

Uncertainty-based Bias Mitigation in Ranking

Task: Increasing fairness in a ranking system with minimal loss in user utility.

Proposed solution: Trade-off fairness for utility where the ranker is most uncertain.

Predictive Uncertainty-based Fair Ranking (PUFR):

- Aim: Increase the exposure of documents of the protected group (here in green)
- Increase the score of protected documents by multiple of standard deviation of predicted score.
- Decrease the score of non-protected documents by multiple of standard deviation.



Maria Heuss, Daniel Cohen, Masoud Mansoury, Maarten de Rijke, and Carsten Eickhoff. Predictive Uncertainty-based Bias Mitigation in Ranking. In CIKM 2023: 32nd ACM International Conference on Information and Knowledge Management. ACM, October 2023

Fairness and Explainability in Learning To Rank

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Results and future research questions:

- Uncertainty improves the ability to effectively *trade-off* fairness and utility
- *RQ*: How to include uncertainty in existing approaches
- RQ: What role does calibration of uncertainty play?
- *RQ*: How can we define *calibration* for *rankers*?

What it means: Uncertainty can guide decisions for trading off conflicting objectives.

- Trade off between compactness (size of explanation) k) and validity/ completeness
- - Trade-off is an attribute of explanations, not just of explanation method
- RQ: Are there trade-offs between other properties of explanations as well?
 - RQ: Can we identify user biases, encoded in the click data through explainability?



Property Trade-offs in Explanations for Rankings

Results and future research questions:

Trade-off between validity and completeness

What it means: Explanations cannot satisfy all desirable properties equally. An *application-based decision* needs to be made.



Task: Creating list-wise explanations for ranking models, focusing on different properties.

Main take-away: There exists a trade-off between two seemingly complementary properties of explanations. Greedy (local) list-wise feature explanations:

Explanation: Small subsets $f \subset F$ of the features Fthat model bases decision on.

Validity (val): How well can model re-construct the ranking, while only seeing f.

Completeness (comp): How well can model re-construct ranking, without seeing f.

Greedily add features that maximize $\lambda \cdot val + (1 - \lambda) \cdot comp$ to the explanation.



Trade-off for Explanation size = 5