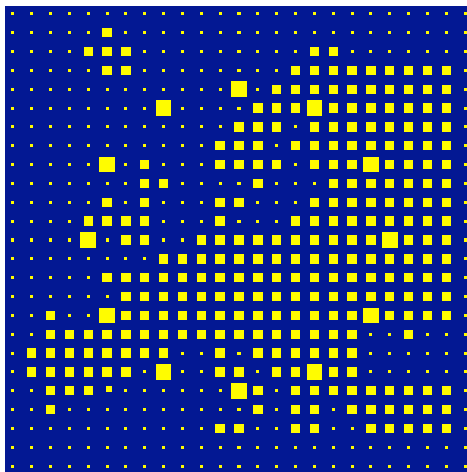


Intro to Statistical MT



EuroMatrix
MT Marathon
Chris Callison-Burch

Various approaches

- Word-for-word translation
- Syntactic transfer
- Interlingual approaches
- Controlled language
- Example-based translation
- Statistical translation

Advantages of SMT

- Data driven
- Language independent
- No need for staff of linguists or language experts
- Can prototype a new system quickly and at a very low cost

Statistical machine translation

- Find most probable English sentence given a foreign language sentence
- Automatically align words and phrases within sentence pairs in a parallel corpus
- Probabilities are determined automatically by training a statistical model using the parallel corpus

Parallel corpus

what is more , the relevant cost dynamic is completely under control .	im übrigen ist die diesbezügliche kostenentwicklung völlig unter kontrolle .
sooner or later we will have to be sufficiently progressive in terms of own resources as a basis for this fair tax system .	früher oder später müssen wir die notwendige progressivität der eigenmittel als grundlage dieses gerechten steuersystems zur sprache bringen .
we plan to submit the first accession partnership in the autumn of this year .	wir planen , die erste beitrittspartnerschaft im herbst dieses jahres vorzulegen .
it is a question of equality and solidarity .	hier geht es um gleichberechtigung und solidarität .
the recommendation for the year 1999 has been formulated at a time of favourable developments and optimistic prospects for the european economy .	die empfehlung für das jahr 1999 wurde vor dem hintergrund günstiger entwicklungen und einer für den kurs der europäischen wirtschaft positiven perspektive abgegeben .
that does not , however , detract from the deep appreciation which we have for this report .	im übrigen tut das unserer hohen wertschätzung für den vorliegenden bericht keinen abbruch .

Probabilities

- Find most probable English sentence given a foreign language sentence

$$p(e|f)$$

$$\hat{e} = \arg \max_e p(e|f)$$

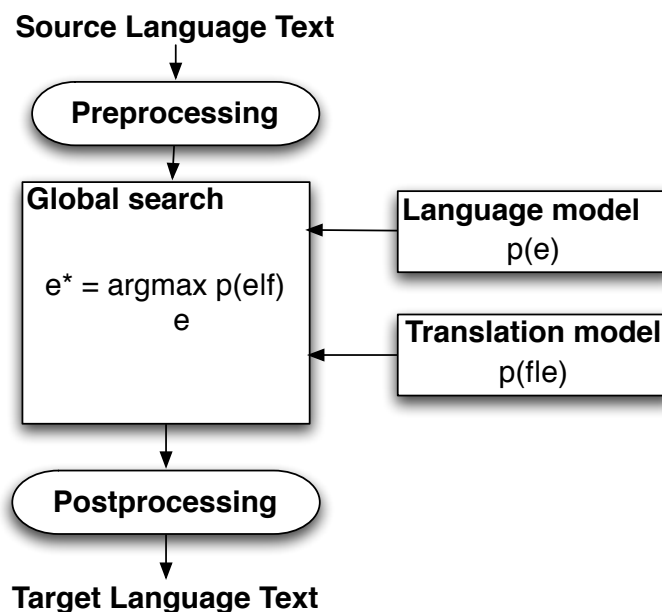
$$p(e|f) = \frac{p(e)p(f|e)}{p(f)}$$

$$\hat{e} = \arg \max_e p(e)p(f|e)$$

What the probabilities represent

- $p(e)$ is the "Language model"
 - Assigns a higher probability to fluent / grammatical sentences
 - Estimated using monolingual corpora
- $p(f|e)$ is the "Translation model"
 - Assigns higher probability to sentences that have corresponding meaning
 - Estimated using bilingual corpora

