







Health Card Retrieval for Consumer Health Search

Jimmy^{1,2}, Guido Zuccon¹, Bevan Koopman³, & **Gianluca Demartini¹**¹University of Queensland (UQ), Australia

²University of Surabaya (UBAYA), Indonesia

³AEHRC / CSIRO , Australia

Presented at:
28th International Conference on Information and Knowledge Management (CIKM)
November 03-07, 2019
Beijing, China



Consumer Health Search (CHS)

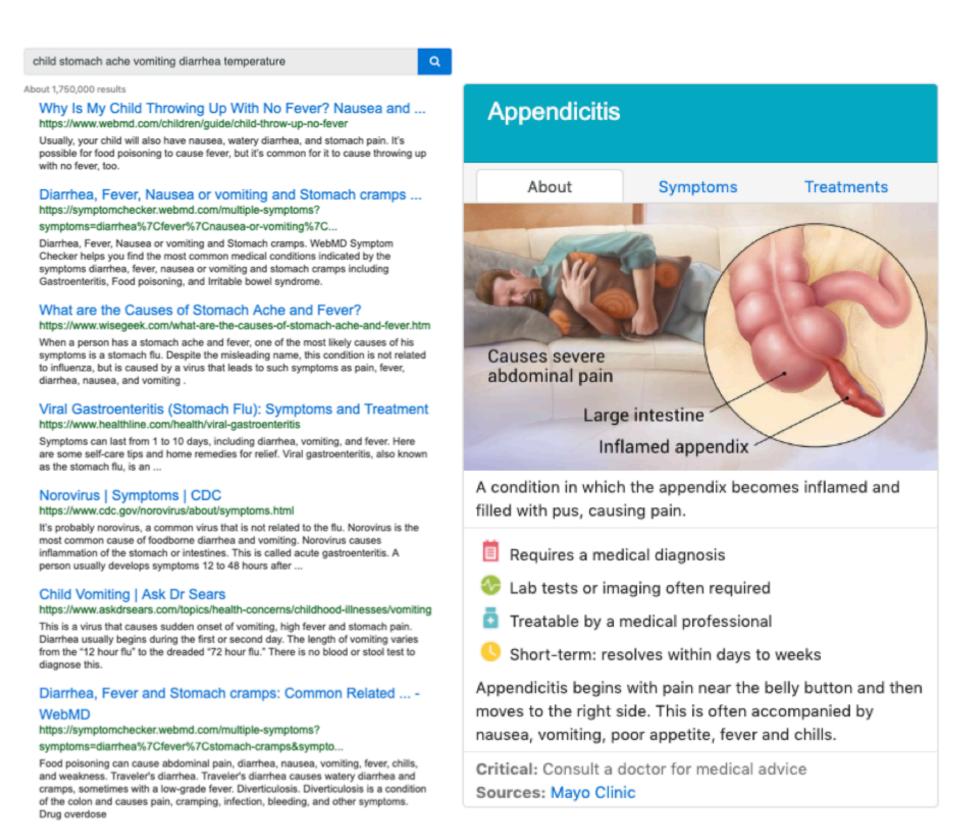
- CHS: people searching for health advice online.
 - 59% of U.S. adults has searched online for health information [Fox & Duggan, 2013].
 - 7% of Google searches are health related ~ 70,000 health queries/minute [Murphy, 2019].
- Search results strongly bias people's health decisions [Pogacar, 2017; White, 2009].
- People struggle to understand health search results and fail to make correct health decisions [Alpay, 2009; Kobayashi & Ishizaki, 2019].
- 59% of self-diagnosers decided **NOT** to confirm their condition with a health professional [Fox & Duggan, 2013].

