Towards Reproducible Shared Tasks



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Motivation

How to advertise your Dataset / Paper?



- □ You are Rich / Famous?
- People will come anyway

Motivation

How to advertise your Dataset / Paper?

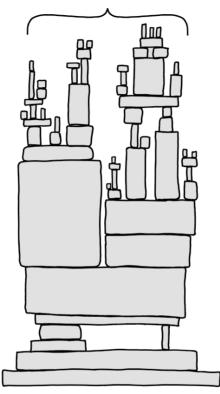




- □ You are Rich / Famous?
- People will come anyway
- □ Organize a shared task (e.g., 2023):
 - Clickbait Spoiling (30 of 83 teams submitted; 24 countries)
 - ValueEval (41 submitted)
 - Pan (32 submitted)
 - Touché (8 submitted)

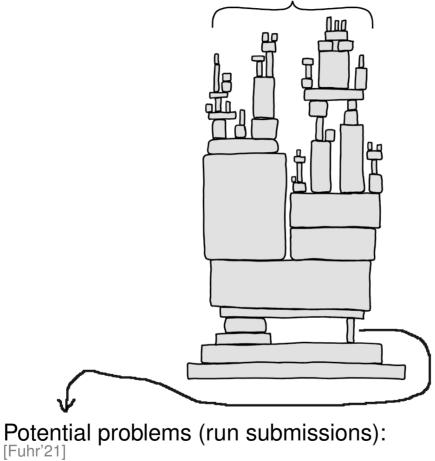
Motivation

Your Shared Task?



Motivation

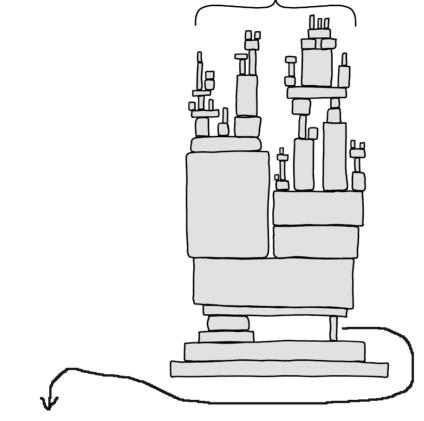
Your Shared Task?



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

Motivation

Your Shared Task?



Potential problems (run submissions): [Fuhr'21]

- □ Problem 1: Internal validity □ Problem 3: Blinded e
- Problem 2: External validity

 Problem 3: Blinded experimentation with LLMs

Problem 1: Internal Validity [Fuhr'21]

Goal

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Goal

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

Problem 1: Internal Validity [Fuhr'21]

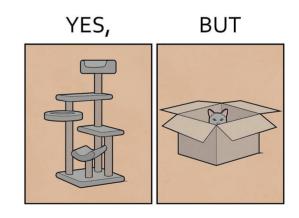
Goal

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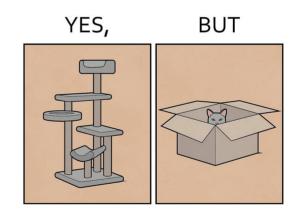


Problem 1: Internal Validity [Fuhr'21]

Goal

The hypothesis is supported by the data.

- Possible problems
 - Wrong baseline
 [Armstrong'09,Lin'18]
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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]

Goal

Repeating an experiment on similar data yields similar observations.

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Possible problems

Non-reproducible results

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Possible problems

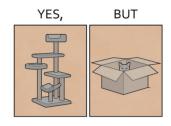
- Non-reproducible results
- **Possible Solutions**
 - TREC Open Runs
 [Voorhees'16]
 - Reproducibility initiatives
 - OSIRRC: Archive artifacts [Arguello'15,Clancy'19]
 - CENTRE: Reimplementation [Ferro'19,Sakai'19]
 - Platforms + documentation
 - CodaLab, EvalAI, PRIMAD, STELLA, TIRA
 - Meta evaluations: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

Problem 2: External Validity [Fuhr'21]

Goal

Repeating an experiment on similar data yields similar observations.

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 - Non-reproducible results
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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



Percy Liang @percyliang

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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TIRA to the Rescue?



Reproducible Shared Tasks with TIRA

Evolution of TIRA

[Gollub'12,Potthast'19,Fröbe'23]

- □ 2005–2011: Pipelines, eval. run submissions, manual software submissions
- □ 2012–2022: Software submissions with virtual machines
- 2023-today: Immutable software submissions with Docker + Git CI/CD
 - Shared task = git repository
 - Software execution = commit

Reproducible Shared Tasks with TIRA

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[Gollub'12,Potthast'19,Fröbe'23]

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Procedure:

- 1. Implement approach in Docker image
- 2. Upload image to dedicated image registry in TIRA
- 3. Your approach is executed in a Kubernetes cluster via a commit



Benefits of TIRA

Blinded Experimentation

- Software executed in sandbox: No internet connection
- □ 2 types of datasets:

Туре	Blinded	Unblinding	Feedback
Validation	Nothing	Direct	Everything
Test	Everything	Manual	√vs X

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Repeat, Replicate, and Reproduce in One Line of Code

Git repository of the shared task can be published after the task

□ SemEval'23: 2 tasks, 83 + 91 reg. teams (active: 31 + 42; Docker: 21 + 7)

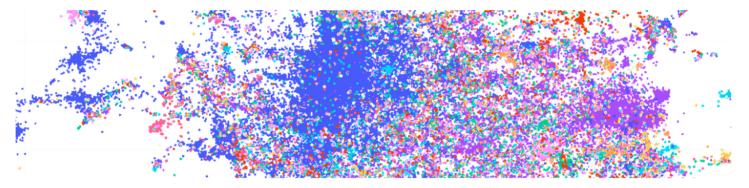
Benefits of TIRA

Run Experiments / Analysis on Confidential Data

Example: Paper for which you cant publish the Dataset

I The Archive Query Log

Mining Millions of Search Result Pages of Hundreds of Search Engines from 25 Years of Web Archives.



