# Pharaoh: a Beam Search Decoder for Phrase-Based Statistical Machine Translation

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# Outline

- Phrase-Based Statistical MT
- Beam Search Decoding
- Experiments
- Advanced Features

## **Machine Translation**

• Task: Make sense of foreign text like

#### 毒品

本册子爲家長們提供實際和有川的關于毋品 的信息,包括如何減少使用非法毒品的危險. 它有助於您和您的家人討論有關毒品的問題. 這本小册子的主要內容已錄在磁帶上,如果您 想索取一盒免費的磁帶(中文), 請在下面的

- Long-standing problem in artificial intelligence
- Ultimately requires syntax, semantics, pragmatics

# Statistical Machine Translation

• Components: Translation model, language model, decoder



#### **Phrase-Based Translation**



- Foreign input is segmented in phrases
  - any sequence of words, not necessarily linguistically motivated
- Each phrase is translated into English
- Phrases are reordered

## **Phrase-Based Systems**

- A number of research groups developed phrase-based systems (RWTH Aachen, USC/ISI, CMU, IBM, JHU, ITC-irst, MIT, ... )
- Systems differ in
  - training methods
  - model for phrase translation table
  - reordering models
  - additional feature functions
- Currently best method for SMT (MT?)
  - top systems in DARPA/NIST evaluation are phrase-based
  - best commercial system for Arabic-English is phrase-based

## Pharaoh

- Translation engine
  - works with various phrase-based models
  - beam search algorithm
  - time complexity roughly linear with input length
  - good quality takes about 1 second per sentence
- Very good performance in DARPA/NIST Evaluation
- Freely available for researchers

http://www.isi.edu/licensed-sw/pharaoh/

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Maria no dio una bofetada a	a la	bruja verde
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#### • Build translation left to right

- select foreign words to be translated



- Build translation left to right
  - select foreign words to be translated
  - find English phrase translation
  - add English phrase to end of partial translation

Maria	no	dio	una	bofetada	a	la	bruja	verde
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- Build translation left to right
  - select foreign words to be translated
  - find English phrase translation
  - add English phrase to end of partial translation
  - mark foreign words as translated



#### • One to many translation



• Many to one translation



• Many to one translation



• Reordering



• Translation finished

# **Translation Options**

Maria	no	dio	una	bofetada	a	la	bruja	verde
<u>Mary</u>	not didnot	give	<u> </u>	slap	t.o by	the	wit.ch green	green witch
	<u>no</u>	slap			to the			
	did_no	t give			t	. <u>o</u>		
			sl	ар		the t	witch	

• Look up possible phrase translations

- many different ways to segment words into phrases
- many different ways to translate each phrase

Maria	no	dio	una	bofetada	a	la	bruja	verde
<u>Mary</u>	not	give	aa_s	<u>slap</u> lap	<u>    t.o    </u> bv	the	wit.ch green	green witch
	no		slap	-	t.o	the		
	did_no	t give			t	0		
					tł	ne		
			sl	ар		the v	witch	



#### • Start with null hypothesis

- e: no English words
- f: no foreign words covered
- p: probability 1

Maria	no	dio	una	bofetada	a	la	bruja	verde
Mary	not 	give	aas	<u>slap</u>	<u>     t.o                               </u>	<u>the</u>	witch green	green witch
	no	slap		t.o	the			
	<u> </u>	t give			t	o		
		slap				the w	witch	

e:	e	: Mary
f:	e	: *
p: 1	P	: .534

- Pick translation option
- Create hypothesis
  - e: add English phrase Mary
  - f: first foreign word covered
  - p: probability 0.534



• Add another hypothesis



• Further hypothesis expansion



- ... until all foreign words covered
  - find best hypothesis that covers all foreign words
  - backtrack to read off translation



- Adding more hypothesis
- $\Rightarrow$  Explosion of search space

## **Explosion of Search Space**

- Number of hypotheses is exponential with respect to sentence length
- $\Rightarrow$  Decoding is NP-complete [Knight, 1999]
- $\Rightarrow$  Need to reduce search space
  - risk free: hypothesis recombination
  - risky: histogram/threshold pruning



