Query Automation for Systematic Reviews

Harry Scells Leipzig University https://scells.me

Sheffield NLP Seminar · November 3, 2023

How do clinicians become informed about how to treat their patients?

> How do governments and institutions make health policy decisions?

How do clinicians become informed about how to treat their patients?

Systematic Reviews

How do governments and institutions make health policy decisions?

Systematic Reviews

- Guide clinical decisions
 Inform practice and policy
- Provide evidence

Systematic Reviews

Critically Appraised Individual Articles

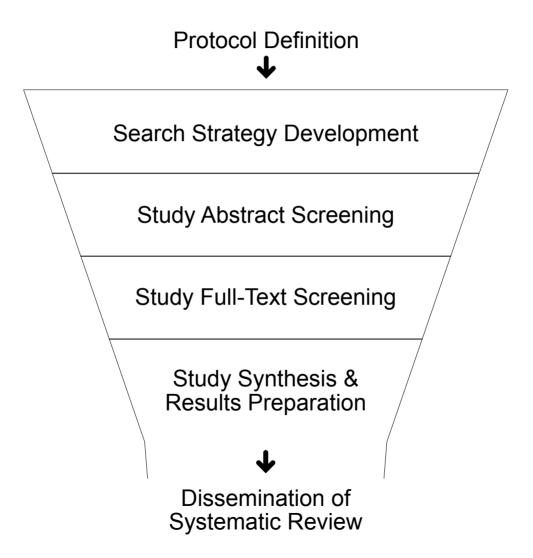
Randomised Controlled Trials (RCTs)

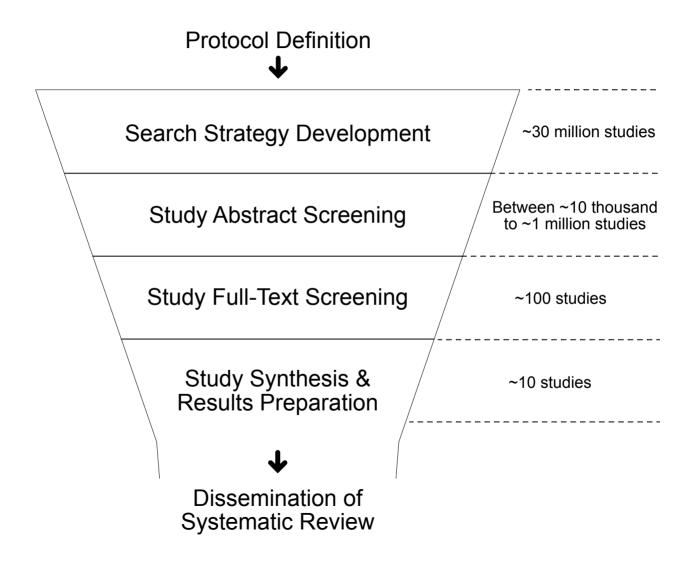
Cohort Studies

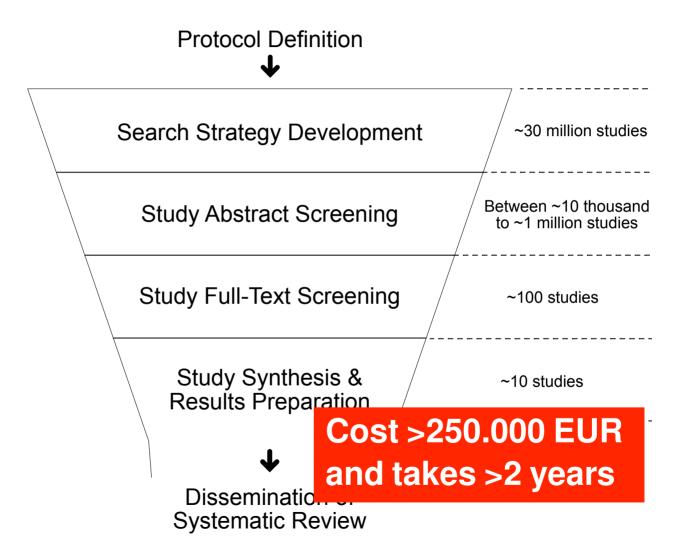
Case-controlled Studies / Case Reports

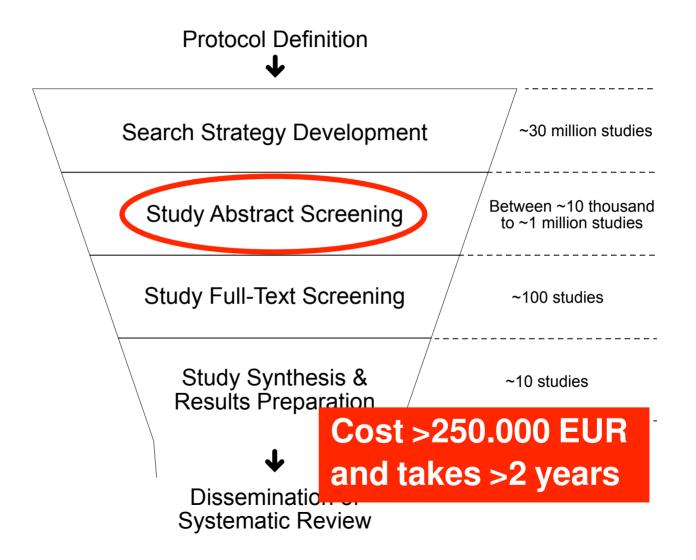
Background Information / Expert Opinion

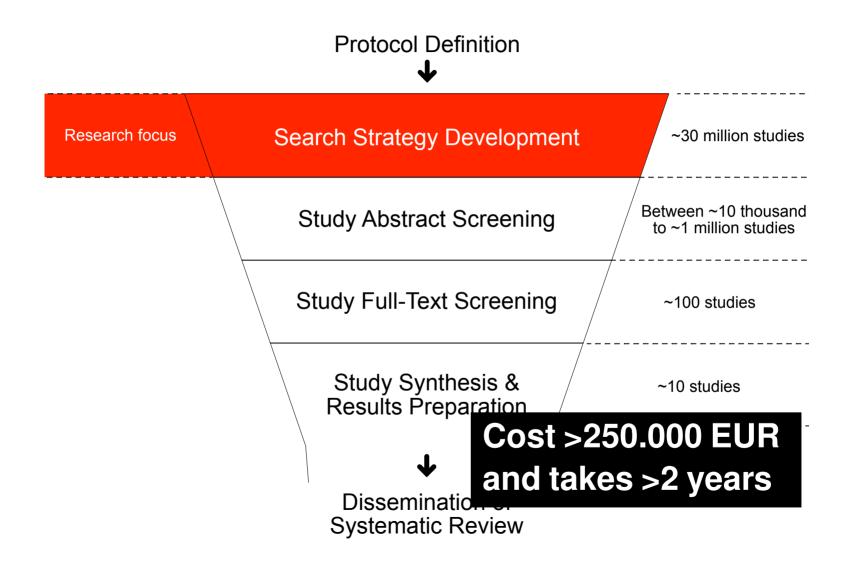
Systematic review creation is hard!





