

Equity of Reproduction:

A reproduction of Amortizing Individual Fairness in Rankings

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The Paper

Asia J Biega, Krishna P Gummadi, and Gerhard Weikum.

Equity of attention: Amortizing individual fairness in rankings.

In The 41st International ACM SIGIR Conference on Research & Development in Information Retrieval, pages 405–414.ACM, 2018.

Problem

Disparity between ranking
relevance and attention

- Define an unfairness metric
- Define a comparative quality metric for a new ranking
- Define an ILP to minimise the unfairness given a constraint on the ranking quality



The Mathematical Model

$$\text{unfairness}(\rho^1, \dots, \rho^m) = \sum_{i=1}^n |A_i - R_i| = \sum_{i=1}^n \left| \sum_{j=1}^m a_i^j - \sum_{j=1}^m r_i^j \right|$$

$$\text{minimize } \sum_{i=1}^n \sum_{j=1}^n |A_i^{l-1} + w_j - (R_i^{l-1} + r_i^l)| \cdot X_{i,j}$$

$$\text{subject to } \sum_{j=1}^k \sum_{i=1}^n \frac{2^{r_i^l} - 1}{\log_2(j+1)} X_{i,j} \geq \theta \cdot IDC@k$$

$$X_{i,j} \in \{0, 1\}, \forall_{i,j}$$

$$\sum_i X_{i,j} = 1, \forall_j$$

$$\sum_j X_{i,j} = 1, \forall_i$$

Our reproduction

We can affirm the unavoidable use of technical devices, and also deny them the right to dominate us, and so to warp, confuse, and lay waste our nature.

- Martin Heidegger

Data collection

Attempt to reproduce the
experiment with the same data

- Synthetic dataset
 - AirBnB
 - StackExchange
-

Approach

Python library

- Model the ILP with PuLP
- Utilise the Gurobi solver through PuLP
- Simulate thousands of rankings and see how the unfairness progresses



