#### DATA DRIVEN MACHINE TRANSLATION

David Talbot 22nd April 2017

Computer Science Club, St. Petersburg, Russia



### what you will learn

- · Why machine translation is hard
- · How to build a machine translation system from data
- · How to evaluate a machine translation system
- · How to improve it using linguistic knowledge/inductive bias

### what you will learn

- · Phrase-based machine translation
  - · Word alignment
  - · Syntax based reordering
  - · Language models
- · Neural machine translation
  - · Word embeddings
  - · Encoder-decoder models
  - · Attention mechanisms
  - · Challenges for NMT

what else you'll learn (beyond mt)

Practical aspects of statistical modelling

- · Estimating statistical models from data
- · Handling hidden variables
- · Introducing inductive bias into statistical models





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- · В конце концов компьютер, который понимает вас так же хорошо, как он понимает вашу маму.

#### word order



"As English not all languages words in the same order put. Hmmmmmm." – Yoda

#### word order

Languages can use very different word order

- · He went to school by train.
- ・彼は電車で学校に行きました。
- · kara wa densha de gakkou ni ikimashita.

Why is reordering a huge problem for MT systems?

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Why is reordering a problem for an MT system?

- A priori set of possible 'reorderings' scales exponentially with sentence length

### morphology



"Rarely is the question asked: Is our children learning?" – George W. Bush

### morphology

Many languages require case marking and agreement

- Перечень различных рекордов скорости, установленных на рельсовых путях, был ...
- Word choices depend on the gender, case, number etc. of other words
- · Morphological agreement can span many tokens

#### information structure

Morphological case and agreement make word order less important

- · The dog bit the hippopotamus.
- · The hippopotamus bit the dog.
- Собака укусила бегемота.
- Собаку укусил бегемот.

### world knowledge

Usually only one interpretation is reasonable for us

- · Stolen painting found by tree.
- · I haven't slept for ten days.
- · I saw a man with a telescope.

### even easy sentences are hard

```
en I've got two brothers.

fr J'ai deux frères.
  (I've two brothers.)

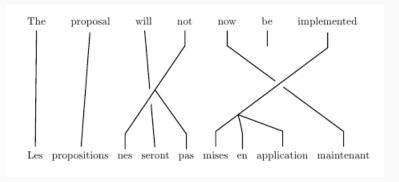
ru У меня два брата.
  (At me [are] two brothers.)

ja 私は 2 人の弟がいます.
  (As for me, two people younger brother there are.)
```

## not a simple machine learning problem

- · High dimensional: vocabulary > 1 million words
- · Sparse: natural language follows a Zipf law
- · Combinatorial: reordering is a priori O(N!)
- Dependencies
- · More than one right answer...
- · Partially observed data

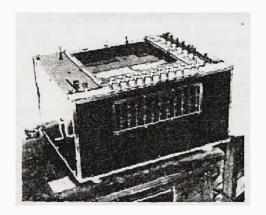
### a word alignment



'The Mathematics of Machine Translation: Parameter Estimation', Brown et al. (1993). '

SOME OF THE STORY SO FAR

# pre-history



1930s Peter Troyanskii and Georges Artsrouni patented mechanical translation devices.

### ww2 cryptography



1940s Shannaon, Weaver, Turing: Information theory, Bayesian inference

### great expectations

1954 Georgetown - IBM experiment Russian to English

Mi pyeryedayem mislyi posryedstvom ryechyi. We transmit thoughts by means of speech.

- · Translated 60 sentences.
- · Claimed that MT would be solved within three or five years.
- · Difference between limited and open domain.