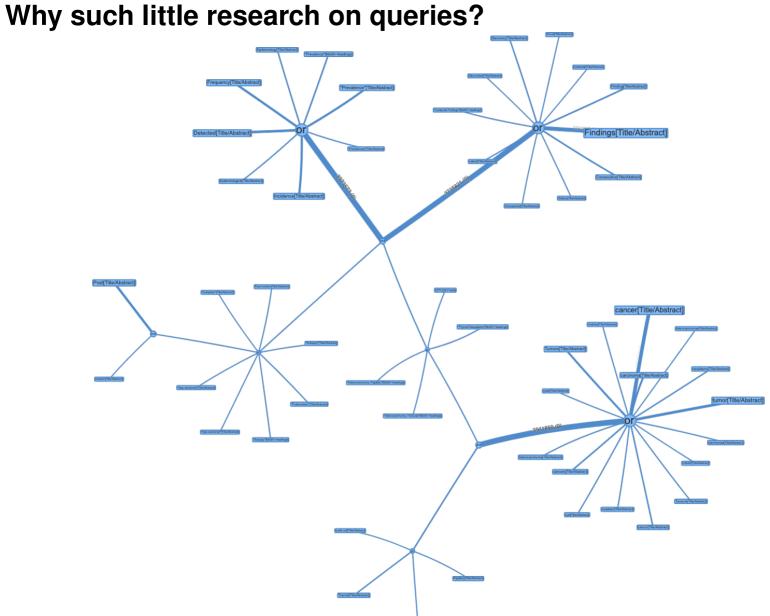






Why such little research on queries?

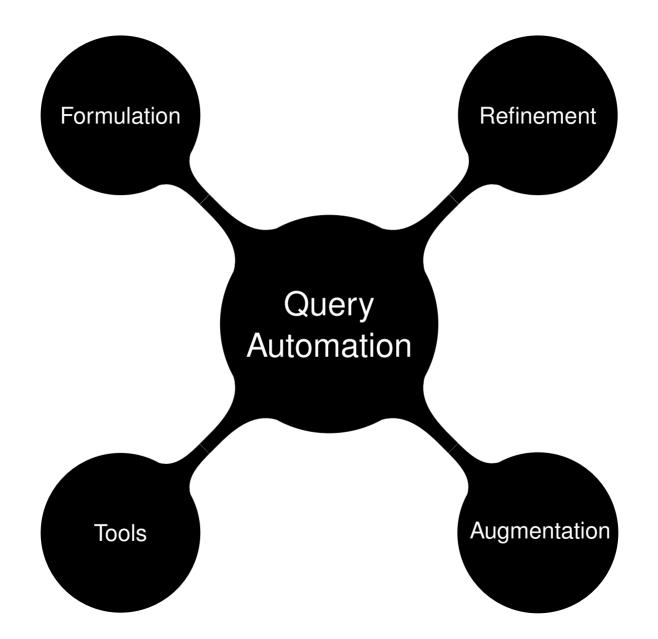
(("Thyroid Neoplasms" [MeSH] OR "Adenocarcinoma, Follicular" [MeSH] OR "Adenocarcinoma, Papillary" [MeSH] OR OPTC OR ((Thyroid[tiab] OR Follicular[tiab] OR Papillary[tiab] OR hurtle cell[tiab]) AND (cancer[tiab] OR cancers[tiab] OR carcinoma[tiab] OR carcinomas[tiab] OR Adenocarcinoma[tiab] OR Adenocarcinomas[tiab] OR neoplasm[tiab] OR neoplasms[tiab] OR nodule[tiab] OR nodules[tiab] OR tumor[tiab] OR tumour[tiab] OR Tumors[tiab] OR Tumours[tiab] OR cyst[tiab] OR cysts[tiab]))) AND ("Autopsy"[MeSH] OR "Autopsy"[tiab] OR "Autopsies" [tiab] OR "Postmortem" [tiab] OR Post-mortem [tiab] OR "step-sectioned"[tiab] OR "step sectioned"[tiab] OR (Post[tiab] AND mortem[tiab])) AND (Prevalence"[MeSH] OR Prevalence"[tiab] OR Prevalences"[tiab] OR Incidence[tiab] OR Epidemiology[tiab] OR Epidemiological[tiab] OR Frequency[tiab] OR Detected[tiab]) AND ("Incidental Findings" [MeSH] OR Incidental [tiab] OR Unsuspected [tiab] OR Discovery[tiab] OR Discoveries[tiab] OR Findings[tiab] OR Finding[tiab] OR Occult[tiab] OR Hidden[tiab] OR Latent[tiab] OR Consecutive[tiab]))

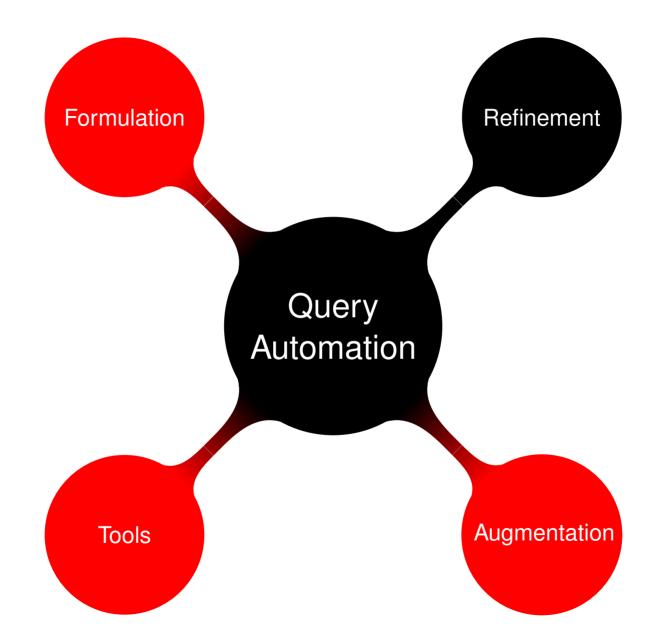


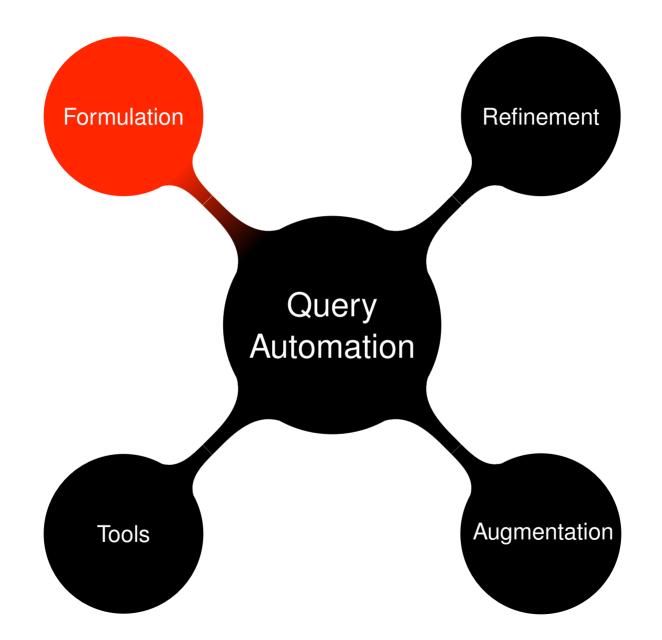
Why are Boolean queries used?

