

Maik Fröbe¹



Christopher Akiki² Martin Potthast²



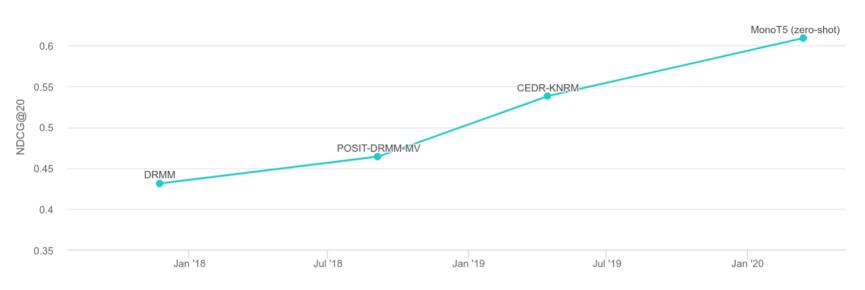


Matthias Hagen¹

Friedrich Schiller University Jena¹ Leipzig University² SPIRE, 8-10 November 2022

webis.de

Motivation: Leaderboard for Retrieval Effectiveness on Robust04

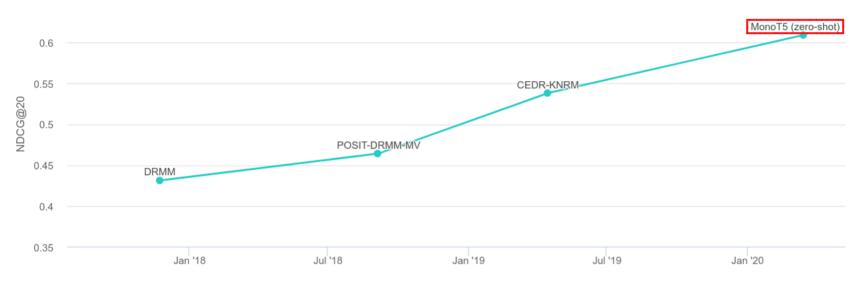


- □ Robust04: 249 test queries with dense judgments
 - Traditional setup with cross-validation



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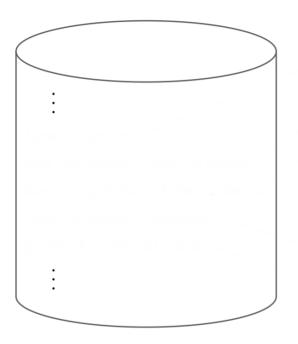
- Robust04: 249 test queries with dense judgments
 - Traditional setup with cross-validation
- MonoT5 (zero-shot)
 - Trained only on MS MARCO (> 10 million queries available)
 - There might be overlapping queries: Is this train—test leakage?



Overlapping Queries for Topic 441 of Robust04

MS MARCO

Robust04



☐ Train on many queries

Title: lyme disease

Description: How do you prevent and treat Lyme disease?

Narrative: Documents that discuss current prevention and treatment techniques for Lyme disease are relevant. Reports of research on new treatments of the disease are also relevant.

Query variants:

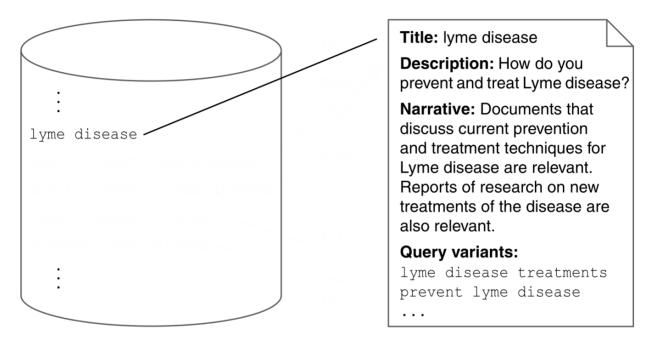
lyme disease treatments prevent lyme disease

□ Test on 249 queries



Overlapping Queries for Topic 441 of Robust04

MS MARCO Robust04



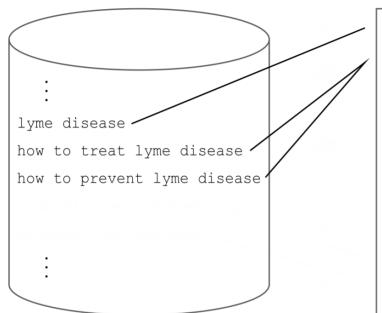
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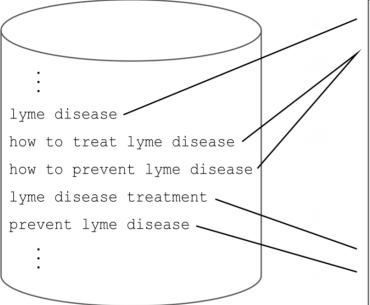
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Overlapping Queries for Topic 441 of Robust04 MS MARCO Robust04

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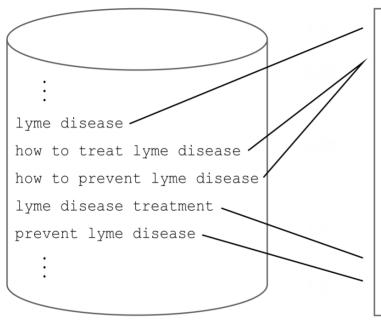
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Overlapping Queries for Topic 441 of Robust04

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Train on many queries

□ Test on 249 queries

Is the evaluation of MonoT5 invalidated by overlapping queries?



