
- Featured Outdoor Essentials**
- Bug Spray
- Patio Furniture
- Picnic Tables






Plant Your Garden


				
Shrubs	Perennials	Annuals	Trees	Edible Garden Plants
				
Indoor Plants	Plant Care	Raised Garden Beds	Plant Support	Garden Tools

Search





all | title | author | genre

[Books](#) | [Groups](#) | [Quotes](#) | [People](#) | [Listopia](#)

- Page 1 of about 4444 results (0.05 seconds)
- 
Short-Cut Math (Dover Books on Mathematics)
 by Gerard W. Kelly
 ★★★★★ 3.77 avg rating — 128 ratings — published 1969 — 7 editions
 - 
Math Coloring Books for Kids: Coloring Book for Minecrafters: The Best Relaxing Color by Number Pixel Art Game - Math Activities Addition Subtraction Workbook (Amazing Jumbo Activity Book Grade 1-3, Ages 2-4, 4-8, 9-12, Boys, Girls, Teen & Adults) (Unoffi
 by Math Coloring Books For Kids, Math Activity Books
 ★★★★★ 2.23 avg rating — 4 ratings — 1 edition
 - 
My Math Books: Think Math!
 by Harcourt School Publishers (Creator)
 ★★★★★ 4.13 avg rating — 15 ratings — published 2006 — 1 edition
 - 
Probability (Young Math Books)
 by Charles F. Linn
 ★★★★★ 3.80 avg rating — 13 ratings — published 1972 — 2 editions
 - 
Read Any Good Math Lately?: Children's Books for Mathematical Learning, K-6
 by David J. Whitin
 ★★★★★ 4.00 avg rating — 12 ratings — published 1992 — 1 edition


[Home](#) | [TV Shows](#) | [Movies](#) | [New & Popular](#) | [My List](#) | [Browse by Languages](#)

Suspenseful Movies

			
---	--	---	--

IDEAS • TECH BIAS

Google Has a Striking History of Bias Against Black Girls

BY **SAFIYA NOBLE** MARCH 26, 2018 4:30 PM EDT

Dr. Safiya U. Noble is the author of *Algorithms of Oppression: How Search Engines Reinforce Racism* and is an assistant professor of communication at the University of Southern California, Annenberg School of Communication & Journalism. She is a partner in *Stratelligence* and co-founder of the *Information Ethics & Equity Institute*.

IDEAS

[ENGINEERING](#) | [NEWS RELEASES](#) | [RESEARCH](#) | [TECHNOLOGY](#)

April 9, 2015

Who's a CEO? Google image results can shift gender biases

[Jennifer Langston](#)

UW News

ARTIFICIAL INTELLIGENCE

LinkedIn's job-matching AI was biased. The company's solution? More AI.

ZipRecruiter, CareerBuilder, LinkedIn—most of the world's biggest job search sites use AI to match people with job openings. But the algorithms don't always play fair.

By **Sheridan Wall & Hilke Schellmann**

June 23, 2021

Fairness and discrimination in retrieval and recommendation systems

[MD Ekstrand](#), [R Burke](#), [F Diaz](#) - ... and Development in Information Retrieval, 2019 - dl.acm.org
... to assessing **fairness** in particular problem settings. • Engaging **fairness**, ... , and identify new research questions on the **fairness** of recommendation systems.
☆ Save 🔄 Cite Cited by 33 Related articles All 2 versions

Measuring fairness in ranked results: An analysis of fairness in recommendation systems

[A Raj](#), [MD Ekstrand](#) - ... on Research and Development in Information Retrieval, 2019 - dl.acm.org
... to be relevant to the user's **information** need. Evaluating **access system**'s behavior beyond accuracy or utility constraints.
☆ Save 🔄 Cite Cited by 4 Related articles 🔄

Addressing bias and fairness in search systems

[R Gao](#), [C Shah](#) - ... on research and development in Information Retrieval, 2019 - dl.acm.org
... of bias in such **systems**. And yet, we lack tools and techniques for bringing **fairness** to search **systems**. At the same time, those tools are often limited.
☆ Save 🔄 Cite Cited by 13 Related articles All 2 versions

Fairness of user clustering in MIMO non-orthogonal multiple access systems

[Y Liu](#), [M Elkashan](#), [Z Ding](#) - ... - IEEE Communications Letters, 2019 - IEEE
... **fairness** issue among single-antenna users, a power allocation scheme is investigated under perfect channel state **information** (...).
☆ Save 🔄 Cite Cited by 186 Related articles All 13 versions

Counteracting bias and increasing fairness in search and recommender systems

[R Gao](#), [C Shah](#) - ... ACM Conference on Recommender Systems, 2020 - dl.acm.org
... on recommender **systems** for all kinds of **information** needs — ... **fairness**-aware analysis, methods for evaluating **fairness**-aware recommender **systems**, and the impact of encoded **fairness** ...
☆ Save 🔄 Cite Cited by 16 Related articles All 3 versions 🔄