Reproducibility -> double check screening

Understandability -> control set size







Query Formulation

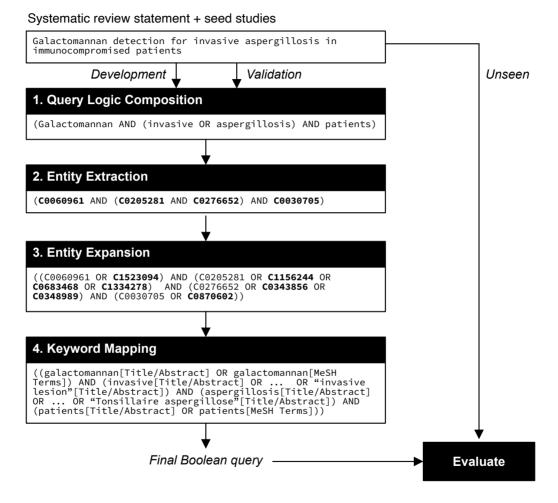
The automatic creation of complex queries for the task of systematic review literature search

Content covered

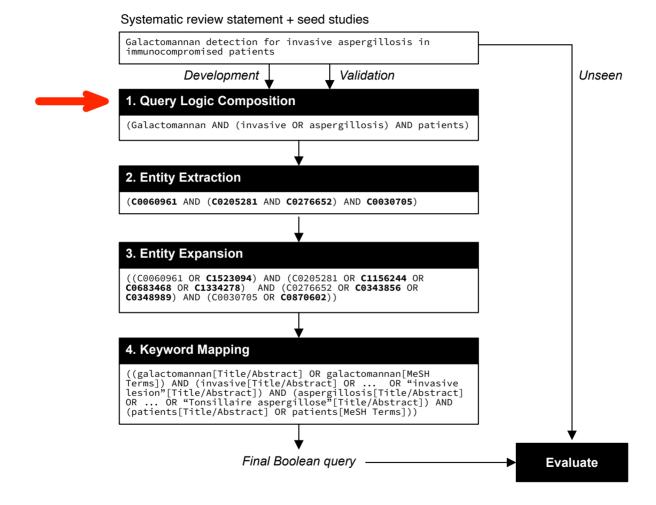
- Shuai Wang, Harrisen Scells, Bevan Koopman, and Guido Zuccon. Can ChatGPT Write a Good Boolean Query for Systematic Review Literature Search? In *Proceedings of the* 46th International ACM SIGIR Conference on Research and Development in Information Retrieval, 2023
- □ Further reading
 - Harrisen Scells, Guido Zuccon, Bevan Koopman, and Justin Clark. Automatic boolean query formulation for systematic review literature search. In *Proceedings of the 29th World Wide Web Conference*, pages 1071–1081, 2020
 - Harrisen Scells, Guido Zuccon, and Bevan Koopman. A computational approach for objectively derived systematic review search strategies. In *Proceedings of the 42nd European Conference on Information Retrieval*, pages 385–398, 2020
 - Harrisen Scells, Guido Zuccon, and Bevan Koopman. A comparison of automatic boolean query formulation for systematic reviews. *Information Retrieval Journal*, pages 1–26, 2020

How humans formulate queries Overview

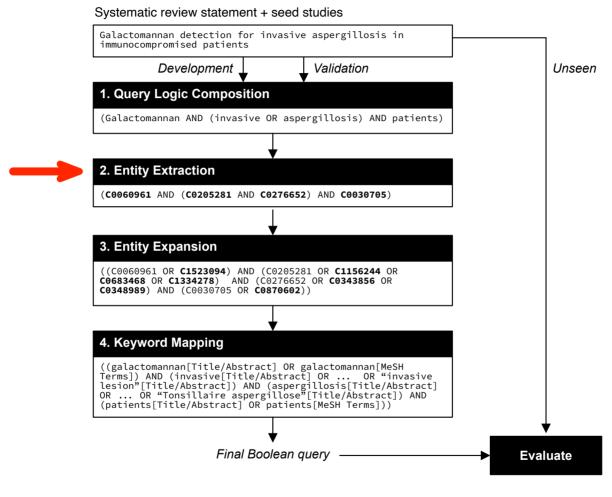
- Conceptual method [Clark 2013] → Human expertise Objective method [Hausner et al. 2012] → More algorithmic
- Both methods → Seed studies



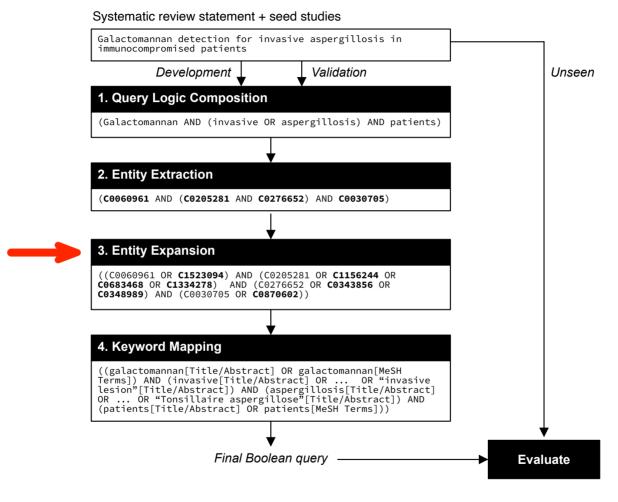
High level concepts → broaden search → iterate until satisfied



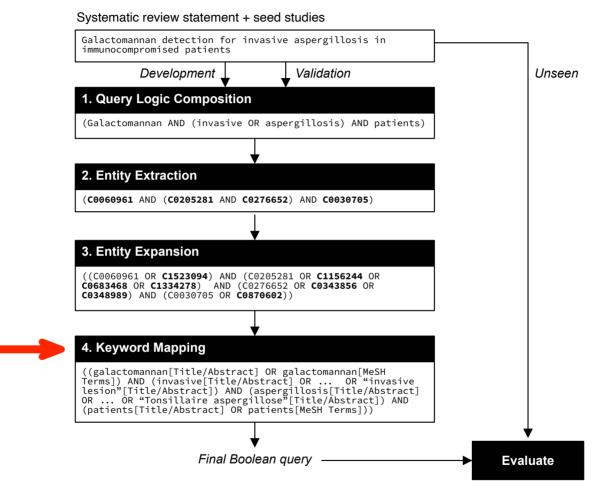
POS tagger → parse grammar & segment words into noun phrases



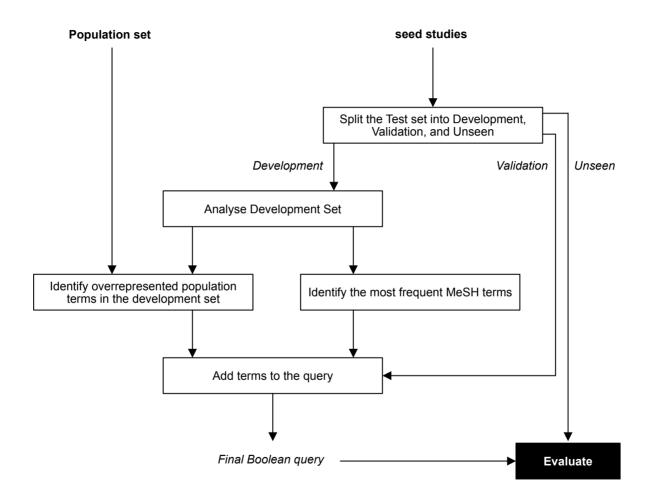
MetaMap → extract CUIs from UMLS ontology



