Lecture 10: Language generation

Overview of Natural Language Generation

Text summarisation

Extractive summarisation

Query-focused multi-document summarisation

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Overview of Natural Language Generation

Language generation

Generation from what?! (Yorick Wilks)

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Generation

Starting points:

- Some semantic representation:
 - logical form (early work)
 - distributional representations (e.g. paraphrasing)
- Formally-defined data: databases, knowledge bases

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- Semi-structured data: tables, graphs etc.
- Numerical data: e.g., weather reports.
- User input in assistive communication.

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Regeneration: transforming text

Statistical machine translation

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- Paraphrasing
- Summarization
- Text simplification

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Components of a generation system

Content selection deciding what information to convey (selecting important or relevant content) Discourse structuring overall ordering, sub-headings etc Aggregation deciding how to split information into sentence-sized chunks

Referring expression generation deciding when to use pronouns, which modifiers to use etc

Lexical choice which lexical items convey a given concept

Realization mapping from a meaning representation (or syntax tree) to a string (or speech)

Fluency ranking discriminate between grammatically / semantically valid and invalid sentences

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Approaches to generation

- Early work (limited domain): hand-written rules for first five steps, grammar for realization
- Templates (limited domain): most practical systems. Fixed text with slots, fixed rules for content selection.
- Statistical (limited domain): components as above, but use machine learning (supervised or unsupervised).

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Regeneration: statistical or mixed.

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Text summarisation ¹

Task: generate a short version of a text that contains the most important information

Single-document summarisation:

- given a single document
- produce its short summary

Multi-document summarisation:

- given a set of documents
- produce a brief summary of their content

¹This part of the lecture is based on Dan Jurafsky's summarisation lecture, and is (quite appropriately) a summary thereof. The full lecture can viewed online at https://class.coursera.org/nlp/lecture/preview

Generic vs. Query-focused summarisation

Generic summarisation:

 identifying important information in the document(s) and presenting it in a short summary

Query-focused summarisation:

 summarising the document in order to answer a specific query from a user

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A simple example of query-focused summarisation

Google	what is natural language processing?							
	Web	Videos	News	Images	Shopping	More -	Search tools	
	About 3	0,200,000 re	esults (0.38	seconds)				
	Natu com inter lang hum Natu https	Natural language processing (NLP) is a field of computer science, artificial intelligence, and computational linguistics concerned with the interactions between computers and human (natural) language. As such, NLP is related to the area of human–computer interaction.						
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	Natur https://	al langua en.wikipedia language p	age proc a.org/wiki/l processing	essing - V Natural_lang (NLP) is a fig	Vikipedia, t guage_proce	he free ssing ▼ r science, art	ificial intelligence,	

and computational linguistics concerned with the interactions between computers and human (natural) languages. As such, NLP is related to the area of human–computer interaction.

Outline of natural language ... - Natural language understanding

Approaches

Extractive summarisation:

- extract important / relevant sentences from the document(s)
- combine them into a summary

Abstractive summarisation:

- interpret the content of the document (semantics, discourse etc.) and generate the summary
- formulate the summary using other words than in the document
- very hard to do!

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Three main components:

 Content selection: identify important sentences to extract from the document

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- Information ordering: order the sentences within the summary
- Sentence realisation: sentence simplification

Content selection – unsupervised approach

- Choose sentences that contain informative words
- Informativeness measured by:
 - tf-idf: assign a weight to each word i in the doc j as

weight(w_i) = $tf_{ij} * idf_i$

 tf_{ij} – frequency of word *i* in doc *j* idf_i – inverse document frequency

$$idf_i = \log \frac{N}{n_i}$$

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N – total docs; n_i docs containing w_i

- mutual information
- log-likelihood ratio (LLR)

Content selection – supervised approach

- start with a training set of documents and their summaries
- align sentences in summaries and documents
- extract features:
 - position of the sentence (e.g. first sentence)
 - sentence length
 - informative words
 - cue phrases
 - etc.
- train a binary classifier: should the sentence be included in the summary?

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Content selection - supervised vs. unsupervised

Problems with the supervised approach:

- difficult to obtain data
- difficult to align human-produced summaries with sentences in the doc
- doesn't perform better than unsupervised in practice

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An example summary

from Nenkova and McKeown (2011):

As his lawyers in London tried to quash a Spanish arrest warrant for Gen. Augusto Pinochet, the former Chilean Dictator, efforts began in Geneva and Paris to have him extradited. Britain has defended its arrest of Gen. Augusto Pinochet, with one lawmaker saying that Chile's claim that the former Chilean Dictator has diplomatic immunity is ridiculous. Margaret Thatcher entertained former Chilean Dictator Gen. Augusto Pinochet at her home two weeks before he was arrested in his bed in a London hospital, the ex-prime minister's office said Tuesday, amid growing diplomatic and domestic controversy over the move.

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Query-focused multi-document summarisation

Example query: "Describe the coal mine accidents in China and actions taken"

Steps in summarization:

- 1. find a set of relevant documents
- 2. simplify sentences
- 3. identify informative sentences in the documents

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- 4. order the sentences into a summary
- 5. modify the sentences as needed

Sentence simplification

- parse sentences
- hand-code rules to decide which modifiers to prune
 - appositives: e.g. Also on display was a painting by Sandor Landeau, an artist who was living in Paris at the time.
 - attribution clauses: e.g. Eating too much bacon can lead to cancer, the WHO reported on Monday.
 - PPs without proper names: e.g. Electoral support for Plaid Cymru increased to a new level.
 - initial adverbials: e.g. For example, On the other hand,
- also possible to develop a classifier (e.g. satelite identification and removal)

Content selection from multiple documents

Select informative and non-redundunt sentences:

- Estimate informativeness of each sentence (based on informative words)
- Start with the most informative sentence:
 - identify informative words based on e.g. tf-idf
 - words in the query also considered informative
- Add sentences to the summary based on maximal marginal relevance (MMR)

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Content selection from multiple documents

Maximal marginal relevance (MMR): iterative method to choose the best sentence to add to the summary so far

- Relevance to the query: high cosine similarity between the sentence and the query
- Novelty wrt the summary so far: low cosine similarity with the summary sentences

$$\hat{s} = \operatorname*{argmax}_{s_i \in D} \left[\lambda sim(s_i, Q) - (1 - \lambda) \max_{s_j \in S} sim(s_i, s_j) \right]$$

Stop when the summary has reached the desired length

Sentence ordering in the summary

- Chronologically: e.g. by date of the document
- Coherence:
 - order based on sentence similarity (sentences next to each other should be similar, e.g. by cosine)
 - order so that the sentences next to each other discuss the same entity / referent
- Topical ordering: learn a set of topics present in the documents, e.g. using LDA, and then order sentences by topic.

Example summary

Query: "Describe the coal mine accidents in China and actions taken"

Example summary (from Li and Li 2013):

(1) In the first eight months, the death toll of coal mine accidents across China rose 8.5 percent from the same period last year.
(2) China will close down a number of ill-operated coal mines at the end of this month, said a work safety official here Monday. (3) Li Yizhong, director of the National Bureau of Production Safety Supervision and Administration, has said the collusion between mine owners and officials is to be condemned. (4) from January to September this year, 4,228 people were killed in 2,337 coal mine accidents. (5) Chen said officials who refused to register their stakes in coal mines within the required time