Fairness issues in AI systems that augment sensory abilities

Addressing social bias in information retrieval

[J Otterbacher](#) - International Conference of the Cross-Language Evaluation Forum, 2018 - Springer
Journalists and researchers alike have claimed that IR systems are socially biased, leading to results to users that perpetuate gender and racial stereotypes. In this position paper, we discuss the impact of social bias on IR systems and propose a framework for addressing social bias in IR systems.
☆ Save 🔄 Cite Cited by 11 Related articles All 4 versions Import into BibTeX

Cognitive biases in search: a review and reflection of cognitive biases in Information Retrieval

[L Azzopardi](#) - ... on human information interaction and retrieval, 2021 - dl.acm.org
... on cognitive **biases** in the field of **Information Retrieval**; and (2) ... studies examined cognitive **biases** in search, ... and negative — of cognitive **biases** in search.
☆ Save 🔄 Cite Cited by 31 Related articles All 5 versions Import into BibTeX

[HTML] Statistical biases in information retrieval metrics for systems

[A Bellogin](#), [P Castells](#), [I Cantador](#) - Information Retrieval Journal, 2017 - Springer
... the application of **Information Retrieval** methodologies to ... **biases** arising from the application of **Information Retrieval** metrics to recommendation tasks, namely sparsity and popularity.
☆ Save 🔄 Cite Cited by 119 Related articles All 8 versions Import into BibTeX

Fairness in information retrieval

[A Lipani](#) - ... on Research and Development in Information Retrieval, 2016 - dl.acm.org
... The offline evaluation of **Information Retrieval** (IR) systems is ... This method is called **pool bias**, which is the ... Thereby, this **bias** affects the evaluation of a system.
☆ Save 🔄 Cite Cited by 17 Related articles All 3 versions Import into BibTeX

[HTML] Bias–variance analysis in estimating true query model parameters

[P Zhang](#), [D Song](#), [J Wang](#), [Y Hou](#) - Information Processing & Management, 2019 - Elsevier
... **bias**–variance tradeoff, which is a fundamental theory in statistics. We formalize the **bias**–variance regarding **retrieval** ... the **bias**–variance tradeoff will occur, and propose a method to address it.
☆ Save 🔄 Cite Cited by 14 Related articles All 8 versions Import into BibTeX

Impact of query sample selection bias on information retrieval system ranking

[M Melucci](#) - 2016 IEEE International Conference on Data Mining, 2016 - IEEE
... We report that the **ranking** ... **bias**, while the **ranking** of the average systems is much more affected. We also report that the measure of **bias** depends on the **retrieval** measure used to **rank** ...
☆ Save 🔄 Cite Cited by 8 Related articles All 2 versions Import into BibTeX

Rank-biased precision for measurement of retrieval effectiveness

[A Moffat](#), [J Zobel](#) - ACM Transactions on Information Systems (TOIS), 2008 - dl.acm.org
... **information retrieval** systems has been proposed. These are typically intended to provide a quantitative single-value summary of a document **ranking** ... metric, **rank-biased** precision, that ...
☆ Save 🔄 Cite Cited by 579 Related articles All 3 versions Import into BibTeX

On the robustness and discriminative power of information retrieval metrics for top-N recommendation

[D Valcarce](#), [A Bellogin](#), [J Parapar](#) - Proceedings of the 12th ACM Conference on Recommender Systems, 2018 - dl.acm.org
... of different **ranking** metrics which were previously used in **Information Retrieval** and are now used ... to the sparsity **bias** and the robustness to the popularity **bias** independently. Sparsity ...
☆ Save 🔄 Cite Cited by 63 Related articles All 3 versions Import into BibTeX 🔄

[HTML] Statistical biases in information retrieval metrics for recommender systems

[A Bellogin](#), [P Castells](#), [I Cantador](#) - Information Retrieval Journal, 2017 - Springer
... In this context, we propose and analyse two approaches to mitigate popularity **bias** on the measured **ranking** quality, providing theoretical and empirical evidence of their effectiveness. ...
☆ Save 🔄 Cite Cited by 119 Related articles All 8 versions Import into BibTeX

cwl_eval: An evaluation tool for information retrieval

[L Azzopardi](#), [P Thomas](#), [A Moffat](#) - ... Development in Information Retrieval, 2019 - dl.acm.org
... We provide implementations of a number of recent metrics, including Time **Biased** Gain, U-**Rank**, and **Rank** at **Rank** **i**. Using these weights the expected utility (EU) of a **ranking** is computed as ...
☆ Save 🔄 Cite Cited by 17 Related articles All 5 versions Import into BibTeX 🔄

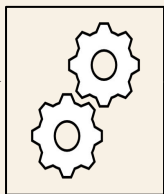
Controlling fairness and bias in dynamic learning-to-rank