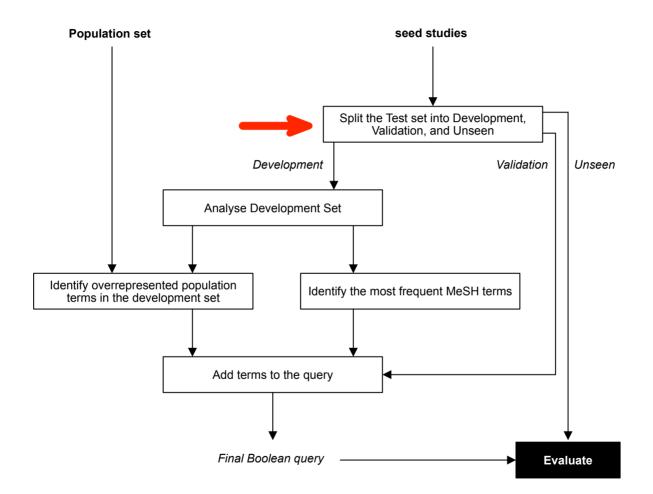
Skipgram model → broaden scope



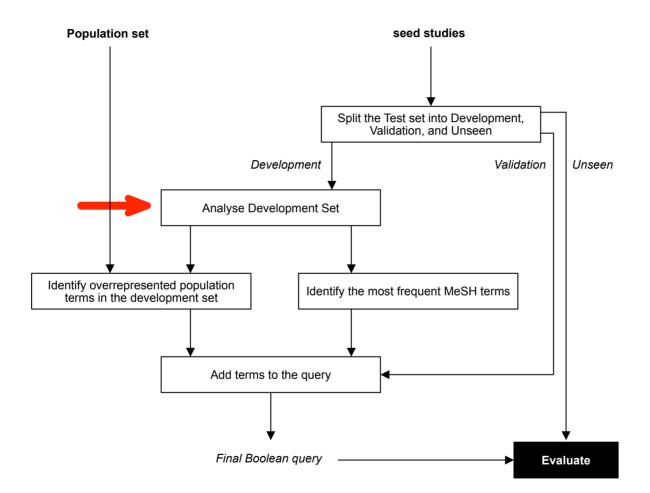
Map concepts (CUIs) to terms



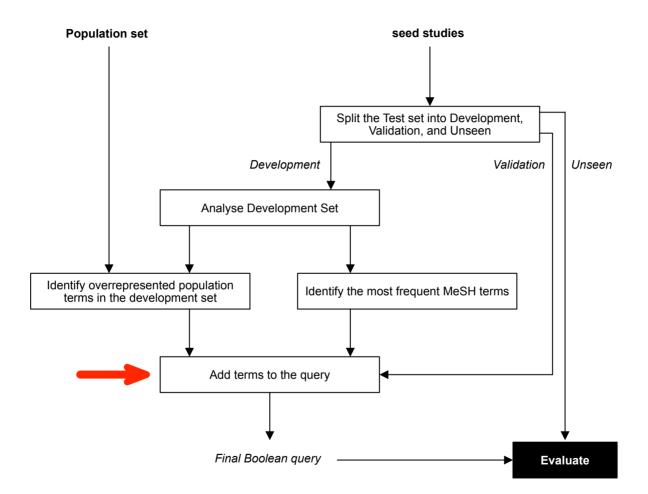
Find prominent terms from docs → Add these terms to query



Extract list of keywords from seed studies



Rank documents using term frequency



Add keywords from documents to query

CLEF TAR [Kanoulas et al. 2017, 2018]

Seed study collection [Wang et al. 2022]

	Precision	Recall	Precision	Recall
Conceptual Objective		0.6996 0.9128	0.0018 0.0057	0.4138 0.5192
num. queries=72			num. queries	=40

Harry Scells 2023

CLEF TAR [Kanoulas et al. 2017, 2018]

Seed study collection [Wang et al. 2022]

	Precision	Recall
Human	0.0207	0.8317
Conceptual	0.0014	0.6996
Objective	0.0002	0.9128

num. queries=72

Precision	Recall
0.0367	0.7366
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PrecisionRecall0.03670.73660.00180.41380.00570.5192

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Objective provides highest recall, but larger trade-offs than conceptual

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Precision Recall

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Objective provides highest recall, but larger trade-offs than conceptual

Both methods are highly sensitive to initial seed studies

Conceptual method better starting point for precision, and objective for recall

Can ChatGPT formulate effective queries?

Hey ChatGPT, formulate a query for me!

Sure, no problem! ;-)

- □ Translation task: prompt → query
- Two methods to formulate queries:
 - Unguided prompt one shot to generate query
 - Guided prompt *follows objective approach*
- Prevalence of Differentiated Thyroid Cancer in Autopsy Studies Over Six Decades: A Meta-Analysis

Unguided prompt examples Simple

For a systematic review titled "{review_title}", can you generate a systematic review Boolean query to find all included studies on PubMed for the review topic?

("Prevalence of Differentiated Thyroid Cancer" AND "Autopsy Studies"AND "Six Decades") AND (systematic review OR meta-analysis)

Unguided prompt examples Detailed

You are an information specialist who develops Boolean queries for systematic reviews. You have extensive experience developing highly effective queries for searching the medical literature. Your specialty is developing queries that retrieve as few irrelevant documents as possible and retrieve all relevant documents for your information need. Now you have your information need to conduct research on "{review_title}". Please construct a highly effective systematic review Boolean query that can best serve your information need.

> ("Differentiated Thyroid Cancer" OR "Papillary Thyroid Cancer" OR "Follicular Thyroid Cancer") AND "Autopsy Studies" AND "Prevalence" AND (("1960/01/01"[PDAT] : "2023/01/30"[PDAT]) OR ("1960"[PDAT]:"2023"[PDAT]))

Unguided prompt examples With Example

You are an information specialist who develops Boolean queries for systematic reviews. You have extensive experience developing highly effective queries for searching the medical literature. Your specialty is developing queries that retrieve as few irrelevant documents as possible and retrieve all relevant documents for your information need. You are able to take an information need such as: "{example_review_title}" and generate valid pubmed queries such as: "{example_review_query}". Now you have the information need to conduct research on "{review_title}", please generate a highly effective systematic review Boolean query for the information need.

("Differentiated Thyroid Cancer"[Mesh] OR (differentiated[All Fields] AND thyroid[All Fields] AND cancer[All Fields]) OR Papillary Thyroid Cancer"[Mesh] OR (papillary[All Fields] AND thyroid[All Fields] AND cancer[All Fields]) OR Follicular Thyroid Cancer"[Mesh] OR (follicular[All Fields] AND thyroid[All Fields] AND cancer[All Fields])) AND ((prevalence[All Fields] OR incidence[All Fields]) AND (autopsy[All Fields] OR post-mortem[All Fields] OR mortuary[All Fields]) AND (study[All Fields] OR studies[All Fields]) AND (six[All Fields] OR 60[All Fields] OR decades[All Fields]))

Previous query formulation results

CLEF TAR [Kanoulas et al. 2017, 2018]

Seed study collection [Wang et al. 2022]

	Precision	Recall
Human	0.0207	0.8317
Conceptual	0.0014	0.6996
Objective	0.0002	0.9128

Precision	Recall
0.0367	0.7366
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Unguided query formulation results

CLEF TAR [Kanoulas et al. 2017, 2018]

	Precision	Recall	Pr
Human	0.0207	0.8317	0.
Conceptual	0.0014	0.6996	0.
Objective	0.0002	0.9128	0.
Simple	0.0543	0.1293	0.
Detailed	0.1166	0.1310	0.
Example	0.0752	0.5035	0.

