

Toward Evaluating the Reproducibility of Information Retrieval Systems with Simulated Users

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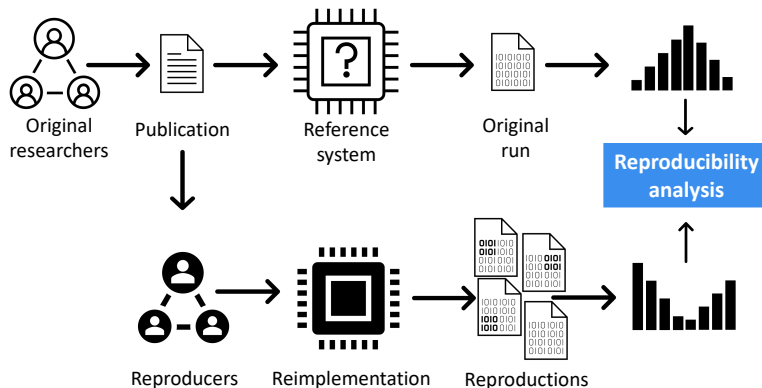
June 20, 2024

⚡ **Issue:** Gap between system- and user-oriented evaluations of retrieval systems.

❓ **Question:** When does a user consider a retrieval system as reproduced?

🔗 **Solution:** Bridge the gap with user simulations for large-scale evaluations!

Reactive reproducibility analysis



repro_eval: A Python Interface to Reproducibility Measures of System-Oriented IR Experiments, Breuer et al., ECIR'21

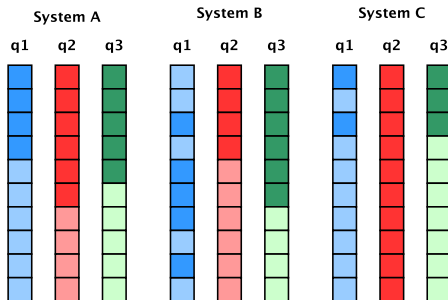
System-oriented information retrieval experiments

Experimental setup of the Cranfield paradigm [1]:

 Document collection

 Topics / queries

 Relevance labels



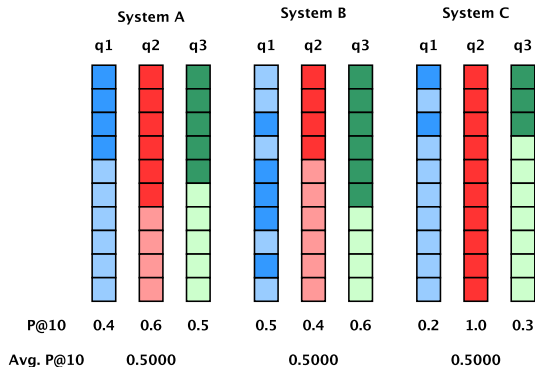
How to Measure the Reproducibility of System-oriented IR Experiments, Breuer et al., SIGIR'20

An in-depth investigation on the behavior of measures to quantify reproducibility, Maistro et al., IPM'23

Retrieval effectiveness

Precision

$$P = \frac{\text{Number of relevant documents retrieved}}{\text{Total number of documents retrieved}}$$



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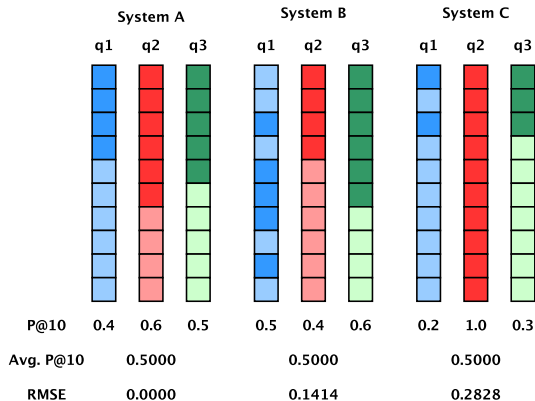
An in-depth investigation on the behavior of measures to quantify reproducibility, Maistro et al., IPM'23

Retrieval effectiveness error

Root mean square error

$$\text{RMSE}(M(r), M(r')) = \sqrt{\frac{1}{n} \sum_{j=1}^n (M_j(r) - M_j(r'))^2}$$

r, r' Original and reproduced runs
 n Total number of queries
 $M(r)$ Vector where each component is the score of an evaluation measure M with respect to the j -th query



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Rank correlation

Kendall's τ

$$\tau_j(r_j, r'_j) = \frac{P - Q}{\sqrt{(P + Q + U)(P + Q + V)}}$$

$$\bar{\tau}(r, r') = \frac{1}{n} \sum_{j=1}^n \tau_j(r_j, r'_j)$$

r_j, r'_j Original and reproduced rankings

n Total number of queries

P, Q Number of dis-/concordant pairs

U, V Number of ties in r_j and r'_j

	System A			System B			System C		
	q1	q2	q3	q1	q2	q3	q1	q2	q3
P@10	0.4	0.6	0.5	0.5	0.4	0.6	0.2	1.0	0.3
Avg. P@10	0.5000			0.5000			0.5000		
RMSE	0.0000			0.1414			0.2828		
KTU	1.000			0.4370			0.0815		