For people who don't like equations



Language Model

- Component that tries to ensure that words come in the right order
- Some notion of grammaticality
- Standardly calculated with a trigram language model, as in speech recognition
- Could be calculated with a statistical grammar such as a PCFG

Trigram language model

- $p(\text{I like bungee jumping off high bridges}) =$
 $p(\text{I} \mid \langle s \rangle \langle s \rangle) *$
 $p(\text{like} \mid \text{I} \langle s \rangle) *$
 $p(\text{bungee} \mid \text{I like}) *$
 $p(\text{jumping} \mid \text{like bungee}) *$
 $p(\text{off} \mid \text{bungee jumping}) *$
 $p(\text{high} \mid \text{jumping off}) *$
 $p(\text{bridges} \mid \text{off high}) *$
 $p(\langle /s \rangle \mid \text{high bridges}) *$
 $p(\langle /s \rangle \mid \text{bridges} \langle /s \rangle)$

Calculating Language Model Probabilities

- Unigram probabilities

$$p(w_1) = \frac{\textit{count}(w_1)}{\textit{total words observed}}$$

Calculating Language Model Probabilities

- Bigram probabilities

$$p(w_2|w_1) = \frac{\textit{count}(w_1w_2)}{\textit{count}(w_1)}$$

Calculating Language Model Probabilities

- Trigram probabilities

$$p(w_3|w_1w_2) = \frac{\text{count}(w_1w_2w_3)}{\text{count}(w_1w_2)}$$

Calculating Language Model Probabilities

- Can take this to increasingly long sequences of n-grams
- As we get longer sequences it's less likely that we'll have ever observed them

Backing off

- Sparse counts are a big problem
- If we haven't observed a sequence of words then the count = 0
- Because we're multiplying the n-gram probabilities to get the probability of a sentence the whole probability = 0

Backing off

$$\begin{aligned} &.8 * p(w_3 | w_1 w_2) + \\ &.15 * p(w_3 | w_2) + \\ &.049 * p(w_3) + \\ &.001 \end{aligned}$$

- Avoids zero probs

Translation model

- $p(f|e)$... the probability of some foreign language string given a hypothesis English translation
- f = Ces gens ont grandi, vécu et oeuvré des dizaines d'années dans le domaine agricole.
- e = *Those people have grown up, lived and worked many years in a farming district.*
- e = *I like bungee jumping off high bridges.*

Translation model

- How do we assign values to $p(f|e)$?

$$p(f|e) = \frac{\text{count}(f, e)}{\text{count}(e)}$$

- Impossible because sentences are novel, so we'd never have enough data to find values for all sentences.

Alignment probabilities

- So we can calculate translation probabilities by way of these alignment probabilities

$$p(f|e) = \sum_a p(a, f|e)$$

- Now we need to define $p(a, f | e)$

$$p(a, f|e) = \prod_{j=1}^m t(f_j|e_i)$$

Calculating $t(f_j|e_i)$

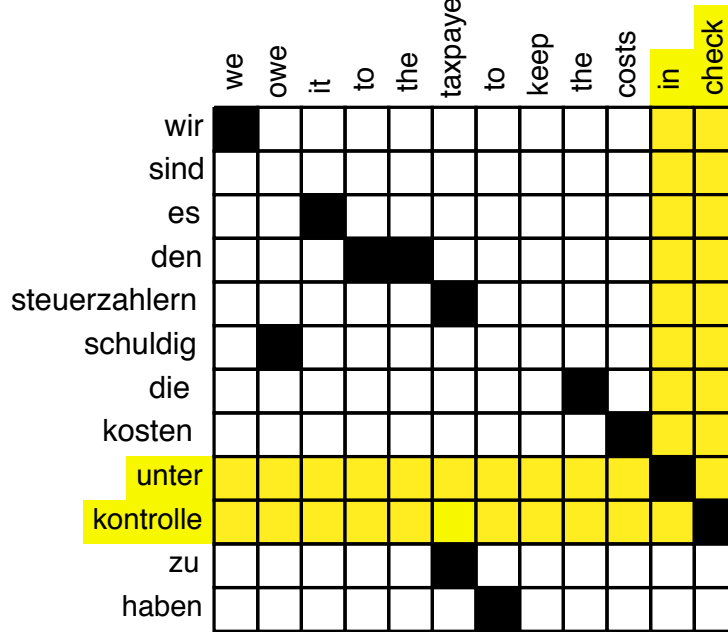
	Those	people	have	grown	up	,	lived	and	worked	many	years	in	a	farming	district	
Ces	■								■							
gens		■														
ont			■													
grandi				■												
,					■											
vécu						■										
et							■									
oeuvre	■	■	■	■	■	■	■	■	■							
des										■						
dizaines											■					
d'												■				
années													■			
dans														■		
le															■	
domaine																■
agricole																■