Precision	Recall		
0.0367	0.7366		
0.0018	0.4138		
0.0057	0.5192		
0.0501	0.0528		
0.0983	0.0394		
0.0283	0.1290		

Unguided query formulation results

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Example	0.0752	0.5035

Recall
0.7366
0.4138
0.5192
0.0528
0.0394
0.1290

Collection contains **seed studies**

→ Use ChatGPT for objective method

Guided prompt examples Step 1

Follow my instructions precisely to develop a highly effective Boolean query for a medical systematic review literature search. Do not explain or elaborate. First, Given the following statement and text from a relevant study, **identify 50 terms or phrases that are relevant**. The terms you identify should be used to retrieve more relevant studies, so be careful that the terms you choose are not too broad. You are not allowed to have duplicates in your list. Statement: {statement}. Text: {text}

1. Differentiated thyroid cancer

- 2. Prevalence
- 3. Autopsy studies
- 4. Meta-analysis
- 5. Occult carcinomas

. . .

50. Parameters.

Guided prompt examples Step 2

... 50. Parameters.

For each item in the list you created in Step 1, **classify it into three categories**: terms relating to health conditions (A), terms relating to a treatment (B), terms relating to types of study design (C). When an item does not fit one of these categories, mark it as (N/A). Each item needs to be categorised into (A), (B), (C), or (N/A).

1. (A) Differentiated thyroid cancer

- 2. (N/A) Prevalence
- 3. (C) Autopsy studies
- 4. (C) Meta-analysis
- 5. (A) Occult carcinomas

50. (N/A) Parameters.

Guided prompt examples Step 3

50. (N/A) Parameters.

. . .

Using the categorised list you created in step 2, create a Boolean query that can be submitted to PubMed which groups together items from each category. For example: ((**itemA1** OR **itemA2** or **itemA3**) AND (itemB1 OR itemB2 OR itemB3) AND (itemC1 OR itemC2 OR itemC3))

(((differentiated[Title/Abstract] OR thyroid[Title/Abstract] OR carcinoma[Title/Abstract] OR papillary[Title/Abstract] OR microcarcinoma[Title/Abstract]) AND (cancer[Title/Abstract] OR incidence[Title/Abstract] OR etiology[Title/Abstract] OR risk[Title/Abstract] OR gender[Title/Abstract] OR hormonal[Title/Abstract] OR nodular[Title/Abstract] OR goiter[Title/Abstract] OR Hashimoto's[Title/Abstract] OR malignancy[Title/Abstract] OR concomitant[Title/Abstract] OR tumor[Title/Abstract] OR infiltrate[Title/Abstract] OR fibrosis[Title/Abstract] OR development[Title/Abstract] OR frequency[Title/Abstract])) AND (autopsy[Title/Abstract] OR surgical[Title/Abstract] OR material[Title/Abstract] OR series[Title/Abstract] OR specimens[Title/Abstract] OR cases[Title/Abstract]))

Guided prompt examples

Step 4

...AND (autopsy[Title/Abstract] OR surgical[Title/Abstract] OR material[Title/Abstract] OR series[Title/Abstract] OR specimens[Title/Abstract] OR cases[Title/Abstract]))

Use your expert knowledge to **refine the query**, making it retrieve as many relevant documents as possible while minimising the total number of documents retrieved. Also add relevant MeSH terms into the query where necessary, e.g., MeSHTerm[MeSH]. Retain the general structure of the query, however, with each main clause of the query corresponding to a PICO element. The final query still needs to be executable on PubMed, so it should be a valid query.

> (((differentiated thyroid cancer[MeSH] OR "differentiated thyroid"[All Fields] OR "thyroid carcinoma"[All Fields] OR "papillary microcarcinoma"[All Fields]) AND (prevalence[All Fields] OR incidence[MeSH] OR "etiology of"[All Fields] OR "risk factors"[All Fields] OR gender[All Fields] OR hormonal[All Fields] OR "nodular goiter"[All Fields] OR "Hashimoto's thyroiditis"[MeSH] OR malignancy[MeSH] OR "concomitant lesion"[All Fields] OR tumor[All Fields] OR infiltrate[All Fields] OR fibrosis[All Fields] OR "early stages of development"[All Fields] OR frequency[All Fields])) AND (autopsy[MeSH] OR surgical[All Fields] OR material[All Fields] OR series[All Fields] OR specimens[All Fields] OR cases[All Fields]))

Unguided query formulation results

	Precision	Recall
Human	0.0367	0.7366
Conceptual	0.0018	0.4138
Objective	0.0057	0.5192
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Guided query formulation results

Seed study collection [Wang et al. 2022]

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ChatGPT is more effective than automatic conceptual and objective methods

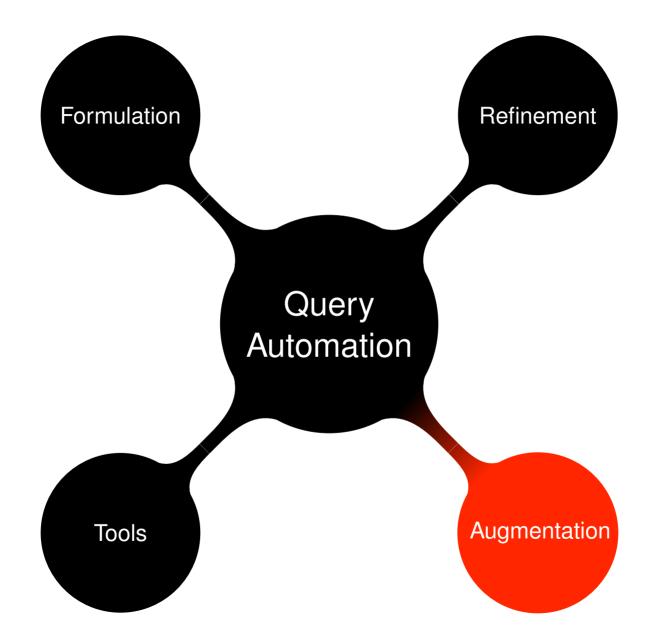
Guided query formulation results

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ChatGPT is more effective than automatic conceptual and objective methods

ChatGPT is highly dependent on prompt and prone to hallucination

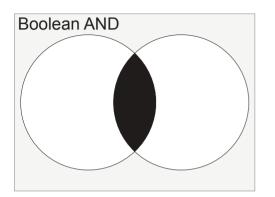


Query Augmentation

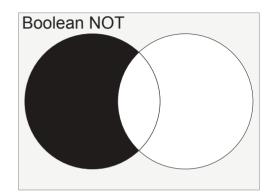
The modification or extension of complex queries in order to improve their effectiveness at the task of systematic review literature search

