Towards Reproducible Shared Tasks in IR



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Motivation: SIGIR '20 Keynote by Norbert Fuhr

# Solid Empirical Evidence is Important!



Norbert Fuhr

Proof by Experimentation? Towards Better IR Research

7

A Shared Task in IR?



A Shared Task in IR?



- Problem 1: Internal validity
- Problem 2: External validity

A Shared Task in IR?



Potential problems: [Fuhr'21]

- Problem 1: Internal validity
- D Problem 2: External validity

 Problem 3: Blinded experimentation with LLMs

Problem 1: Internal Validity [Fuhr'21]

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- Possible problems
  - □ Wrong baseline [Armstrong'09,Lin'18]
  - Formulate hypothesis after experiments [Fuhr'21]

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- Possible solutions
  - Centralized leaderboards
    - E.g., Run uploads to EvaluateIR [Armstrong'09]
  - Task-specific leaderboards
    - E.g., MS MARCO, MIRACL [Lin'22,Zhang'22]

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Problem 1: Internal Validity [Fuhr'21]

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### The hypothesis is supported by the data.

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  - Wrong baseline
     [Armstrong'09,Lin'18]
  - Formulate hypothesis after experiments [Fuhr'21]
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"EvaluateIR never gained traction, and a number of similar efforts following it have also floundered" [Lin'18]

Problem 2: External Validity [Fuhr'21]

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Repeating an experiment on similar data yields similar observations.

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Possible problems

- Non-reproducible results
- **Possible Solutions** 
  - TREC Open Runs
    [Voorhees'16]
  - Reproducibility initiatives
    - OSIRRC: Archive artifacts [Arguello'15,Clancy'19]
    - CENTRE: Reimplementation [Ferro'19,Sakai'19]
  - Platforms + documentation
    - CodaLab, EvalAI, PRIMAD, STELLA, TIRA
  - Meta evaluations: BEIR
     [Thakur'21]

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- □ 19 of 69 runs (Problems: 11)
- 2015: 8 systems archived
   2019: 1 system fully reproducible
   [Lin'19]
- Limited adoption of jig + CIFF [Clancy'19]
- Additional effort
- Evaluations on subsets
- Often sparse judgments

Problem 3: Blinded Experimentation with LLMs



I worry about language models being trained on test sets. Recently, we emailed support@openai.com to opt out of having our (test) data be used to improve models. This isn't enough though: others running evals could still inadvertently contribute those test sets to training.

...

Problem 3: Blinded Experimentation with LLMs



Percy Liang @percyliang

I worry about language me emailed support@openai. used to improve models. T could still inadvertently co



Horace He @cHHillee

I suspect GPT-4's performance is influenced by data contamination, at least on Codeforces.

Of the easiest problems on Codeforces, it solved 10/10 pre-2021 problems and 0/10 recent problems.

Tweet übersetzen

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triangle! brute force, geometry, math	2 2	Actions data structures, greedy, implementation, math	4
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### TIREx to the Rescue?



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TIREx does "one thing": Integrate Existing Tools

TIRA

□ Reproducible shared tasks: Software submissions + blinded experiments

ir\_datasets

□ Unified + random data access: Documents + queries + rel. judgments

PyTerrier

Declarative reproducibility pipelines

### TIREx to the Rescue?



### TIREx does "one thing": Integrate Existing Tools

# TIRA Improves internal validity No Leakage for LLMs Reproducible shared tasks: Software submissions + blinded experiments ir\_datasets Unified + random data access: Documents + queries + rel. judgments PyTerrier

Declarative reproducibility pipelines

### **TIREx: Overview**

- Organizer provides (private) Docker image with ir\_datasets integration
- Participants provide Docker images with retrieval approaches



Covers a shared task end-to-end

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Advantages of software submissions via Docker:

- $\hfill\square$  Executed in a sandbox  $\rightarrow$  blinded experimentation
- $\hfill\square$  Re-execution on same/similar data  $\rightarrow$  improves reproducibility + replicability

# **TIREx: Feasibility Study**

### 50 Transferrable Retrieval Models in TIRA

- Derived from tira-starters from 4 starters
- Retrieve against default text in ir\_datasets
- $\hfill\square$  Selecting suitable baseline  $\rightarrow$  improves internal validity
- Diversification of pools for shared tasks with few participants

Framework	Туре	Description	Systems	
BEIR [78]	Bi-Encoder	Dense Retrieval	17	
ChatNoir [7]	BM25F Retrieval	Elasticsearch Cluster	1	
ColBERT@PT [55]	Late Interaction	Pyterrier Plugin	1	
DuoT5@PT [71]	Cross-Encoder	Pairwise Transformer	3	
PyGaggle [59]	Cross-Encoder	Pointwise Transformer	8	
PyTerrier [64]	Lexical	Traditional Baselines	20	
$\sum = 6 = 4$ frameworks + 2 forks				

