

Phrase-based Statistical Machine Translation



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(Lecture on decoding algorithm, B Pouliquen, Gian course, Varanasi)

Original slides available at

<http://ufal.mff.cuni.cz/mtm16/files/14-phrase-based-smt-ulrich-germann.pdf>

Decoding

based on slides originally by P. Koehn, edited by M. Huck (and possibly others)

Given the model, find the best translation

$$\mathbf{e}_{\text{best}} = \operatorname{argmax}_{\mathbf{e}} p(\mathbf{e} | \mathbf{f})$$

We use the “Viterbi approximation”

$$(a, \mathbf{e})_{\text{best}} = \operatorname{argmax}_{(a, \mathbf{e})} p(a, \mathbf{e} | \mathbf{f})$$

- This is a search problem - a big one.
 - Dynamic programming
 - Approximation (beam search)
 - Model restrictions (reordering)

Decoding

Maria	no	dio	una	bofetada	a	la	bruja	verde
Mary	not	give	a	slap	to	the	witch	green
	did not		a slap		by		green witch	
	no		slap		to the			
	did not give				to			
					the			
				slap		the witch		

- many different ways to *segment* the input sentence into phrases
- many different ways to *translate* each phrase

Hypothesis Expansion

Maria	no	dio	una	bofetada	a	la	bruja	verde
Mary	not	give	a	slap	to	the	witch	green
	did not		a slap		by		green witch	
	no		slap		to the			
	did not give				to			
					the			
				slap		the witch		

```
e:
f: -----
p: 1
```

- Start with **empty hypothesis**
 - e: no English words
 - f: no foreign words covered
 - p: probability 1

Hypothesis Expansion

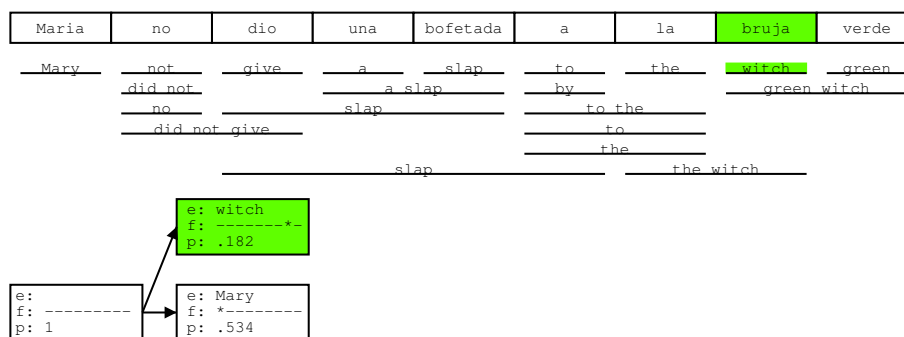
Maria	no	dio	una	bofetada	a	la	bruja	verde
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Mary	not	give	a	slap	to	the	witch	green
	did not		a slap		by		green witch	
	no		slap		to the			
	did not give				to			
					the			
				slap		the witch		

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

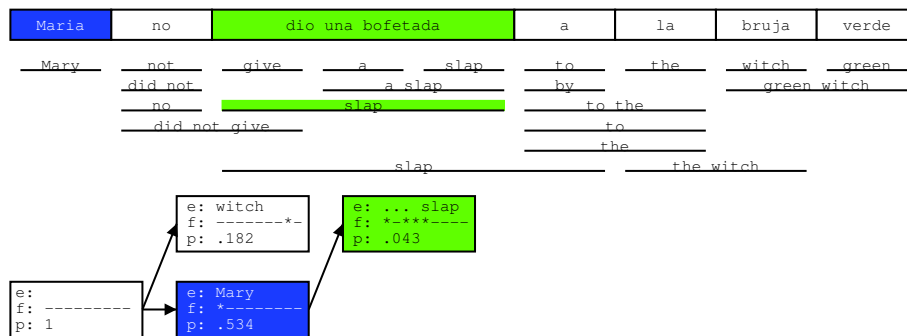
- Pick *translation option*
- Create *hypothesis*
 - e: add English phrase Mary
 - f: first foreign word covered
 - p: probability 0.534