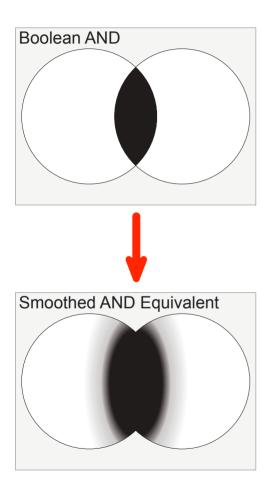
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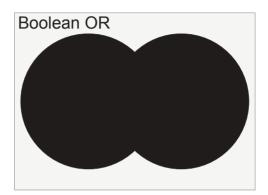
- Harrisen Scells, Ferdinand Schlatt, and Martin Potthast. Smooth Operators for Effective Systematic Review Queries. In *Proceedings of the 46th International ACM SIGIR Conference on Research and Development in Information Retrieval*, 2023
- □ Further reading
 - Harrisen Scells, Guido Zuccon, Bevan Koopman, Anthony Deacon, Leif Azzopardi, and Shlomo Geva. Integrating the framing of clinical questions via PICO into the retrieval of medical literature for systematic reviews. In *Proceedings of the 26th International Conference on Information and Knowledge Management*, pages 2291–2294, 2017
 - Harrisen Scells and Guido Zuccon. You can teach an old dog new tricks: Rank fusion applied to coordination level matching for ranking in systematic reviews. In *Proceedings* of the 42nd European Conference on Information Retrieval, pages 399–414, 2020

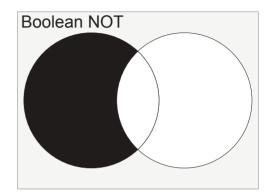


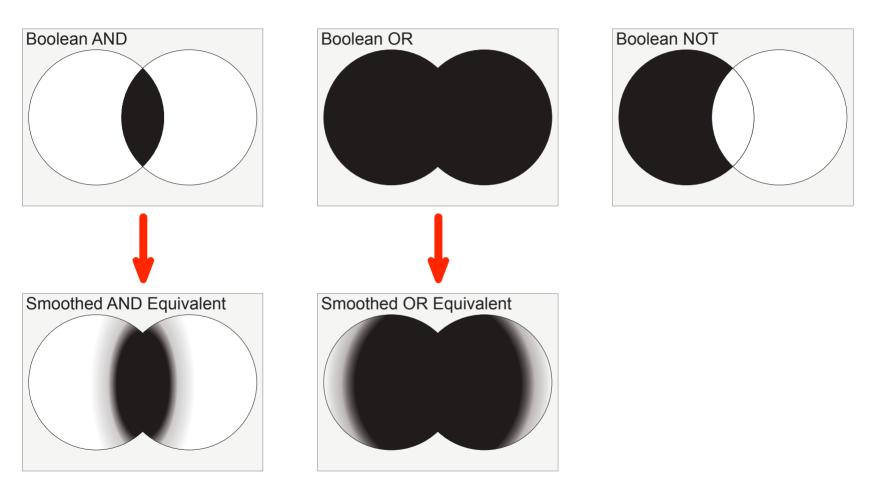


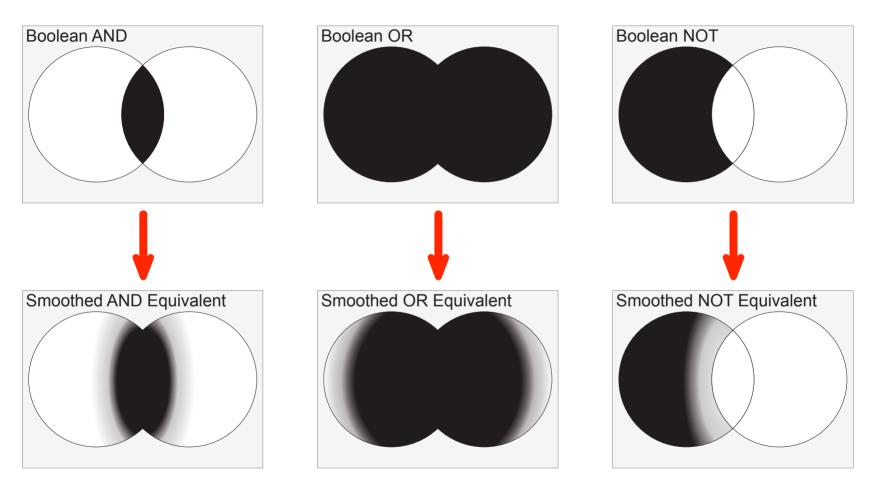


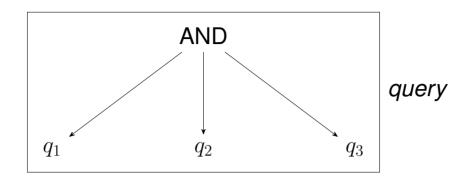


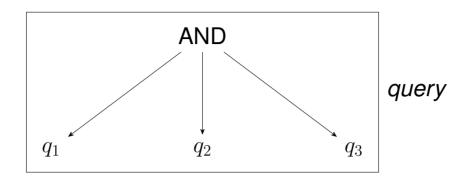


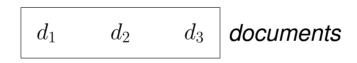


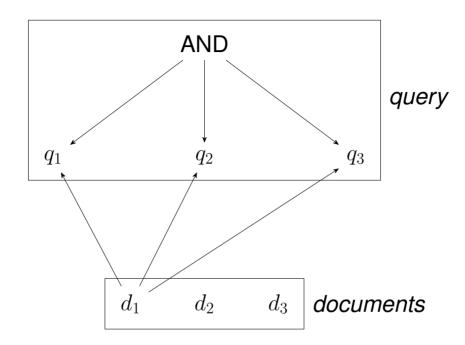


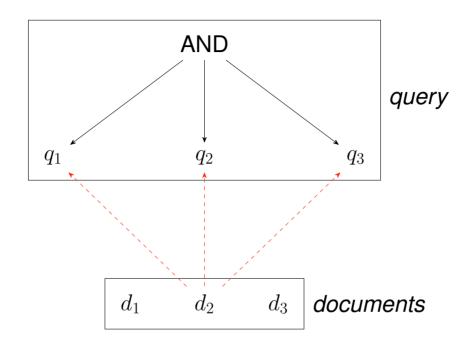


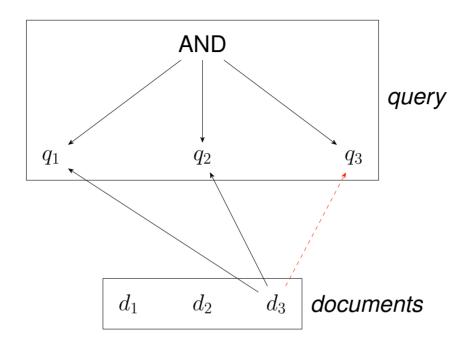












Theory Smoothing result sets

 $\begin{array}{ccc} q & q \\ \uparrow & \uparrow \\ d & d \end{array}$

 $P(d|q) \Rightarrow$ extent to which d should belong to q

$$P(d|q) = \frac{P(d)P(q|d)}{P(q)}$$

Theory Smoothing result sets

 q_1 q_2 q_3

$$P(d|q_1, \dots, q_k) = \frac{P(d) \prod P(q_i|d)}{P(d) \prod P(q_i|d) + P(\overline{d}) \prod P(q_i|\overline{d})}$$

Theory Smoothing result sets

 q_1 q_2 q_3

$$P(d|q_1, \dots, q_k) = \frac{P(d) \prod P(q_i|d)}{P(d) \prod P(q_i|d) + P(\overline{d}) \prod P(q_i|\overline{d})}$$