# **TIREx: Feasibility Study**

### 32 Exchangeable Benchmarks in TIRA

 $\Box$  Models can be transferred to new corpora  $\Rightarrow$  improves external validity

Corpus			Included Benchmarks			
Name	Docs.	Size	Details	#		
Args.me	0.4 m	8.3 GB	Touché 2020–2021 [9, 10]	2		
Antique	0.4 m	90.0 MB	QA Benchmark [47]	1		
ClueŴeb09	1.0 b	4.0 TB	Web Tracks 2009–2012 [22–25]	4		
ClueWeb12	731.7 m	4.5 TB	Web Tracks [29, 30], Touche [9, 10]	4		
ClueWeb22B	200.0 m	6.8 TB	Touché 2023 [8] (ongoing)	1		
CORD-19	0.2 m	7.1 GB	TREC-COVID [85, 90]	1		
Cranfield	1,400	0.5 MB	Fully Judged Corpus [27, 28]	1		
Disks4+5	0.5 m	602.5 GB	TREC-7/8 [87, 88], Robust04 [81, 82]	3		
Gov	1.2 m	4.6 GB	Web Tracks 2002–2004 [32–34]	3		
Gov2	25.2 m	87.1 GB	TREC TB 2004–2006 [18, 21, 26]	3		
Medline	3.7 m	5.1 GB	Trec Genomics [48, 49], PM [73, 74]	4		
MS MARCO	8.8 m	2.9 GB	Deep Learning 2019–2020 [35, 36]	2		
NFCorpus	3,633	30.0 MB	Medical LTR Benchmark [12]	1		
Vaswani	11,429	2.1 MB	Scientific Abstracts	1		
WaPo	0.6 m	1.6 GB	Core 2018	1		
$\Sigma = 15$ corpora	1.9b	15.3 TB		32		

# **TIREx: Feasibility Study**

Initial Leaderboards: 1600 runs

- □ Running all 50 models on all benchmarks took 1 Week
- □ See https://github.com/tira-io/ir-experiment-platform
- □ Additional use-cases: LTR, QPP, etc.

Teaser of results:

Observe system preferences on TREC DL 2019

Benchmark

Use repro\_eval to measure the proportion of reproducible preferences
[Breuer'20,Breuer'21]

Rank Succ.

TREC DL 2020	1	85.2
Touché 20 (Task 2)	2	81.0
Touché 21 (Task 2)	3	72.6
Web Track 2004	4	72.1
CORD-19	5	70.0
Terabyte 2006	10	62.1
<b>TREC PM 2017</b>	15	53.4
Terabyte 2005	20	42.2
TREC PM 2018	25	33.2
Cranfield	30	28.8

# **TIREx: Conclusion**

Integration of existing tools

□ TIRA, ir\_datasets, PyTerrier

Better benefit/effort ratio then previous approaches for shared tasks?

- One software submission, evaluation on many datasets
- Evaluate on datasets to which you dont have access

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Future Work

- □ Move to generative IR (integration of Alpaca)
- □ Integration of cloud infrastructure of the Open Search Foundation
- Render SERPs with DiffIR
- We would be happy to help you with your shared task!

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### Backup: SemEval'23 ValueEval Demo (1)

# Human Value Detection Demo

Demo for the Adam Smith human value detector by Schroter et al. (2023) [paper under review], which performed best in the ValueEval'23 co ensemble of three models that performed best in the ablation tests. [code: original, docker image, server docker image]

Enter an argument in the text area and click on submit. After a few seconds, the detected value categories will be highlighted in the value ta



### Backup: SemEval'23 ValueEval Demo (2)



personal

Security

societal

Conformity:

rules

Tradition

€<sup>ace</sup>

# **Backup: Limitations**

- Computational resources.
   Potential Solution:
  - Hybrid submissions: Run upload, Software submission only for plausibility checks
  - —
  - OSF infrastructure
- □ How to avoid big ensembles?
- Evaluation measures required that combine efficiency with effectiveness?
- New iteration of the IRF?

### **Backup: Use in Teaching**

- □ Cover the "full cycle" with students in IR exercises?
  - We do this next term

### **Backup: Definition of Multi-Stage Software**

	🗟 TIRA			Admin	Forum	•	۹	≡	200
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# Figure 3: The definition of a full-rank retrieval software in TIRA that consists of two modularized components.

### **Backup: Full-Rank**

```
pipeline = tira.pt.retriever(
    '<task-name>/<user-name>/software',
    dataset
)
advanced_pipeline = pipeline >> advanced_reranker
```

#### Listing 1: Full-Rank Retrieval from a complete corpus.

### **Backup: Load Submissions**

```
first_stage = tira.pt.from_submission(
    '<task-name>/<user-name>/<software>',
    dataset='<dataset>'
)
advanced_pipeline = first_stage >> advanced_reranker
```

Listing 3: Re-Rank a run created by a software submission.