Hypothesis Expansion



- Add another *hypothesis*

Hypothesis Expansion



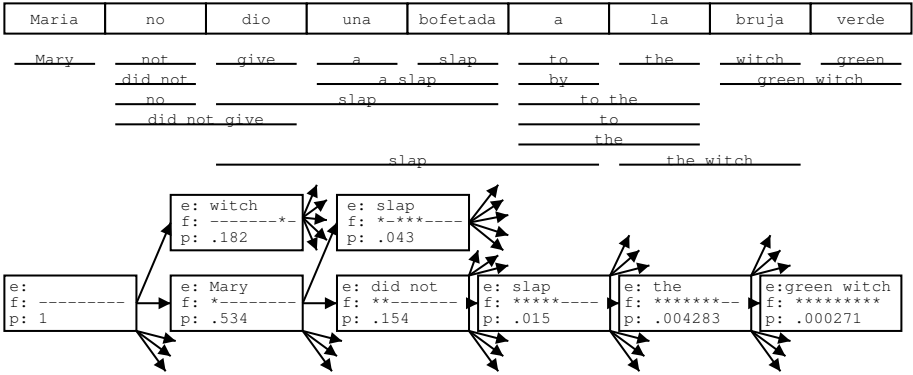
- Further *hypothesis expansion*

Hypothesis Expansion



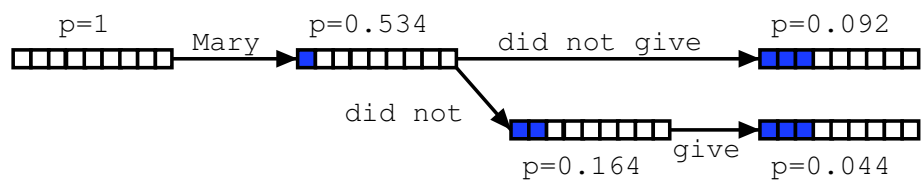
- ... until all foreign words covered
 - find *best hypothesis* that covers all foreign words
 - *backtrack* to read off translation

Hypothesis Expansion



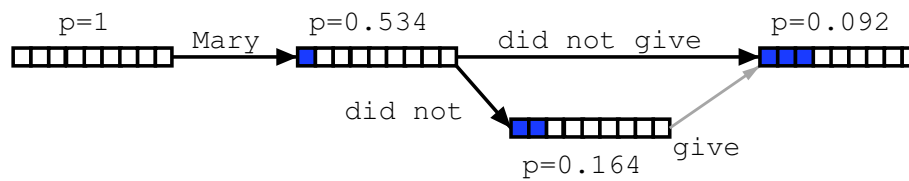
Adding more hypothesis \Rightarrow *Explosion* of search space

Hypothesis Recombination



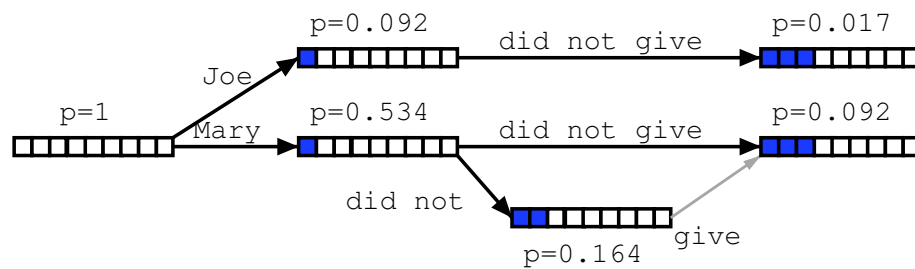
- Different paths to the *same* partial translation

Hypothesis Recombination



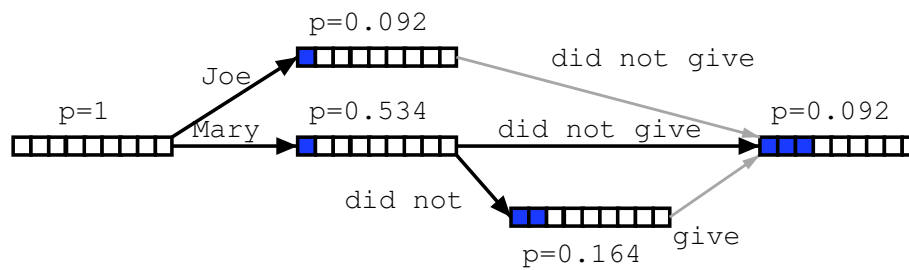
- Different paths to the same partial translation
- ⇒ *Combine paths*
- *drop weaker path*
 - keep pointer from weaker path (for lattice generation)