Benefits of Health Card for CHS

 Entity cards have been used by commercial search engines to present coherent, easy to understand and trustworthy information

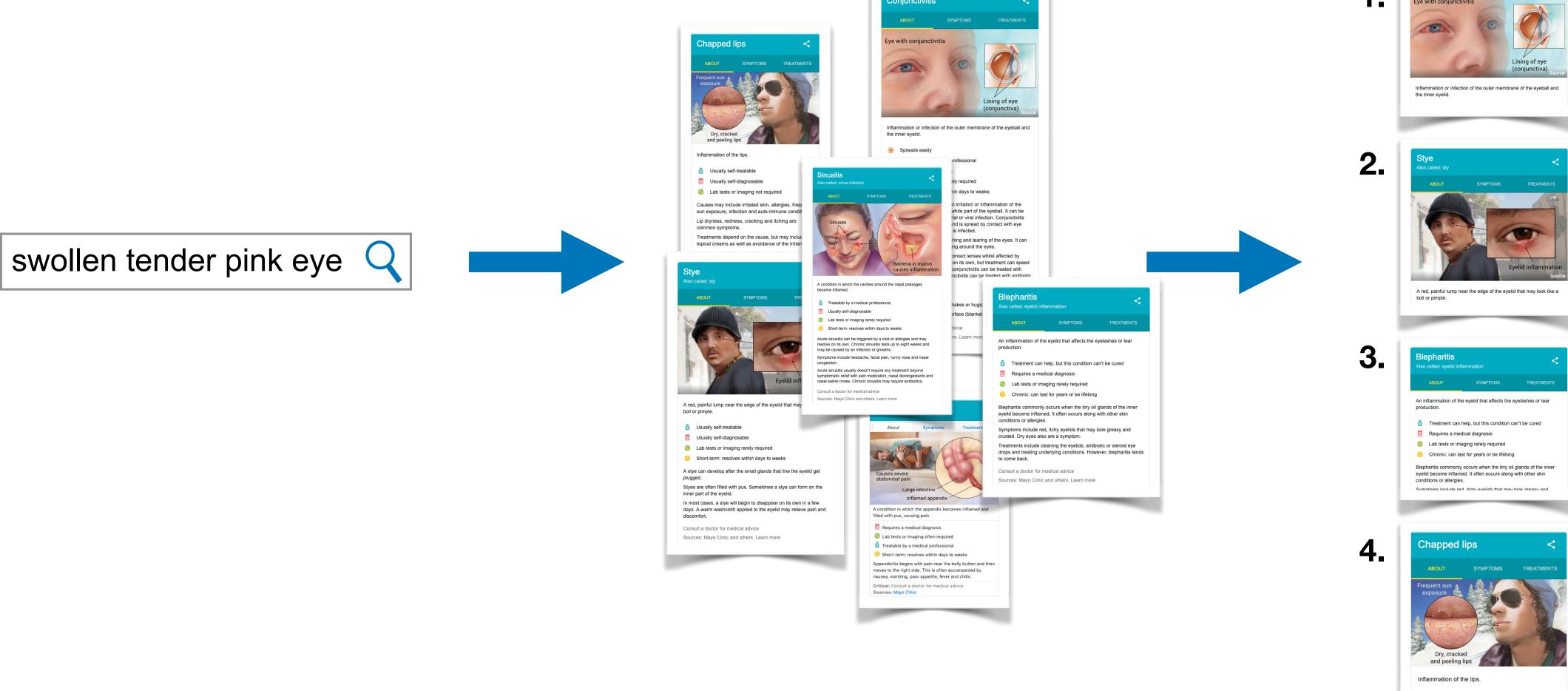
(Radlinski, 2015).

- The appearance of a health card is triggered by queries that contain a health condition name or its aliases.
- Health cards significantly increases the number of correct diagnosis [Jimmy et al., 2019].