Experiments

Synthetic dataset

100 subjects

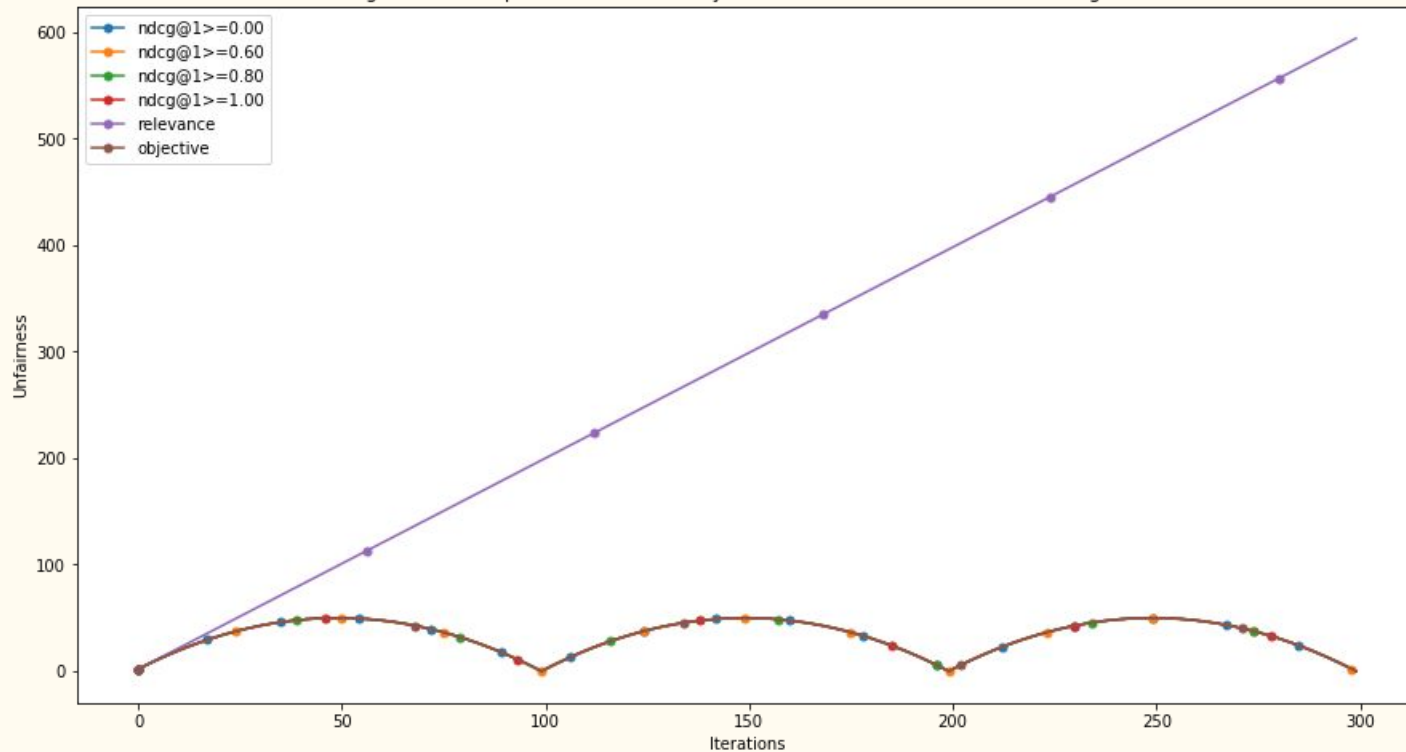
Singular attention

Geometric attention

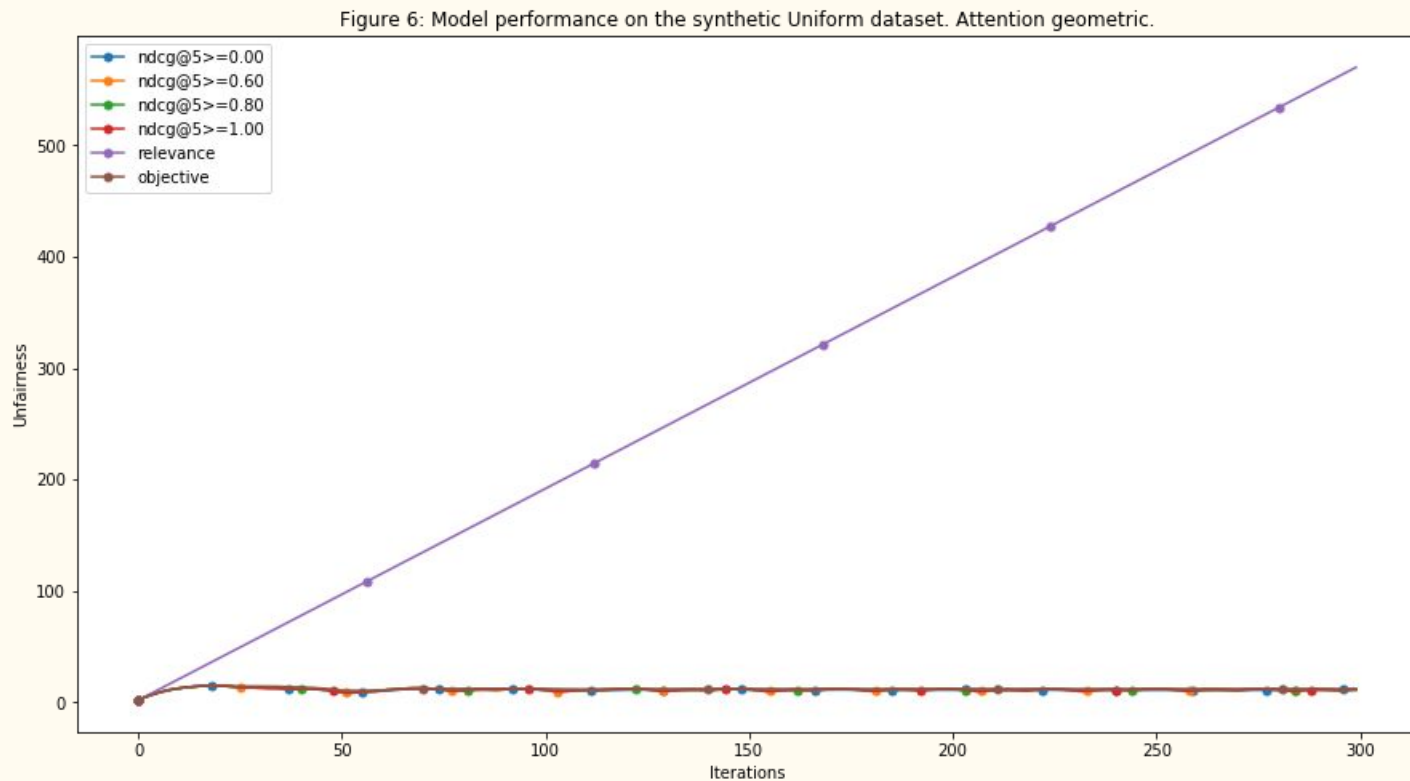
- Uniform
- Linear
- Exponential

Uniform dataset, singular attention

Figure 3: Model performance on the synthetic Uniform dataset. Attention singular.



Uniform dataset, geometric attention



AirBnB dataset

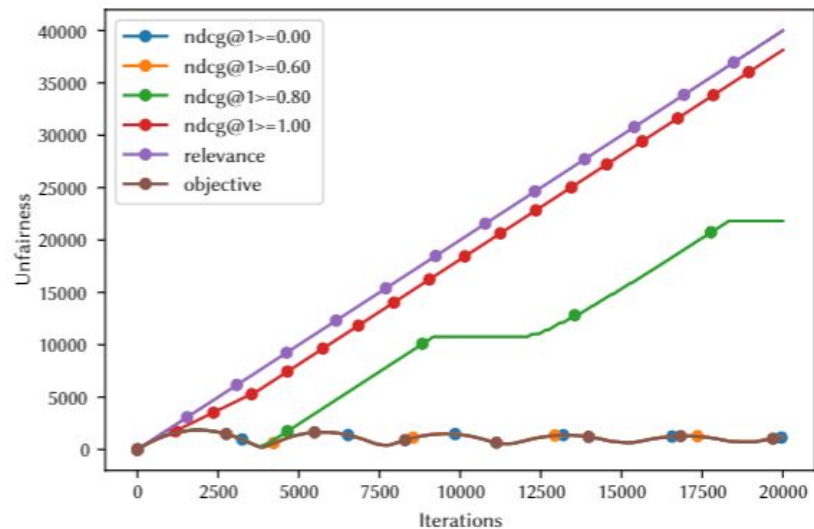
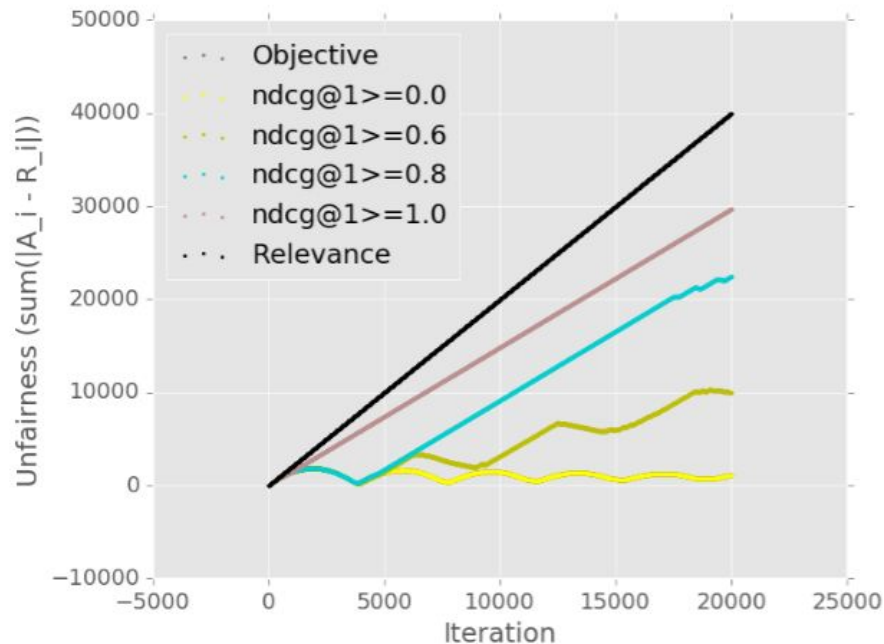
Singular attention

Single-query

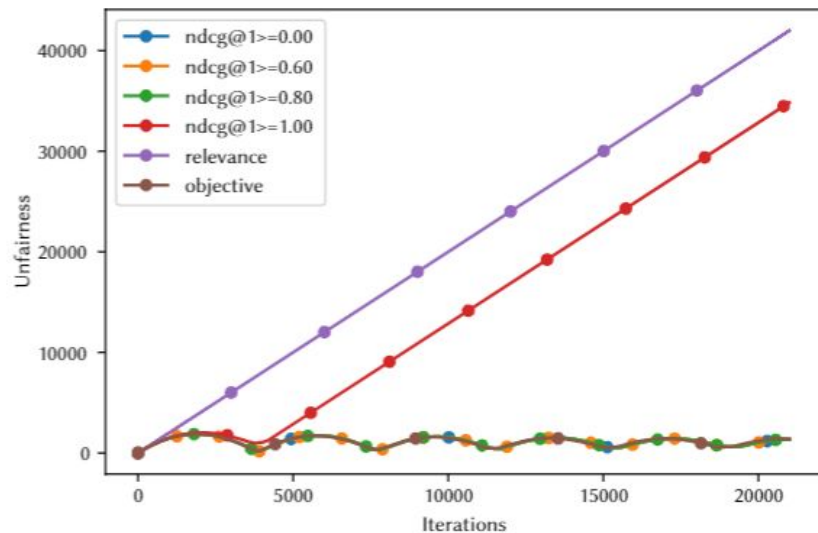
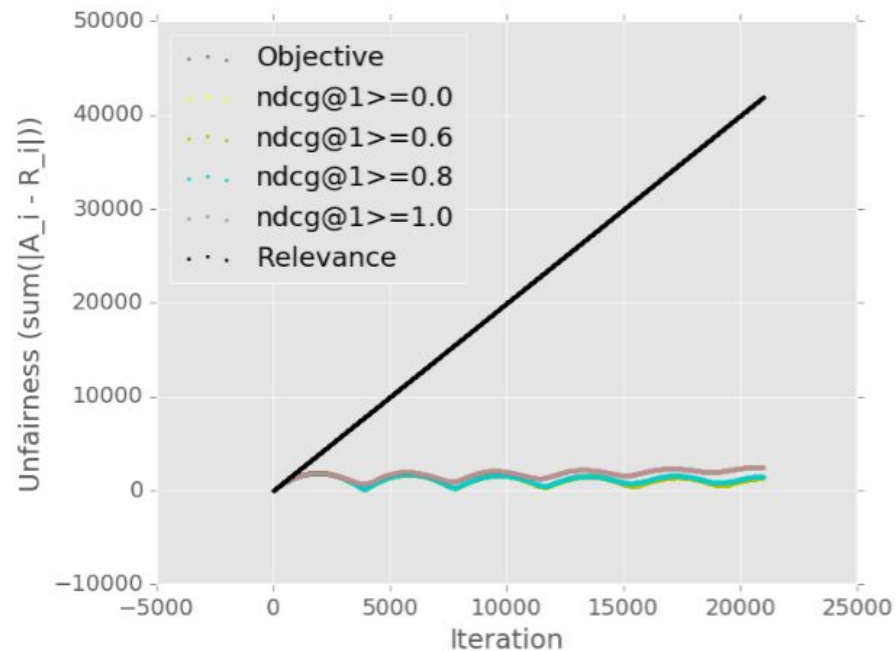
Multi-query

