

Triple-based Background Knowledge Ranking for Document Enrichment

Muyu Zhang, Bing Qin, Ting Liu, and Mao Zheng

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Our Task

To find and rank relevant background knowledge in the form of triple

Input: one source document and a large set of background Knowledge in the form of tripleOutput: *Top N* relevant background knowledge





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EXAMPLE









Source Document

S₁: Coalition may never know if Iraqi
 president Saddam Hussein survived a U.S.
 air strike yesterday.







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The key is: Background Knowledge!

But, these knowledge is available for human **NOT FOR COMPUTERS!**



Our Aim!!!

Background Knowledge:

"Saddam, liveIn, Baghdad"

"Iraqi, hasCapital, Baghdad"

"Saddam, hasChild, Qusay"

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MOTIVATION



Previous researches

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Baghdad

From Wikipedia, t

This article : Baghdad (Arabic: coterminous Baghd largest city in I city in Western A

<u>-City name</u> <u>-History</u> <u>-Main sights</u> <u>-Economy</u> <u>-Culture</u> <u>-Sport</u> <u>-Major streets</u>



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Not relevant



Why triple?

 We use background knowledge in the form of triple: "argument₁, predicate, argument₂"



Less noise and less ambiguity

• So we focus on finding and ranking on these triples





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OUR SOLUTION ...



Questions

- Where knowledge comes from
- How to rank these knowledge



- Existing knowledge bases
 - YAGO (Hoffart et al., 2013)
 - 447,000,000 facts formed as "argument₁, predicate, argument₂" and partly manually edited
 - E.g. "Iraqi, hasCapital, Baghdad"



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 - E.g. "Iraqi, hasCapital, Baghdad"
- Automatically extracted knowledge
 - Reverb (Etzioni et al., 2011)
 - Take raw text as input and automatically extract knowledge formed as "argument₁, predicate, argument₂"
 - E.g. "Saddam, return to live in, Baghdad"



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- YAGO: lexically matched facts
- Automatic extraction: knowledge extracted from relevant documents



Questions

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 Source document consists of multiple information, which can be extracted as triples



sd-node: source document information *bk-node*: background knowledge



 For certain background knowledge in the form of triple (*bk-node*), the relevance to source document is converted into relevance to its *sd-nodes*.





 We present *sd-nodes* and *bk-nodes* together, then propagate relevance score from sd-nodes to bk-nodes



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$$\begin{aligned} WebJaccard(p,q) = \\ \left\{ \begin{array}{ll} 0 & \text{if } H(p \cap q) \leq C \\ \frac{H(p \cap q)}{H(p) + H(q) - H(p \cap q)} & \text{otherwise.} \end{array} \right. \end{aligned}$$

H(*P*) indicates the number of pages returned by search engine, given the query *P*.



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- F3: How important are these *sd-nodes*?

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Combine them together...



- Iterative relevance propagation over the graph
 - Iterative propagation

$$\vec{W'} = \vec{W} \times P$$

= $\vec{W} \times \begin{bmatrix} p(1,1) & p(1,2) & \cdots & p(1,n) \\ p(2,1) & p(2,2) & \cdots & p(2,n) \\ \cdots & \cdots & \cdots & \cdots \\ p(n,1) & p(n,2) & \cdots & p(n,n) \end{bmatrix}$





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for every element

$$w'_{i} = w_{1} \cdot p(1, i) + w_{2} \cdot p(2, i) + \dots + w_{n} \cdot p(n, i)$$
$$= \sum_{k \in N} w_{k} \cdot p(k, i)$$
$$= \sum_{k \in N} w_{k} \cdot \left(\frac{r(i, j) \times \delta(i, j)}{\sum_{k \in N} r(k, j) \times \delta(k, j)}\right)$$



Iterative relevance propagation over the graph

Propagation probability

$$p(i,j) = \frac{r(i,j) \times \delta(i,j)}{\sum_{k \in N} r(k,j) \times \delta(k,j)}$$



where

$$\delta(i,j) = \begin{cases} 1 & \text{if } (i,j) \in E \\ 0 & \text{otherwise} \end{cases}$$



- Iterative relevance propagation over the graph
 - Stop when a global stage is achieved
 - Rank all the background knowledge according to their relevance scores
 - Output the ranked list of background knowledge





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EXPERIMENTS



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 - Baseline: compute relevance between *background knowledge* and source document by accumulating relevance to *sd-nodes*



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- Different Setups
 - The effect of automatic extraction of *source document*





- Different Setups
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CONCLUSION & FUTURE WORK



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- There are always some knowledge gaps in documents
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- There are always some knowledge gaps in documents
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- Our model extracts source documents and background knowledge automatically -- useful in real applications

- To further improve the ranking performance
- Automatic evaluation, instead of manual annotation
- To apply these background knowledge in real tasks





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Thanks

Q&A