Leaves two estimations:

□ P(d) → Probability of a document
 □ P(q_i|d) → Probability of a query given a document

Theory

Smoothing result sets

P(d) → ratio of queries that retrieve d $P(d) = \frac{|\{\forall q_i \in q : d \in D_{q_i}\}|}{|q|}$

Theory

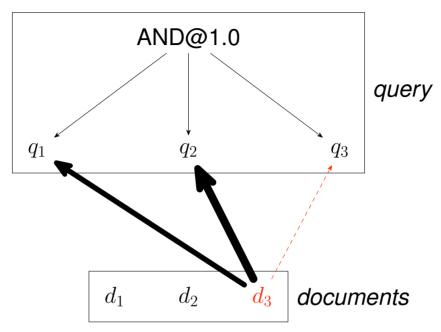
Smoothing result sets

P(d) → ratio of queries that retrieve d $P(d) = \frac{|\{\forall q_i \in q : d \in D_{q_i}\}|}{|q|}$

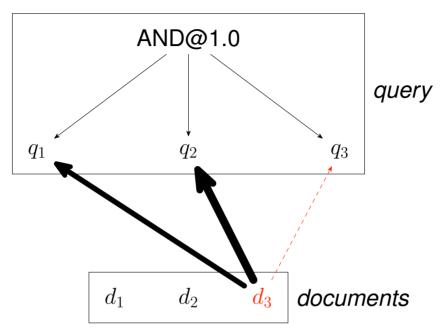
 $P(q_i|d) \rightarrow \text{relevance between } q_i \text{ and } d$

$$P(q_i|d) = 1 - \frac{pos(q_i, d)}{|D_{q_i}|}$$

Theory Implementing smooth operators

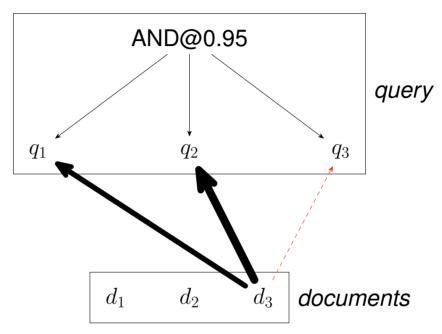


Theory Implementing smooth operators



 $P(d_3|q_1, q_2, q_3) = 0.95$

Theory Implementing smooth operators



 $P(d_3|q_1, q_2, q_3) = 0.95$

	Recall	Precision	nDCG@100
Boolean operators	0.7149	0.0362	-
BM25 Title	0.7149	0.0362	0.0972
smooth Boolean equivalents	0.7149	0.0362	0.2486

Seed study collection [Wang et al. 2022]

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smooth Boolean equivalents	0.7149	0.0362	0.2486

Only using smooth operators for ranking is already considerably better than using BM25

	Recall	Precision	nDCG@100
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BM25 Title	0.7149	0.0362	0.0972
smooth Boolean equivalents	0.7149	0.0362	0.2486
AND@0.99	0.7206	0.0019	0.2331
AND@0.9	0.7658	0.0002	0.2180

Seed study collection [Wang et al. 2022]

	Recall	Precision	nDCG@100
Boolean operators	0.7149	0.0362	-
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smooth Boolean equivalents	0.7149	0.0362	0.2486
AND@0.99	0.7206	0.0019	0.2331
AND@0.9	0.7658	0.0002	0.2180

Smoothing AND operators increases recall at the cost of precision and ranking effectiveness

	Recall	Precision	nDCG@100
Boolean operators	0.7149	0.0362	-
BM25 Title	0.7149	0.0362	0.0972
smooth Boolean equivalents	0.7149	0.0362	0.2486
AND@0.99	0.7206	0.0019	0.2331
AND@0.9	0.7658	0.0002	0.2180
OR@0.01	0.3078	0.0366	0.1604
OR@0.1	0.0612	0.0486	0.0620

Seed study collection [Wang et al. 2022]

	Recall	Precision	nDCG@100
Boolean operators	0.7149	0.0362	-
BM25 Title	0.7149	0.0362	0.0972
smooth Boolean equivalents	0.7149	0.0362	0.2486
AND@0.99	0.7206	0.0019	0.2331
AND@0.9	0.7658	0.0002	0.2180
OR@0.01	0.3078	0.0366	0.1604
OR@0.1	0.0612	0.0486	0.0620

Smoothing OR operators increases precision at the cost of recall and ranking effectiveness

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Predicting the smoothness using features struck middleground between smooth OR and AND

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However: using ground truth information, more effective queries are possible

Seed study collection [Wang et al. 2022]

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Predictor	0.3876	0.0364	0.1651
Oracle	0.7437	0.0440	0.2885
PubmedBERT	0.7148	0.0363	0.2252
BERT	0.7149	0.0363	0.2447
DistilBERT	0.7118	0.0362	0.2449

Seed study collection [Wang et al. 2022]

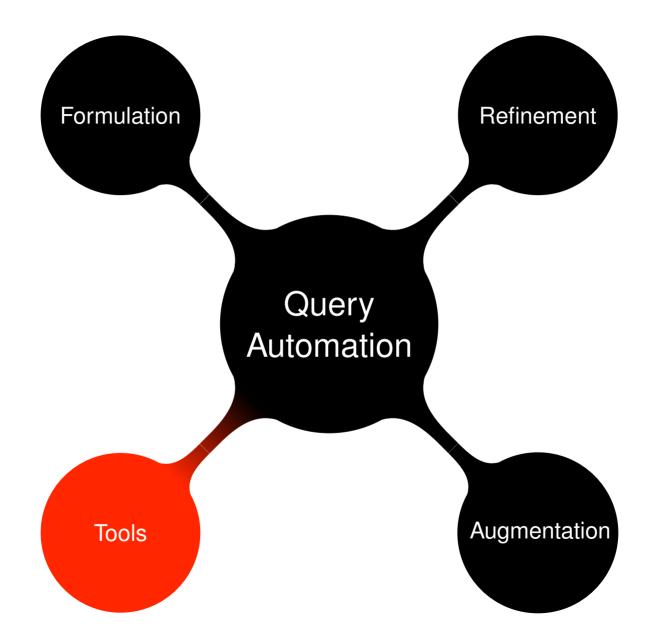
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More advanced neural ranking models did not have any effect on ranking effectiveness

CLEF TAR [Kanoulas et al. 2018]

	Recall	Precision	nDCG@100
Boolean operators	0.8344	0.0204	-
BM25 Title	0.8344	0.0204	0.0232
smooth Boolean equivalents	0.8344	0.0204	0.1995
Predictor	0.6205	0.0206	0.1698
Oracle	0.8487	0.0211	0.2125
BERT	0.8344	0.0204	0.2191
ECNU_RUN1	0.5147	0.0490	0.2440
ECNU_RUN2	0.3831	0.0539	0.1368
ECNU_RUN3	0.5147	0.0490	0.2438
sheffield-bm25	0.4525	0.0095	0.1197
sheffield-boolean	0.3048	0.0061	0.0562
sheffield-tfidf	0.2572	0.0059	0.0154
[Wu et al. 2018, Alharbi et al. 2018]			

Harry Scells 2023



Tools

Software to support systematic review and information retrieval practitioners in undertaking query automation