[M Malik](#), [A Singh](#), [J Wang](#), [T Lochbihari](#) - ... in Information Retrieval, 2020 - dl.acm.org

Bias

- Difficult to define
- Domain dependent
- Systematic and unfair discrimination against certain **individual** or **group** entities by denying opportunity and assigning unfair outcomes
- Group (Sensitive Attributes) and Individual Fairness

User Request



1



2



3



4



5



6



Exposure

Attention

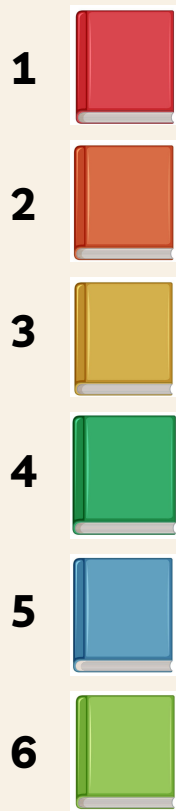


Consumer

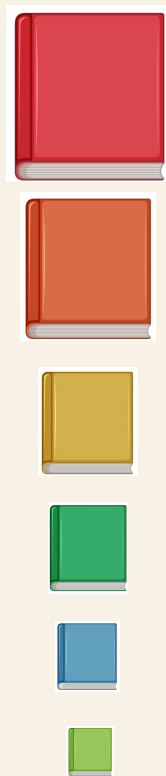
Ranked Items

Providers

Disparate Exposure



Ranked List



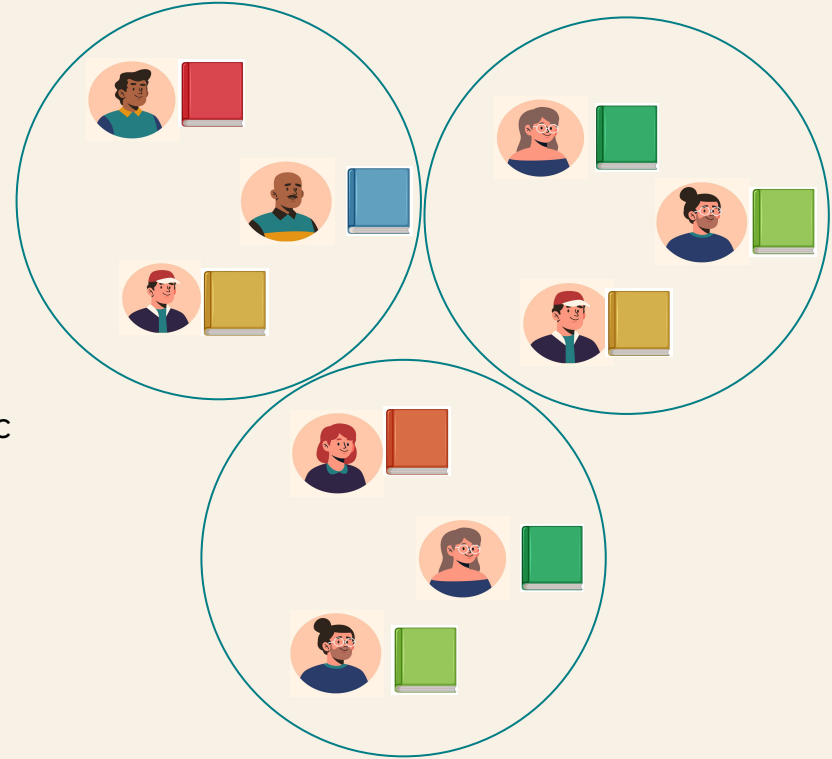
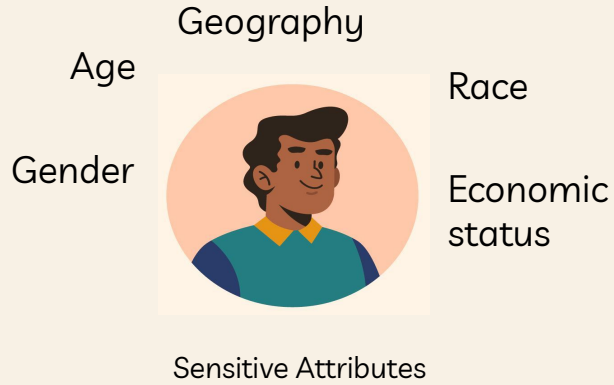
Attention



Fairness Positioning



Provider Fairness



Group Fairness

Fair Ranking Metrics

PreF Δ

(Yang et. al.; SSDBM '17):

AWRF

(Sapienzynski et. al.; WWW'19)

EEL, EED, EER

Diaz et.al.; CIKM'20)