- Counting! I told you probabilities were easy!

$$= \frac{\text{count}(f_j, e_i)}{\text{count}(e_i)}$$

- worked... fonctionné, travaillé, marché, oeuvre
- 100 times total 13 with this f. 13%

Phrase Translation Probabilities

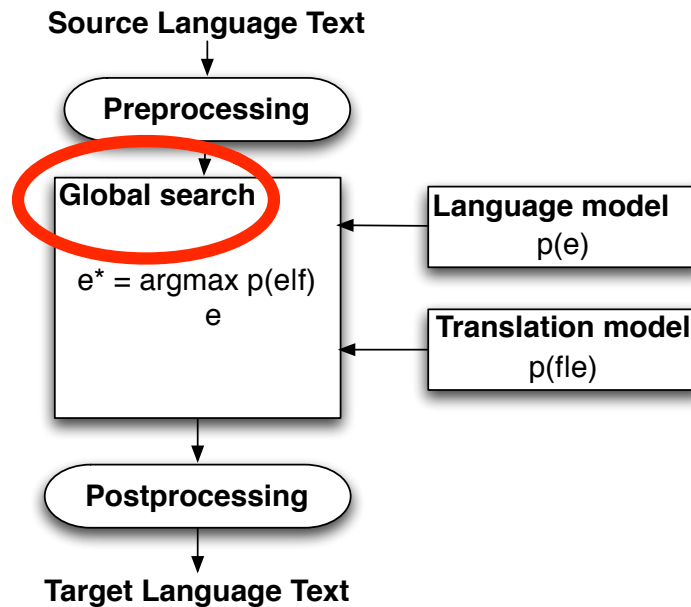


Phrase Table

- Exhaustive table of source language phrases paired with their possible translations into the target language, along with probabilities

das thema	the issue	.51
	the point	.38
	the subject	.21

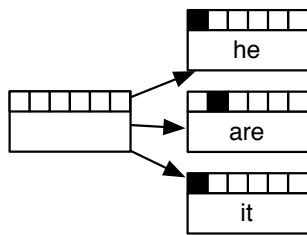
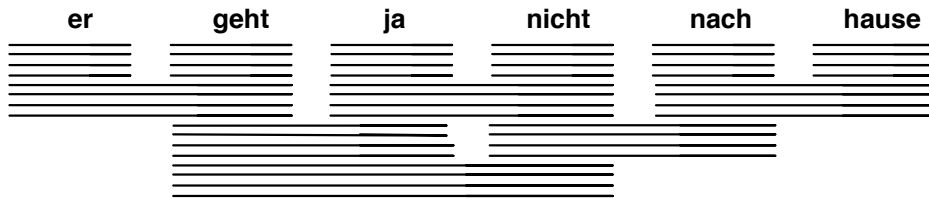
``Diagram Number 1''