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Retrieval method

BM25 [2] is a strong and common baseline, also implemented in many industrial applications.

$$s(d, q) = \sum_{t \in q} \log \left(\frac{N - df_t + 0.5}{df_t + 0.5} \right) \cdot \frac{tf_{td}}{k_1 \cdot \left(1 - b + b \cdot \left(\frac{L_d}{L_{avg}} \right) \right) + tf_{td}}$$

d Document $d \in D$

q Query

t Term contained in query q

N Number of all documents in collection D

tf_{td} Term frequency of term t in document d

df_t Document frequency of term t in collection D

L_d Length of the document d

L_{avg} Average length of the documents in collection D

b **controls the impact of document length normalization**

k_1 **controls the saturation point of term frequency normalization**

Experimental setup

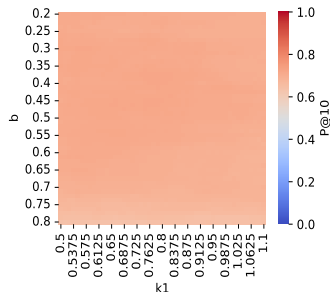
Simulating system reproductions:

⚙️ **Original system:** BM25 with $b = 0.5$ and $k_1 = 0.8$ as the reference system

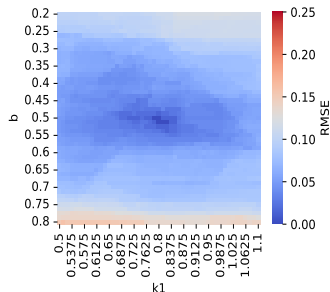
⚙️ **Reproduced systems:** A total of 2,400 “reproduced” BM25-based systems with different b and k_1 parameters simulating reimplementations of the reference system

📊 **Dataset:** TREC-COVID [4] test collection comprising 191,175 documents, 50 topics, and 69,318 relevance judgments

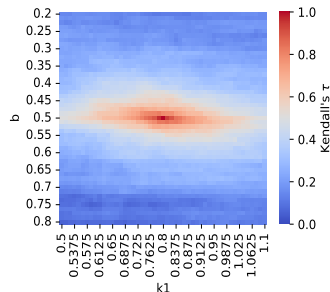
Experimental results



(a) Retrieval effectiveness



(b) Retrieval effectiveness error



(c) Rank correlation

Figure: Heatmaps showing the average retrieval effectiveness in terms of P@10, the error between topic scores in terms of RMSE, and the correlation between the rankings in terms of Kendall's τ . Each patch is averaged over 50 queries given by the TREC-COVID collection and based on BM25 rankings with different b and k_1 parameters. RMSE and Kendall's τ are determined wrt. the center patch with $b = 0.5$ and $k_1 = 0.8$.

What is a successful reproduction?

👍 Same retrieval effectiveness:

On average, the retrieval effectiveness in terms of $P@10$ is the same.

🗨️ Users would see different rankings as in the original experiment:

What does a different ranking mean for the user in terms of reproducibility?

❓ Users can compensate for changes or deteriorations in the rankings [5]:

Is bit- or listwise reproducibility really a hard requirement?

User simulations

Real-life user studies are too costly and time-intensive:

Think of $2,401 \text{ systems} \times 50 \text{ queries} = 120,050 \text{ experiments!}$

User simulations are a cost- and time-efficient solution:

- ✓ No participants need to be recruited,
- ✓ User behavior is controllable,
- ✓ Reproducible interactions with the system,
- ✓ No (cognitive) biases or learning effects,
- ✓ Simulate different devices (e.g., mobile vs. desktop).

Validation of query variations and click behavior:

Validating Simulations of User Query Variants, Breuer et al., ECIR'22

Validating Synthetic Usage Data in Living Lab Environments, Breuer et al., JDIQ'24

- ✎ Always document and report (hyper-)parameters as detailed as possible!
- 📊 Reproducibility can be quantified at different levels, from different perspectives.
- 👥 User simulations for estimating the implications of reproduced real-life applications.
- 🎯 Future work needs to evaluate the fidelity of the simulations and the user model.

Thank you!

Thank you for your attention.

Questions?



<https://github.com/breuert/acmrep24>



<https://doi.org/10.5281/zenodo.10931438>

References I

- [1] D. Harman, *Information Retrieval Evaluation* (Synthesis Lectures on Information Concepts, Retrieval, and Services). Morgan & Claypool Publishers, 2011.
- [2] S. E. Robertson, S. Walker, S. Jones, M. Hancock-Beaulieu, and M. Gatford, “Okapi at TREC-3,” in *TREC*, ser. NIST Special Publication, vol. 500-225, National Institute of Standards and Technology (NIST), 1994, pp. 109–126.
- [3] M. G. Kendall, *Rank Correlation Methods*. Oxford, England: Griffin, 1948.
- [4] E. M. Voorhees, T. Alam, S. Bedrick, *et al.*, “TREC-COVID: constructing a pandemic information retrieval test collection,” *SIGIR Forum*, vol. 54, no. 1, 1:1–1:12, 2020.
- [5] W. R. Hersh, A. Turpin, S. Price, *et al.*, “Do batch and user evaluation give the same results?” In *SIGIR*, ACM, 2000, pp. 17–24.