RCES: Rapid Cues Exploratory Search Using Taxonomies For COVID-19

Wei Li
wei.li@adaptcentre.ie
Dublin City University

Rishi Choudhary
rishi.choudhary@tudublin.ie
Technological University Dublin

Arjumand Younus
arjumand.younus@ucd.ie
University College Dublin

Bruno Ohana
bruno.ohana@adaptcentre.ie
Trinity College Dublin

Nicole Baker
nicole.baker@adaptcentre.ie
Trinity College Dublin

Brendan Leen
brendan.leen@hse.ie
Health Services Executive, Ireland

M. Atif Qureshi
muhammatatif.qureshi@tudublin.ie
ADAPT Centre, Technological
University Dublin

ABSTRACT

To assist the COVID-19 focused researchers in life science and healthcare in understanding the pandemic, we present an exploratory information retrieval system called RCES. The system employs a previously developed EVE (Explainable Vector-based Embedding) model using DBpedia and an adopted model using MeSH taxonomies to exploit concept relations related to COVID-19. Various expansion methods are also developed, along with explanations and facets that collectively form rapid cues for a valuable navigational and informed user experience.

CCS CONCEPTS

• Information systems → Search interfaces; • Applied computing → Document searching.

KEYWORDS

Exploratory Search, Query Expansion, Facets, Explanations, Rapid Cues, COVID-19

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1 INTRODUCTION

The COVID-19 pandemic threw the scientific community into an emergency response mode leading to a plethora of scientific literature on the topic within a brief time frame [6]. Keeping up with the massive amount of ever-increasing literature generated at an

exponential rate [7] is challenging. There is an imperative demand for tools to facilitate navigation of scientific publications related to COVID-19 [2]. Some search and recommendation tools were developed by various research groups to address this demand [3, 9]. However, to the best of our knowledge, systems allowing deep and insightful exploration into COVID-19 concepts are not available; and neither do they provide associated explanations for the retrieved results.

Pursuing a study of the scientific literature is a complex phenomenon, and besides an effective information retrieval (IR) model, there is an additional aspect of exploratory search that needs to be incorporated [8]. The nature of COVID-19 as a virus where researchers, clinicians, virologists, epidemiologists, and health policymakers know very little make the information need concerning the virus very unclear and non-specific [4]. In such a scenario, an exploratory search that supports navigation through facets [10] helps with the organisation of results in a structure while allowing techniques that expand a query, thereby increasing the recall or the search space. Overall, both navigation and explainable query expansion techniques form the basis of what we call rapid cues in this work.

Essentially, on top of rapid cues we propose a human-centric, exploratory semantic retrieval system (via browser), Rapid Cues Exploratory Search System (RCES). In RCES, the user begins the exploration process by defining a query concerning the information need concepts. The system retrieves a connection between the queried concepts and COVID-19 related topics, supporting visual exploratory search by generating a list of suggested relationships and associations useful for researchers in life sciences and healthcare areas. To achieve this, we take advantage of taxonomies in medical science and previous research on explainable embeddings using our model [5]. The main idea of using these resources is to increase the retrieval effectiveness by employing concept relations within our exploratory search process. These relations are extracted from an existing linked data source such as DBpedia¹ and MeSH². In particular, for a given query, rich association concepts related to COVID-19 are extracted from the initial query by using mapping

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¹<https://www.dbpedia.org/>

²<https://www.ncbi.nlm.nih.gov/mesh/>

functions over taxonomies (explained in more detail in Section 3). Furthermore, facets are extracted using retrieved results to navigate the scientific publications related to the query topic effectively.

Lastly, and most importantly, this is a first of its kind COVID19 literature search project in which an official health and personal social services body of Ireland is directly involved. Their input also informed some of the case-studies and expert assessment efforts around those case-studies.

2 ARCHITECTURE AND METHODOLOGY

This section introduces the architecture of RCES and the methodologies behind its various functionalities. Figure 1 shows that the exploratory search system consists of four components:

2.1 Data Preparation

This component is responsible for crawling various COVID-19 related articles from websites periodically, such as PubMed³, medRxid⁴, Elsevier⁵ and arXiv⁶ to prepare the corpus for the system. Duplication between articles is also removed at this stage, and this is done by means of checking the authors and titles of the articles.

