# Natural Language Understanding, Generation, and Machine Translation

Lecture 22 and 23: Paraphrase Acquisition and Generation

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# What Is Paraphrasing?



# What Is Paraphrasing?

Sentences or phrases that convey approximately the same meaning using different words (Bhagat and Hovy, 2012).

#### Words

fortunate lucky favourable advantageous opportune timely convenient felicitous

#### Phrases

for example one such example an istance in particular including specifically however more concretely

#### Sentences

The parrot is dead It's not pinin, it's passed on! This parrot is no more! It has ceased to be! It's gone to meet its maker! It's a late parrot! It's a stiff!

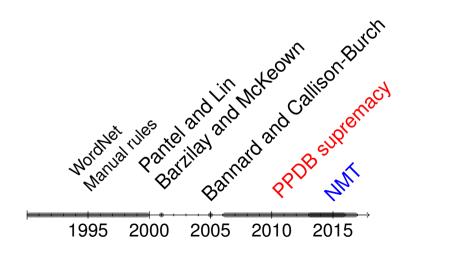
### Why Should we Care?

Recognizing and generating paraphrases is an important component in many natural language processing applications:

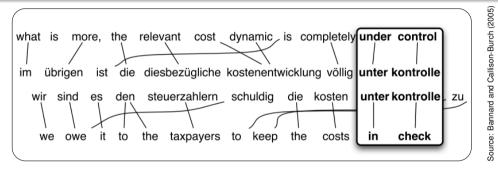
- Question Answering (Fader et al., 2013)
- MT Evaluation (Denkowski and Lavie, 2014)
- Semantic Parsing (Berant and Liang, 2014)
- Simplification (Xu et al., 2016)
- Summarization (Woodesend and Lapata, 2012)
- Style transfer (Xu et al., 2012; Pavlick and Nenkova, 2015)
- Grammar correction (Dahlmeier and Tou Ng, 2011)

### The Past

### A Very Brief History of Paraphrasing



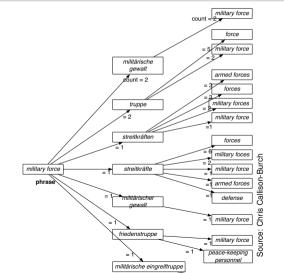
### Bilingual Pivoting: An Example



- No direct path from en  $\rightarrow$  en; indirect path from en  $\rightarrow$  de and de  $\rightarrow$  en
- If two English phrases are aligned with the same phrase in another language they are likely to be a paraphrase:

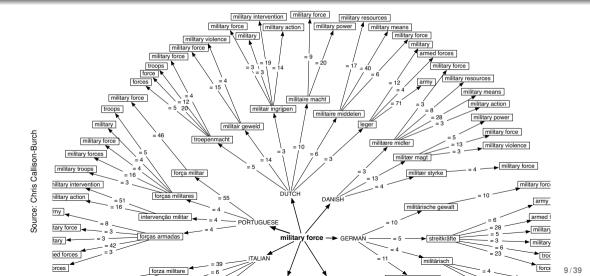
$$P(e_2|e_1) = \sum_{F} P(e_2|f)P(f|e_1)$$

### Bilingual Pivoting: An Example

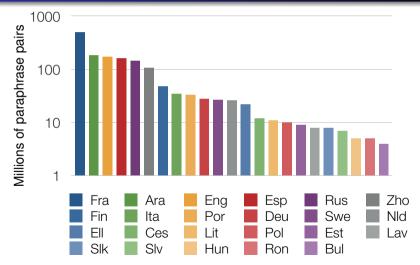


- paraphrases extracted from phrase table of SMT system
- lexical and syntactic (aka rules)
- enriched with annotations for specific tasks (e.g., entailment, simplification)
- http://paraphrase.org

