LIA at TREC 2011 Web Track

Online resources combination and Wikipedia thematic graphs

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Introduction

- 2 main ideas experimented this year
- Wikipedia contextual expansion
- Online resource combination
- All query expansion related
- Showed to be effective on book search at INEX 2010

Introduction (cont.)

- Building a Wikipedia thematic and query-driven graph
 - Expansion terms
 - Anchor texts
 - Linking topic-related articles together
- Combining information from 2 online resources
 - Google web search
 - Wikipedia

Outline

- Introduction
- Online resources combination
- Wikipedia thematic graphs
- Conclusions

Online resources combination

- Using external sources of information for web search
- Using multiple sources of information [Diazzoo6]
- Query expansion with each resource taken separately...
- ... but how about combining them?

- Two online resources queried on July, 2011:
 - Wikipedia API
 - Google API
- Wikipedia English contained by the ClueWeb
- Google « contains » the ClueWeb (to a certain extent)

- Using the API of each resource to retrieve their first ranked document
- Topic 111: lymphoma in dogs
 - Wikipedia best ranked article:

http://en.wikipedia.org/wiki/Lymphoma_in_animals

Google best ranked page: http://www.caninecancer.com/Lymphoma.html

- Computing an entropy measure for each word w of a Wikipedia article $\mathcal W$:

$$H_{\mathcal{W}}(w) = -\sum_{w \in \mathcal{W}} p_{\mathcal{W}}(w) \cdot \log p_{\mathcal{W}}(w)$$

- Selecting the top 20 words
- Expanding the query with the selected words, using their entropy to weigh them inside the query
- Example of words extracted from the article Lymphoma in Animals:

```
0.14528344114469402
                     lymphoma
0.07659549574716934
                     cats
0.05273200888401322
                     dogs
0.04716238190543428
                     cell
0.041381311223068006 treatment
0.039401174024736065 lymph
0.039401174024736065 chemotherapy
0.03739176127955678
                     veterinary
0.03535111847579512
                     proceedings
0.03535111847579512
                     common
0.03535111847579512
                     disease
0.03535111847579512
                     symptoms
0.03327701121637184
                     loss
0.031166860398936912 nodes
0.031166860398936912 gastrointestinal
0.029017655680289543 prognosis
0.026825837286266783 remission
```

...

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- Each resource counts as a single word in the expansion

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Wikipedia

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- Indri for indexing and searching
- Krovetz stemming and stopwords removal
- No spam filter
 - How does the LM behave on the entire web?
 - No weight optimization, entire non-parametric system
- 2 submitted runs on category A, 1 run on category B

Run	Resources	MAP	P@20	nDCG@20	ERR@20
SDM (unofficial)	-	0.1111	0.1270	0.0963	0.0409
liaQEWikiA GooA (unofficial) liaQEWikiGoA liaQEGoo	Wiki	0.1323**	0.2500***	0.1567**	0.0519**
	Google	0.1438***	0.2140***	0.1868***	0.0825***
	Wiki + Google	0.1566 ***	0.2780***	0.1978***	0.0765***
	Wiki + Google (cat B)	0.1321	0.3260	0.2228 ***	0.0952 ***
NoSpamSDM	-	0.1651 0.1628	0.3370	0.2390	0.1216
NoSpamWikiGoA	Wiki + Google		0.3600	0.2635	0.1230

Table 1: Comparison of the retrieval performance of the four submitted runs and two additional runs. We use two sided paired wise Wilcoxon test (* : p < 0.1; ** : p < 0.05; *** : p < 0.01) to determine significant differences with baseline.