Hypothesis Recombination



- Recombined hypotheses do *not* have to *match completely*
- No matter what is added, weaker path can be dropped, if:
 - *last $n - 1$ English words match* (matters for language model)
 - *foreign word coverage vectors match* (affects future path)

Hypothesis Recombination



- Recombined hypotheses do not have to match completely
- No matter what is added, weaker path can be dropped, if:
 - last $n - 1$ English words match (matters for language model)
 - foreign word coverage vectors match (effects future path)

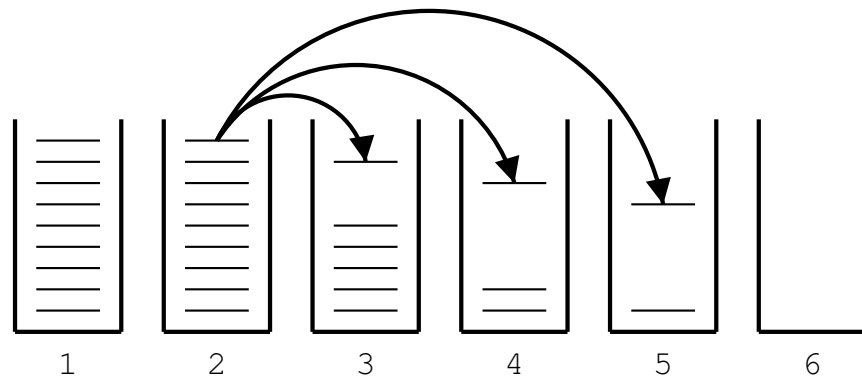
⇒ *Combine paths*

Beam Search

heuristically *discard* weak hypotheses early

- it is better to organize hypotheses in stacks (actually: priority queues), e.g. by
 - *same* foreign words covered
 - *same number* of foreign words covered
- compare hypotheses in stacks, discard bad ones
 - **histogram pruning**: keep top k hypotheses in each stack (e.g., $k=100$)
 - **threshold pruning**: keep hypotheses that are at most α times the cost of best hypothesis in stack (e.g., $\alpha = 0.001$)

Hypothesis Stacks



- Organization of hypotheses into stacks
 - here: based on *number of foreign words* translated
 - during translation all hypotheses from one stack are expanded
 - expanded hypotheses are placed into stacks

Comparing Hypotheses

- Comparing hypotheses with *same number of foreign words covered*

Maria no dio una bofetada a la bruja verde

┌───┐
└─┐
e: Mary did not
f: **-----
p: 0.154

**better
partial
translation**

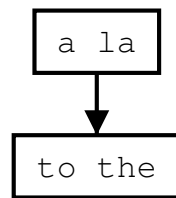
┌───┐
└─┐
e: the
f: -----**--
p: 0.354

**covers
easier part
--> lower cost**

- Hypothesis that covers *easy part* of sentence is preferred
⇒ Need to consider **future cost** of uncovered parts

Future Cost Estimation

Step 1: estimate future cost for each *translation option*

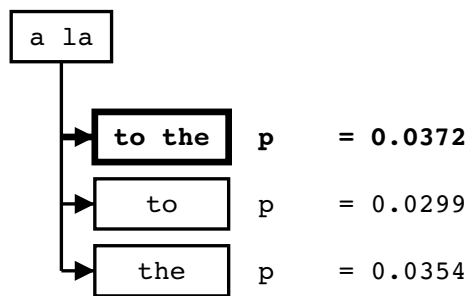


- look up translation model cost
- estimate language model cost (no prior context)
- ignore reordering model cost

$$\Rightarrow \text{LM} * \text{TM} = p(\text{to}) * p(\text{the}|\text{to}) * p(\text{to the}|\text{a la})$$