Content covered

 Harrisen Scells and Martin Potthast. pybool_ir: A Toolkit for Domain-Specific Search Experiments. In Proceedings of the 46th International ACM SIGIR Conference on Research and Development in Information Retrieval, 2023

Further reading

- Harrisen Scells, Daniel Locke, and Guido Zuccon. An information retrieval experiment framework for domain specific applications. In *Proceedings of the 41st Annual International ACM SIGIR Conference on Research and Development in Information Retrieval*, pages 1281–1284, 2018
- Harrisen Scells and Guido Zuccon. Searchrefiner: A query visualisation and understanding tool for systematic reviews. In *Proceedings of the 27th International Conference on Information and Knowledge Management*, pages 1939–1942, 2018
- Hang Li, Harrisen Scells, and Guido Zuccon. Systematic review automation tools for end-to-end query formulation. In *Proceedings of the 43rd Annual International ACM SIGIR Conference on Research and Development in Information Retrieval*, pages 25–30, 2020

- Slow/limited search APIs
- Specific indexing and document processing
- Complex query languages

Domain-specific search has high barrier to entry

- Slow/limited search APIs
- Specific indexing and document processing
- Complex query languages

```
from pybool_ir.experiments.collections import load_collection
from pybool_ir.experiments.retrieval import RetrievalExperiment
from ir_measures import *
import ir_measures
```

```
# Automatically downloads, then loads this collection.
col = load_collection("ielab/sysrev-seed-collection")
```

Evaluate the run using ir_measures. ir_measures.calc_aggregate([SetP, SetR, SetF], col.qrels, run)

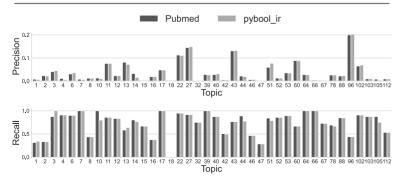
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col = load_collection("ielab/sysrev-seed-collection")
```

```
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```

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Domain-specific search has high barrier to entry

- Slow/limited search APIs
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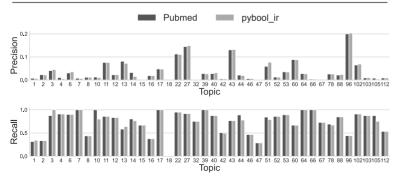
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- Domain-specific indexing
 - ir_datasets
 - Arbitrary JSON
- Extend query syntaxes
 - Smooth operators
 - Faster demo prototyping
- Compatibility with pyserini
 - − index → search

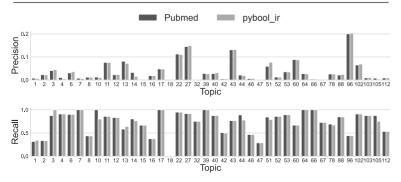
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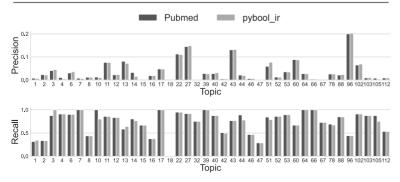
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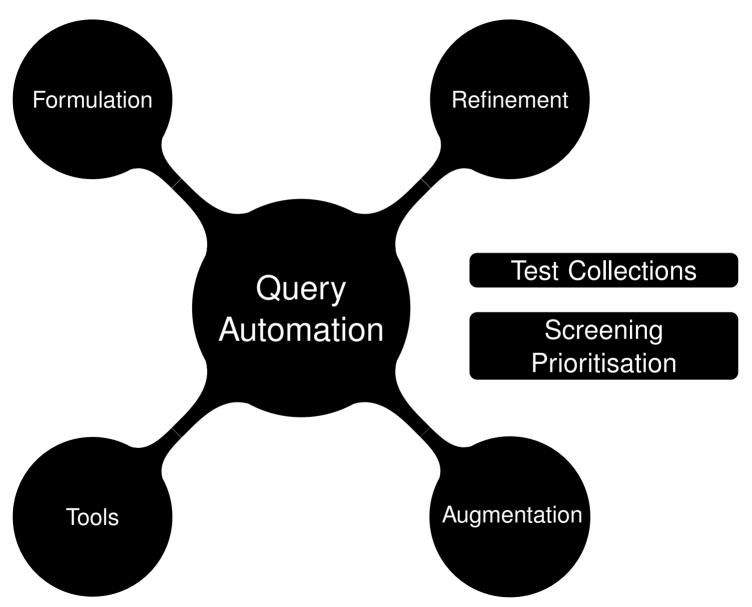
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Outlook

- □ **Money** → Can cost upwards of 250,000 Euros
- \Box **Time** \rightarrow Can take over a year

[McGowan and Sampson, 2005]

Outlook

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[McGowan and Sampson, 2005]

Does hydrocloroquin treat COVID-19?

Should everyone wear a mask during the COVID-19 pandemic?

- Priority for decision making?
- □ Level of uncertainty in the literature?
- □ Frequency of new studies is high?

Conclusion

Next steps

- Evaluating generative IR Is information relevant/correct/readable?
- Making query development easier
 Formulation, refinement, augmentation
- Furthering tools to enable research Reproduction with pybool_ir

ve	Is hydroxychloroquine useful in treating people with CO	VID-19, or in preventing	g infection in people who I	nave been exposed t	o the virus? Searc
Answer HCQ for people infected with COVID-19 has little or no effect on the risk of death and probably no effect on progression to mechanical vesicition. Adverse events adverse events were locut. No turther this of hydroxychioroquine or chioroquine for teatment should be carried using a start of the short high the drug is effective modeling and the short high the drug is effective modeling prevention of infection, and emain trails examining prevention of infection, and emain unmation of the short high standards to provide umatibility of the short high standards to provide umatibility of the short of the short of the drug provide trails examining prevention of infection, and emain umatibility of the short of the original start of the provide umatibility of the short of the short of the short of the short of the umatibility of the short	Answer	Population	Intervention	Control	Outcome
	COVID-19 Exposure to Covid-19 Humans	Hydrochloroquine Chloroquine	Antiviral agent	Time to clinical improvement Adverse event Hospitalisation Death	
	in protecting people from infection, although this is not excluded entirely. It is probably sensible to complete trials examining prevention of infection, and ensure these are carried out to a high standard to provide	Search Results Hydroxychloroqui multicenter random	ine in the treatment of CO ized controlled study	VID-19: a	Flow Diagram
	View statistical analysis Confidence		of hydroxychloroquine a hylaxis for Covid-19.	860 m	
	Confident about results for how many people died and moderately confident about how many needed	Hydroxychloroqui mild-to-moderate C	Hydroxychloroquine with or without azithromycin in mild-to-moderate Covid-19.		ssed excluded
	mechanical ventilation. Moderately confident about the unwanted effects of hydroxychloroquine treatment, but less confident about results for serious unwanted effects; these results might change with further	Hydroxychloroqui mild Covid-19: a ra	ine for early treatment of a andomized-controlled trial.	89963	ot articles sed for billy
evidence. View study characteristics	Efficacy of hydroxy COVID-19: results	of a randomized clinical tri	with ial. 14 st inclus		

Envisioned outcomes

- Faster and less expensive systematic reviews
- Fully automated evidence synthesis
- Tools for librarians and researchers to automate evidence creation

Stay in touch

- □ @hscells
- □ https://scells.me
- □ harry.scells@uni-leipzig.de