- Boston (3924 subjects)
- Geneva (2209 subjects)
- Hong Kong (4513 subjects)

Boston dataset, single-query



Boston dataset, multi-query

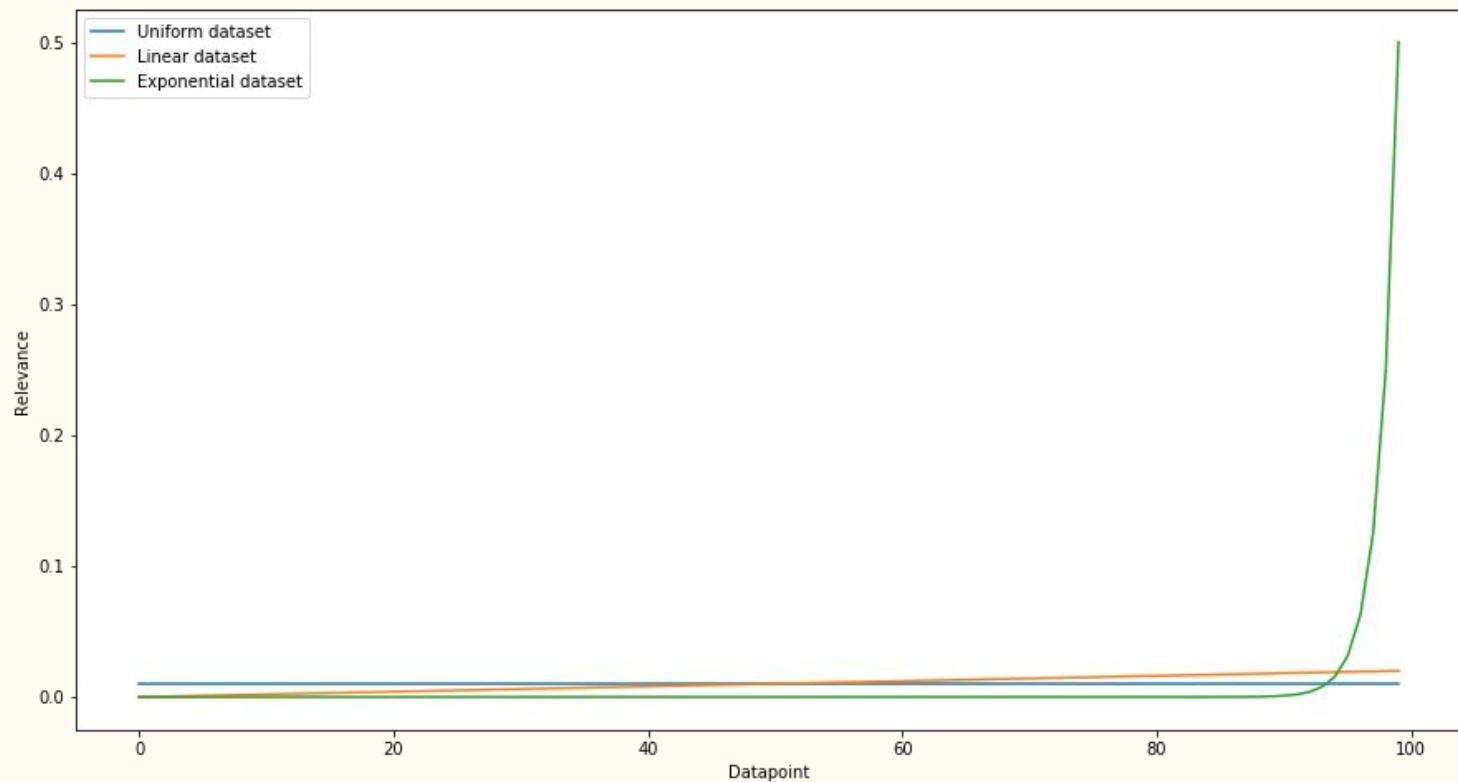


Conclusion

- Synthetic dataset - completely reproduced
- AirBnB dataset - trends reproduced
- Algorithm - successfully implemented
- Actual unfairness reduction - under debate

