

# A Proposal for an LSR-Benchmark for Efficiency and Effectiveness-oriented Evaluations

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Additional Collaborators Welcome :)

University of Jena

University of Kassel

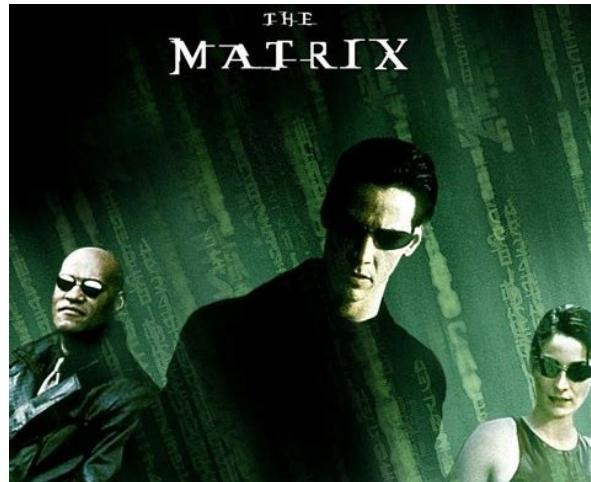
ISTI-CNR

University of Tübingen

University of Pisa

# A Proposal for an LSR-Benchmark

Learned Sparse Retrieval in a Nutshell



Example document:

Neo lives in a simulation.

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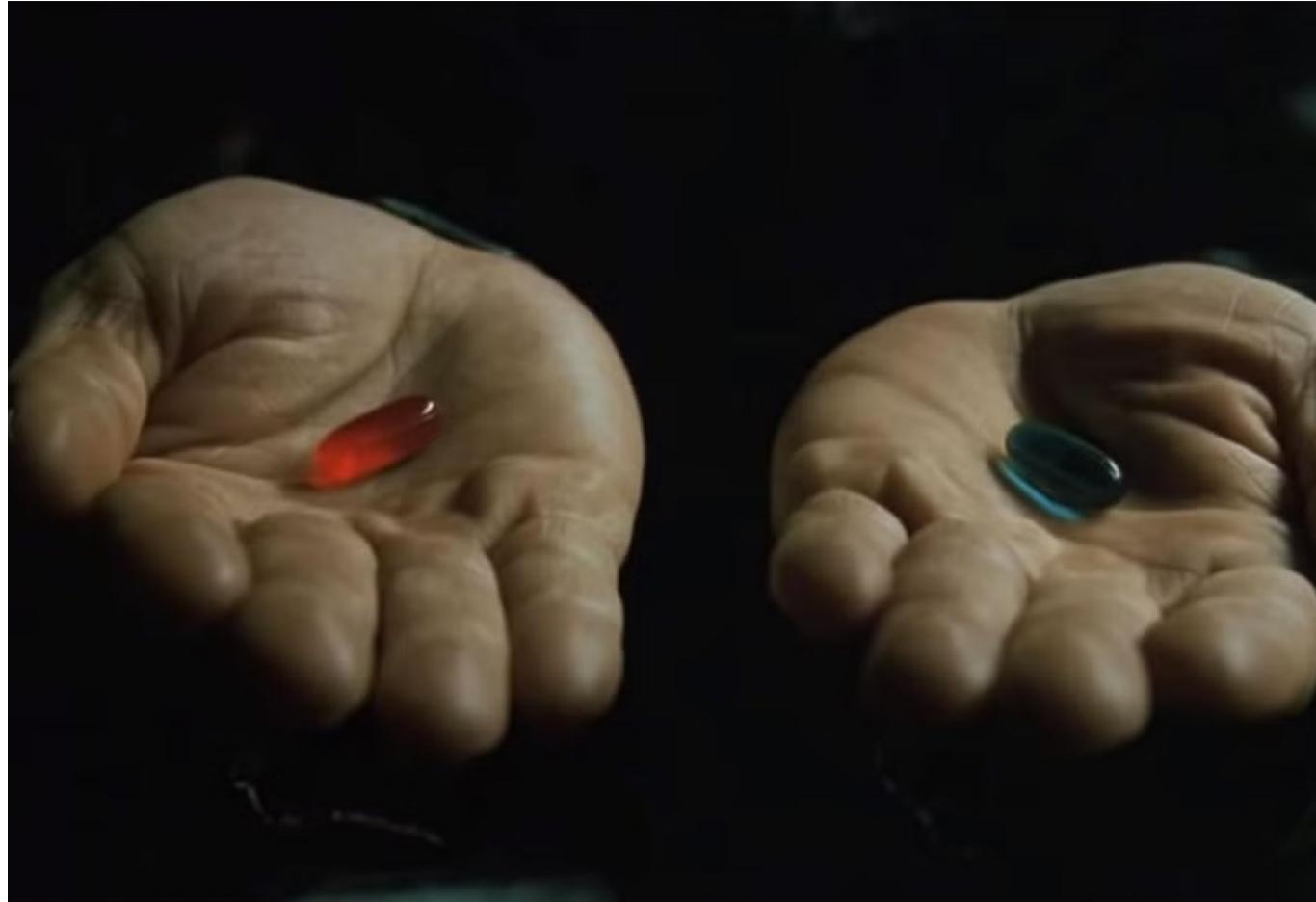
**Learn** how to expand and remove terms

Interesting trade-offs to study:

- ❑ Efficiency: Inverted index can be used
  - But: different term distributions compared to lexical retrieval
- ❑ Effectiveness: model learned for specific task
  - Different retrieval tasks might require different term expansion/removal

# A Proposal for an LSR-Benchmark

Should we focus our Evaluation on Efficiency or Effectiveness?



from the Matrix Movie

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## Effectiveness-Oriented LSR Experiments

### Inputs

- Document texts
- Query texts

### Focus

- Model** to embed documents/queries

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Goal: Enable Synergies between both for holistic evaluations

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## What Evaluation Corpora Should we use?