FAIR

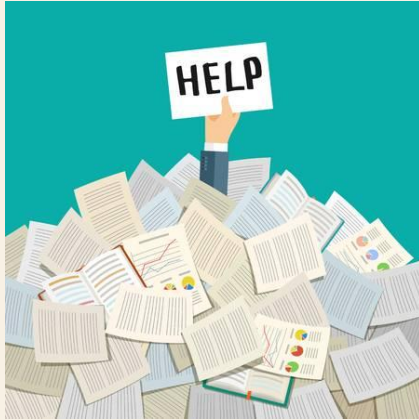
(Zehlike et.al.; CIKM'17)

IAA

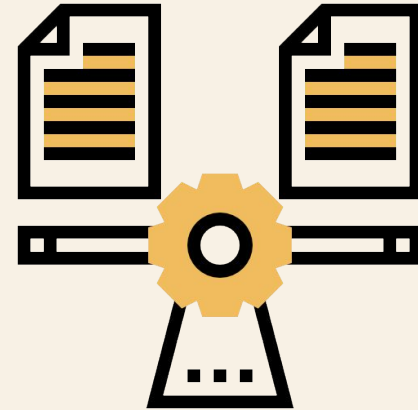
(Biega et. al.; SIGIR'18)

DP, EUR, RUR

(Singh et.al.; KDD'18)



**Several Fair
Ranking Metrics**



**No Comparative and
Comprehensive
Analysis**

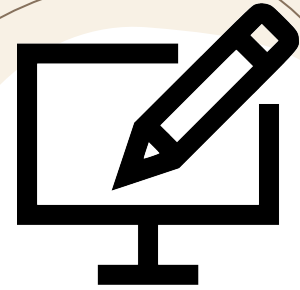
Why is the Problem a Problem?



Finding suitable Metrics



Differences among the
Metrics



Implementation in
Real-world IAS dataset

Research Questions

RQ1. What are the conceptual differences among the fair ranking metrics?

RQ2. What is needed to apply these metrics to real IAS?

RQ3. What are the design decisions and parameters involved, and how sensitive are the resulting metrics to those decisions?

RQ4. What are the empirical differences in how these metrics assess the relative fairness of different recommendation algorithms or retrieval runs?

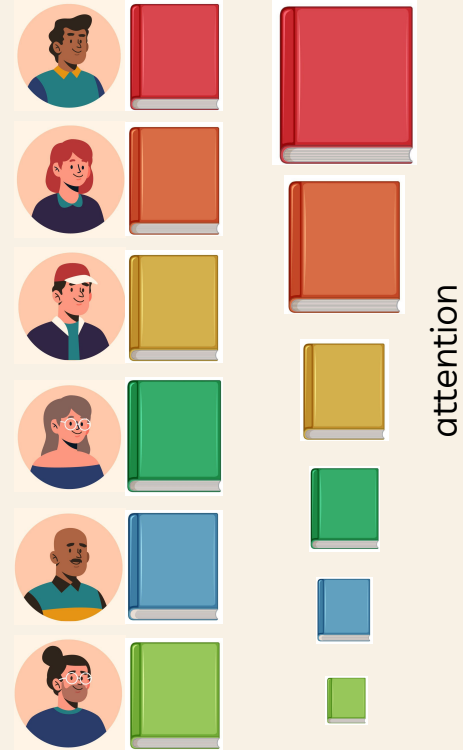
Research Tasks

Conceptual
Analysis of
Fair Ranking
Metrics

Implementing
Fair Ranking
Metrics in
Real-World
IAS Datasets

Sensitivity
Analysis

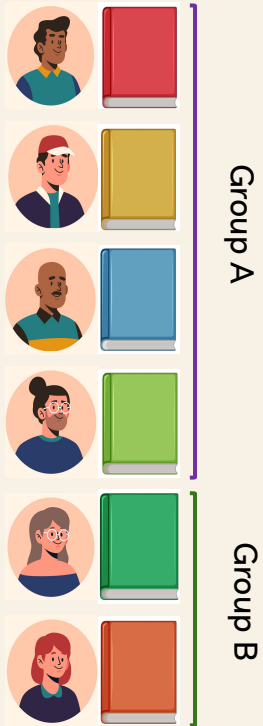
Metrics Design Decomposition



Fairness Goal

What does it mean to be fair?