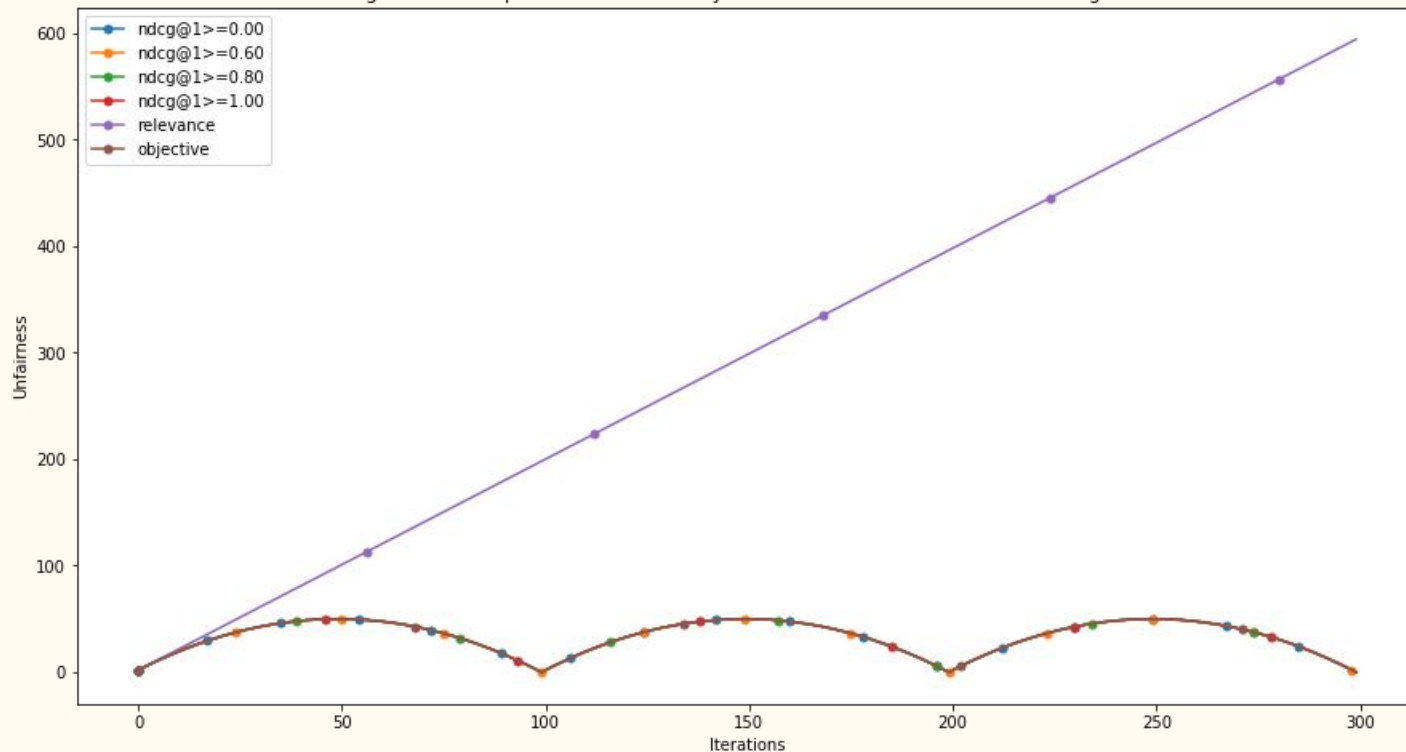


Synthetic dataset

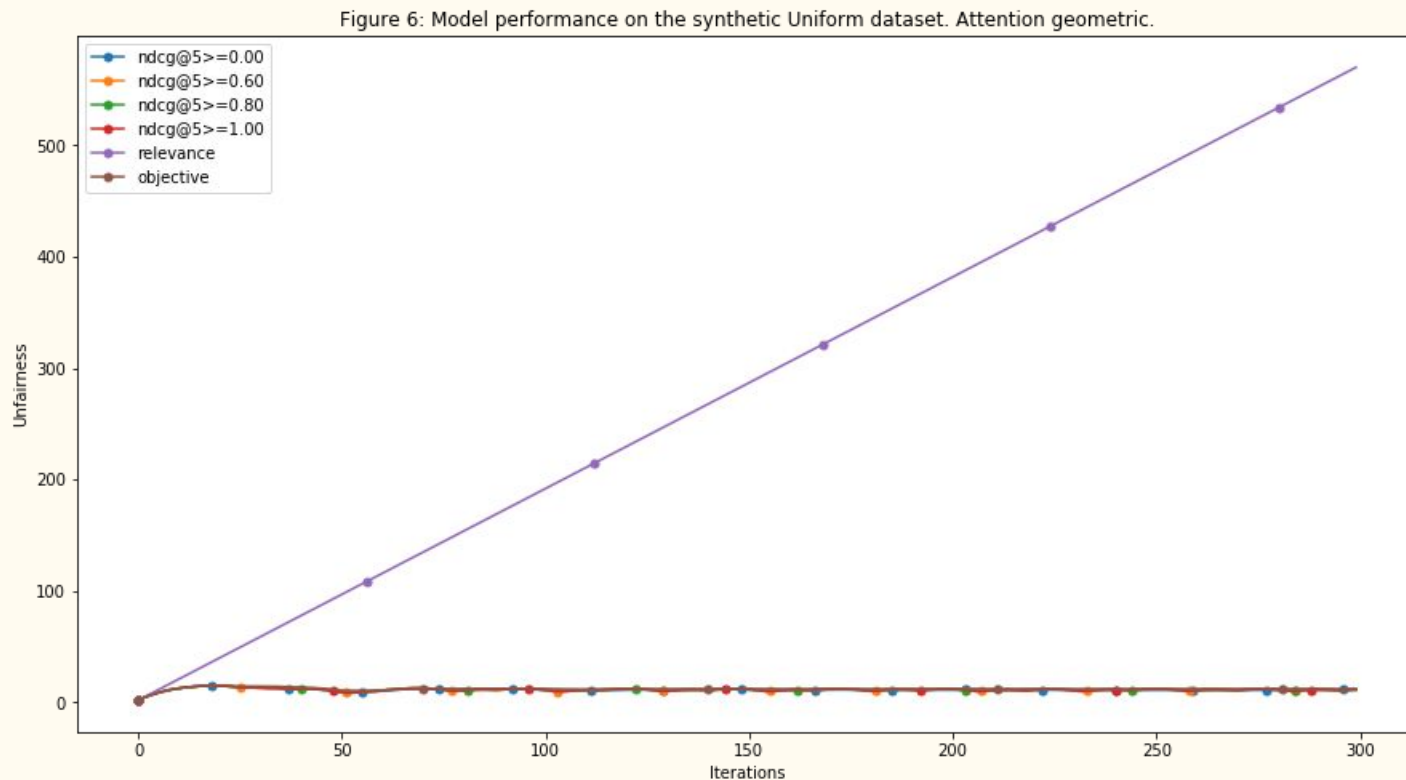


Uniform dataset, singular attention

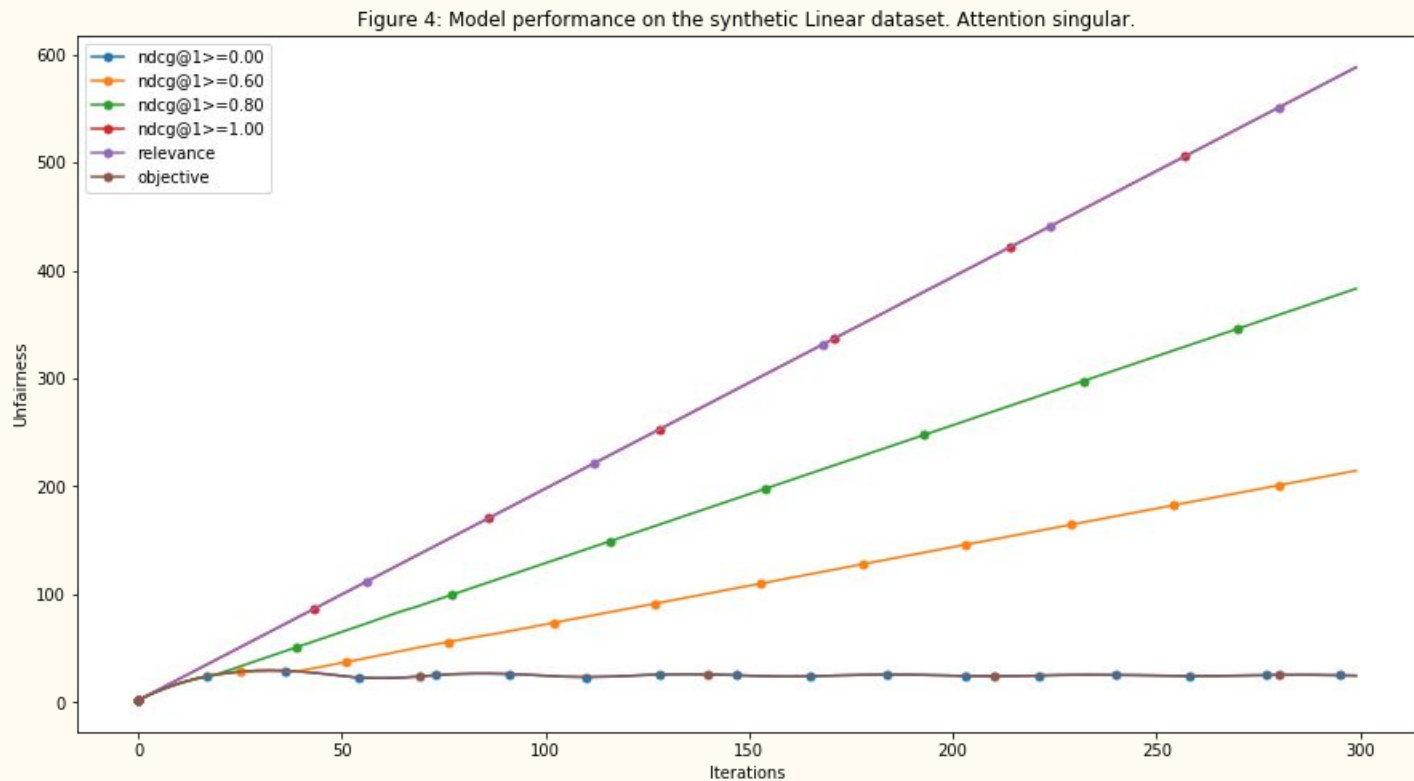
Figure 3: Model performance on the synthetic Uniform dataset. Attention singular.



