### Text Summarisation for Evidence Based Medicine

### Diego Mollá

Centre for Language Technology, Macquarie University

IIT Patna, 16 December 2012



### Contents

Evidence Based Medicine What is Evidence Based Medicine? FBM and NIP A Corpus for Summarisation

Text Summarisation Sentence Extraction Cohesion Check Balance and Coverage

Proposals for Text Summarisation Single-document Summarisation Optimisation and Summarisation



# About us: Macquarie University





# About us: Centre for Language Technology

http://www.clt.mq.edu.au

# Core Staff (\* involved in the AISRF project)

- ▶ Prof. Robert Dale
- \* Prof. Mark Johnson
- \* A. Prof. Mark Dras
- ► A. Prof. Steve Cassidy
- \* Dr. Diego Molla-Aliod
- ▶ Dr. Rolf Schwitter



# About Us: Research Group on Natural Language Processing of Medical Texts

http://web.science.mq.edu.au/~diego/medicalnlp/

#### **Active Members**

Diego Mollá Senior lecturer at Macquarie University.

Abeed Sarker PhD student at Macquarie University.

Sara Faisal Shash Masters student.

#### Past Members

María Elena Santiago-Martínez Research programmer.

Patrick Davis-Desmond Masters student.

Andreea Tutos Masters student.



# About Me: Diego Mollá-Aliod

# Some Highlights

- ► MSc (1992), PhD (1996) University of Edinburgh.
- ► ExtrAns and WebExtrAns projects at University of Zurich.
- AnswerFinder project and Medical NLP research at Macquarie University.



# Research interests

- ► Question Answering.
- Summarisation.
- ▶ Information Extraction.



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#### Evidence Based Medicine

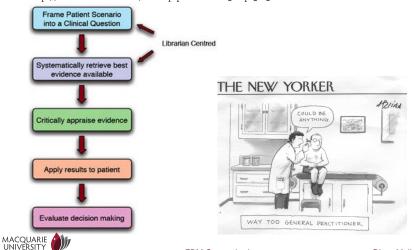


http://laikaspoetnik.wordpress.com/2009/04/04/evidence-based-medicine-the-facebook-of-medicine/



# Suggested Steps in EBM

http://hlwiki.slais.ubc.ca/index.php?title=Five\_steps\_of\_EBM



# PICO for Asking the Right Question

	1	2	3	4
	Patient or Problem	Intervention (a cause, prognostic factor, treatment, etc.)	Comparison Intervention (if necessary)	Outcomes
Tips for Building	Starting with your patient, ask "How would I describe a group of patients similar to mine?" Balance precision with brevity.	Ask "Which main intervention am I considering?"	Ask "What is the main alternative to compare with the intervention?"  Again, be specific	Ask "What can I hope to accom- plish?", or "What could this exposure really affect?" Again, be specific
Example	"In patients with heart failure from dilated cardiomyopathy who are in sinus rhythm"	"would adding anticoagulation with warfarin to standard heart failure therapy"	"when compared with standard therapy alone"	"lead to lower mortality or morbidity from thromboembolism. Is this enough to be worth the increased risk of bleeding?"



#### Where to search for external evidence?

- 1. Evidence-based Summaries (Systematic Reviews):
  - ► The Cochrane Library (http://www.thecochranelibrary.com/).
  - ► EBM Online (http://ebm.bmj.com).
  - UptoDate (http://www.uptodate.com).
  - ▶ ...

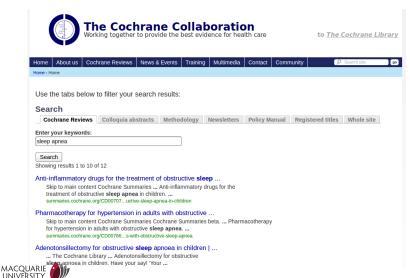


### Where to search for external evidence?

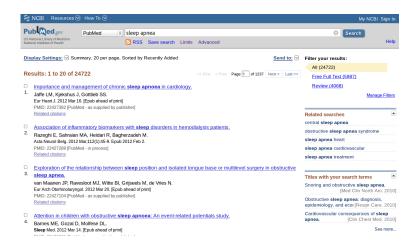
- 1. Evidence-based Summaries (Systematic Reviews):
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  - ► EBM Online (http://ebm.bmj.com).
  - ▶ UptoDate (http://www.uptodate.com).
- 2. Search the Medical Literature:
  - ► E.g. PubMed (http://www.ncbi.nlm.nih.gov/pubmed/).