Fairness Goal

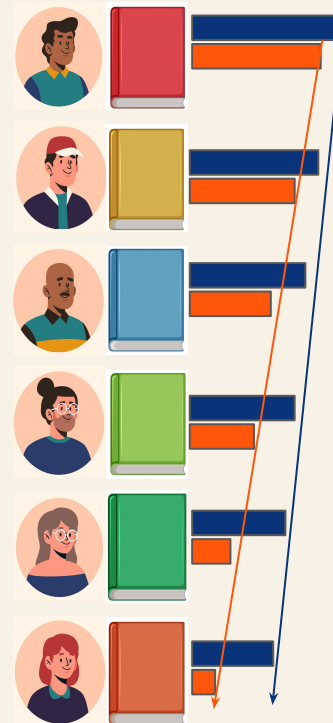


PreF Δ , FAIR, AWRP, DP, EED

Item position should not be affected by membership

Statistical Parity

Relevance
Exposure/Attention



IAA, EUR, RUR, EEL, EER

Exposure/attention should be proportional to relevance

Equal Opportunity

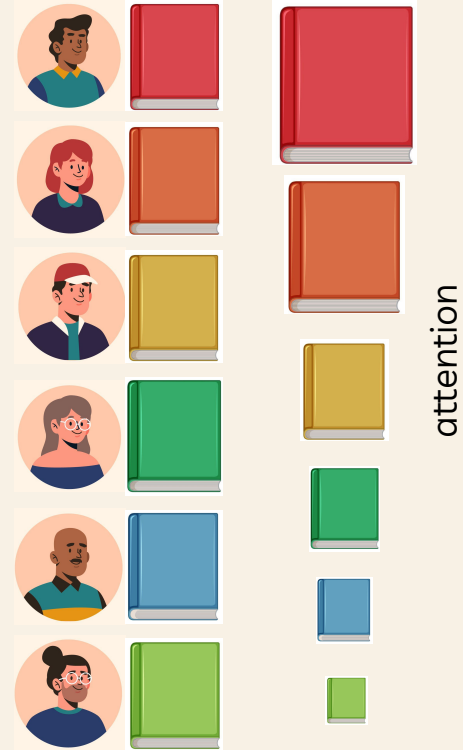
Metrics Design Decomposition

Fairness Goal

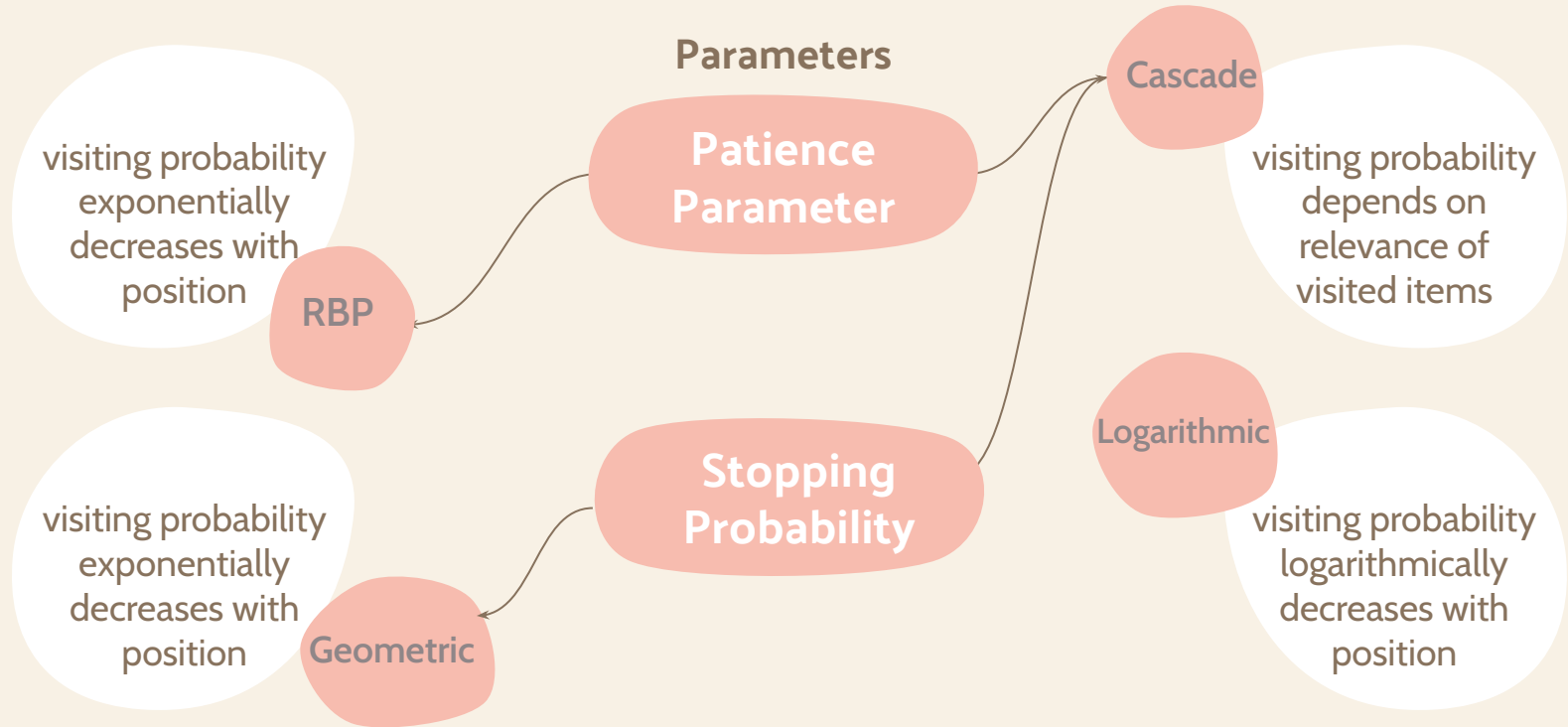
What does it mean to be fair?