Uniform dataset, geometric attention

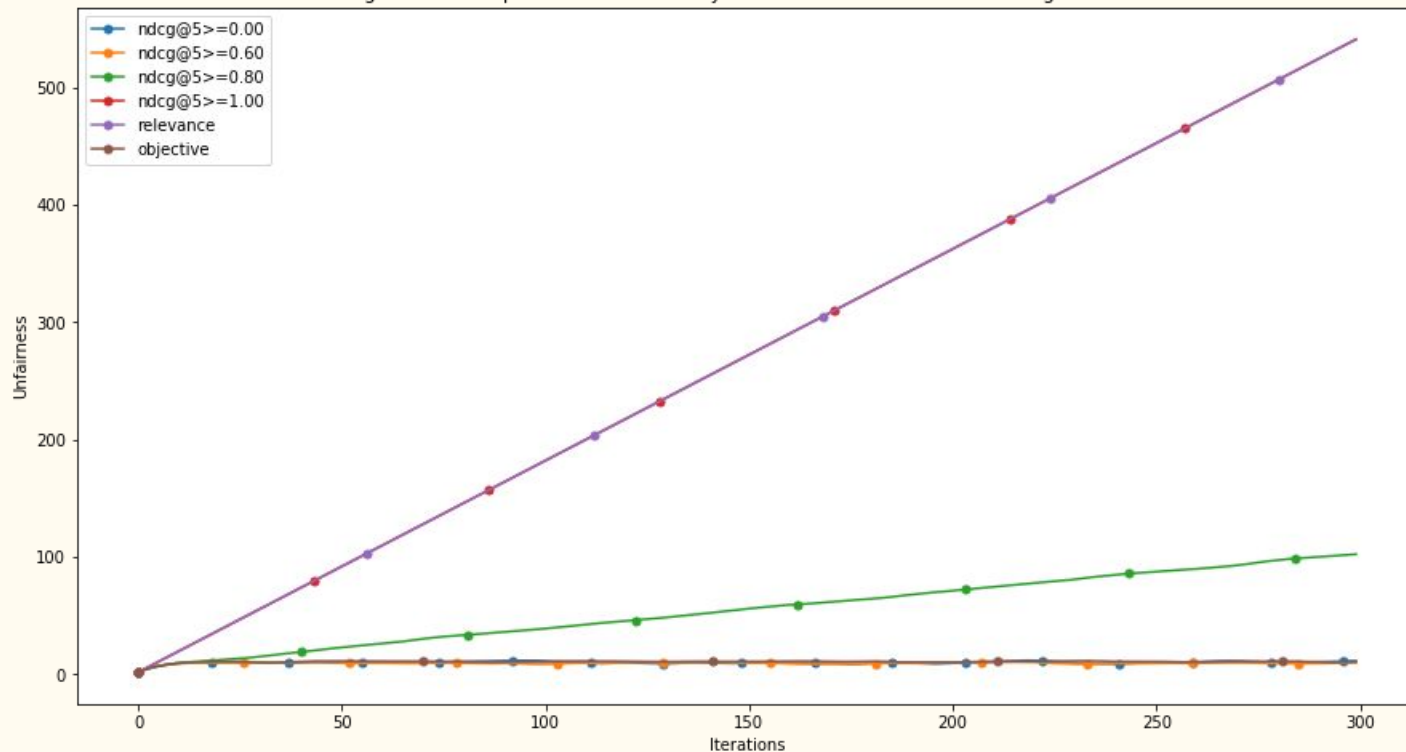


Linear dataset, singular attention

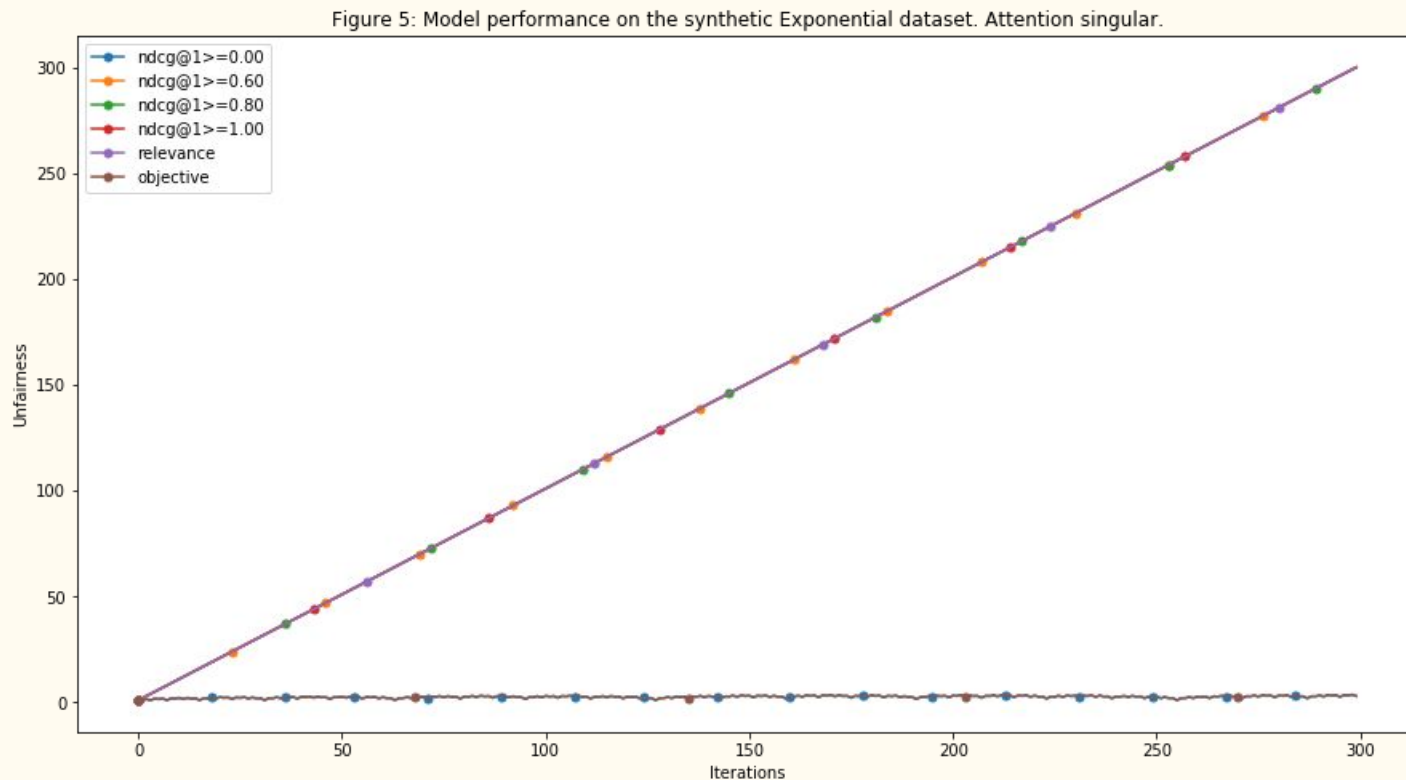


Linear dataset, geometric attention

Figure 7: Model performance on the synthetic Linear dataset. Attention geometric.