# Searching Cochrane

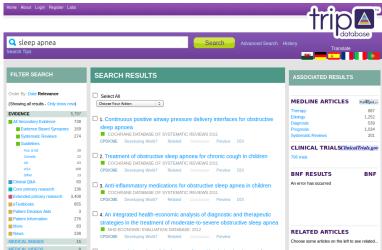


# Searching PubMed





# Searching the Trip Database





# Appraising the Evidence

### The SORT Taxonomy

- Level A Consistent and good-quality patient-oriented evidence.
- Level B Inconsistent or limited-quality patient-oriented evidence.
- Level C Consensus, usual practise, opinion, disease-oriented evidence, or case series for studies of diagnosis, treatment, prevention, or screening.



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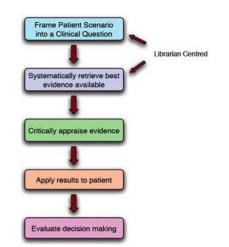
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# Where can NLP Help?

- ► Questions:
  - ► Help formulate answerable questions.
  - From natural question to PICO frames?
  - Question analysis and classification.





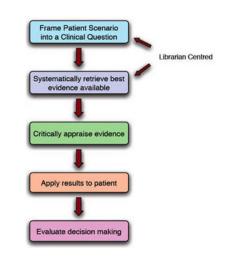
# Where can NLP Help?

#### ► Questions:

- ► Help formulate answerable questions.
- From natural question to PICO frames?
- Question analysis and classification.

#### Search:

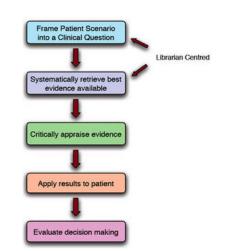
- Retrieve and rank relevant literature.
- Extract the evidence-based information.
- Summarise the results.





# Where can NLP Help? (II)

► Appraisal: Classify the evidence.





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# Where's the Corpus for Summarisation?

### Summarisation Systems

- CENTRIFUSER/PERSIVAL: Developed and tested using user feedback (iterative design).
- ▶ SemRep: Evaluation based on human judgement.
- ▶ Demner-Fushman & Lin: ROUGE on original paper abstracts.
- ► Fiszman: Factoid-based evaluation.



# Where's the Corpus for Summarisation?

### Summarisation Systems

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## Corpora

- Several corpora of questions/answers available.
- ► Answers lack explicit pointers to primary literature.
- Medical doctors want to know the primary sources.

# Journal of Family Practice's "Clinical Inquiries"

#### Which treatments work best for hemorrhoids?

#### Fyidence-based answer

Excision is the most effective treatment for thrombosed external hemorrhoids (strength of recommendation [SOR]: B, retrospective studies). For prolapsed internal hemorrhoids, the best definitive treatment

is traditional hemorrhoidectomy (SOR: A, systematic reviews). Of nonoperative techniques, rubber band ligation produces the lowest rate of recurrence (SOR: A, systematic reviews).

#### Evidence summary

External hemorrhoids originate below the dentate line and become acutely painful with thrombosis. They can lapse (TABLE).

#### For thrombosed external

hemorrhoids, surgery works best Few studies have evaluated the best treatment for thrombosed external Many studies have evaluated the best 53% in the group receiving fiber.8

340 patients who underwent outpa- continence, tient excision of thrombosed external

ported a low recurrence rate of 6.5% at a mean follow-up of 17.3 months.2 A prospective, randomized con-

trolled trial (RCT) of 98 patients treatcause perianal pruritus and excoriation ed nonsurgically found improved pain because of interference with perianal relief with a combination of topical hygiene, Internal hemorrhoids become nifedipine 0.3% and lidocaine 1.5% symptomatic when they bleed or pro- compared with lidocaine alone. The NNT for complete pain relief at 7 days

#### Conventional hemorrhoidectomy beats stanling

hemorrhoids. A retrospective study treatment for prolapsed hemorrhoids. of 231 patients treated conservatively A Cochrane systematic review of 12 When surgical hemorrhoidectomy or surgically found that the 48.5% RCTs that compared conventional is recommended of patients treated surgically had a hemorrhoidectomy with stapled hem. The American Society of Colon and lower recurrence rate than the conservative group (number needed to treat I to III bemorrhoids found a lower fluid and fiber intake for all patients [NNT]=2 for recurrence at mean fol- rate of recurrence (follow-up ranged with symptomatic hemorrhoids, For low-up of 7.6 months) and earlier reso- from 6 to 39 months) in patients who grade I to III hemorrhoids, the society lution of symptoms (average 3.9 days had conventional hemorrhoidectomy states that banding is usually most efcompared with 24 days for conserva- (NNT=14),4 Conventional hemorrhoid- fective. When office treatments fail, the Another retrospective analysis of in decreased bleeding and decreased in-rhoidectomy (SOR: B).

