Information Retrieval in an Infodemic: The Case of COVID-19 Publications

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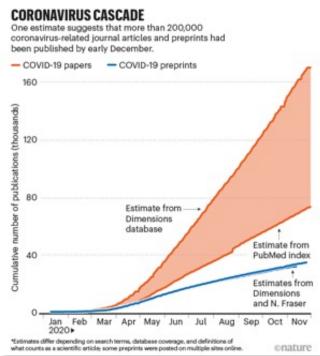


Outline

- Introduction
- TREC COVID
- Information retrieval model
- Challenge results
- Discussion

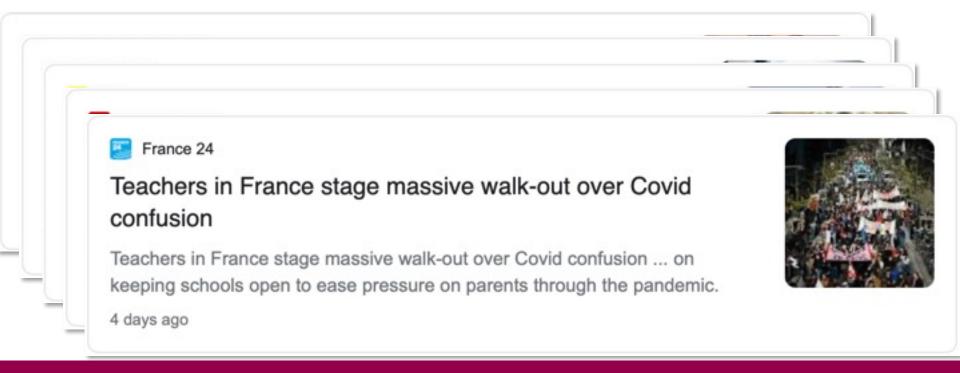
Introduction

- The COVID-19 pandemic resulted in an explosive surge of activities within scientific communities and across many disciplines
 - Stakeholders are unable to keep up with the fastevolving body of knowledge disseminated
- **Infodemic**: overabundance of information online and offline with often negative impacts on the population
 - Confusion and desensitization among audiences



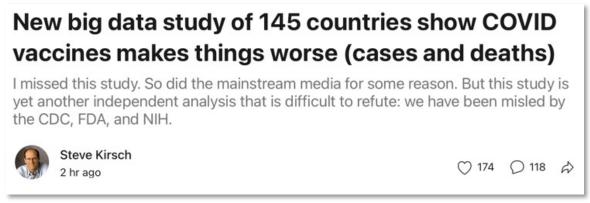
Else H. How a torrent of COVID science changed research publishing-in seven charts. Nature. 2020:553-.

COVID Infodemic



COVID Infodemic

- More than 200 Covid-19 papers have been retracted
 - Elementary calculation errors, lack of transparency, conclusions not supported by the data, etc.



*https://retractionwatch.com/retracted-coronavirus-covid-19-papers/

WHO Framework for Managing the COVID-19 Infodemic



Action area 1: strengthening the scanning, review and verification of evidence and information



Action area 2: strengthening the interpretation and explanation of what is known, fact-checking statements, and addressing misinformation



Action area 3: strengthening the amplification of messages and actions from trusted actors to individuals and communities that need the information

Objective

To investigate an **information retrieval model** supported by deep language models **to improve search and discovery** of COVID-19 scientific literature

The TREC-COVID challenge

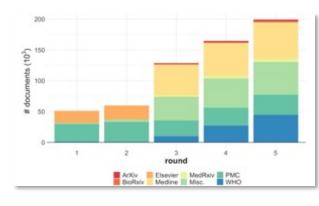
- A query set capturing relevant search questions of researchers during the pandemic
- Run in 5 rounds with a total of more than 50 teams
- Started with 30 topics in round 1 and added 5 new topics in each new round
 - 50 topics in round 5

The TREC-COVID challenge

Topic	Query	Question	Narrative
1	Coronavirus origin	What is the origin of COVID- 19?	Seeking a range of information about the SARS-CoV-2 virus's origin, including its evolution, animal source, and first transmission into humans
25	Coronavirus biomarkers	Which biomarkers predict the severe clinical course of 2019-nCOV infection?	Looking for information on biomarkers that predict disease outcomes in people infected with coronavirus, specifically those that predict severe and fatal outcomes
50	mRNA vaccine coronavirus	What is known about an mRNA vaccine for the SARS-CoV-2 virus?	Looking for studies specifically focusing on mRNA vaccines for COVID-19, including how mRNA vaccines work, why they are promising, and a