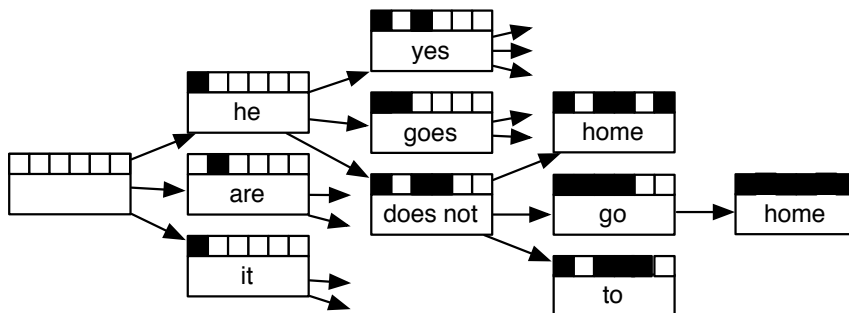
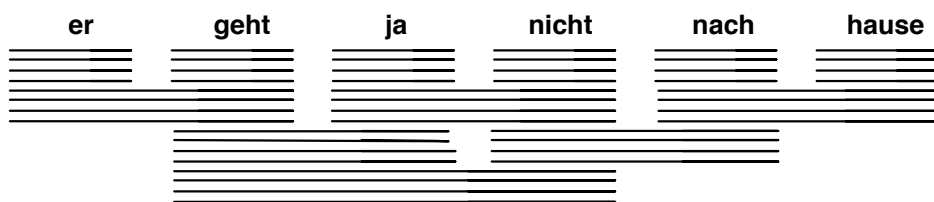
The Search Process AKA ``Decoding''

- Look up all translations of every source phrase, using the phrase table
- Recombine the target language phrases that maximizes the translation model probability * the language model probability
- This search over all possible combinations can get very large so we need to find ways of limiting the search space

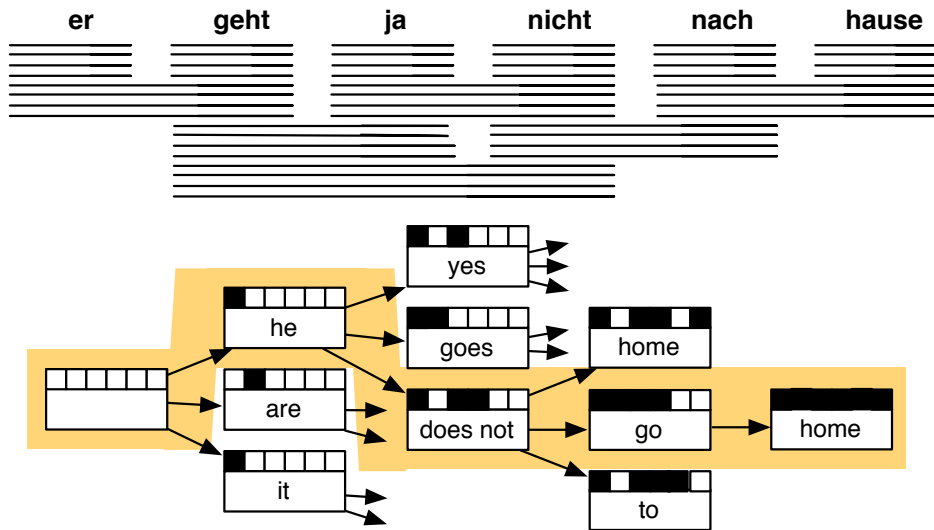
Search



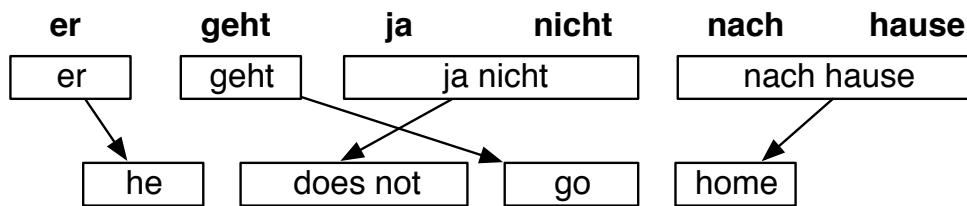
Search



Search



Best Translation



The Search Space

- In the end the item which covers all of the source words and which has the highest probability wins!
- That's our best translation

$$\hat{e} = \arg \max_e p(e)p(f|e)$$

- And there was much rejoicing!

Wrap-up: SMT is data driven

- Learns translations of words and phrases from parallel corpora
- Associate probabilities with translations empirically by counting co-occurrences in the data
- Estimates of probabilities get more accurate as size of the data increases

Wrap-up: SMT is language independent

- Can be applied to any language pairs that we have a parallel corpus for
- The only linguistic thing that we need to know is how to split into sentences, words
- Don't need linguists and language experts to hand craft rules because it's all derived from the data

Wrap-up: SMT is cheap and quick to produce

- Low overhead since we aren't employing anyone
- Computers do all the heavy lifting / statistical analysis of the data for us
- Can build a system in hours or days rather than months or years