hemorrhoids under local anesthesia re- studies, including some that were of hours old and expectant treatment with

lower quality, showed a higher recurrence rate at 1 year with stapled hemorrhoidectomy than with conventional

#### Nonoperative techniques?

Consider rubber band ligation A systematic review of 3 poorquality trials comparing rubber band ligation with excisional hemorrhoidectomy in patients with grade III hemorrhoids found that excisional hemorrhoidectomy produced better long-term symptom control but more immediate postoperative complications of anal stenosis and hemorrhage,6 Rubber band ligation had techniques of sclerotherapy and infra- grade III and IV hemorrhoids.10 ■ red coagulation.7

#### Fiber supplements help relieve symptoms

A Cochrane systematic review of 7 RCTs enrolling a total of 378 patients with grade I to III hemorrhoids evaluated the effect of fiber supplements on pain, itching, and bleeding. Persistent hemorrhoid symptoms decreased by

ectomy showed a nonsignificant trend society recommends surgical hemor-

The society recommends excision of A second systematic review of 25 thrombosed hemorrhoids less than 72

#### Classification of sympto internal hemorrhoid

GRADE DESCRIPTION Hemorrhoids do not prote Hemorrhoids protrude witl reduce spontaneously Hemorrhoids protrude and by hand Hemorrhoids are permane Source: Madoff RD. et al. Gestroentero/cov. 2004.10

hemorrhoids that present early, Surgical hemorrhoidectomy should be rethe lowest recurrence rate at 12 months served for when conservative treatment compared with the other nonoperative fails and for patients with symptomatic

- 1. Greenspool, Williams SR, Young HA, et al. Thrombased external hemomboids: outcome after conservative or surgical management. Dis Colon Rectum. 2004:47:1493-1498.
- Jonnen J. Bach S, Stubinger SH, et al. Excision of thrombosed external hemorrhoids under local anesthesia: a retrospective evaluation of 340 patients. Dis Colon Rectum. 2003:46:1226-1231
- 3. Perrotti P. Antropoli C. Molino D. et al. Conservative treatment of acute thrombosed external hemortholds with topical nifedipine. Dis Colon Rectum 4. Javaraman S. Colouboun PH. Malthaner RA. Sta-
- pled versus conventional surpery for hemorrhoids Cochrane Databese Syst Rev. 2006:(4):CD005383. 5. Tlandra JJ. Chan MK. Systematic review on the procedure for prolapse and hemorrhoids (stapled hemorholdopexy). Dis Colon Rectum
- Systematic review of randomized trials comparing rubber band ligation with excisional haemorrhoid ectomy, Br J Surg. 2005;92:1481-1487.
- 7. Johanson JF, Rimm A. Optimal nonsurgical treatment of hemorrhoids: a comparative analysis of infrared conquistion, rubber band ligation and injection scientherapy. Am J Gastroenterpl 1992:87:1600-1606.
- Alonso-Coello P. Guyatt G, Heels-Ansdell D, et al Laxatives for the treatment of hemorrhoids. Cochrane Database Syst Rev. 2005(4):CD004649.



tive treatment).1

# The XMI Contents I

```
<record id="7843">
<url>http://www.jfponline.com/Pages.asp?AID=7843&amp;issue=September_2009&amp;UID=</url>
<question>Which treatments work best for hemorrhoids?</question>
<answer>
 <snip id="1">
   <sniptext>Excision is the most effective treatment for thrombosed
external hemorrhoids.</sniptext>
    <sor type="B">retrospective studies</sor>
   < long id = "1_1" >
      <longtext>A retrospective study of 231 patients treated
      conservatively or surgically found that the 48.5% of patients
      treated surgically had a lower recurrence rate than the
      conservative group (number needed to treat [NNT]=2 for
      recurrence at mean follow-up of 7.6 months) and earlier
      resolution of symptoms (average 3.9 days compared with 24 days
      for conservative treatment). 
      <ref id="15486746" abstract="Abstracts/15486746.xml">Greenspon
      J. Williams SB. Young HA .et al. Thrombosed external
      hemorrhoids: outcome after conservative or surgical
      management. Dis Colon Rectum. 2004; 47: 1493-1498.</ref>
    </long>
    long id="1_2">
      <longtext>A retrospective analysis of 340 patients who underwent
      outpatient excision of thrombosed external hemorrhoids under
      local anesthesia reported a low recurrence rate of 6.5% at a
```