### Bilingual Pivoting: An Example



### PPDB: The Paraphrase Database



NMT and PARANET Sentence Compression Question Answering

### The Present

### **Neural Machine Translation**

	Time	Company	NMT Implementation	Framework	Characteristic
Source: Synced Review	2016.09	Google	GNMT	Tensorflow	State of the art industrial implementation of the "attentional encoder-decoder networks" model
	2016.11	Microsoft	No technical details disclosed		
	2017.05	Facebook	Fairseq	Torch	Used CNN to replace RNN
	2017.06	Google	Transformer	Tensorflow/ Tensor2Tensor	Solely attention based NMT
	2017.07	Amazon	Sockeye		MXNet

Other companies took part in the NMT R&D include: IBM, NVIDIA, SYSTRAN

### NMT and Multiple Pivots

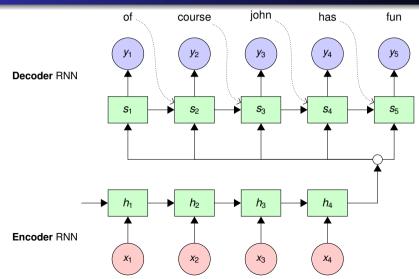
#### Neural machine translation

- context-aware, places greater emphasis on semantics;
- allows for multiple pivots with multiple models;
- assigns a probability to arbitrary sequences.

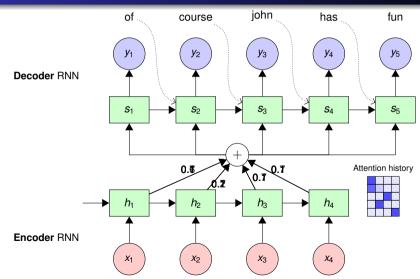
#### Relevance to paraphrases

- represent paraphrases in continuous space;
- NMT learns embeddings for words and phrases naturally;
- estimate semantic relatedness between text segments of arbitrary length;
- generate candidate paraphrases for any source input.

### **Encoder-Decoder Architecture**

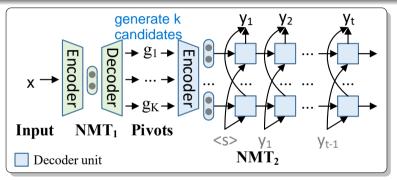


### **Encoder-Decoder with Attention**



### **PARANET**

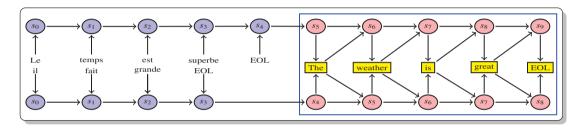
source to pivots to source



Source is translated into *K*-best list of foreign pivots  $G_X = \{g_1, \dots, g_K\}$ .

$$p(x'|\mathcal{G}_x) = \prod_{t=1}^{|x'|} p(y_t|y_{< t}, \mathcal{G}_x) = \prod_{t=1}^{|x'|} \sum_{k=1}^{K} p(g_k|x) p(y_t|y_{< t}, g_k)$$

### PARANET: Multiple Pivots



- Late-weighted combination (Firhat et al., 2016)
- Pivot sentences are encoded separately by two encoders
- At each time step, two decoders produce a probability distribution over all words, which are then combined (in the yellow square)
- From this combined distribution a word is chosen, which is then given as input to each decoder.

### Pivoting over Multiple Languages

- Pivot over sentences from multiple language pairs
- Translate x into K-best German  $\mathcal{G}^{D\mathcal{E}}$  and French  $\mathcal{G}^{FR}$  sentences

• Then average distributions 
$$P(x'|\mathcal{G}^{DE},\mathcal{G}^{FR}) = \prod_{t=1}^{|x'|} p(y_t|y_{< t},\mathcal{G}^{DE},\mathcal{G}^{FR})$$