- Each resource improves baseline on its own
- Combining the resources achieves best MAP
- The ClueWebo9 category B is a « purest » sub-collection

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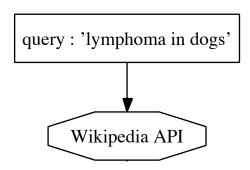
- Several approaches build a complete Wikipedia graph α priori [Milne2008, Coursey2009, Yeh2009]
- No contextual information
- Building a query-oriented Wikipedia graph could highlight more thematic relations between articles

Given a query (topic 111):

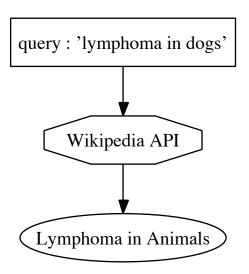
lymphoma in dogs

query: 'lymphoma in dogs'

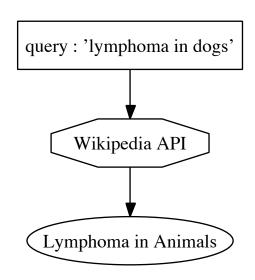
- Given a query (topic 111): lymphoma in dogs
- Query Wikipedia API



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```
lymphoma, dogs, cell, treatment, lymph, chemotherapy, gastrointestinal, prognosis...
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lymphoma, dogs, cell, treatment, lymph, chemotherapy, gastrointestinal, prognosis...
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- Compute entropy of each word and select top 20
- Intersect anchor texts with selected words and select top 2

```
[
  lymphoma, dogs, cell, treatment, lymph,
  chemotherapy, gastrointestinal,
  prognosis...
]
```

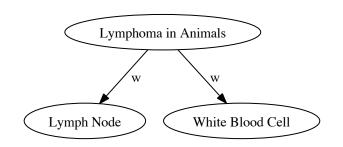
```
[
lymph node,
white blood cell,
gastrointestinal,
remission...
]
Lymph Node
White Blood Cell
```

- Compute entropy of each word and select top 20
- Intersect anchor texts with selected words and select top k
- Instantiate graph with the first article and two linked sub-articles

```
[
    lymphoma, dogs, cell, treatment, lymph,
    chemotherapy, gastrointestinal,
    prognosis...
]
```

```
[
lymph node,
white blood cell,
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```
#weiaht (
   1.0 #weight (0.85 #combine (lymphoma in dogs)
                                                                   Sequential Dependance
                0.1 #combine ( #1(lymphoma in) #1(in dogs) )
                                                                   Model [Metzler2005]
                0.05 #combine ( #uw8(lymphoma in) #uw8(in dogs) ) )
   1.0 #weight ( 0.14528344114469402 lymphoma
                0.07659549574716934 cats
                                                                   Lymphoma in Animals
                0.05273200888401322 dogs ... )
   0.5 #combine
                 #weight ( 0.18625434418721107 lymph
                                                                   Lymph Node
                           0.09701593966960126 nodes ...)
                  #weight ( 0.09773065837466471 prognosis
                           0.05711842409792383 disease ... )
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liaQEWikiA liaQEWikiAnA	Wiki Wiki graph	0.1323 ** 0.1218	0.2500*** 0.2610 ***	0.1567** 0.1630 **	0.0519** 0.0606 **
NoSpamWikiAnA	Wiki graph	0.1259	0.3110	0.2167	0.1008

Table 2: Comparison of the retrieval performance of the Wikipedia thematic graph approach and simple Wikipedia expansion. We use two sided paired wise Wilcoxon test (* : p < 0.1; ** : p < 0.05; *** : p < 0.01) to determine significant differences with baseline.

- Thematic graphs improves early precision
- Also performs better for graded metrics
- Slight loss in average precision

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Conclusions

Experiments with the use of two online resources

- Generation of a thematic graph for expanding the query
- Combining the resources largely improves the results, especially early precision

Ongoing work

- Reproducing results with offline resources
 - System with on-board indexes
 - Using the non-spammed ClueWebo9 as a clean web resource

Run	Resources	MAP	P@20	nDCG@20	ERR@20
SDM	-	0.1111	0.1270	0.0963	0.0409
indexWikiA liaQEWikiA	Wiki index Wiki API	0.1095 0.1323	0.2600 0.2500	0.1737 0.1567	0.0766 0.0519

Table 3: Comparison of query expansion using the Wikipedia API or an index of a Wikipedia dump. Retrieval is performed on the entire ClueWeb09 without spam-filtering.

Thank you for your attention

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