The CORD-19 dataset

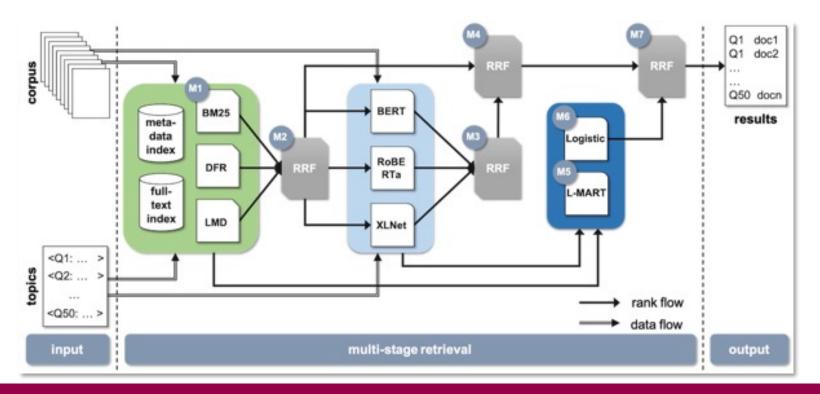
- **CORD-19 dataset**: Effort to gather publications, preprints, and reports related to the coronaviruses and acute respiratory syndromes from the AI2
- Large and dynamically growing semi-structured dataset from various sources
 - PubMed, PubMed Central (PMC), WHO, bioRxiv, medRxiv, arXiv, etc.
- Content: metadata, including title, abstract, and authors, among others, and the full text or link to full-text files when available



Teams

Team	Institution	Country
CSIROmed	Commonwealth Scientific and Industrial Research Organisation	Australia
anserini	University of Waterloo	Canada
covidex	University of Waterloo	Canada
xj4wang	University of Waterloo	Canada
HKPU	Hong Kong Polytechnic University	China
mpiid5	Max Planck Institute for Informatics	Germany
UCD_CS	University College Dublin	Ireland
uogTr	Glasgow Terrier Team	Scotland
Elhuyar_NLP_team	Elhuyar Foundation	Spain
risklick	ours	Switzerland
unique_ptr	Google Research	US
SFDC	Salesforce	US
udel_fang	University of Delaware	US
UlowaS	University of Iowa	US

Multistage retrieval methodology



First-stage retrieval

Classic query-document probabilistic weighting models

$$w(t,d,D) = tf(t,d) \cdot idf(t,D)$$

$$= \frac{f(t,d) \cdot (k_1+1)}{f(t,d) + k_1 \cdot \left(1 - b + b \cdot \frac{|d|}{avg_l}\right)} \cdot \log\left(\frac{|D|}{n_t}\right)$$

DFR

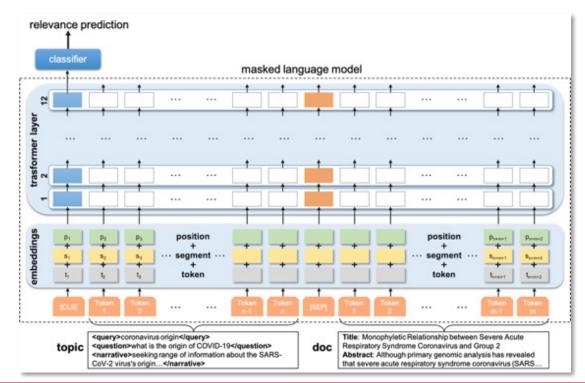
$$w(t, d, D) = k \cdot \log p_M(t \in d|D)$$

• LMD

$$w(t,d,D) = \frac{|d|}{|d|+\mu} \cdot p(t|d) + \frac{\mu}{|d|+\mu} \cdot p(t|D)$$

Second-stage reranking

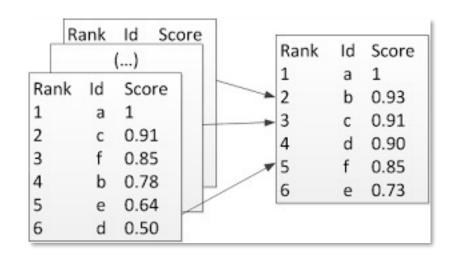
- Models
 - BERT
 - RoBERTa
 - XLNet
- Training set
 - 46k annotated
 query-doc pairs



Combining model results

- Reciprocal rank fusion (RRF)
- Given a set of documents D to be sorted and a set of ranking files $R = \{r_1...r_n\}$, each with a permutation on 1...|D|, the aggregated score is computed using the following equation:

$$s(q,d,R) = \sum_{i=1}^{n} \frac{1}{k+r_i(q,d)}$$

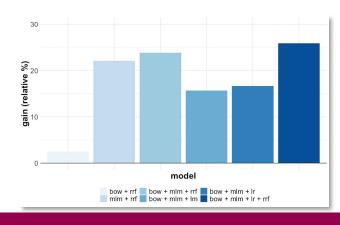


Results

Run	Name	Description
1	bm25	Run based on the baseline BM25 model using the metadata index
2	bow + rrf	An RRF combination of BM25, DFR, and LMD models computed against the metadata and full-text indices
3	mlm + rrf	An RRF combination of BERT, RoBERTa, and XLNet models applied to run 2
4	bow + mlm + rrf	An RRF combination of runs 2 and 3
5	bow + mlm + lm	A LambdaMART-based model using features from the individual models used to create runs 2 and 3
6	bow + mlm + lr	A logistic regression model using features from the individual models used to create runs 2 and 3
7	bow + mlm + lr + rrf	An RRF combination of runs 2, 3, and 6