Browsing Model

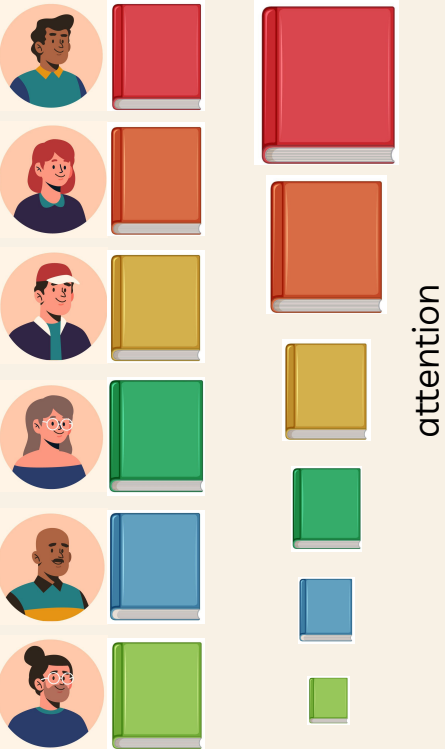
How to measure position weight?



Browsing Models



Metrics Design Decomposition



Fairness Goal

What does it mean to be fair?

Target Exposure

Compare system exposure with what?

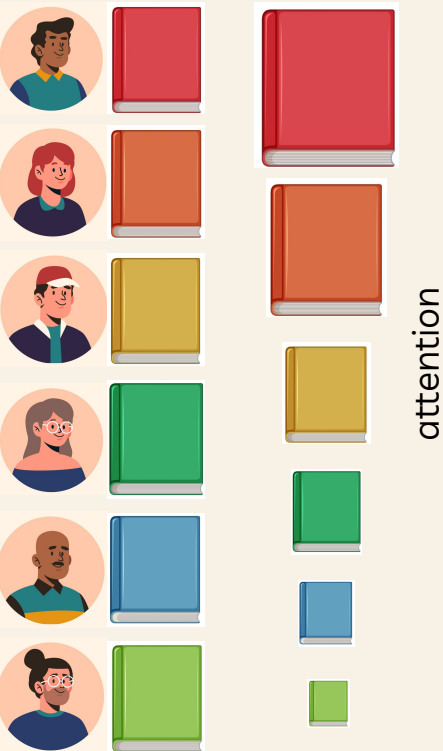
Browsing Model

How to measure position weight?

Target Exposure

- Population estimator
 - From full ranking
 - Configured
- Ideal exposure based on relevance
- Estimated utility (Predicted relevance)

Metrics Design Decomposition



Fairness Goal

What does it mean to be fair?

Target Exposure

Compare system exposure with what?

Group Membership

Does it allow multinomial and soft group association?

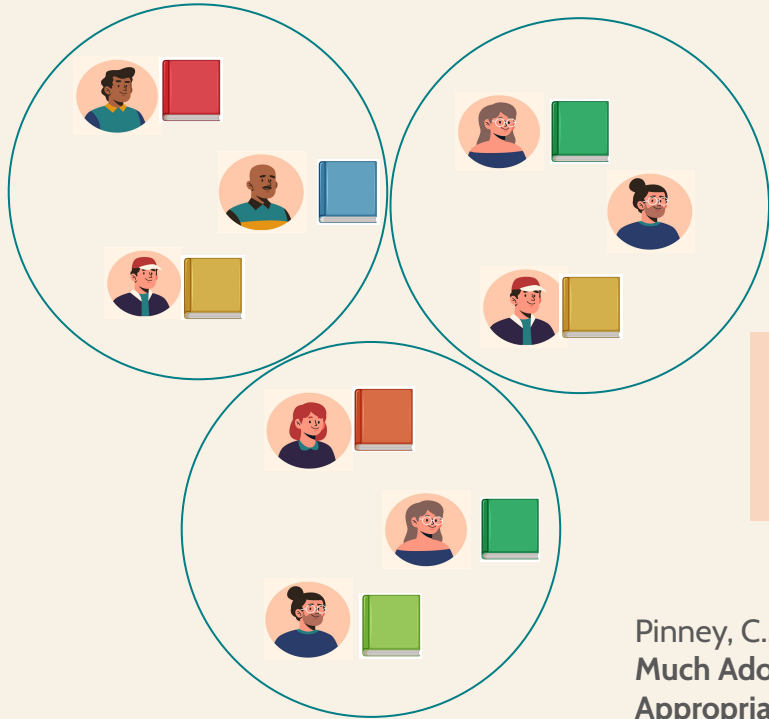
Browsing Model

How to measure position weight?

Relevance

How to incorporate relevance?