2.2 Indexing

Within this system component, we build the index. This index is updated periodically in a streaming fashion to cope efficiently with the increasing volume of COVID-19 related articles. The dataset gathered is dated from Dec 2019 to date. Moreover, we also index the data from taxonomies at this stage for efficiency purposes, i.e., MeSH and DBpedia. The crawled articles are indexed in Elasticsearch⁷. The use of Elasticsearch helps incorporate custom query expansion methods and different filters within the search interface in an easy-to-use and intuitive manner. Furthermore, it allows a full-text search over a collection of documents, thereby enabling near real-time search.

Finally, post-processing is applied to the crawled articles at index time to enable named entity recognition and other value-added information.

2.3 Query Expansion

This component is a critical part of RCES. It is the engine behind the generation of expansion terms and their associated explanations for exploratory search. It allows to increase the recall when the search space is limited to address the information need. As previously mentioned in Section 1, RCES supports rapid cues for ease of exploration into various aspects of COVID-19 related topics. In the following, we present the explanation of taxonomies and the associated methods we apply on top of those taxonomies for the suggestion of candidate query expansion terms and their associated explanations:

- MeSH (Medical Subject Headings): This refers to the NLM⁸ controlled vocabulary thesaurus used for indexing articles

for PubMed. For query expansion within our system, we perform a matching of the query terms against MeSH taxonomy elements to extract the top five relevant entity terms from MeSH. These MeSH entities are then used in a pseudo-relevance feedback [1] mechanism but with a slight modification that the four most frequent terms within each MeSH entity are used as pseudo-feedback documents. Finally, all frequent terms representing the top entities are aggregated and scored according to frequency while expanding a query. Also, these terms are then presented as explanations.

- MeSH Synonyms: MeSH is essentially a collection of medical subject headings, along with synonyms (called entity terms) within the taxonomy. We use those indicated synonyms against an exact matching entity with the query to retrieve synonyms as expanded terms.
- DBpedia: DBpedia is a large multi-domain ontology derived from Wikipedia. We exploit DBpedia structure through our previously proposed model [5], which essentially produces explainable vector embeddings using mappings between Wikipedia concept (article) and the corresponding Wikipedia category structure. As Figure 2 shows, the query is first mapped to a Wikipedia concept and then an explainable concept vector⁹ is derived through its associativity with Wikipedia categories. Finally, using the vector's associativity scores, top-10 Wikipedia categories that match most strongly to the queried terms are retrieved as expansion terms and explained.
- DBpedia Synonyms: Like MeSH headings and sub-headings, DBpedia includes redirects of Wikipedia articles that we treat as synonyms for the input query. We use the matching of the query against a Wikipedia article to expand terms using Wikipedia redirects.
- Rule-based approach: This approach basically combines custom rules that can be specified by medical practitioners and healthcare researchers (who are users of the system) for their unique information needs. Here, the user uploads custom-defined taxonomy that RCES then uses to expand terms.

2.4 Retrieval Rank and Presentation

This component is responsible for ranking and presenting the documents retrieved in response to a query. Once users input a query, the most relevant documents are retrieved by standard *BM25* retrieval model, where $k1=1.2$ and $b=0.75$, with the search output including the following information (as shown in Figure 3):

- Expansion Terms' Explanation: associated explanations for the expansion terms corresponding to the selected query together with the scores assigned to them via our pseudo-relevance feedback mechanisms.
- Expansion Taxonomy: a choice of taxonomies to choose from amongst those mentioned previously in the Query Expansion component. The selected taxonomy is then used to select expansion terms, and it is possible to select more than one taxonomy at once.

³<https://pubmed.ncbi.nlm.nih.gov/>

⁴<https://www.medrxiv.org/>

⁵<https://www.elsevier.com/>

⁶<https://arxiv.org/>

⁷<https://www.elastic.co/elasticsearch>

⁸National Library of Medicine in United States

⁹Explainable because each vector dimension is associated with a label derived from a Wikipedia category

Figure 1: RCES system Architecture diagram

Figure 2: Mapping of concept (article) to Wikipedia categories.