# disappointment

#### 1966 ALPAC report

- · Concluded that MT was too expensive and ineffective
- Recommended that research focus on tools to help human translators

### statistical renaissance

1993 Brown et al., 'The mathematics of statistical machine translation'

The Fundamental Equation of Statistical Machine Translation

$$\begin{split} \hat{\mathbf{e}} &= \underset{\mathbf{e}}{\text{argmax}} \, \text{Pr}(\mathbf{e}|\mathbf{f}) \\ &= \underset{\mathbf{e}}{\text{argmax}} \, \text{Pr}(\mathbf{e}) \text{Pr}(\mathbf{f}|\mathbf{e}) \end{split}$$

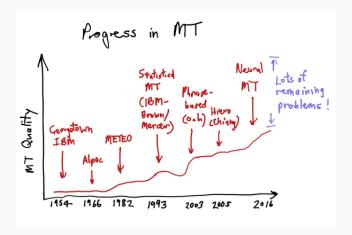


2000s Huge amounts of naturally occuring parallel data



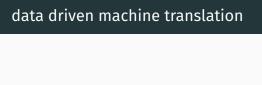
GPUs with LSTMs and other robust recurrent neural networks

## progress so far



(From Chris Manning's slides)

A PLAN



 $\cdot$  Specify a simple statistical model of translation

#### data driven machine translation

- · Specify a simple statistical model of translation
- · Learn the parameters of the model from data

#### data driven machine translation

- · Specify a simple statistical model of translation
- · Learn the parameters of the model from data
- · Use linguistic analysis to inform and constrain the model

arallel corpora
· Translated documents from governments, newspapers, etc.

#### parallel corpora

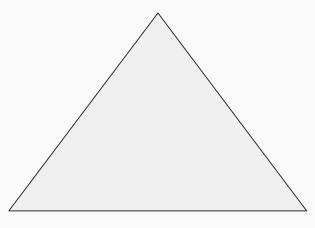
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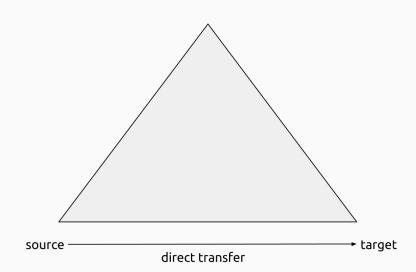
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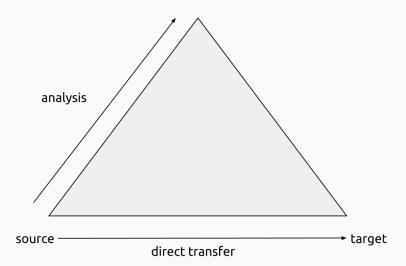
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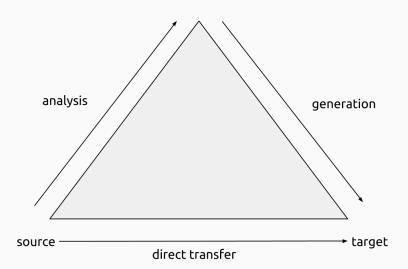
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- · What's wrong with the data?
  - · It's often noisy
  - · It's in the wrong domains (mostly)
  - · It's only partially observed
  - · There's not enough of it!

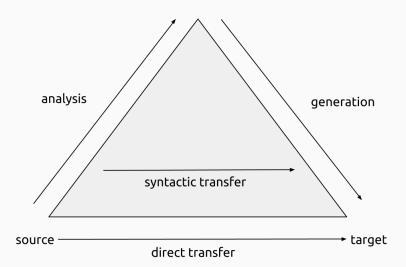


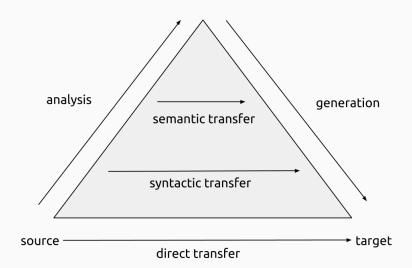
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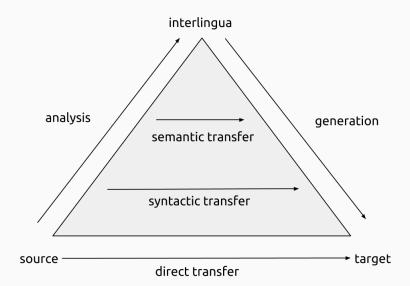














#### hands-on evaluation

#### Compare Bing, Google and Yandex Translate

- · Work in pairs
- Compare sentences from random wikipedia articles (en<->ru)
- · Add the source sentence, translations and judgements to this spreadsheet:

https://goo.gl/TcG5MZ