Consumer Health Question Answering Using Off-the-Shelf Components

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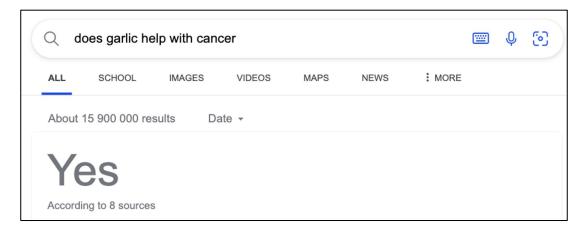
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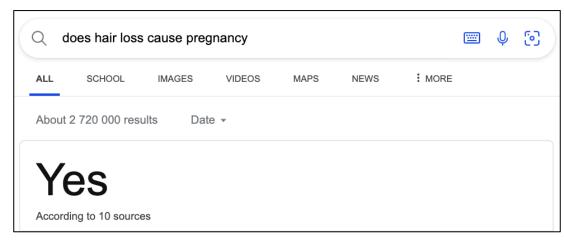


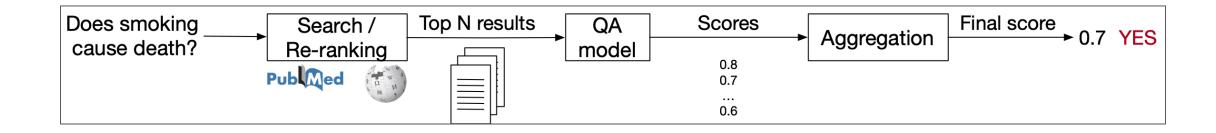
45th European Conference on Information Retrieval (ECIR 2023)

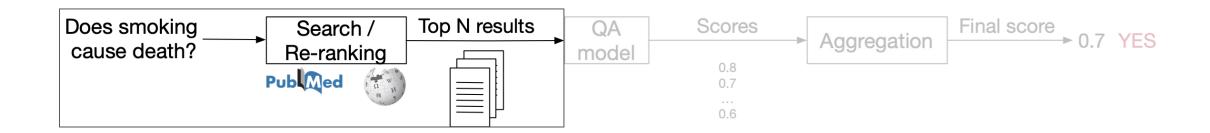
Motivation

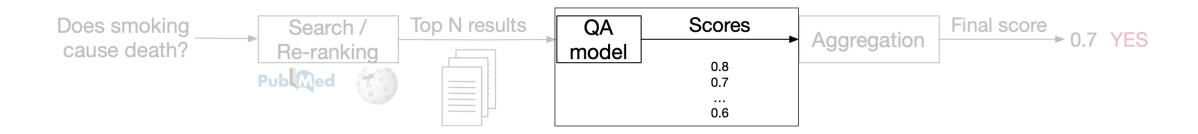
- About half of top-ranked search engines' results may provide incorrect answers to health-related questions [White, R.; SIGIR'13]
- It is important to provide the answers based on **reliable sources**
- We build our system from ready-touse third-party blocks

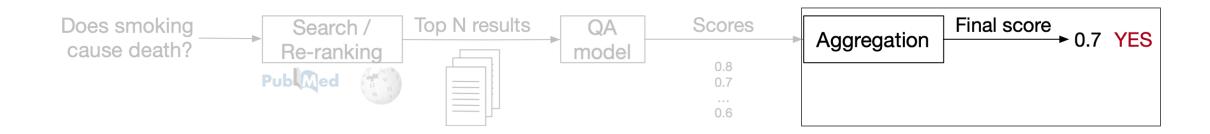












Experiments

Retrievers

- PubMed
 - Native PubMed Search
 - Google Search over PubMed
 - Google's BioMed Explorer
- Wikipedia
 - Wikipedia API
 - Google Search over Wikipedia

QA Models

- RoBERTa-large
 - Fine-tuned on BoolQ
- BioLinkBERT-large
 - Fine-tuned on PubMedQA
 - Fine-tuned on BioASQ

Score Aggregation

- Top-1 document's score
- Plain average over top 10 documents
- Weighted average over top 10 documents

Evaluation Data

• We use a collection of 113 yes/no health questions:

Source	# Questions (yes/no)				
TREC (2019, 2021)	84 (42/42)				
Yandex Log	15 (7/8)				
Health Beliefs in Twitter	14 (12/2)				

• Examples:

Question	Keyword Query	Label	Evident PubMed ID	
Can caffeine help people with asthma?	caffeine asthma	1	20091514	
Does aloe help for a runny nose?	aloe help runny nose	0	32924222	
Does milk cause acne?	milk causes acne	1		

Evaluation Results

	Retriever Query		RoBERTa-large (BoolQ)		BioLinkBERT (PubMedQA)			BioLinkBERT (BioASQ)			
			top 1	avg	wavg	top 1	avg	wavg	top 1	avg	wavg
PubMed	PubMed	keywords	0.65	0.66	0.66	0.61	0.62	0.61	0.56	0.58	0.58
		question	0.61	0.62	0.62	0.57	0.58	0.58	0.58	0.48	0.48
	Google	keywords	0.79	0.73	0.73	0.59	0.66	0.68	0.71	0.65	0.69
		question	0.71	0.76	0.77	0.64	0.64	0.67	0.65	0.57	0.62
	BioMed	keywords	0.72	0.75	0.74	0.63	0.64	0.66	0.58	0.74	0.71
	Explorer	question	0.74	0.77	0.77	0.60	0.69	0.69	0.72	0.71	0.73
Wikipedia	Wikipedia	keywords	0.81	0.77	0.78	0.68	0.65	0.68	0.58	0.57	0.59
		question	0.65	0.68	0.68	0.57	0.56	0.58	0.48	0.48	0.47
	Google	keywords	0.80	0.79	0.82	0.63	0.72	0.68	0.55	0.57	0.57
		question	0.75	0.75	0.75	0.57	0.67	0.64	0.52	0.53	0.56

The values represent ROC-AUC scores

Conclusion

- Our solution exploits evidence search in PubMed and Wikipedia for open-domain health question answering
- The Google Search over Wikipedia combined with RoBERTa-large model achieves
 AUC of 0.82 on 113 test yes/no questions
- The proposed approach does not use annotated data directly and does not require training on the target data or task
- Our source code and data are publicly available on GitHub¹



¹ https://github.com/apugachev/consumer-health-qa

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- A huge gratitude to the organizing committee for providing the student support



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