Graph-based Concept Weighting for Medical Information Retrieval

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Document Representation for IR

Document Representation for IR



Document Representation for IR



Why Medical IR?

- Vocabulary mismatch
 - hypertension \approx high blood pressure
- Interdependence between terms



(Patel et al, 2007)











Bag-of-Concepts Model for Medical IR

Convert Terms to Concepts

"human immunodeficiency virus" "T-lymphotropic virus" "HIV" "AIDS"

Convert Terms to Concepts

"human immunodeficiency virus"

"T-lymphotropic virus" 86406008
"HIV"
"AIDS"

Convert Terms to Concepts



"esophageal reflux"



235595009 Gastroesophageal reflux
 "esophageal reflux"
 196600005 Acid reflux or oesophagitis
 47268002 Reflux
 249496004 Esophageal reflux finding



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Index & retrieval on "bag-of-concepts" (Koopman et al, 2012)

Graph-based Term Weighting

Example Medical Doc

"The patient is a 32-year-old female with a past medical history significant for a prior history of peptic ulcer disease who presents with a complaint of right lower dental pain. The patient states that she was started on recent dental procedures, on a right lower molar, over the past few months, including a recent root canal, at which time she had a temporary filling placed."

Document Term Graph











 $w(t, d) = idf(t) * S(v_t)$

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 $R(d,q) = \sum w(t,d)$ $t \in q$

Graph-based Concept Weighting

 $w(c, d_c) = idf(c) * S(v_c)$ $R(d_c, q_c) = \sum w(c, d_c)$ $c \in q_c$

Concept c $w(c, d_c) = idf(c) * S(v_c)$ $R(d_c, q_c) = \sum w(c, d_c)$ $c \in q_c$

• Document is a graph of SNOMED CT concepts

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- SNOMED CT ontology is also a graph
- Concepts "connectedness" in SNOMED CT indicator of importance in medical domain

 Adjust original concept weight by the "background" importance of concept in medical domain:

 $w(c, d_c) = idf(c) * S(v_c) * \log(|\mathcal{V}_s(c)|)$

 $w(c, d_c) = idf(c) * S(v_c) * \log(|\mathcal{V}_s(c)|)$ document

 $w(c, d_c) = idf(c) * S(v_c) * \log(|\mathcal{V}_s(c)|)$ document corpus

 $w(c, d_c) = idf(c) * S(v_c) * \log(|\mathcal{V}_s(c)|)$ document corpus domain

Empirical Evaluation

Test Collection

- TREC 2011 Medical Records
 Track
 - I00,866 clinical records
 - 34 clinical queries + qrels
- Entire collection converted to SNOMED-CT concepts using MetaMap

Baselines + Models • terms-tfidf

• terms-tfidf

concepts-tfidf

- terms-tfidf
- concepts-tfidf
- terms-graph

- concepts-tfidf
- terms-graph

- terms-tfidf
- concepts-tfidf
- terms-graph

- concepts-graph
- concepts-graphsnomed

• terms-tfidf

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concepts-graph

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Bpref, Precision@10

Retrieval Results

Run	Bpref	Prec@10
terms-tfidf	0.4722	0.4882
concepts-tfidf	0.4993	0.5176
terms-graph	0.4393	0.4882
concepts-graph	$0.5050 \ (+15\%)$	$0.5441 \ (+11\%)$
concepts-graph-snomed	$0.5245 \ (+19\%)$	0.5559 (+14%)

Query reduction.

$w(c, d_c) = idf(c) * S(v_c) * \log(|\mathcal{V}_s(c)|)$

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• 34 queries, 448 query concepts

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- 34 queries, 448 query concepts
- 127 (28%) excluded

Effect of Reduction

% excluded concepts (IV(c)I=1) for each query

 Concept-based representations show improvements over terms representations

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- Integrating formal background knowledge into data-driven approaches to IR

TREC Medtrack'I2

Run	infAP
terms-tfidf	0.1685
concepts-tfidf	0.2027
terms-graph	0.1394
concepts-graph	0.2072
concepts-graph-snomed	0.2123