# Statistical Machine Translation Lecture 5 Syntax-Based Models

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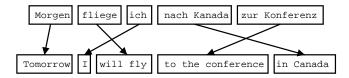
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#### Phrase-Based Translation Model



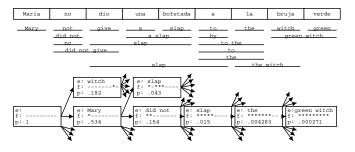
- Foreign input is segmented in phrases
  - any sequence of words, not necessarily linguistically motivated
- Each phrase is translated into English
- Phrases are reordered

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# Search Space for Decoding Too Big



- Explosion of search space
- ⇒ Pruning, Beam Search

#### Outline

• Reminder: Modeling and Decoding

• Why Syntax?

• Yamada and Knight: translating into trees

• Wu: tree-based transfer

• Chiang: hierarchical transfer

• Koehn: clause structure

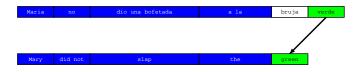
Other approaches

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### **Decoding**

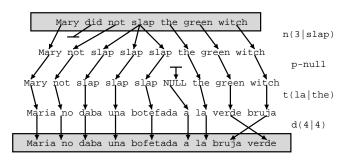


Decoding process builds an English translation left to right,
 by picking foreign phrases to translate into English phrases

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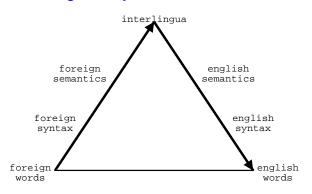
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### **Word-Based Translation Model**



- Translation process is broken up into small step: word translation, reordering, duplication, insertion
- Decoding can be done similarly to phrase-based decoding

### The Challenge of Syntax



The classical machine translation pyramid

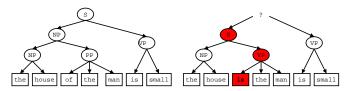
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### Syntactic Language Model

- ullet Good syntax tree o good English
- · Allows for long distance constraints



· Left translation preferred by syntactic LM

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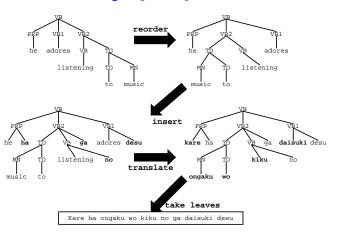
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# Yamada and Knight [2001]

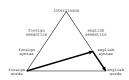


## **Advantages of Syntax-Based Translation**

- Reordering for syntactic reasons
  - e.g., move German object to end of sentence
- Better explanation for function words
  - e.g., prepositions, determiners
- · Conditioning to syntactically related words
  - translation of verb may depend on subject or object
- Use of syntactic language models

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### String to Tree Translation



- Use of English syntax trees [Yamada and Knight, 2001]
  - exploit rich resources on the English side
  - obtained with statistical parser [Collins, 1997]
  - flattened tree to allow more reorderings
  - works well with syntactic language model

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# **Reordering Table**

Original Order	Reordering	p(reorder original)
PRP VB1 VB2	PRP VB1 VB2	0.074
PRP VB1 VB2	PRP VB2 VB1	0.723
PRP VB1 VB2	VB1 PRP VB2	0.061
PRP VB1 VB2	VB1 VB2 PRP	0.037
PRP VB1 VB2	VB2 PRP VB1	0.083
PRP VB1 VB2	VB2 VB1 PRP	0.021
VB TO	VB TO	0.107
VB TO	TO VB	0.893
TO NN	TO NN	0.251
TO NN	NN TO	0.749

ongaku

kiku

**Decoding as Parsing** 

Chart Parsing

PRP he

kare

### **Decoding as Parsing**

Chart Parsing



- Pick Japanese words
- Translate into tree stumps

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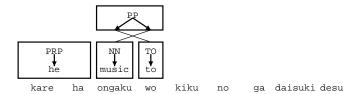
• Pick Japanese words

Translate into tree stumps

ga daisuki desu

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### **Decoding as Parsing**

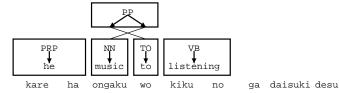


• Adding some more entries...

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# **Decoding as Parsing**



Combine entries

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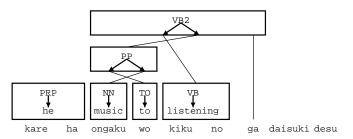
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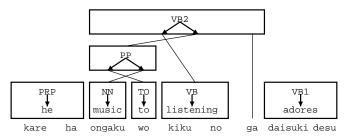
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# **Decoding as Parsing**

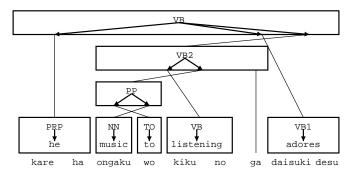


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# **Decoding as Parsing**



## **Decoding as Parsing**



· Finished when all foreign words covered

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#### Is the Model Realistic?