Model performance

Run	Model	NDCG@20	P@20	Bpref	MAP	# rel
1	bm25	0.6320	0.6440	0.5021	0.2707	6533
2	bow + rrf	0.6475	0.6650	0.5174	0.2778	6695
3	mlm + rrf	0.7716	0.7880	0.5680	0.3468	6963
4	bow + mlm + rrf	0.7826	0.8050	0.5616	0.3719	7006
5	bow + mlm + lm	0.7297	0.7460	0.5759	0.3068	6834
6	bow + mlm + lr	0.7375	0.7450	0.5719	0.3439	6976
7	bow + mlm + lr + rrf	0.7961	0.8260	0.5659	0.3789	6939



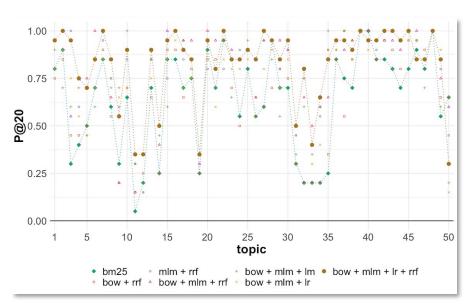
Challenge performance

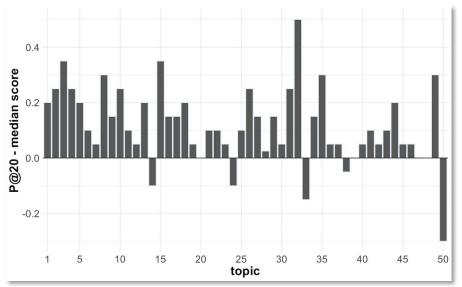
	Team	NDCG@20	P@20	Bpref	MAP
	unique_ptr	0.8496	0.8760	0.6378	0.4731
	covidex	0.8311	0.8460	0.5330	0.3922
	Elhuyar_NLP_team	0.8100	0.8340	0.6284	0.4169
)	risklick (ours)	0.7961	0.8260	0.5759	0.3789
	udel_fang	0.7930	0.8270	0.5555	0.3682
	CIR	0.7921	0.8320	0.5735	0.3983
	uogTr	0.7921	0.8420	0.5709	0.3901
	UCD_CS	0.7859	0.8440	0.4488	0.3348
	sabir	0.7789	0.8210	0.6078	0.4061
	mpiid5	0.7759	0.8110	0.5873	0.3903

Neural re-ranking *vs.* ranking fusion

Model	NDCG@20	P@20	Bpref	MAP	# rel
BERT	0.6209	0.6430	0.5588	0.2897	6879
RoBERTa	0.6261	0.6440	0.5530	0.2946	6945
XLNet	0.6436	0.6570	0.5644	0.3064	6926
mlm + rrf	0.7716	0.7880	0.5680	0.3468	6963

Topic performance analyses

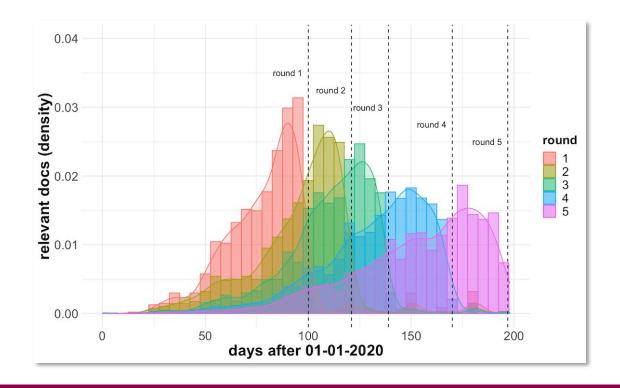




Topic performance analyses

Topic ID	Information	Included	Excluded
11	guidelines for triaging patients infected with coronavirus	diagnosis ("early recognition of coronavirus", "RT-PCR testing of SARS-CoV-2 for hospitalized patients clinically diagnosed",)	"telephone triage of patients with respiratory complaints"
12	best practices in hospitals and at home in maintaining quarantine	hospital preparedness ("improving preparedness for", "preparedness among hospitals",)	"home-based exercise note in Covid-19 quarantine situation"

Time-dependent relevance analyses



Conclusion

- The use of the multistage retrieval approach significantly improved the search results of COVID-related literature
 - Gain in performance of 25.9% in terms of the NDCG@20 metric compared to a bag-of-words baseline
- The **ensemble of masked language models** brought the highest performance gain to the search pipeline
- The proposed information retrieval pipeline can provide a potential solution to help stakeholders search and find the relevant information in the unique situation caused by the COVID-19 pandemic
 - Very competitive results as judged by the official leaderboard of the challenge

Thank you