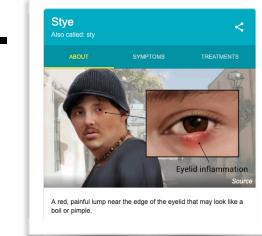


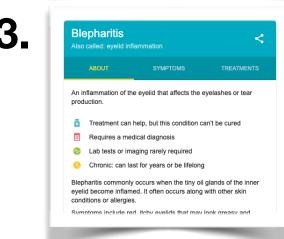
Research Question

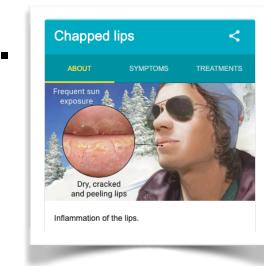
How effective are current retrieval models in ranking health cards based on their relevance to a person's health query?





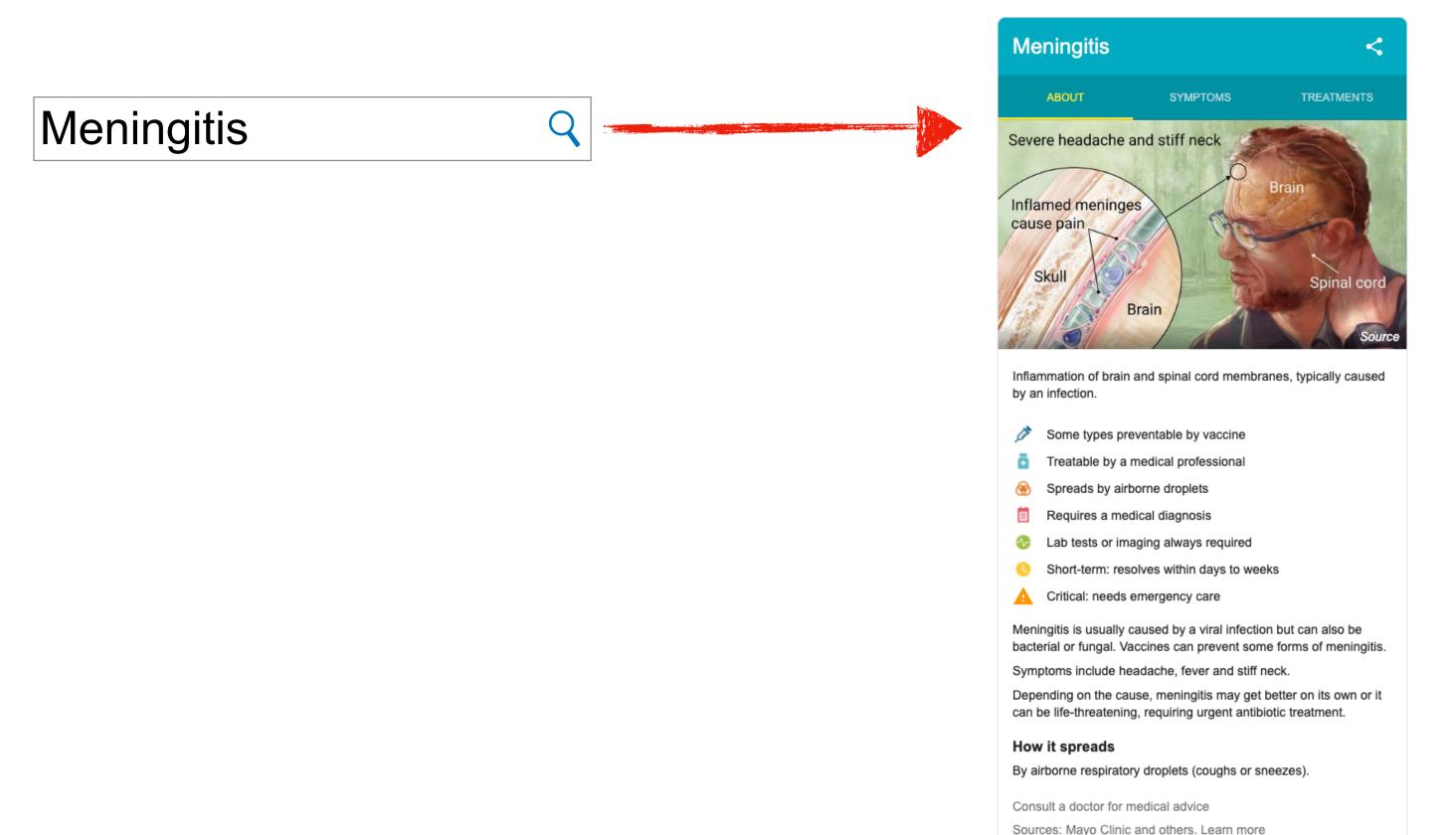






Current Practice in Health Entity Retrieval

Current commercial search engines present health entity cards for 'navigational queries' that target a specific health condition.

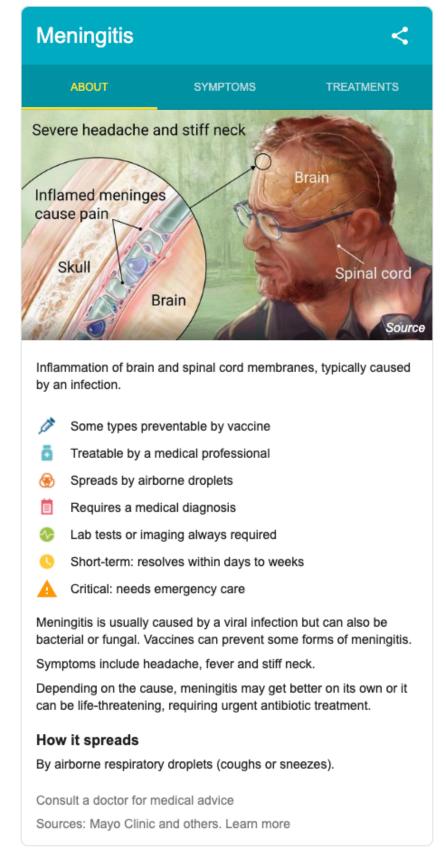


Symptoms Based Retrieval

We are the first to consider methods in retrieving health entities given a symptom based query.

headache fever light sensitivity neck stiffness Q





Evaluated Retrieval Methods

We empirically evaluate four general entity retrieval methods adapted to the problem of ranking health cards:

- BM25F
- Fielded Sequential Dependence Model (FSDM)
- Mixture Language Model (MLM)
- Learning to Rank (LeToR)

Methodology

Health scenarios

We use 41 self-diagnostic health scenarios [Semigran et al., 2015].

Queries