#### The XML Contents II

```
mean follow-up of 17.3 months.
     <ref id = "12972967" abstract="Abstracts/12972967.xml">Jongen J,
      Bach S, Stubinger SH, et al. Excision of thrombosed external
      hemorrhoids under local anesthesia: a retrospective evaluation
      of 340 patients. Dis Colon Rectum. 2003: 46: 1226-1231.
    </long>
   <long id="1_3">
     <longtext>A prospective . randomized controlled trial (RCT) of 98
      patients treated nonsurgically found improved pain relief with a
      combination of topical nifedipine 0.3% and lidocaine 1.5% compared
      with lidocaine alone. The NNT for complete pain relief at 7 days was
      3. < / longtext >
     <ref id="11289288" abstract="Abstracts/11289288.xml">Perrotti P,
      Antropoli C, Molino D, et al. Conservative treatment of acute
      thrombosed external hemorrhoids with topical nifedipine. Dis
      Colon Rectum, 2001: 44: 405-409.</ref>
   </long>
 </snip>
</answer>
</record>
```



# Components of the Corpus

Question Direct extract from the source.

Answer Split from the source and manually checked.

Evidence Extracted from the source.

Additional text Manually extracted from the source and massaged.

References PMID looked up in PubMed (automatic and manual procedure).



# Corpus Statistics

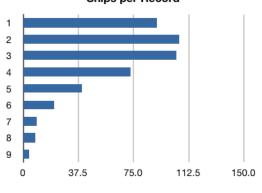
## Size

- ▶ 456 questions ("records").
- ▶ 1,396 answer parts ("snips").
- ▶ 3,036 answer justifications ("longs").
- ▶ 3,705 references:
  - 2,908 unique references.
  - ▶ 2,657 XML abstracts from PubMed.



# Answer parts per Question



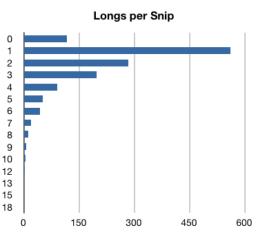


Avg=3.06





# Answer justifications per answer part

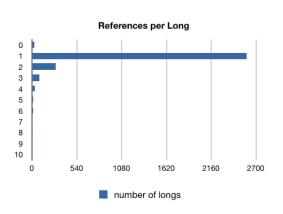


Avg=2.17





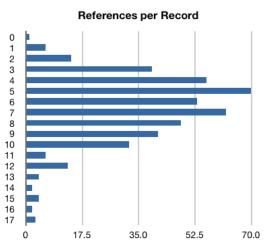
# References per answer justification



Avg=1.22



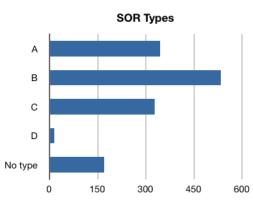
# References per question

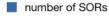


Avg=6.57



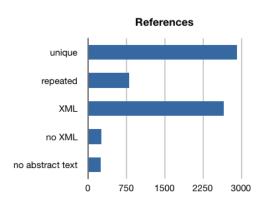
# **Evidence Grade**







# References



number of references



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#### What is Document Summarisation

# Summarisation (or automatic abstracting)

A summary is a text that is produced from one or more texts, that contains a significant portion of the information of the original text(s), and that is no longer than half of the original text(s). (Hovy, 2003)



# What for?

► For busy people to read the summary instead of the full text.



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# What is Document Summarisation Good For?

- ► For busy people to read the summary instead of the full text.
  - → informative summary



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# What is Document Summarisation Good For?

- ▶ For busy people to read the summary instead of the full text.
  - → informative summary
- ► For researchers, web surfers, ... to read the summary to decide if it is worth to read the original text.



- ▶ For busy people to read the summary instead of the full text.
  - → informative summary
- For researchers, web surfers, . . . to read the summary to decide if it is worth to read the original text. → indicative summary



- ► For busy people to read the summary instead of the full text.
  - → informative summary
- ► For researchers, web surfers, . . . to read the summary to decide if it is worth to read the original text.  $\rightarrow$  indicative summary
- ▶ To avoid having to type out an abstract for a technical report when the publisher requests it.