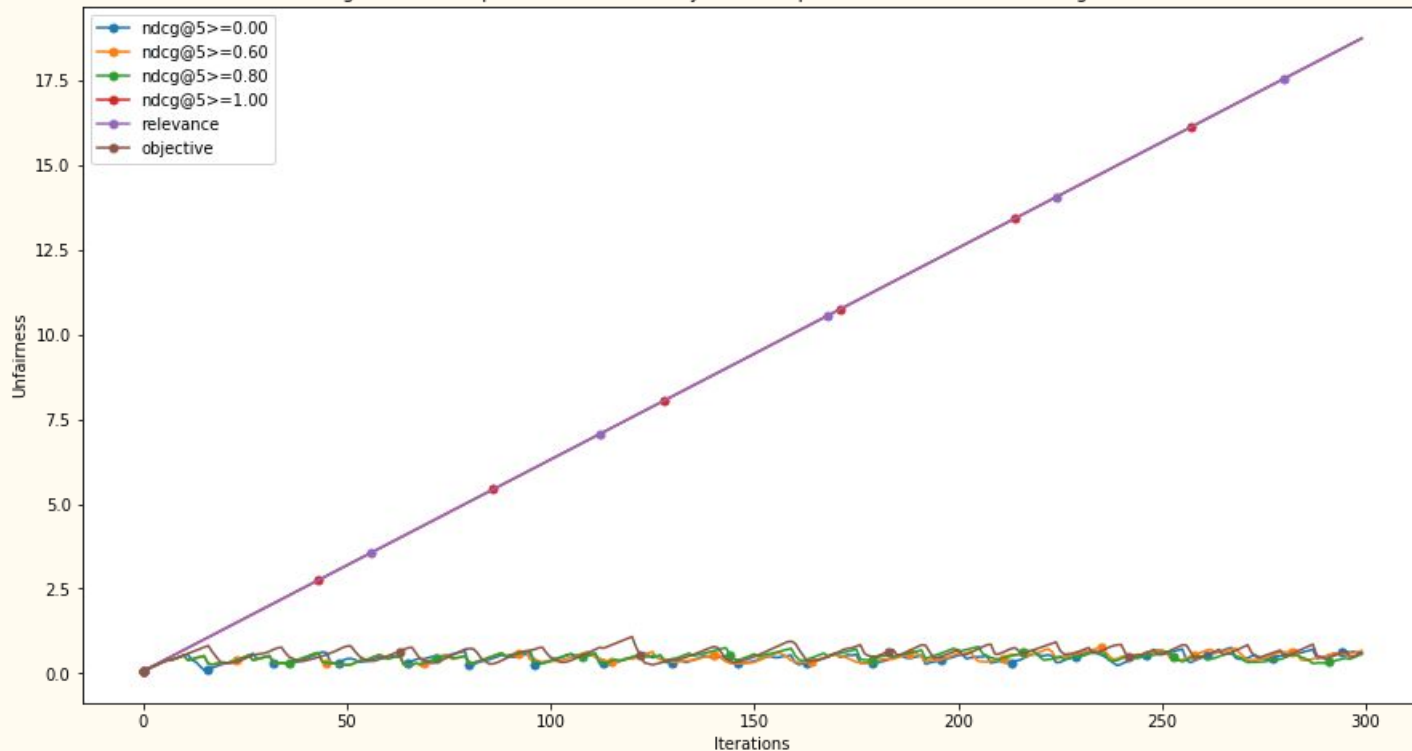


Exponential dataset, singular attention

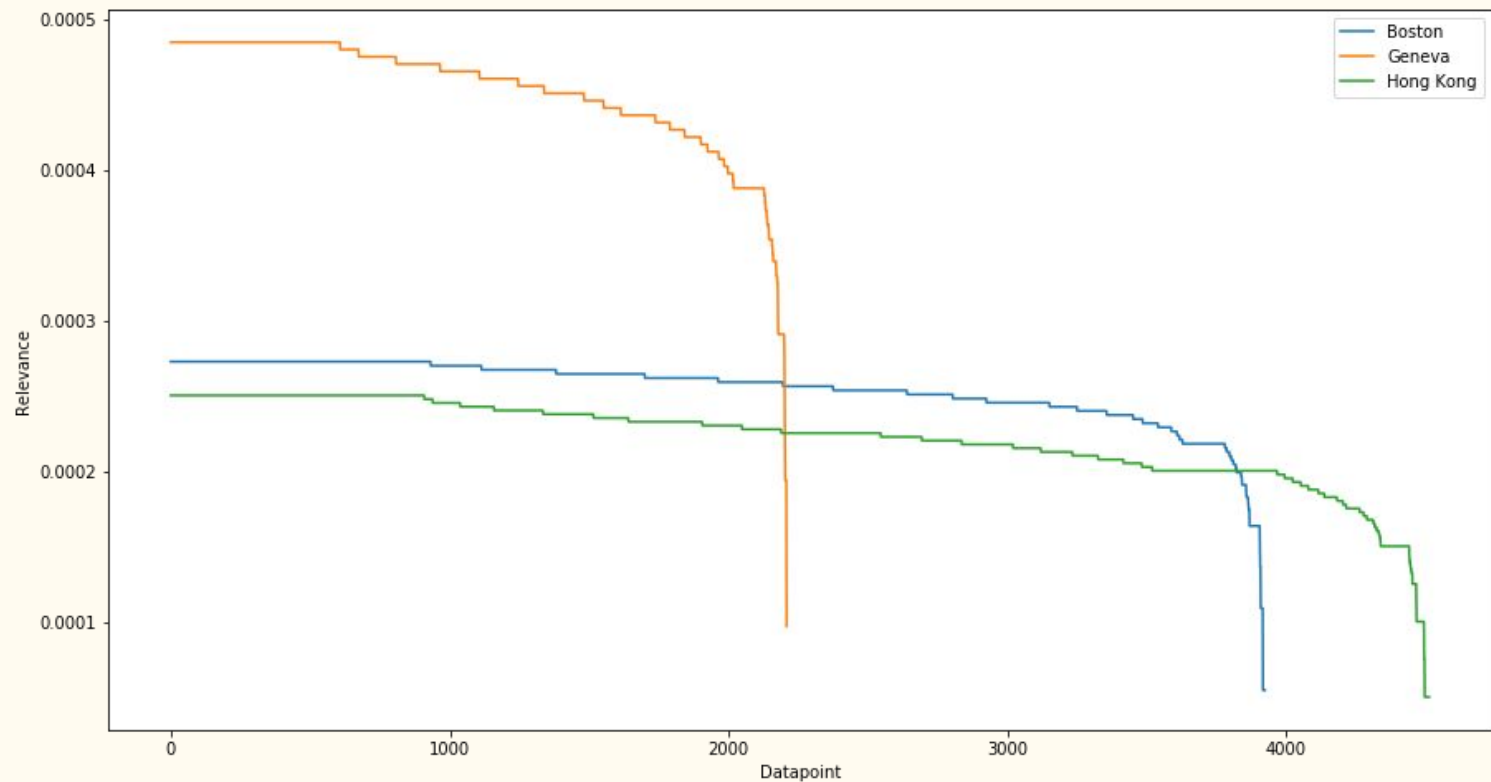


Exponential dataset, geometric attention

Figure 8: Model performance on the synthetic Exponential dataset. Attention geometric.