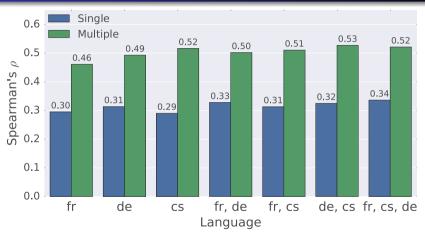
$$p(y_t|y_{< t},\mathcal{G}^{DE}) = \sum_{k=1}^{K} p\left(g_k^{\mathcal{D}\mathcal{E}}|x\right) p\left(y_t|y_{< t},g_k^{\mathcal{D}\mathcal{E}}\right)$$

$$p(y_t|y_{< t},\mathcal{G}^{FR}) = \sum_{k=1}^{K} p\left(g_k^{\mathcal{F}\mathcal{E}}|x\right) p\left(y_t|y_{< t},g_k^{\mathcal{F}\mathcal{E}}\right)$$

$$p(y_t|y_{< t},\mathcal{G}^{FR}) = \lambda_1 \sum_{k=1}^{K} p\left(g_k^{\mathcal{F}\mathcal{E}}|x\right) p\left(y_t|y_{< t},g_k^{\mathcal{F}\mathcal{E}}\right)$$

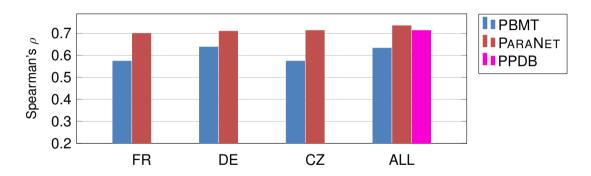
$$p(y_t|y_{< t},\mathcal{G}^{DE},\mathcal{G}^{FR}) = \lambda_1 \sum_{k=1}^{K} p\left(g_k^{\mathcal{D}\mathcal{E}}|x\right) p\left(y_t|y_{< t},g_k^{\mathcal{D}\mathcal{E}}\right) + \lambda_2 \sum_{k=1}^{K} p\left(g_k^{\mathcal{F}\mathcal{E}}|x\right) p\left(y_t|y_{< t},g_k^{\mathcal{F}\mathcal{E}}\right)$$

# Results: One vs Many Pivots



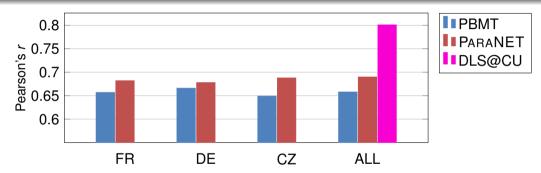
- PPDB 2.0 dataset: humans rate (1-5) paraphrase pairs (Pavlick et al. 2015);
- Spearman's  $\rho$  between ratings and PARANET's  $P(e_2|e_1)$ .

### Results: How Good are the Paraphrases?



- Supervised scoring model (Pavlick et al., 2015)
- Features: sentence length,1-4 gram string similarity, paraphrase probability  $p(e_2|e_1)$ , the language model score, cosine distance of paraphrase vectors, attention scores.

### Results: How Good are the Embeddings?



- SemEval-2015 Task 2: Semantic Textual Similarity
- Predict similarity between two sentences on a 1–5 scale
- News headlines, image captions, student answers, answers to question in public forums and sentences expressing committed belief

### What about Generation?

WikiAnswers (Fader et al., 2013)

Beauty is not in the eye of the beholder.

Beauty is not in the mind of the viewer.

What is the importance of employee satisfaction in an organization? What is the significance of staff satisfaction at an organisation?

2000 Leagues under the Sea (Barzilay and McKeown, 2001)

"I owed myself this revenge!" Said the Captain to the Canadian.

"I am indebted to this revenge!" the captain told the Canadian.

"That's what I've been telling you Ned"

"That's what I said, Ned".

### Results: What about Generation?

#### Multiple Translation Chinese (MTC) corpus

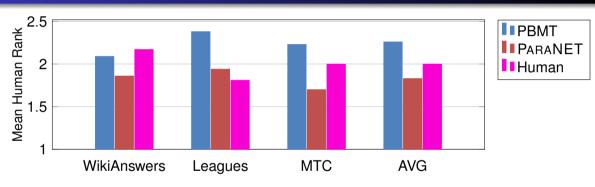
We will accelerate the drafting of telecommunications legalization, amend the law of post and the regulations governing wireless telecommunications.

We will speed up the design of telecommunications, change the law and regulations governing wireless telecommunication.

(London, AP) The British government is working on resolving the increasingly serious problems of street crimes and will strengthen patrolling police.