- ▶ For busy people to read the summary instead of the full text.
  - $\rightarrow$  informative summary
- ► For researchers, web surfers, ... to read the summary to decide if it is worth to read the original text. → indicative summary
- ► To avoid having to type out an abstract for a technical report when the publisher requests it. → not realistic



# Examples I

#### Original Text

Abraham Lincoln was born in Kentucky on February 12th, 1809. His family moved to Indiana when he was eight years old. His mother died when he was ten. Abraham Lincoln loved to read. He only went to school for a few months. He had to study and learn on his own. From his childhood, Abraham Lincoln was known as a hard worker. He worked on a farm and in a store and on a boat. He studied law and practiced for many years in Illinois. He married Mary Todd Lincoln. Together they had four sons. Abraham Lincoln was elected as the 16th President of the United States in 1860. He did many things as President. Many people think he was the best American President of all time. He is most remembered for freeing the slaves. He was President of the United States during the time the Civil War was fought. The Civil War was fought between the Northern and Southern states. He was known as a great speaker. Some of his most famous speeches include the Emancipation Proclamation - in which he freed the slaves - and the Gettysburg Address, which starts "Four score and seven years ago ..." While attending a play on April 14th, 1865, Abraham Lincoln was shot by John Wilkes Booth. Lincoln died the next morning. Lincoln's birthday is celebrated in February of each year.



# Examples II

#### Word's Autosummarize

Abraham Lincoln was born in Kentucky on February 12th, 1809.

Abraham Lincoln loved to read.

Abraham Lincoln was elected as the 16th President of the United States in 1860.

The Civil War was fought between the Northern and Southern states.

While attending a play on April 14th, 1865, Abraham Lincoln was shot by John Wilkes Booth.

Lincoln died the next morning.

#### **MEAD**

Abraham Lincoln was born in Kentucky on February 12th 1809.

His family moved to Indiana when he was eight years old.

From his childhood Abraham Lincoln was known as a hard worker.

Abraham Lincoln was elected as the 16th President of the United States in 1860.

While attending a play on April 14th 1865 Abraham Lincoln was shot by John Wilkes Booth.



# An Ideal Document Summarisation System

# **Understanding Stage**

 $Document(s) \rightarrow Knowledge base$ 



# An Ideal Document Summarisation System

# **Understanding Stage**

 $Document(s) \rightarrow Knowledge base$ 

#### Generation Stage

Knowledge base  $\rightarrow$  Summary



# A Compromise Solution

# Sentence Extraction

Document  $\rightarrow$  Sentence candidates



# A Compromise Solution

# Sentence Extraction

 $Document \rightarrow Sentence candidates$ 

## Cohesion Check

Sentence candidates → Coherent text



# A Compromise Solution

# Sentence Extraction

 $Document \rightarrow Sentence candidates$ 

## Cohesion Check

Sentence candidates  $\rightarrow$  Coherent text

### Balance and Coverage

Coherent text  $\rightarrow$  Summary



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# A Compromise Solution

#### Sentence Extraction

 $Document \rightarrow Sentence candidates$ 

This is what most commercial and free summarisers do

#### Cohesion Check

Sentence candidates  $\rightarrow$  Coherent text

## Balance and Coverage

Coherent text  $\rightarrow$  Summary



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# General Approach

#### For each sentence . . .

- 1. Look for clues to its importance.
- 2. Compute a score for the sentence based on the clues found.
- 3. Select all sentences whose scores exceed some threshold.
  - Or select the highest scoring sentences up to a certain total.



# The Frequency-keyword Approach

- 1. Compute the keywords of the document:
  - Ignore the function words by using a stop word list.
  - Sort all remaining words according to frequency or measures such as tf.idf (next slide).
  - ► Select the top words (say, the top 5%).



# The Frequency-keyword Approach

- 1. Compute the keywords of the document:
  - Ignore the function words by using a stop word list.
  - Sort all remaining words according to frequency or measures such as tf.idf (next slide).
  - ► Select the top words (say, the top 5%).
- Score the document sentences according to the presence of keywords:
  - ► Simple keyword count.
  - ▶ Weighted keyword count (keyword weights for each sentence).
  - Looking for keyword clusters in the sentence.