- **Filters:** These are features of a particular granularity that can be applied over the retrieved set of articles. Since RCES searches over a corpus of medical articles¹⁰ which fall into trials or reviews and can be related to dosage, drugs, duration, form, route, or strength and hence, the filters help in narrowing down the search. Other filters that users can choose are specific expansion methods, i.e. MeSH, MeSH synonyms, DBpedia, DBpedia synonyms and a rule-based approach. Furthermore, users can also filter search results based on the time period in advance of search to reduce retrieval output size by essentially limiting the publications during a certain period.
- **Facets:** these are essentially the extracted topical terms representing information extracted from the retrieved result set to bring a structure to a retrieved set of articles (and this is done by adopting biased PageRank model for extraction of topics from retrieved result set). Also, de-duplication is performed in the facet terminology (such as flu and influenza are treated as one) by exploiting semantic synonyms extracted through available taxonomies in the system (i.e., DBpedia, MeSH, and user-defined). Furthermore, using these facets, the users can re-rank the search results by selecting different facets, and this will then enable them to explore specific aspects of the input query.
- **Articles:** A list of articles retrieved in response to a query. Note that this list is a ranked list presented in the same manner as in traditional search engines. Moreover, when

the facets are used to re-rank the articles in our interface, we highlight those parts of article titles where these facets specifically appear.

3 DEMONSTRATION

A video demonstration of RCES is available at <https://youtu.be/67yCmwWAjN8> and below we present a brief overview of various features.

3.1 Case Study: Unique Aspects of COVID-19

To illustrate the use of RCES, we present a few examples of concept discovery concerning COVID-19. We enter some unique query concepts obtained from the Health Service Executive in Ireland¹¹.

The first of these is depicted in Figure 3, where the query “INFLUENZA VACCINE AND COVID-19” is entered. As per RCES’s support of a rule-based approach, the logical AND operator provides users with additional control over the exploratory search. As shown in Figure 3, the search output includes both the candidate expansion terms’ explanation, facets, and a list of retrieved articles. The expansion terms that led to a larger search space via query expansion component highlight useful relationships between influenza vaccine and COVID-19; particularly, the concepts such as vaccine derived from MeSH and animal virology derived from DBpedia explain the interplay of virus-related concepts. Furthermore, the facets are dynamically extracted from the top 200 retrieved result set. The various synonyms of the facets are combined to reduce any redundancy in facets, thereby depicting the same topic and combined through extracted synonyms from taxonomies. The extracted facets effectively highlight different aspects of the influenza vaccine’s effect concerning COVID-19, and in this case, these terms are “infection”, “disease”, “severity”, and “protein”. As shown in Figure 4, after selecting facet “disease” from the output in Figure 3, the top-ranked articles change accompanied with this selection.

3.2 Expert Assessment

The following list shows some queries and information needs that were defined and used to design the system; we also mention the top two expansion terms and selected facets corresponding to each query. Note that the choice of queries reflects an aspect of COVID-19 necessary for healthcare practitioners to explore in order to have a deeper understanding of the virus, and a senior healthcare

¹⁰In this case medical articles specifically related to COVID-19 form our corpus

¹¹<https://www.hse.ie/>

Available Expansion Methods

Expansion: ☒ MeSH ☐ MeSH Synonyms ☒ DBpedia ☐ DBpedia Synonyms ☒ Rule Based

From: 14/02/2021 To: 23/02/2021

INFLUENZA VACCINE AND COVID-19

Filters: Dosage ☐ Drugs ☐ Duration ☐ Form ☐ Route ☐ Strength ☐

Article Type: ☐ Trials ☐ Reviews ☐

Expansion Terms Explanation

MeSH vaccine: 1.0, influenza: 0.15, hepatitis: 0.15, diphtheria: 0.15, tetanus: 0.15, haemophilus: 0.14, virus: 0.13, use: 0.07, varicella: 0.07, pertussis: 0.07

DBpedia Animal virology: 1.0, Coronaviridae: 0.99, Vaccination: 0.3, Virology: 0.29, Vaccines: 0.28, Henipavirus: 0.25, Nidovirales: 0.25, Animal viral diseases: 0.21, Influenza: 0.17, Influenza vaccines: 0.13

Rule Based (influenza vaccine) AND (covid-19 OR severe acute respiratory syndrome coronavirus 2 OR ncov OR 2019 ncov OR sars-cov-2 OR coronavirus)

Facets: Zoom in Concepts

Showing 200 of 357 results

Covid-19 Cov-2 SARS Vaccination Infection Disease Coronavirus Influenza

Virus Potential Patient Use Analysis Severity Protein View All

Ranked Retrieval Result (highlighting facet)