Might MonoT5 Benefit From Overlapping Queries?

MonoT5

- 3 billion parameters sequence-to-sequence model
- \Box The query q and the document d are embedded in a input sequence:

Query: q Document: d Relevant:

Documents ranked by the probability that the next token is "true"



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WHEN YOU TRAIN PREDICTIVE MODELS ON INPUT FROM YOUR USERS, IT CAN LEAK INFORMATION IN UNEXPECTED WAYS.



Candidates for Leaking Queries

- Nearest-neighbor search for overlapping queries
- Sentence-BERT embeddings for all MS MARCO and ORCAS queries
- Exact cosine similarity nearest-neighbor search with Faiss



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Verification of Candidates for Leaking Queries

- Manually review of the 5 most similar candidates per topic above threshold
- Identified query reformulation types:

Туре	Queries
Identical	187
Generalization	124
Specialization	228
Reformulation	182
Different Topic	106



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172 of 249 test queries from Robust04 occur in MS MARCO (69%)



Impact of Leaking Queries: Experimental Setup

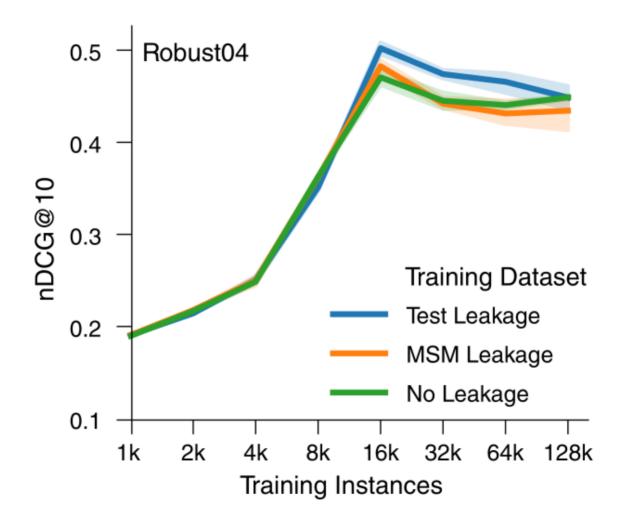
- Models trained on dedicated datasets to assess train—test leakage
- □ Varying training set sizes: 1,000 to 128,000 instances
- Each model trained five times on each dataset

Training Datasets

- No Leakage
 - Random non-leaking queries
 - balanced between MS MARCO and ORCAS
- MS MARCO Leakage
 - 500 random manually verified leaking queries from MS MARCO
 - supplemented by no-leakage queries
- □ Test Leakage
 - 500 queries from the actual test data
 - supplemented by no-leakage queries
 - Meant as an "upper bound" for any train—test leakage effect



Effectiveness of Retrieval Models



Effectiveness of Retrieval Models

Multiple models in five-fold cross-validation setup

Model		nDCG@10 on R04	
	No Leakage	MS MARCO Leakage	Test Leakage
Duet	0.201	0.198	0.224 [†]
KNRM	0.194	0.214^{\dagger}	0.309^\dagger
${\sf monoBERT}$	0.394	0.373^{\dagger}	0.396
monoT5	0.461	0.457	0.478^\dagger
PACRR	0.382	0.364 [†]	0.391

Effectiveness of Retrieval Models

Increase in rank-offset between leaked relevant and non-relevant documents

Model MS MARCO Leakage		kage Test Leakage	
Duet	6.378 ±32.15	0.809 ±17.69	
KNRM	$0.640{\scriptstyle~\pm 19.22}$	$1.335{\scriptstyle~\pm 11.75}$	
monoBERT	0.692 ± 17.97	3.886 ± 20.39	
monoT5	$\textbf{0.443} \pm 8.60$	3.443 ± 19.96	
PACRR	0.043 ±19.30	1.952 ±17.71	

Takeaways

- Possible train—test leakage for models trained on MS MARCO
 - Potential to invalidate experiments
 - Default in PyTerrier/Pyserini/PyGaggle often trained on MS MARCO
 - Only few training instances overlap: Impact measurable, but negligible
- Future work:
 - Effects on Dense Retrieval models
 - Practical consequences for real search engines



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Thank You!