# Finding Most Informative Sentences

# tf.idf to find keywords

▶ Term Frequency (tf): Words that are very frequent in a document are more "important".

$$tf(w) = \#$$
 times word wis in document

▶ Inverse Document Frequency (idf): Words that appear in many documents are less "important".

$$idf(w) = \log \frac{\# \text{ documents}}{\# \text{ documens that contain word} w}$$



# The Biased Keyword Approach

# Title and headings biased

Compute a list of keywords on the basis of document structure:

- select candidates from titles and headings only, or
- candidates from titles and headings have more importance:
  - e.g. they are counted as being more frequent.



# The Biased Keyword Approach

### Title and headings biased

Compute a list of keywords on the basis of document structure:

- select candidates from titles and headings only, or
- candidates from titles and headings have more importance:
  - e.g. they are counted as being more frequent.

# Query biased (customised summaries)

Use the user's query to determine the keyword's weights:

- ▶ the user's query determines all the keywords, or
- ▶ the user's query introduces additional keywords or updates the weights of existing keywords.



#### The Location Method

# Observation

First and last sentence of a paragraph are usually most central to the theme of a text.

Increase the score of a sentence according to its position in the paragraph:

- Beginning of paragraph.
- End of paragraph.



# Cues, Indicator Phrases I

#### Cues

- ► Certain words (not necessarily keywords) provide an indication of the importance of the sentence.
- ▶ Use these words to determine the sentence score:
  - bonus words increase the sentence score:
    - ▶ "greatest", "significant"
  - stigma words decrease the sentence score:

FRM Summarisation

"hardly", "impossible", "now"



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# Cues, Indicator Phrases II

#### Indicator Phrases

Indicator phrases are specific phrases or patterns of phrases that can be used to determine the sentence importance:

- ▶ "The main aim of the present paper is ..."
- ▶ "The purpose of this article is . . . "
- "In this report, we outline ..."
- "Our investigation has shown that ..."



#### Relational Criteria

- 1. Build a semantic structure for the document:
  - sentences are vertices
  - inter-sentence links are edges
    - ► Rhetorical links (ellaboration, sequence, etc)
    - Cooccurrence of keywords
    - ▶ . .
- Use the link structure to determine the most important sentences
- ► Degree of the vertex
- ► Eigenvalues (PageRank style)





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## Textual Cohesion

- ► Lack of cohesion results in "odd" extracts.
- Sentences include references to other sentences:
  - ► Anaphoric reference:
  - "John saw Mary. She was talking over the phone"
  - Rhetorical connectives:
  - ▶ " So, the following example ..."
  - Lexical or definite reference:
  - ▶ "I saw a man with a book. The book was . . . "
- Possible solutions:

Aggregation Add preceding sentences until there are no external references

Deletion Remove the difficult sentences.

Modification Alter the sentences to eliminate or disguise the problem.



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# Balance and Coverage

- We need to process the selected sentences in order to produce a real abstract:
  - Delete redundant sentences.
  - Harmonise tense and voice of verbs.
  - Ensure balance and proper coverage.
- Combination of information extraction and text generation.
- Need to consider text structure:
  - ► Each sentence plays a role in the text and in relation with the other sentences.
- Problem to address:
  - Lack of balance and coverage:
    - Missing important information.
    - ► Too much emphasis on less important information.



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# Single-document Summarisation

## Input

- Question.
- Document Abstract.

## Output

- Extractive summary that answers the question.
- ► Target summary is the annotated answer justification ("long").
- Evaluated using ROUGE-L with Stemming.



# General Approach (Sarker et al., CBMS 2012)

#### In a Nutshell

- 1. Gather statistics from the best 3-sentence extracts.
  - Exhaustive search to find these best extracts.
- 2. Build three classifiers, one per sentence in the final extract.
  - ▶ Classifier 1 based on statistics from best 1st sentence.
  - ▶ Classifier 2 based on statistics from best 2nd sentence.
  - Classifier 3 based on statistics from best 3rd sentence.



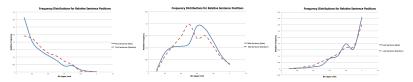
#### The Statistics Gathered

- 1. Source sentence position.
- 2. Sentence length.
- 3. Sentence similarity.
- 4. Sentence type.



#### 1. Source Sentence Position

- ► Compute relative positions (0 ...1).
- ▶ Create normalised frequency histograms  $f_1, f_2, ..., f_{10}$ .
- Score every relative position in bin i with its bin frequency:  $S_{pos}(i) = f_{bin(i)}$ .