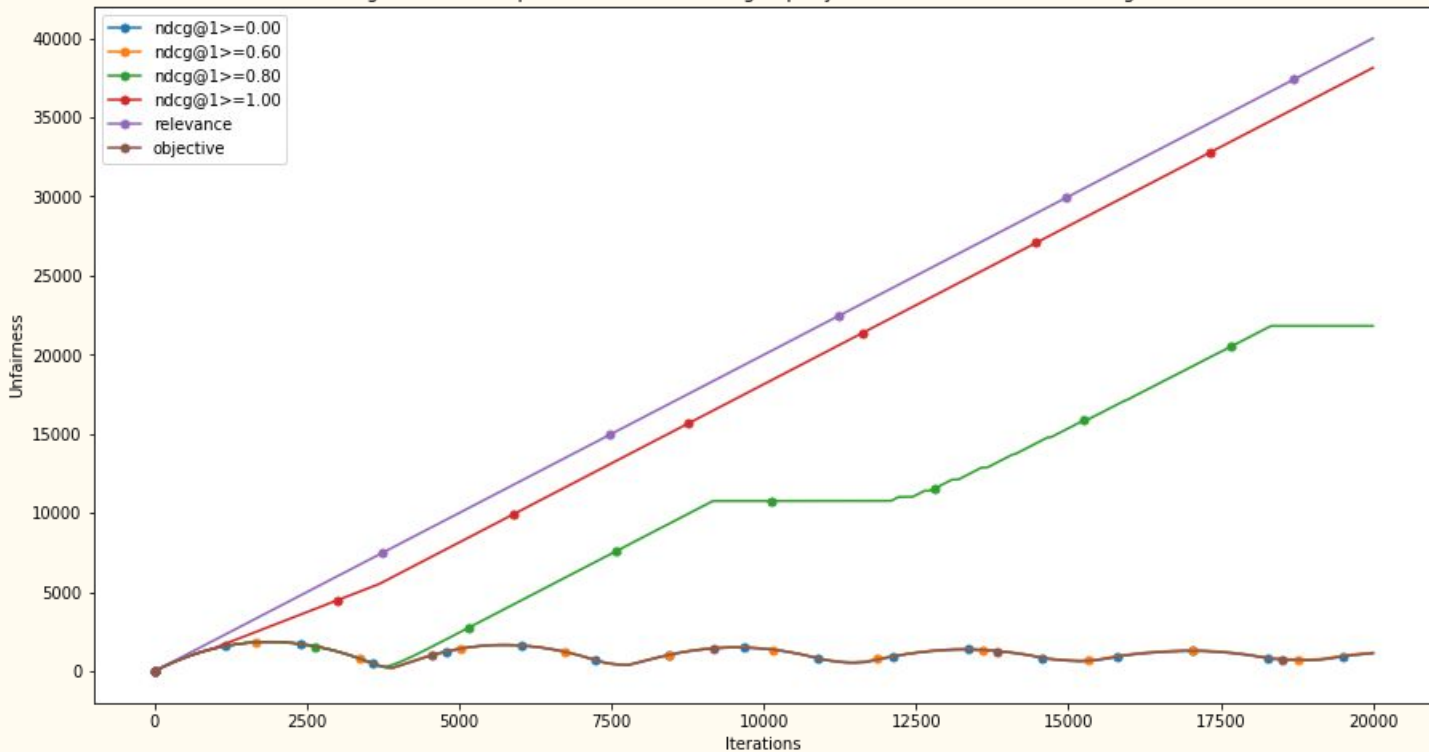


AirBnB Relevances



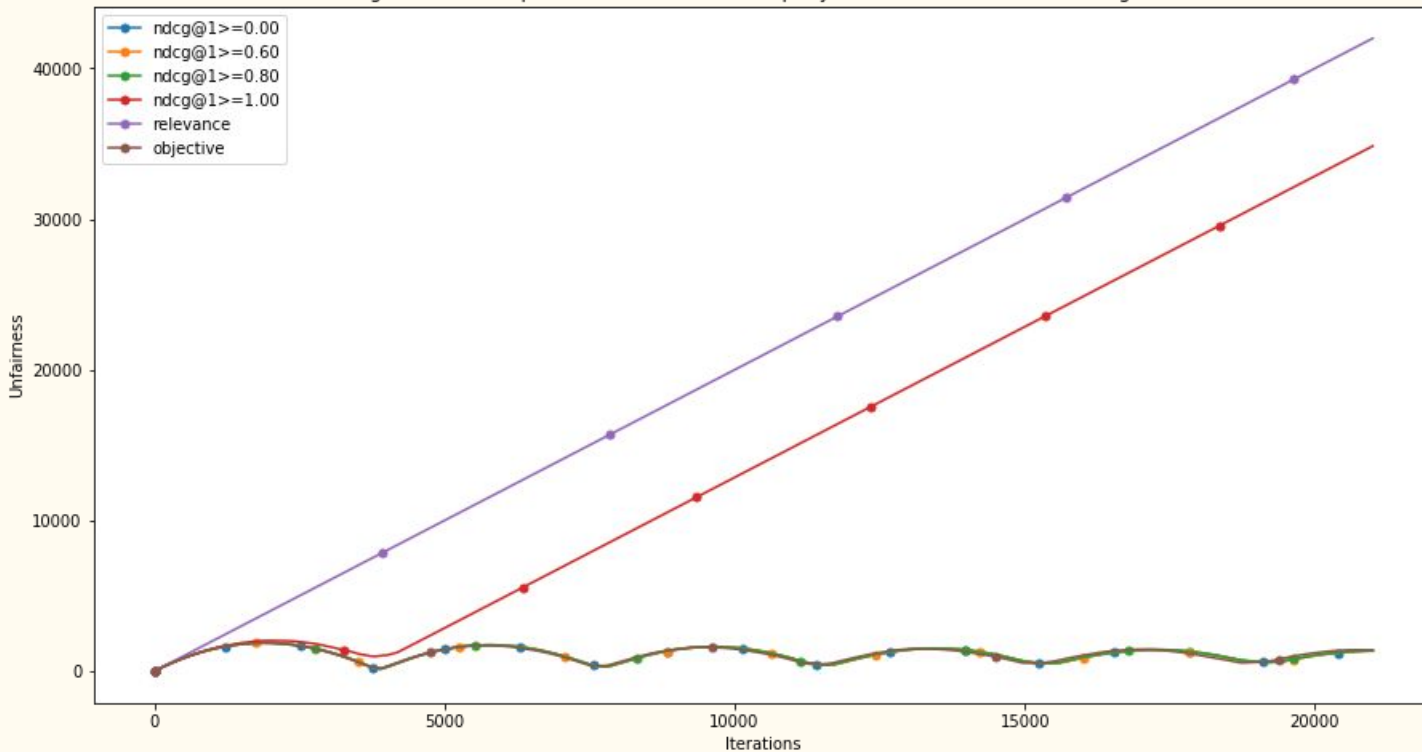
Boston dataset, single-query

Figure 10: Model performance on the single-query Boston dataset. Attention singular.

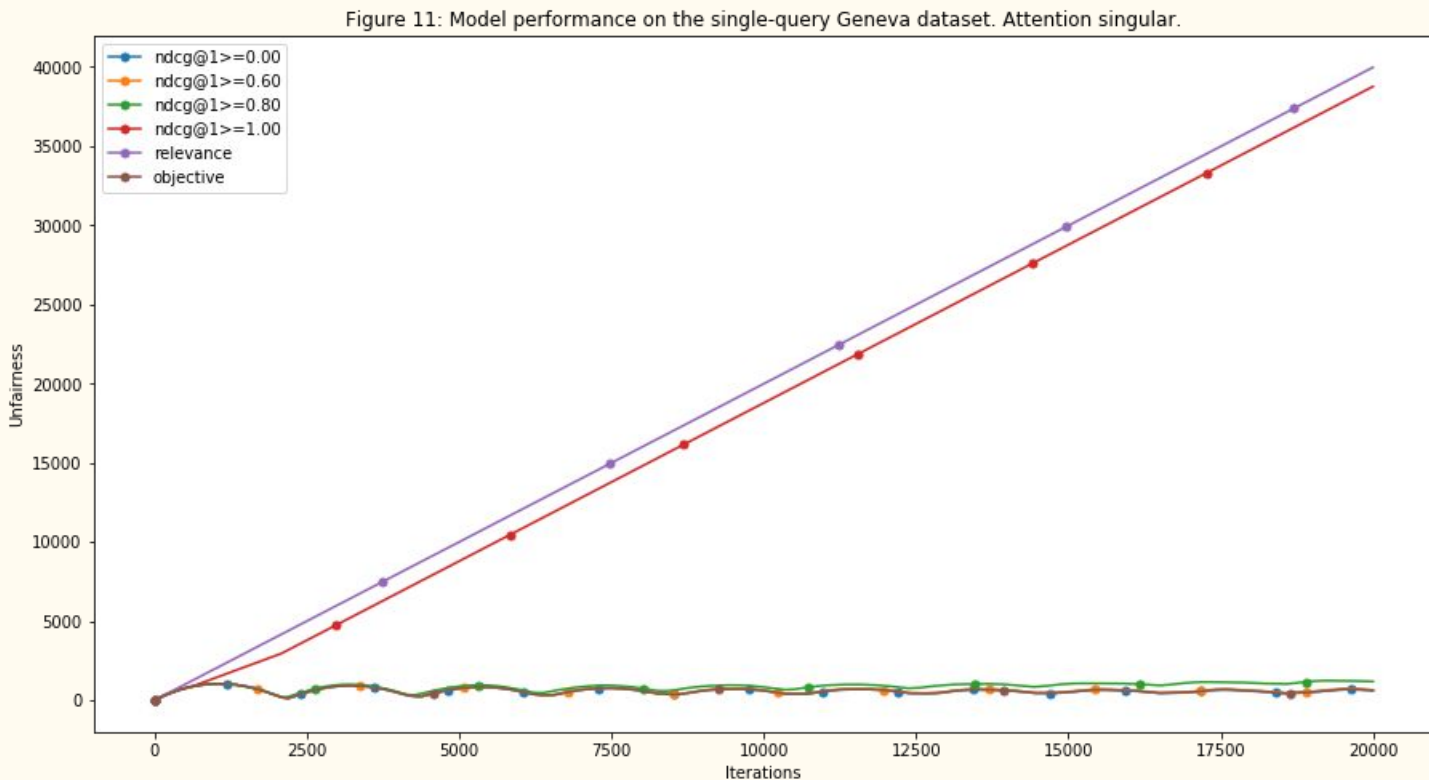


Boston dataset, multi-query

Figure 13: Model performance on the multi-query Boston dataset. Attention singular.