- We collect queries from Amazon Mechanical Turk workers by asking them to search to make self-diagnosis for a health scenario.
- For quality control, we pay \$0.2 (USD) for completing a task with a bonus of \$0.5 for a correct diagnosis.
- Worker selection: released 12 tasks to select good workers to work on the remaining tasks.

Methodology: Health Card Collection

- We crawl pages within the "Diseases and Conditions" sections from Mayo Clinic.
- We collect 1,142 health cards. We found this is comparable to Bing's health cards count: 1,330 health cards.
- Each health card contains fielded information, including: name, aliases, overview, symptoms, and treatments.
- We make our topic, queries and health card collection publicly available for other researchers to use.

Evaluation

- One correct health card for each task.
- Success at rank 1 (S@1): how many times the correct health card is retrieved at rank 1?

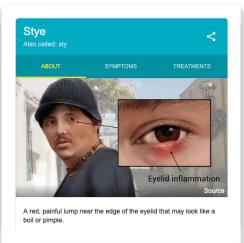


1.



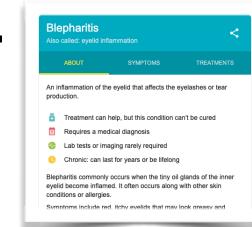


2.



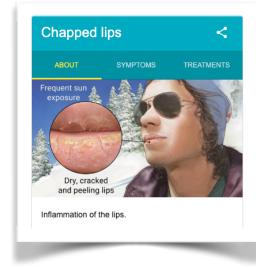


3.