## 2. Sentence Length

Reward larger sentences and penalise shorter sentences:

#### Normalised sentence length

$$S_{len}(i) = \frac{I_s - I_{avg}}{I_d}$$

Is: sentence length

lavg: average sentence length in the corpus

Id: document length



## 3. Sentence Similarity

### Sentence Similarity

- ▶ Lowercase, stem, remove stop words.
- Build vector of tf.idf with remaining words and UMLS semantic types.
- $CosSim(X, Y) = \frac{X.Y}{|X||Y|}$

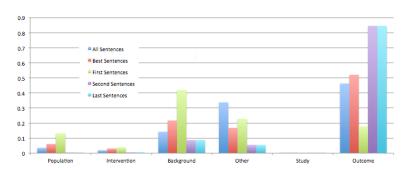
## Maximal Marginal Relevance (Carbonell & Goldstein, 1998)

Reward sentences similar to the query and penalise those similar to other summary sentences.

$$MMR = \lambda(CosSim(S_i, Q)) - (1 - \lambda)max_{S_i \in S}(CosSim(S_i, S_j))$$



## 4. PIBOSO (Kim et al. 2011) I



- 1. Classify all sentences into PIBOSO types (a variant of PICO).
- 2. Generate normalised frequency histograms of resulting PIBOSO types.



# 4. PIBOSO (Kim et al. 2011) II

### Position independent

$$S_{PIPS}(i) = \frac{P_{best}}{P_{all}}$$

#### Position dependent

$$S_{PDPS}(i) = \frac{P_{pos}}{P_{hest}}$$

Phest:

proportion of this PIBOSO type among all best summary sentences.

 $P_{all}$ :

proportion of this PIBOSO type among all sentences.

 $P_{pos}$ :

proportion of this PIBOSO type among all best summary sentences at

this position.

#### Classification

#### Edmunsonian Formula

$$S_{S_i} = \alpha S_{rpos_i} + \beta S_{len_i} + \gamma S_{PIPS_i} + \delta S_{PDPS_i} + \epsilon S_{MMR_i}$$

- ▶ MMR is replaced with cosine similarity for first sentence.
- In case of ties, the sentence with greatest length is chosen.
- Parameters are fine-tuned through exhaustive search (grid search) using training set.

$$\alpha = 1.0, \ \beta = 0.8, \ \gamma = 0.1, \ \delta = 0.8, \ \epsilon = 0.1, \ \lambda = 0.1.$$



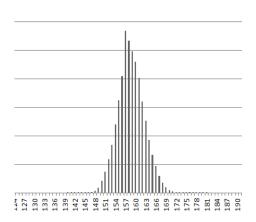
## Percentile-based Evaluation (Ceylan et al. 2010) I

We compare against all possible 3-sentence extracts in the test set.

- 1. Bin all possible three-sentence combinations of each abstract.
  - ▶ 1.000 bins.
- 2. Normalise the resulting histograms.
- Combine all histograms.
  - convolution.
- 4. The result approximates the probability density distribution of all three-sentence summaries in all abstracts.



# Percentile-based Evaluation (Ceylan et al. 2010) II





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## Systems

- L3 Last three sentences.
- O3 Last three PIBOSO outcome sentences.
  - R Random.
  - O All outcome sentences.
  - PI Sentence position independent.
- PD Sentence position dependent (our proposal).



## Results

System	F-Score	95% CI	Percentile (%)
L3	0.159	0.155-0.163	60.3
O3	0.161	0.158-0.165	77.5
R	0.158	0.154-0.161	50.3
0	0.159	0.155-0.164	60.3
PI	0.160	0.157-0.164	69.4
PD	0.166	0.162-0.170	97.3



#### Towards Multi-document Summarisation

- Evidence suggests that a two-step process is promising (Sarker et al., ALTA 2012).
  - 1. Single document summarisation.
  - 2. Multi-document summarisation from the single-document summaries
- ► Traditional clustering techniques seem to produce good clustering of references (Shash & Molla, unpublished).
- ▶ We are still looking at means to obtain the answer parts.
  - ► Topics as cluster centroids.
  - Overlap with the question.