London, AP The British government is working to resolve the increasingly serious problems of street crime and will strengthen patrols.

### Results: What about Generation?



- Human evaluation using Mechanical Turk
- Participants see source sentence and rank paraphrases produced by PBMT,
   PARANET, and Humans.
- 100 sentences per domain, 5 judgments per item

### Text Rewriting, No Problem!

**Sentence compression:** produces a summary of a single sentence by using less words, preserving the most important information, and remaining grammatical.

#### Sentence compression: main idea

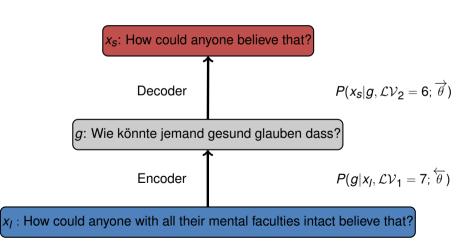
- use ParaNet as back-bone for rewriting;
- Pivot through multiple languages extracting latent compression information;
- Incorporate length information into the decoder.

#### Length embedding (Kikuchi et al., 2016)

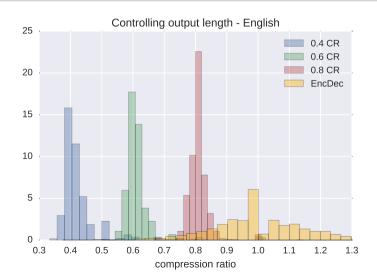
Initialise the decoder, not only with the encoder's hidden state but also a length embedding, which is then scaled according to the target length.

# **Pivoting and Compression**

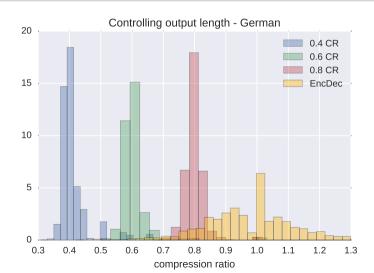
$$P(x_s|x_l, g, \mathcal{LV}_1, \mathcal{LV}_2) \triangleq P(g|x_l, \mathcal{LV}_1)P(x_s|g, \mathcal{LV}_2)$$



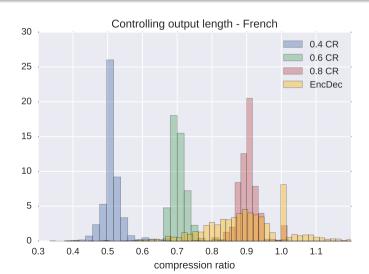
### Parameterizing via Length



# Parameterizing via Length



### Parameterizing via Length



### Compression Output in English

It was through him that the scandal broke in 2011, in an in-depth investigation into corruption related to road construction contracts in Quebec, to which the liberal Prime Minister at the time, Jean Charest, had consented only reluctantly. (39 tokens)

He met him. (4 tokens)

The scandal was revealed by him due to an in-depth investigation into the Quebec Treaties. (16 tokens)

The scandal was revealed by him due to an in-depth investigation into the contracts related to the contracts for roadworks contracts. (22 tokens)

The scandal was revealed by him due to an in-depth investigation into the roadworks contracts in Quebec, which the then Prime Minister reluctantly agreed to. (27 tokens)

### Compression Output in German

Wie Sie sicher aus der Presse und dem Fernsehen wissen, gab es in Sri Lanka mehrere Bombenexplosionen mit zahlreichen Toten. (21 tokens)

Sie wissen aus der Presse und dem Fernsehen, dass es in Sri Lanka eine Reihe von Bombenexplosionen und -Morden gegeben hat. (23 tokens)

Sie wissen aus Presse und Fernsehen, dass es in Sri Lanka zahlreiche Bombenanschläge und Morde gegeben hat. (19 tokens)

Die Presse und das Fernsehen haben zahlreiche Bombenexplosionen und Morde in Sri Lanka. (13 tokens)

In Sri Lanka sind zahlreiche Bombenexplosionen aufgetreten. (8 tokens)