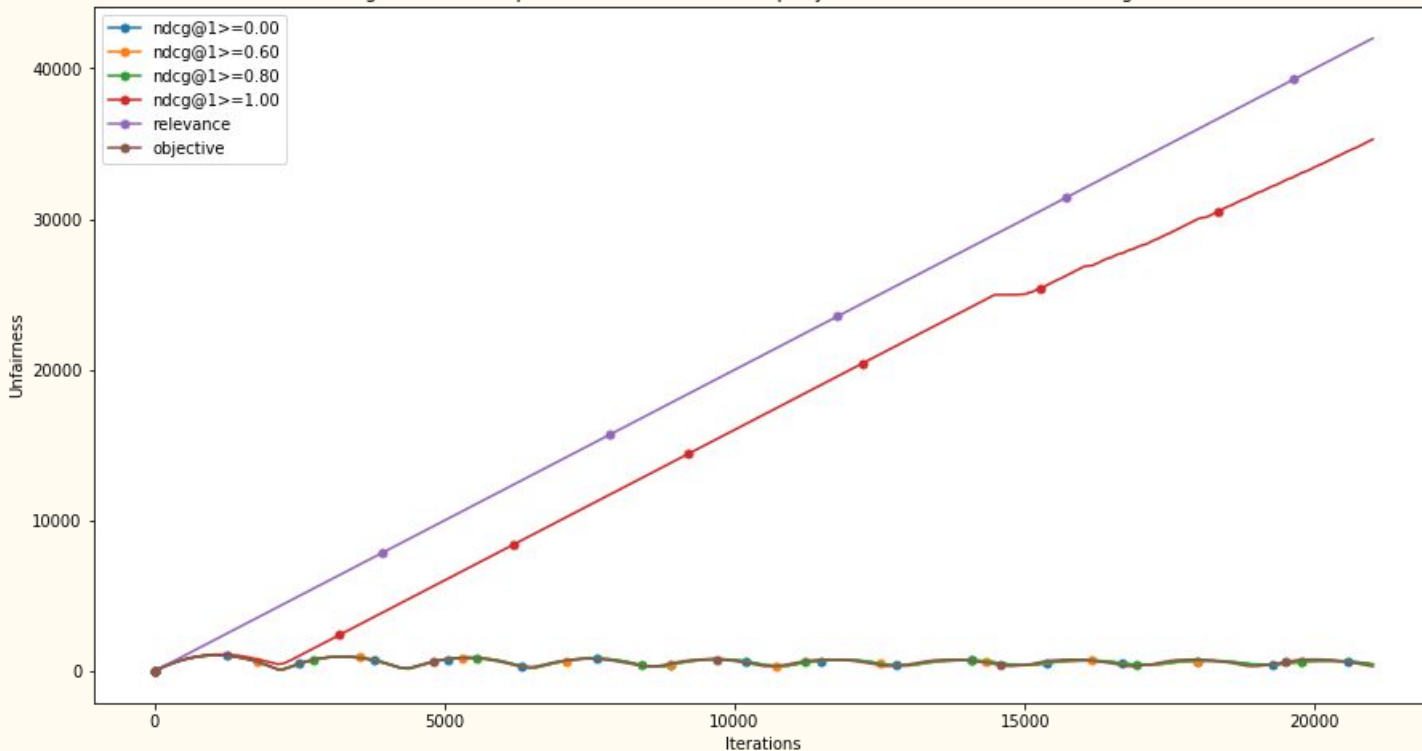


Geneva dataset, single-query



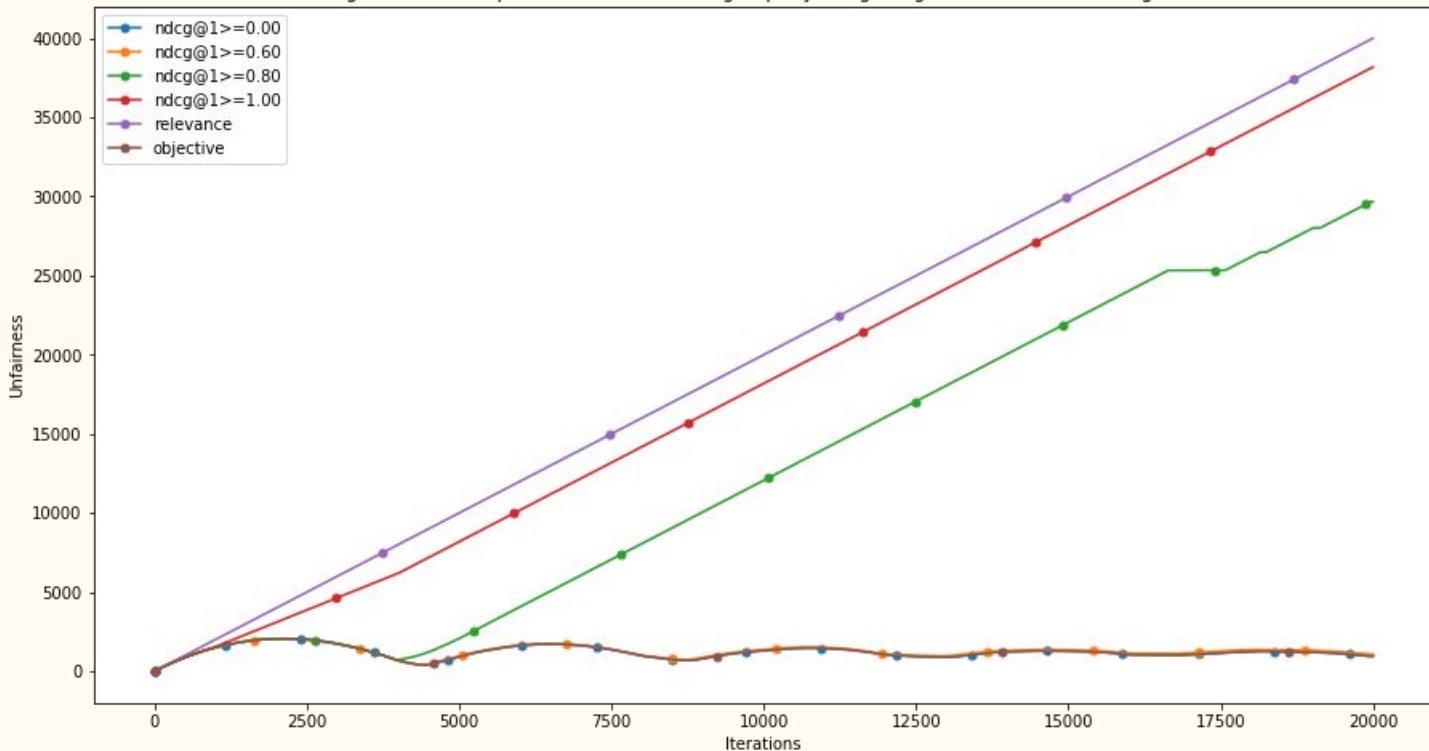
Geneva dataset, multi-query

Figure 14: Model performance on the multi-query Geneva dataset. Attention singular.



Hong Kong dataset, single-query

Figure 12: Model performance on the single-query Hong Kong dataset. Attention singular.



Hong Kong dataset, multi-query

Figure 15: Model performance on the multi-query Hong Kong dataset. Attention singular.