### Contents

#### Evidence Based Medicine

What is Evidence Based Medicine EBM and NLP
A Corpus for Summarisation

#### Text Summarisation

Sentence Extraction Cohesion Check Balance and Coverage

### Proposals for Text Summarisation

Single-document Summarisation

Optimisation and Summarisation



## Optimisation and NLP

### Many NLP tasks are based on optimisation

- ► Text classification: minimise the classification error.
- ▶ Part of speech tagging: Find the optimal sequence of labels.
- ▶ Parsing: Find the most likely parse.
- ► Machine translation: Dual optimisation.
  - ▶ The target sentence must keep the most meaning.
  - ► The target sentence must try to follow the language model of the target language.



## Optimisation in Machine Learning

### Many ML tasks are about optimising parameters

$$\arg\min_{\theta} J(h_{\theta}(X), Y)$$

- J Cost (error) function.
- $\theta$  Machine Learning parametres.
- $h_{\theta}(X)$  Hypothesis function.
  - X Inputs.
  - Y Observed results.



## Optimisation in Classification Tasks

### Hypothesis function in Logistic Regression

Sigmoid (logistic) function

$$h_{ heta}(X) = rac{1}{1 + e^{-\sum_i heta_i x_i}} = rac{1}{1 + e^{- heta^T X}}$$

### Cost function in Logistic Regression

Cross entropy

$$J(h_{\theta}(X), Y) = -\frac{1}{m} \left[ \sum_{i=1}^{m} y^{(i)} \log(h_{\theta}(x^{(i)}) + (1 - y^{(i)}) \log(1 - h_{\theta}(x^{(i)})) \right]$$



## Query Based Summarisation I

A summary sentence  $S_i$  must maximise its similarity with the question

$$\arg\max_{S} \sum_{i} CosSim(S_{i}, Q)$$

A summary sentence  $S_i$  must minimise its similarity with other summary sentences

$$\arg\min_{S} \sum_{i,j} CosSim(S_i, S_j)$$



## Query Based Summarisation II

## Maximal Marginal Relevance (Carbonell & Goldstein, 1998)

#### Greedy approach

▶ Each iteration, select the sentence  $S_i$  with highest MMR score.

$$MMR = \lambda(CosSim(S_i, Q)) - (1 - \lambda) \max_{S_i \in S} (CosSim(S_i, S_j))$$



## Single-document Summarisation

### Information contents in summary sentences must be maximal

▶ The sum of all weighted concepts in a summary must be maximal.

$$\arg\max_{S} \sum_{c \in C_S} w(c)$$

▶ This is the knapsack problem (NP-hard).

## Readability in summary sentences must be maximal

- ▶ We've seen some aspects of readability above . . .
- ▶ ... now we need to express them as a problem of optimisation.



# Query-based Multi-document Abstractive Summarisation

### The summary must fit the question topics best

- 1. Fitting to cluster centroids.
- 2. Topic modelling (LDA) variants.

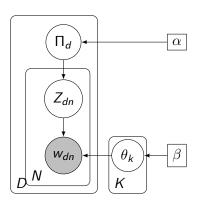
### The summary must be most readable

Find the word extracts that are most likely produced by language models:

- 1. Word sequences (e.g. 2-grams).
- 2. Best likely parse (e.g. the *k*-minimum spanning tree in a graph).



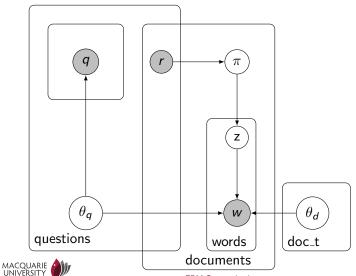
#### Latent Dirichlet Allocation



- Π<sub>d</sub>: topic probability distribution for document d.
- ► Z<sub>dn</sub>: actual topic selected for word *n* in document *d*.
- $\theta_k$ : word probability distribution for topic k.
- $\alpha, \beta$ : hyperparameters.



# Possible LDA Variant for Query-focused Summarisation



## Summary

- ► Evidence Based Medicine (EBM) is an important problem that medical doctors face.
- ► EBM can benefit from Natural Language Processing (NLP) in general, and text summarisation in particular.
- Text summarisation, like many NLP tasks, relies on optimisation.
- We need expertise on optimisation techniques!



## Summary

- Evidence Based Medicine (EBM) is an important problem that medical doctors face.
- ► EBM can benefit from Natural Language Processing (NLP) in general, and text summarisation in particular.
- ► Text summarisation, like many NLP tasks, relies on optimisation.
- We need expertise on optimisation techniques!

#### Questions?

Further information about our research:

http://web.science.mq.edu.au/~diego/medicalnlp/