1 Impact of the Influenza Vaccine on COVID-19 Infection Rates and Severity

Feb. 22, 2021 10.1016/j.jgic.2021.02.012 Elsevier

Background With a unique influenza season occurring in the midst of a pandemic, there is interest in assessing the role of the influenza vaccine in COVID-19 susceptibility and severity. Methods In this retrospective cohort study, patients receiving a laboratory test for COVID-19 were identified. The primary outcome was comparison of positive COVID-19 testing in those who received the influenza vaccine versus those who did not. Secondary end points in patients testing positive for COVID-19 included mortality, need for hospitalization, length of stay, need for intensive care, and mechanical ventilation. Results A total of 27,201 patients received laboratory testing for COVID-19. The odds of testing positive for COVID-19 was reduced in patients who received an influenza vaccine compared to those who did not (odds ratio 0.76, 95% CI 0.68 to 0.86; $P < 0.001$). Vaccinated patients testing positive for COVID-19 were less likely to require hospitalization (odds ratio, 0.58, 95% CI 0.46 to 0.73; $P < 0.001$), or mechanical ventilation (odds ratio, 0.45, 95% CI 0.27 to 0.78; $P = 0.004$) and had a shorter hospital length of stay (risk ratio, 0.76, 95% CI 0.65 to 0.89; $P < 0.001$). Conclusion Influenza vaccination is associated with decreased positive COVID-19 testing and improved clinical outcomes and should be promoted to reduce the burden of COVID-19.

2 Corona- Cov-2 (COVID-19) and ginseng: Comparison of possible use in COVID-19 and influenza

Feb. 18, 2021 10.1016/j.jgic.2020.12.005 Elsevier

In the 1918 influenza pandemic, more than 95% of mortalities were ascribed to bacterial pneumonia. After the primary influenza infection, the innate immune system is attenuated, and the susceptibility to bacteria is increased. Subsequent bacterial pneumonia exacerbates morbidity and increases the mortality rate. Similarly, COVID-19 infection attenuates innate immunity and results in pneumonia. In addition, the current pneumococcal conjugate vaccine may have limited defense against secondary pneumococcal infection after influenza infection. Therefore,

Figure 3: Example of presentation for query "INFLUENZA VACCINE AND COVID-19"

5(default) of 1(facet)

Covid-19 Cov-2 SARS Vaccination Infection Disease Coronavirus Influenza

Virus Potential Patient Use Analysis Severity Protein View All

5 of 1 How Should the World Pay for a Coronavirus Disease (COVID-19) Vaccine?

Feb. 19, 2021 10.1016/j.jval.2020.12.008 Elsevier

The potential health and economic value of a vaccine for coronavirus disease (COVID-19) is self-evident given nearly 2 million deaths, "collateral" loss of life as other conditions go untreated, and massive economic damage. Results from the first licensed products are very encouraging; however, there are important reasons why we will likely need second and third generation vaccines. Dedicated incentives and funding focused explicitly on nurturing and advancing competing second and third generation vaccines are essential. This article proposes a collaborative, market-based financing mechanism for the world to incentivize and pay for the development of, and provide equitable access to, second and third generation COVID-19 vaccines. Specifically,

Figure 4: Example of re-ordering of retrieval results by selecting facet

practitioner served as an expert assessor for classifying results as relevant or irrelevant with and without the:

- SARS-CoV-2 + OPEN WATER SYSTEMS: Expansion terms are water and information from MeSH while systems and animal virology from DBpedia. Facets are Angiotensin, Water and System. Precision@5 with expansion methods is 0.8 as opposed to 0.6 without the methods.
- COVID-19 + TRANSMISSION + HEALTH WORKER TO PATIENT: Expansion terms are transmission and patient from MeSH while workforce and animal virology from DBpedia. Facets are Care, Syndrome and Acute. Precision@5 with expansion methods is 0.6 as opposed to 0.2 without the methods.
- SEQUELAE + COVID-19 + CRITICALLY ILL: Expansion terms are coronavirus and infection from MeSH while medical terminology and animal virology from DBpedia. Facets are Hydroxychloroquine, Medicine and Treatment. Precision@5 with expansion methods is 0.8 as opposed to 0.6 without the methods.

Each of the query, their associated expansion terms and the corresponding facets, as mentioned above, complement the exploration of search results with aspects related to query topic and COVID-19, and hence, allows access to deeper and varied aspects of retrieval.

4 CONCLUSION

In this paper, we present an exploratory semantic retrieval system. The up to date COVID-19 related articles are employed in the search engine to help researchers understand the pandemic better for medicinal needs. Various taxonomic resources and methods defined on top of those resources are applied to explore-exploit the search space. The pseudo-relevance feedback mechanisms over MeSH and the operations performed over embeddings derived from Wikipedia categories help improve the search effectiveness while also providing an enhanced user experience.

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