- ❑ TREC-style evaluation corpora are ideal
  - Reliable evaluations: Pooling for judgments ( $> 500$  judgments per query)
- ❑ Still, TREC-style corpora are hard
  - Often very large (e.g., hundreds of GB to a few TB)

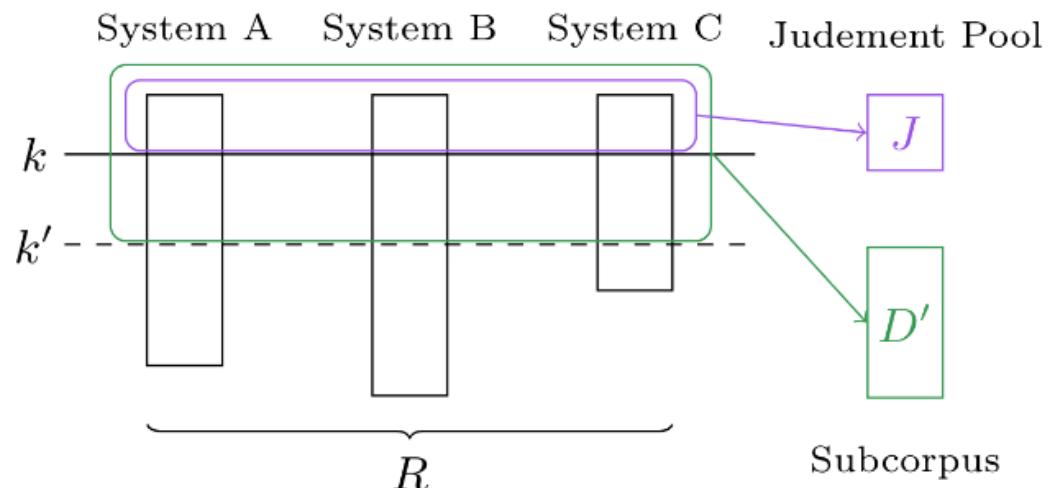
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## Our Solution: Corpus Subsampling for Fast and Green Evaluation

[ECIR'25]



# A Proposal for an LSR-Benchmark

## How big are the Corpus Subsamples?

[ECIR'25]

Corpus	Complete			Subsampled		
	Docs.	$\notin_J$	Size	Docs.	$\notin_J$	Size
ClueWeb09	1.0 b	99 %	4.0 TB	0.3 m	73 %	<b>0.9 GB</b>
ClueWeb12	0.7 b	99 %	4.5 TB	0.1 m	72 %	<b>0.5 GB</b>
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- ❑ Of texts of evaluation corpora for effectiveness-oriented evaluations
- ❑ Embeddings for efficiency-oriented evaluations

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## Custom Post-Processing

- ❑ Re-mapped document IDs to UUIDs to work on custom slices/dices
- ❑ Splitting of documents into passages as in MS MARCO v2.1

# A Proposal for an LSR-Benchmark

We prepare an `ir_datasets` like API for frequent use-cases

Step 1: Embedding

```
dataset = lsr_benchmark.load('<IR-DATASETS-ID>')

# process the document texts:
with tracking():
    for doc in dataset.docs_iter(embedding=None):
        doc # namedtuple<doc_id, segments.text>
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## Step 2: Indexing and Retrieval

```
dataset = lsr_benchmark.load('<IR-DATASETS-ID>')

# process the document texts:
with tracking():
    for doc in dataset.docs_iter(embedding='<EMBEDDING-MODEL>', passage_aggregation="first-passage"):
        doc # namedtuple<doc_id, embedding>
        # todo: index document
```

# A Proposal for an LSR-Benchmark

## Conclusions

- ❑ Goal: lsr benchmark for holistic evaluations of efficiency/effectiveness
- ❑ We use corpus subsampling to enable sharing of corpora and embeddings
- ❑ We have a vertical prototype for the 2009 Web Track on the ClueWeb09
- ❑ There is much room for improvement and collaboration



### lsr-benchmark



# Combining Teaching and Research in IR

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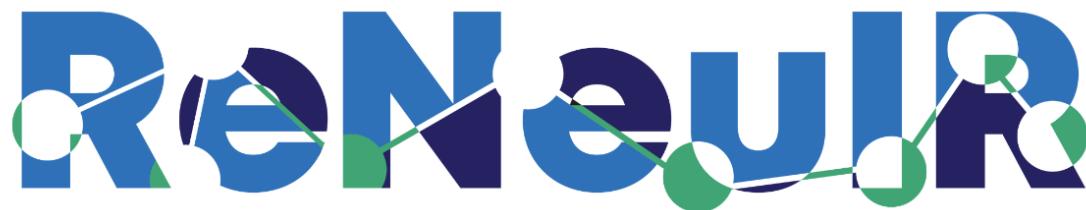
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- ❑ ECIR paper of some form

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lsr-benchmark



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Thank you!