Group Membership



Multinomial Protected Attributes

Non-Binary Groups, such as gender

Soft Group Association

Partial or mixed group membership such as race

Pinney, C., Raj, A., Hanna, A., & Ekstrand, M. D. (2023)
Much Ado About Gender: Current Practices and Future Recommendations for Appropriate Gender-Aware Information Access.
To appear in CHIIR 2023 proceedings.
<https://arxiv.org/abs/2301.04780>

Summary of Fair Ranking Metrics

Metric(s)	Goal	Weighting	Relevance	Binomial?
PreFd	Each prefix representative of whole ranking	✗	✗	Dep on d
FAIR	Each prefix matches target distribution	✗	✗	✓
AWRF	Weighted representation matches population	Geometric	✗	✗
DP	Exposure equal across groups	Logarithmic	✗	✓
EUR	Exposure proportional to relevance	Logarithmic	✓	✓
RUR	Discounted gain proportional to relevance	Logarithmic	✓	✓
IAA	Exposure proportional to predicted relevance	Geometric	Predicted	✗
EEL, EER	Exposure matches ideal (from relevance)	Cascade, Geom	✓	✗
EED	Exposure well-distributed	Cascade, Geom	✗	✗

Statistical Parity

AWRF

(Sapienzynski et. al, WWW'19)

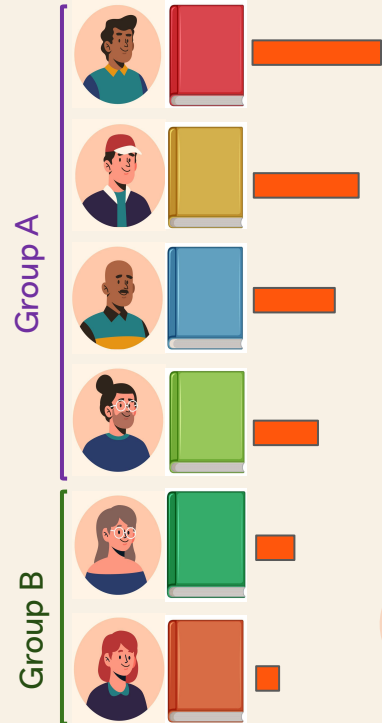
Expected cumulative exposure (Group B x position weight) $\geq p$

Target distribution is the group distribution in entire ranked list (true demographics)

- no relevance information
- geometric attention decay
- non-binary group membership
- uses a target distribution to compare

PreF Δ (Yang et. al, SSDBM'17) and FAIR (Zehlike et. al, CIKM'17) differ in measuring position weight and allowing multinomial groups.

Exposure/Attention



Sequences of Ranking



Statistical Priority

DP, EED

Equal Opportunity

IAA, EUR, RUR, EER, EEL

Equal Opportunity

EE^*

(Diaz et. al, CIKM'20)

EEL(Expected Exposure Loss): $\|target-system\|_2$

EER (Expected Exposure Relevance): Exposure-relevance distribution

- stochastic ranking
- rbp & cascade attention decay
- non-binary group membership

IAA (Biega et. al, SIGIR'18) differs in weighting strategy, group membership, and relevance

EUR, RUR (Singh et. al, SIGKDD'18) differs in weighting strategy and group membership