Start now by running your custom analysis/experiment, scraping your own query log, or just look at our example files.

How you could set it up

What would you need?

- □ A (small) public dev/train dataset
- □ A (private) test dataset
- A dockerized baseline
- A Tutorial: How to run the baseline on dev?
- Evaluation measures (we already have many)

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What we often do:

- □ Step 1: Publish Paper with Baseline + (confidential) Dataset
- □ Step 2: Run shared task

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What we often do:

- □ Step 1: Publish Paper with Baseline + (confidential) Dataset
- □ Step 2: Run shared task
- We often add this to proposals
- □ We combine this with teaching:
 - IR course in Leipzig
 - NLP course in Jena and Weimar

Conclusion

TIRA allows experiments / shared tasks on confidential data with software submissions

- Improved Reproducibility
- Blinded Experimentation

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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 - □ Shared tasks over multiple university courses
 - Currently in discussion with 5 courses

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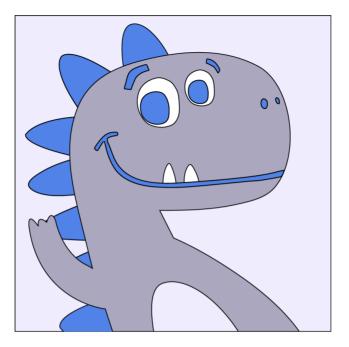
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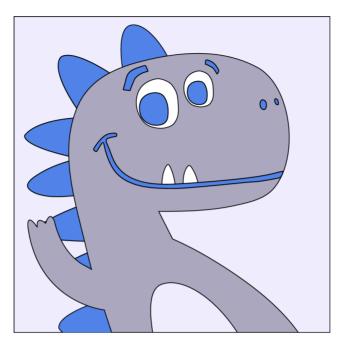
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Example: TIREx



Example: TIREx



TIREx does "one thing": Integrate Existing Tools

TIRA

□ Reproducible shared tasks: Software submissions + blinded experiments

ir_datasets

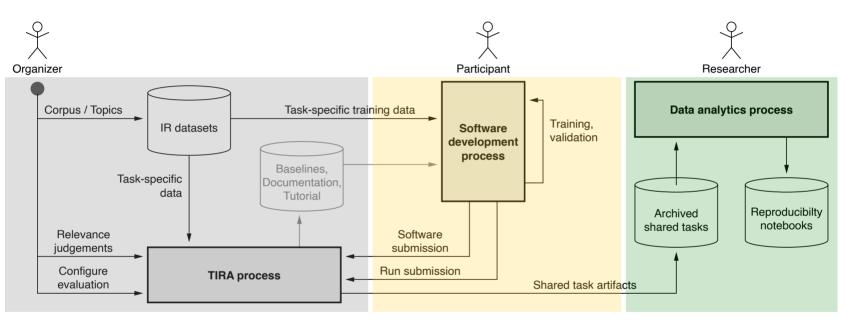
□ Unified + random data access: Documents + queries + rel. Judgments

PyTerrier

³² 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

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		
PyGaggle [59] 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		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		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		
$\sum = 15 \text{ corpora}$	1.9 b	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.

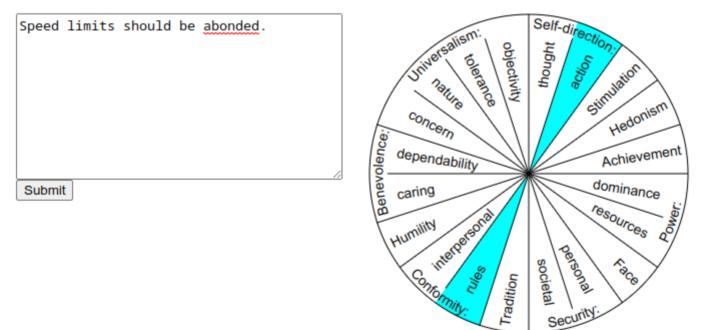
TREC DL 2020	1	85.2
Touché 20 (Task 2)	2	81.0
Touché 21 (Task 2)	3	72.6
Web Track 2004	4	72.1
CORD-19	5	70.0
Terabyte 2006	10	62.1
TREC PM 2017	15	53.4
Terabyte 2005	20	42.2
TREC PM 2018	25	33.2
Cranfield	30	28.8

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

€^{ace}

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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	Docker Image								
	registry.tira.io/tira-ir-starter/pyterrier:0.0.1 \$;	ADD (CONTAI	NER	

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.