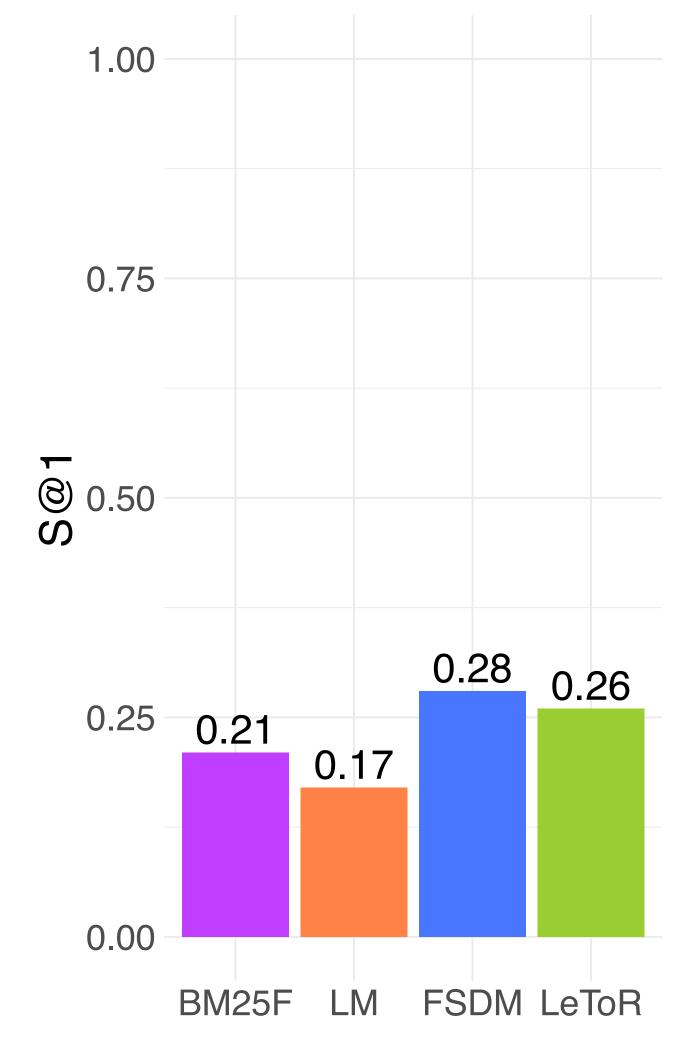


4.



Success @ 1 (n= 584 queries)

- All four methods that we investigated are **not effective** for self-diagnostic, often under-specified queries.
- FSDM (the best) only successfully retrieve the correct health card at rank 1 for 28% of the 584 queries.



Effectiveness Beyond S@1

- Success at rank 4 (S@4): how many times the correct health card is retrieved at rank 4?
- Reciprocal Rank (RR): consider the rank of the correct health card.



1.



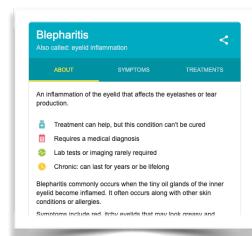




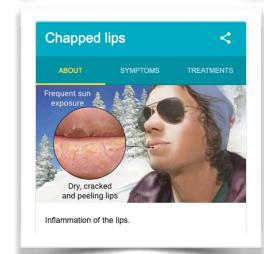




3.

