

BiasMeter: On Measuring Bias in Online Information

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MOTIVATION



... but non-transparent algorithms return information that is 'believed' to be useful

... guiding our decisions in various domains



... the convenience and effectiveness of such services has limited our information seeking abilities and made us overly dependent on them



We live in echo chambers and filter bubbles due to personalization



TYPES OF BIAS

We consider **bias** in terms of the results of a service regarding a *topic (e.g. politics)*





METRICS

DEFINITION 1 (INDIVIDUAL USER BIAS)

An online information provider (OIP) is individual user unbiased if for any pair of users u_1 , u_2 it holds:

 $D_R(R_{u_1}, R_{u_2}) \le D_u(u_1, u_2)$

- User independent
- Over a number of differentiating attributes
 political party, against /
 in favor, artists / scientists

A Web Search engine that returns results in favor of a specific political party

• Different users

 Over a number of protected attributes
race, sex, nationality, religion

A job recommendation engine that returns lowered paid results for women

BIASMETER

Protected Attributes P		Differentiating Aspects A	
User Population U		Topic T	

where R_{u_1} and R_{u_2} are the result lists received by u_1 , u_2 resp. and D_R and D_u appropriate rankings and users distance functions

DEFINITION 2 (GROUP USER BIAS)

An OIP is group user unbiased if it holds:

 $D_R(R_P, R_Q) \leq \varepsilon$

where R_P is the union of the result lists seen by the members of the protected attribute and R_Q is the union of the result lists seen by the members of the non protected attribute for some small $\varepsilon \ge 0$

DEFINITION 3 (CONTENT BIAS)

An OIP is group user unbiased if it holds:

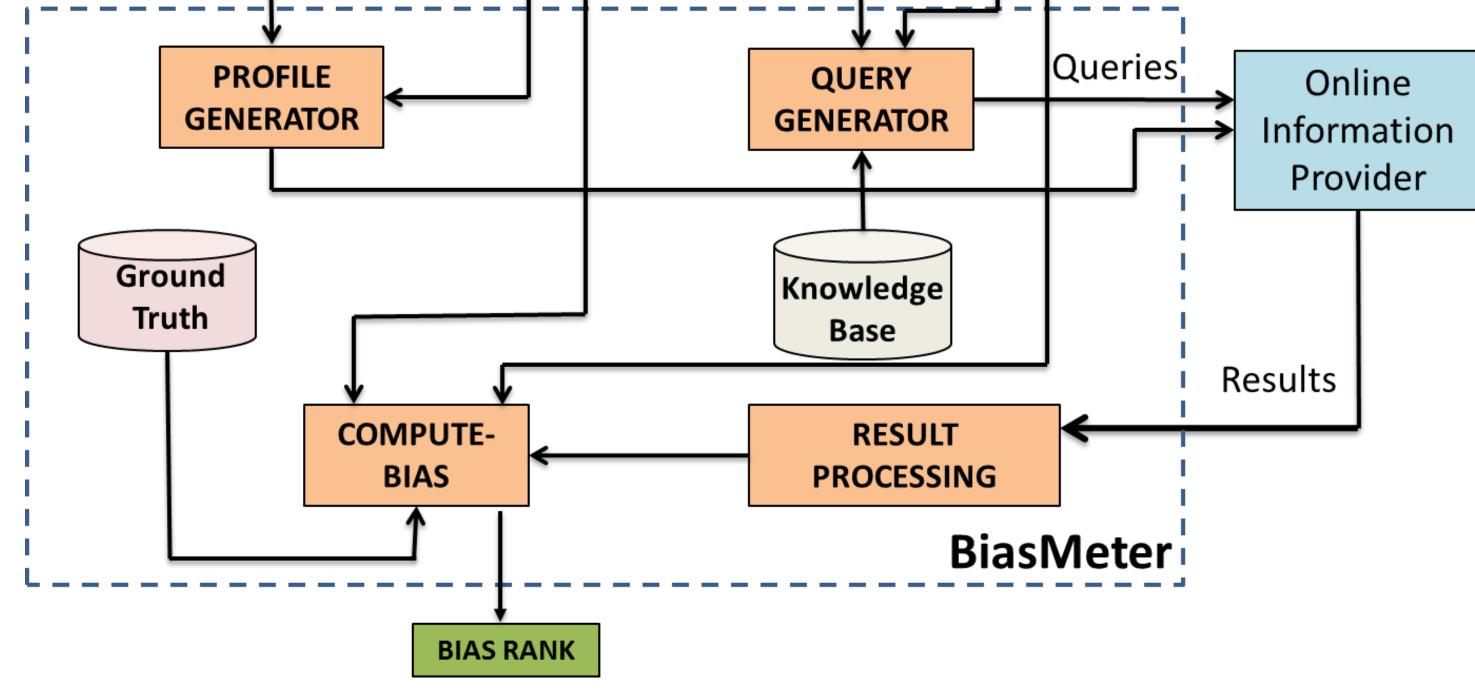
 $D_R(R_u, R_T) \leq \varepsilon$

where R_T is the "ideal unbiased ranking" for some small $\varepsilon \geq 0$

Research Challenges

1. Ground Truth

- human-in-the-loop (crowd-sourcing)
- comparative evaluation of various services



2. Multi-faceted and concrete Bias metrics

- take into account the correlation of attributes
- 3. Engineering and Technical Challenges
 - generation of huge samples of user profiles
 - data mining / data integration / machine learning
 - knowledge representation / entity detection

4. Auditing Algorithms

- access to the internals of services
- co-operation with law and policy makers



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