• Different paths to the same partial translation



• Different paths to the same partial translation

#### $\Rightarrow$ Combine paths

- drop weaker hypothesis
- keep pointer from worse path



- Recombined hypotheses do not have to match completely
- No matter what is added, weaker path can be dropped, if:
  - last two English words match (matters for language model)
  - foreign word coverage vectors match (effects future path)



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#### $\Rightarrow$ Combine paths

# Pruning

- Hypothesis recombination is not sufficient
- $\Rightarrow$  Heuristically discard weak hypotheses
  - Organize Hypothesis in stacks, e.g. by
    - same foreign words covered
    - same number of foreign words covered (Pharaoh does this)
    - same number of English words produced
  - Compare hypotheses in stacks, discard bad ones
    - histogram pruning: keep top n hypotheses in each stack (e.g., n=100)
    - threshold pruning: keep hypotheses that are at most  $\alpha$  times the cost of best hypothesis in stack (e.g.,  $\alpha$  = 0.001)

# **Comparing Hypotheses**

Comparing hypotheses with same number of foreign words covered



- Hypothesis that covers *easy* part of sentence is preferred
- $\Rightarrow$  Need to consider future cost

#### **Future Cost Estimation**

- Estimate cost to translate remaining part of input
- Step 1: find cheapest translation options
  - find cheapest translation option for each input span
  - compute translation model cost
  - estimate language model cost (no prior context)
  - ignore reordering model cost
- Step 2: compute cheapest cost
  - for each contiguous span:
  - find cheapest sequence of translation options
- Precompute and lookup
  - precompute future cost for each contiguous span
  - future cost for any coverage vector: sum of cost of each contiguous span of uncovered words
  - $\rightarrow$  no expensive computation during run time

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## **Experiments**

- Decoder has to be evaluated in terms of search errors
  - translation errors not due to search errors are a challenge to the translation model
  - do not rely on search errors for good translation quality!
- Experimental setup
  - German to English
  - Europarl training corpus (30 million words)
  - 1500 sentence test corpus (avg. length 28.9 words)
  - 3 Ghz Linux machine, needs 512 MB RAM
  - Focus: illustrate trade-off speed / search errors
- Not measuring true search error
  - it is not tractable to find truly best translation
  - ightarrow relative to best translation found with high beam and different settings

# **Threshold Pruning**

Threshold	0.0001	0.001	0.01	0.05	0.08
Time per Sentence	149 sec	119 sec	70 sec	27 sec	18 sec
Search Errors	-	+0%	+0%	+0%	+0%
Threshold	0.1	0.15	0.2	0.3	
Time per Sentence	15 sec	13 sec	10 sec	7 sec	
Search Errors	+1%	+3%	+6%	+12%	

- Low ratio of search errors for threshold  $\alpha \leq 0.1$
- Results depend on weights for models

# **Histogram Pruning**

Beam Size	1000	200	100	50	20	10	5
Time	15s	15s	14s	10s	9s	9s	7s
Search Errors	+1%	+1%	+2%	+4%	+8%	+20%	+35 %

• Low ratio of search errors for beam size  $n \geq 200$ 

### **Translation Table Entries per Input Phrase**

T-Table Limit	1000	500	200	100	50	20	10	5
Time	15.0s	7.6s	3.8s	1.9s	0.9s	0.4s	0.2s	0.1s
Search Errors	+1%	+1%	+1%	+1%	+1%	+2%	+7%	+18%

- Low ratio of search errors for limit of  $\geq 50$  entries in the translation table for each source language phrase
- About 1 second per sentence (30 words per second)
- Your mileage may vary

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#### Word Lattice Generation



- Search graph can be easily converted into a word lattice
  - can be further mined for n-best lists
  - ightarrow enables reranking approaches
  - $\rightarrow$  enables discriminative training



#### XML Interface

- Er erzielte <NUMBER english='17.55'>17,55</NUMBER>
  Punkte .
  - Add additional translation options
    - number translation
    - noun phrase translation [Koehn, 2003]
    - name translation
  - Additional options
    - provide multiple translations
    - provide probability distribution along with translations
    - allow bypassing of provided translations

# **Thank You!**

• Questions?