Relevance
Exposure/Attention

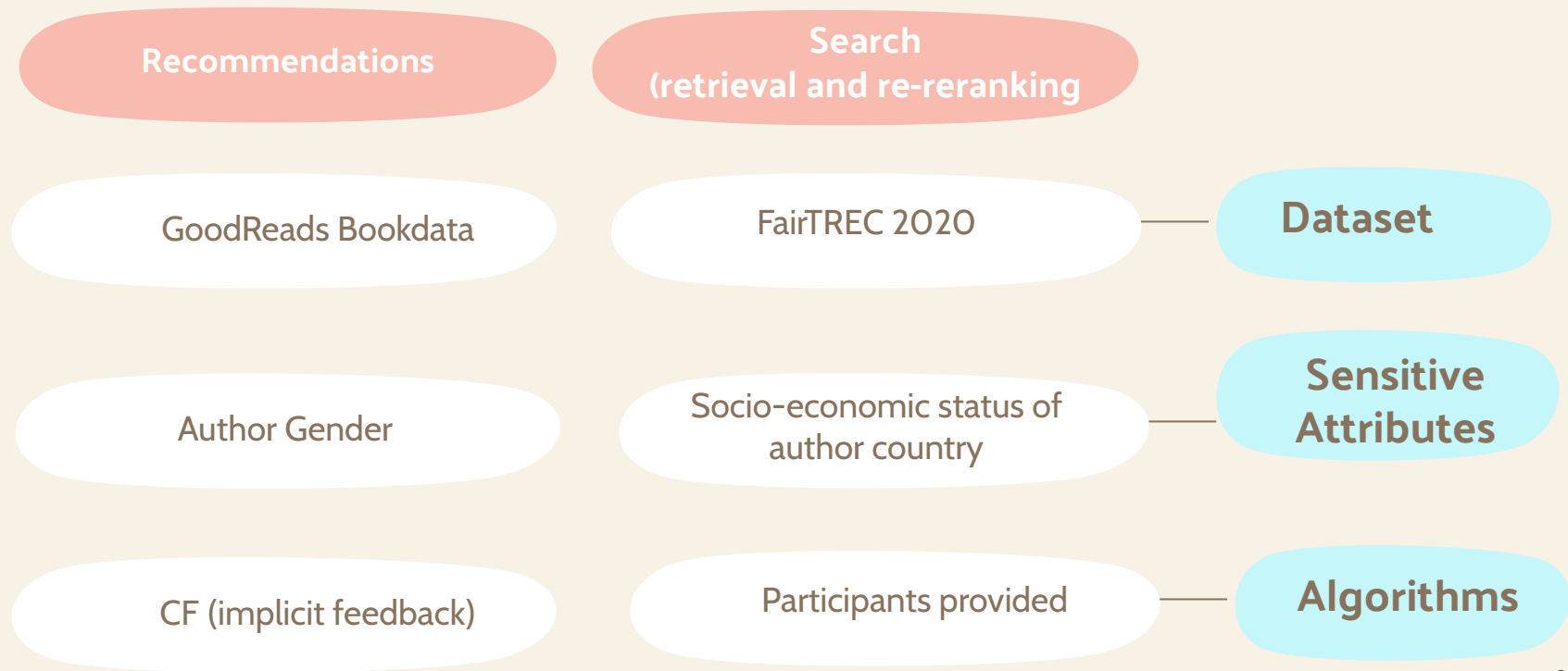


Task 1 Findings

Task 1: Conceptual Analysis Fair Ranking Metrics

- Metrics are conceptually similar with common components like relevance, browsing model, aggregation, target exposure
- Metrics differs in their design choices and fairness assumption
- Metrics with same goal can have different design choices

Implementing the Metrics



Challenges in Implementation



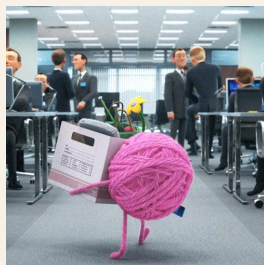
Missing Relevance Information

IAA, EE*, DP, EUR, RUR



Missing Group Label

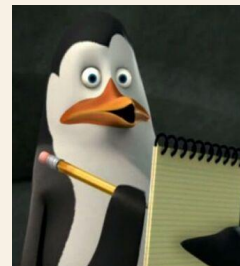
All the metrics



Extreme Imbalance

- **Pre Δ** and **RUR**: suffer from missing data (sparsity) problem
- Reformulated ratio-based metric to smoothed log ratio

AWRF, IAA, DP, EUR, RUR, EE*



Parameter Setting

Pre Δ , FAIR, IAA, DP, EUR, RUR



Soft Group Association
Non-binary groups

Task 2 Findings

Task 2: Implementing Fair Ranking Metrics in Real-World Datasets

- Missing data, missing relevance information, ranked list size are crucial/delicate factors in implementing metrics.
- Metrics with similar fairness goals differ in their ease of implementations

Sensitivity Analysis

Ranked-list size

- No effect on metrics for FairTREC
- Ratio-based metrics and FAIR showed sensitivity

Weighting Strategy

- Default parameters
- EEL and logRUR showed high sensitivity

Parameter Settings

- Almost all metrics showed sensitivity
- logRUR is extremely sensitive

Task 3 Findings

Task 3: Sensitivity Analysis

- Metrics differ in their sensitivity towards external factors.
- High sensitivity towards design choices add complexity in the usability of metrics

Key Findings

Defining metrics in unified framework

- Metrics are surprisingly similar

Implement the metrics in same experimental setup

- Missing data, missing relevance information, ranked list size are crucial/delicate factors in implementing metrics.

Sensitivity Analysis

- Metrics differ in their sensitivity towards external factors.

Recommendations

**Allow multinomial
protected
attributes**

**Allow soft group
association**

**Sensitivity towards
design choices**

Single-list metrics

FAIR, AWRF

AWRF

AWRF

AWRF

**Demographic Parity in
Sequence**

DP, EED

EED

EED

EED

**Equal Opportunity in
Sequence**

EUR, RUR, IAA, EER, EEL

EER, EEL

EER, EEL

EER, EEL, IAA

Research Directions

- Simulation study to understand the impact of crucial factors in metric implementation.
- Incorporating various browsing models
- Missing label
- Missing or sparse relevance
- Ambiguous or multiple group association
- Robust, explainable, and efficient metric design



THANK You!

Credits: This presentation template was created by **Slidesgo**, including icons by **Flaticon**, and infographics & images by **Freepik**