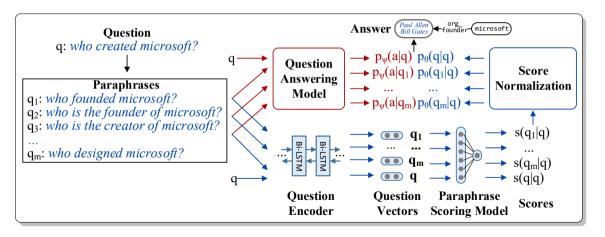
### Paraphrases for Question Answering

An Old Idea: paraphrase question, submit paraphrases to QA module (Duboue and Chu-Carroll, 2006; Narayan et al., 2016; Chen et al., 2016; Fader et al., 2013; Berant et al., 2014; Bordes et al., 2014a,b; Dong et al., 2015).

A Common Problem: generated paraphrases often contain inappropriate candidates; scoring takes place independently of QA module.

**A New Solution:** train paraphrase model jointly with QA-model for end tasks.

#### **Problem Formulation**



### **Problem Formulation**

$$p\left(a|q
ight) = \sum_{q' \in H_q \cup \{q\}} egin{array}{c} oldsymbol{p}_{\psi}\left(a|q'
ight) & oldsymbol{p}_{ heta}\left(q'|q
ight) \ & ext{QA Model} & ext{Paraphrase Model} \end{array}$$

 $H_q$  is the set of paraphrases for question q;  $\psi$ ,  $\theta$  parameters of QA and paraphrase scoring model.

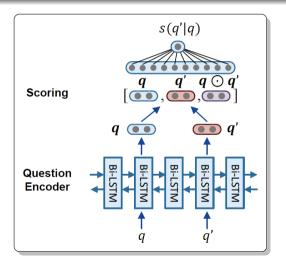
### **Training**

$$\max \sum_{(q,a) \in \mathcal{D}} \log p(a|q)$$

#### Inference

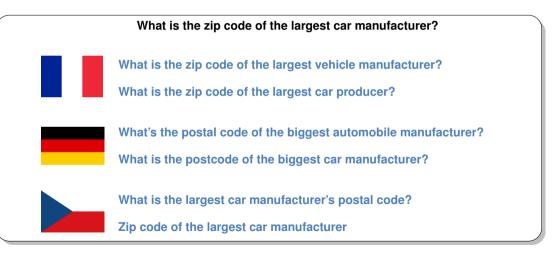
$$\hat{a} = rg \max_{a' \in C_a} p(a'|q)$$

### Paraphrase Model

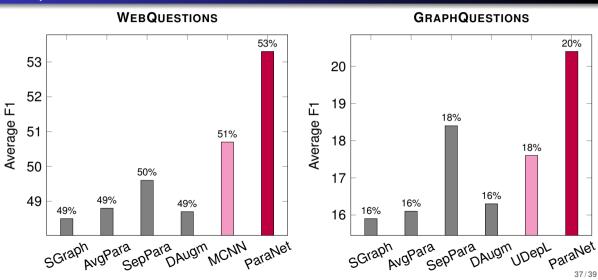


- We need to generate paraphrases for every question.
- Various options (rules, WordNet, PPDB).
- We turn to neural Machine Translation.
- Any NMT engine will do, we use http://opennmt.net/

### English Paraphrases via Different Pivots



# Experimental Results



### Conclusions



- There's tons of data in multiple languages!
- New models can utilize this data for specific tasks.
- Sentence compression: NMT is the compression model.
- QA: the model uses paraphrases without NMT.
- As NMT gets better our paraphrasing models will get better too.

### Ways to Say I Love You

#### French

Je vous aime.
J'adore vous.
Je t'aime beaucoup.
Je vous aime
beaucoup.
Je vous aime bien.
Je vous remercie.
Je t'aime bien.

#### German

Ich liebe Dich.
Ich liebe Euch.
Ich liebe Sie.
Ich liebe Dich selbst.
Ich liebe Dich sehr.
Ich liebe dich
jedenfalls.
Ich liebe dich doch.

#### English

I love you guys.
I do love you.
I just love you.
Me love you.
I love thee.
I really love you.
I love you too.
Well, I love you.