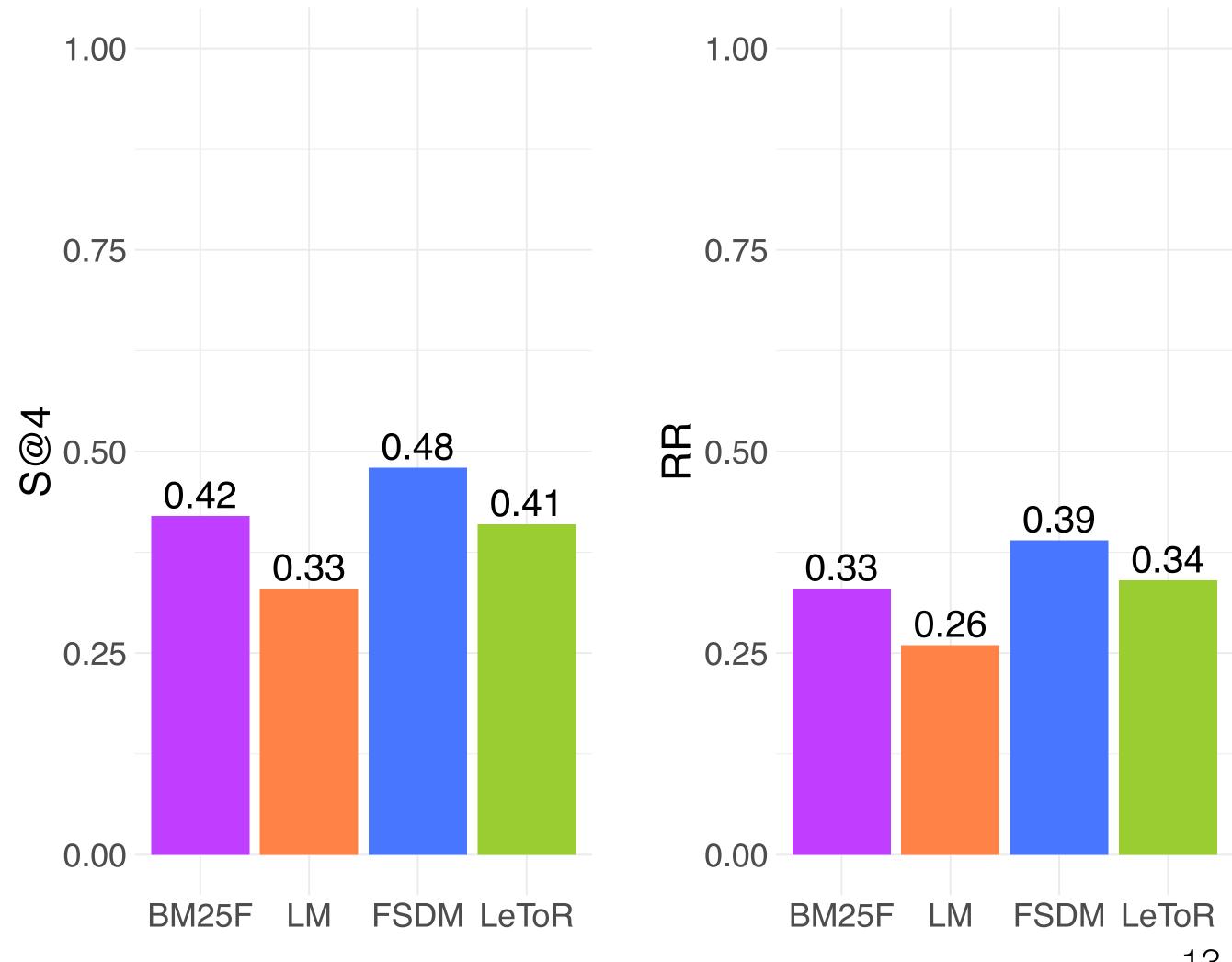






Effectiveness Beyond S@1

- More probability in presenting the correct health card within top 4.
- FSDM (the best) successfully retrieve the correct health card at top 4 for 48% of the 584 queries.
- On average, the correct health cards are found between rank 2 to 3.