Future Cost Estimation

Step 2: find *cheapest cost* (highest probability) among translation options

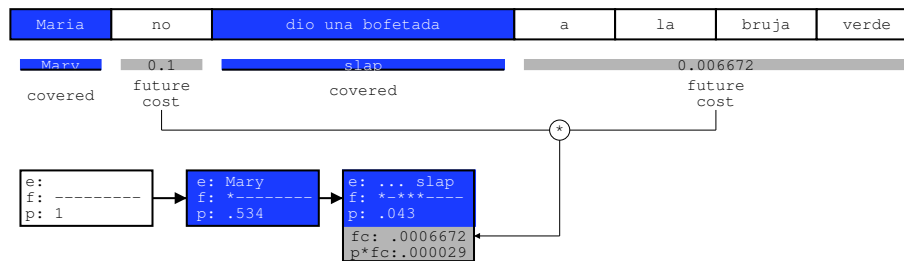


Future Cost Estimation

Step 3: Find *lowest future cost* for each possible span

- Cost of translation option for that span, *or*
 - Sum of costs of covering subspans
- ⇒ Pre-compute future costs, bottom up., via dynamic programming.

Future Cost Estimation: Application

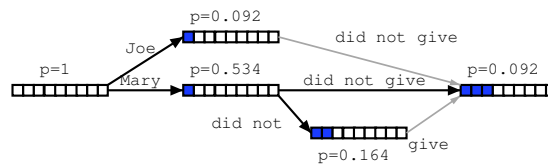


- Use future cost estimates when *pruning* hypotheses
- For each *uncovered continuous span*:
 - look up *future costs* for each maximal contiguous uncovered span
 - *add* to actually accumulated cost for translation option for pruning

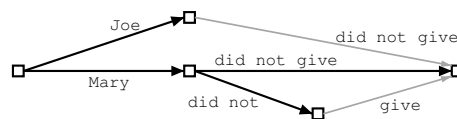
Limits on Reordering

- Reordering may be **limited**
 - **Monotone** translation: No reordering at all
 - Only phrase movements of at most d words
- Reordering limits *speed* up search (polynomial instead of exponential)
- Current reordering models are weak, so limits *improve* translation quality

Word Lattice Generation



- Search graph can be easily converted into a word lattice
 - can be further mined for N-best lists
 - ⇒ enables reranking approaches
 - ⇒ enables discriminative training



Sample N-Best List

- Simple N-best list:

Translation	Reordering	LM	TM	WordPenalty	Score
this is a small house	0	-27.0908	-1.83258	-5	-28.9234
this is a little house	0	-28.1791	-1.83258	-5	-30.0117
it is a small house	0	-27.108	-3.21888	-5	-30.3268
it is a little house	0	-28.1963	-3.21888	-5	-31.4152
this is an small house	0	-31.7294	-1.83258	-5	-33.562
it is an small house	0	-32.3094	-3.21888	-5	-35.5283
this is an little house	0	-33.7639	-1.83258	-5	-35.5965
this is a house small	-3	-31.4851	-1.83258	-5	-36.3176
this is a house little	-3	-31.5689	-1.83258	-5	-36.4015
it is an little house	0	-34.3439	-3.21888	-5	-37.5628
it is a house small	-3	-31.5022	-3.21888	-5	-37.7211
this is an house small	-3	-32.8999	-1.83258	-5	-37.7325
it is a house little	-3	-31.586	-3.21888	-5	-37.8049
this is an house little	-3	-32.9837	-1.83258	-5	-37.8163
the house is a little	-7	-28.5107	-2.52573	-5	-38.0364
the is a small house	0	-35.6899	-2.52573	-5	-38.2156
is it a little house	-4	-30.3603	-3.91202	-5	-38.2723
the house is a small	-7	-28.7683	-2.52573	-5	-38.294
it 's a small house	0	-34.8557	-3.91202	-5	-38.7677
this house is a little	-7	-28.0443	-3.91202	-5	-38.9563
it 's a little house	0	-35.1446	-3.91202	-5	-39.0566
this house is a small	-7	-28.3018	-3.91202	-5	-39.2139

Summary

- Left-to-right decoding as search
- Hypothesis recombination
- Pruning
- Future cost estimation
- Word lattices and n -best lists