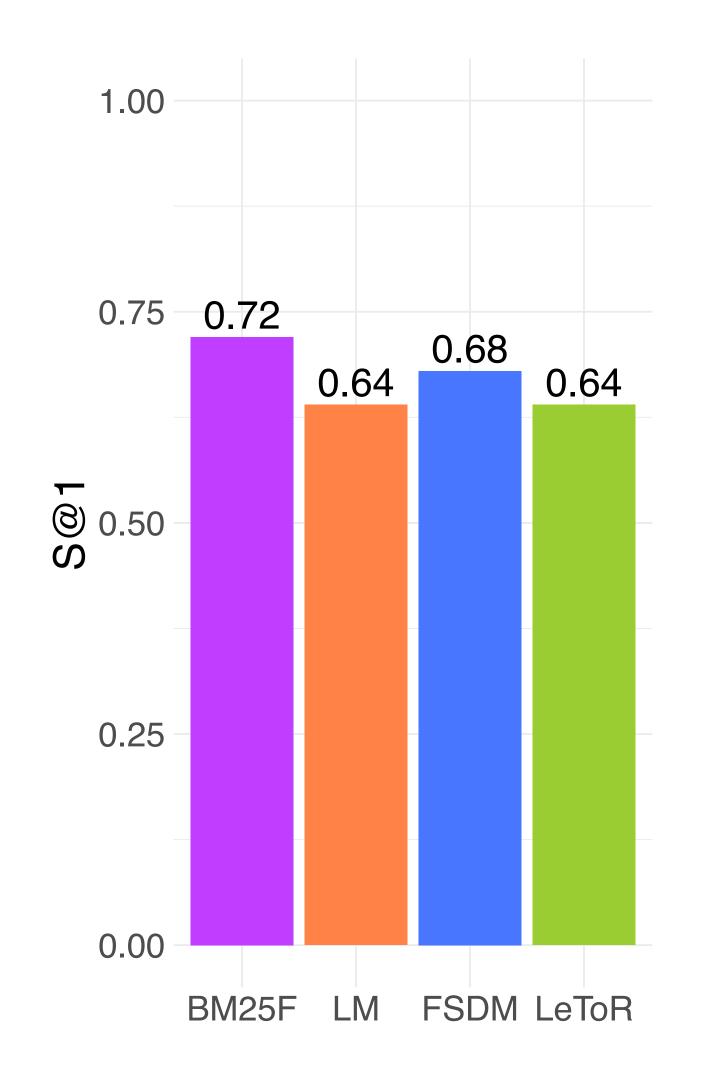


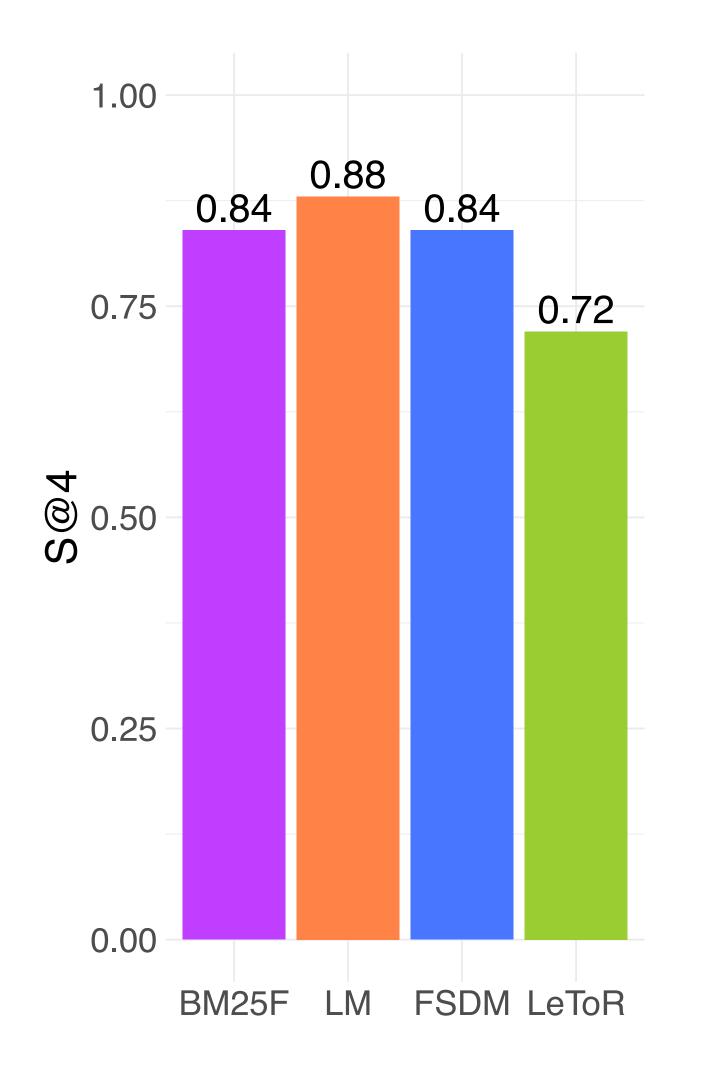
Queries with Navigational Intent

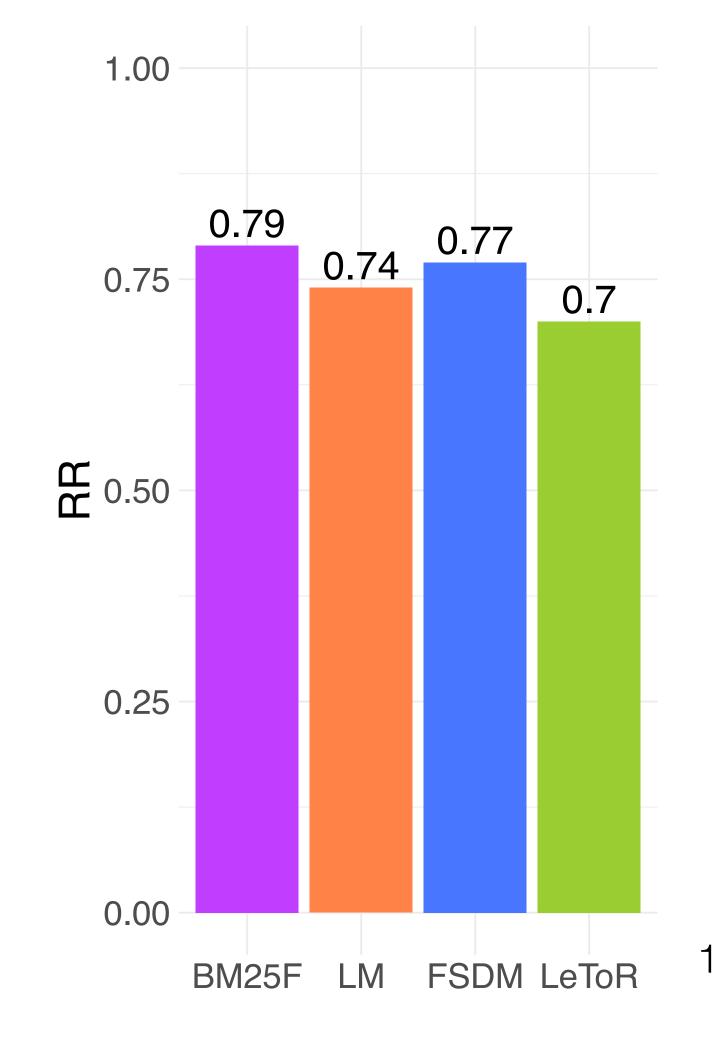
- Navigational Queries target a specific health condition.
- Some navigational queries are a part of the user strategy to identify the correct diagnosis.
- For example, people search for 'migraine' when they attempt to self-diagnose the symptoms of 'meningitis'.

Effectiveness for Navigational Queries (n=25)

The four retrieval models are more effective for the majority of navigational queries.







Contributions

- We investigated methods to retrieve health cards based on observation of symptoms for self-diagnosis.
- We introduced LeToR features specific to health cards that statistically increase the effectiveness LeToR.
- We assembled and released the first test collection of health cards containing information for 1,142 health conditions, and 584 query variations for 41 self-diagnosis search tasks.

Questions?