- Do English trees match foreign strings?
- Crossings between French-English [Fox, 2002]
  - 0.29-6.27 per sentence, depending on how it is measured
- Can be reduced by
  - flattening tree, as done by [Yamada and Knight, 2001]
  - detecting phrasal translation
  - special treatment for small number of constructions
- Most coherence between dependency structures

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# **Syntax Trees**



English binary tree

### Yamada and Knight: Training

- Parsing of the English side
  - using Collins statistical parser
- EM training
  - translation model is used to map training sentence pairs
  - EM training finds low-perplexity model
  - → unity of training and decoding as in IBM models

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### **Inversion Transduction Grammars**

- Generation of both English and foreign trees [Wu, 1997]
- Rules (binary and unary)
  - $-A \to A_1 A_2 ||A_1 A_2||$
  - $A \rightarrow A_1 A_2 || A_2 A_1$
  - $-A \rightarrow e \| f$
  - $-A \rightarrow e \| *$
  - $-A \rightarrow * || f$
- ⇒ Common binary tree required
  - limits the complexity of reorderings

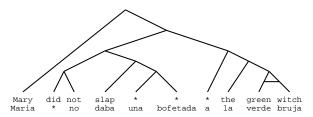
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# Syntax Trees (2)



Spanish binary tree

### Syntax Trees (3)



· Combined tree with reordering of Spanish

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### Chiang: Hierarchical Phrase Model

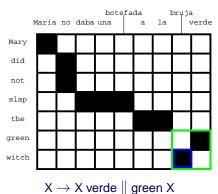
- Chiang [ACL, 2005] (best paper award!)
  - context free bi-grammar
  - one non-terminal symbol
  - right hand side of rule may include non-terminals and terminals
- Competitive with phrase-based models in 2005 DARPA/NIST evaluation

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# Learning Hierarchical Rules



#### **Inversion Transduction Grammars**

- Decoding by parsing (as before)
- Variations
  - may use real syntax on either side or both
  - may use multi-word units at leaf nodes
- Reordering constraints of ITG used in phrase-based systems

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# Types of Rules

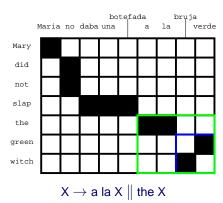
- Word translation
  - $X \rightarrow$  maison  $\parallel$  house
- Phrasal translation
  - X → daba una bofetada | slap
- Mixed non-terminal / terminal
  - $X \rightarrow X$  bleue  $\parallel$  blue X
  - $\ \, X \rightarrow ne \, X \, pas \, \| \, not \, X$
  - X  $\rightarrow$  X1 X2  $\parallel$  X2 of X1
- Technical rules
  - $S \rightarrow S X \parallel S X$
  - $\ S \to X \parallel X$

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# Learning Hierarchical Rules



#### Details

- Too many rules
  - ightarrow filtering of rules necessary
- Efficient parse decoding possible
  - hypothesis stack for each span of foreign words
  - only one non-terminal  $\rightarrow$  hypotheses comparable
  - length limit for spans that do not start at beginning

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### Clause Level Restructuring

- Why clause structure?
  - languages differ vastly in their clause structure (English: SVO, Arabic: VSO, German: fairly free order; a lot details differ: position of adverbs, sub clauses, etc.)
  - large-scale restructuring is a problem for phrase models
- Restructuring
  - reordering of constituents (main focus)
  - add/drop/change of function words
- ACL 2005 paper [Collins, Koehn, Kucerova]

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# Reordering When Translating

```
PPER-SB
              Ich
     VAFIN-HD
              werde
Ihnen
                                          will
     PPER-DA
                                          VOII
     NP-OA
               ART-OA
               ADJ-NK
                       entsprechenden
                                           corresponding
               NN-NK
                        Anmerkungen
                                           comments
     VVFIN
               aushaendigen
                                          pass on
S-MO KOUS-CP
               damit
                                          so that
     PPER-SB
               Sie
                                          you
     PDS-OA
               das
                                          that
     ADJD-MO
               eventuell
                                          perhaps
     PP-MO
               APRD-MO
                        bei
               ART-DA
                         der
                                           the
               NN-NK
                         Abstimmung
     VVINF
               uebernehmen
                                          include
ġ.
```

- Reordering when translating into English
  - tree is flattened
  - clause level constituents line up

### Syntax-Aided Phrase-Based MT [Koehn]

- Approach:
  - stick with phrase-based system
  - special treatment for special syntactic problems
- Noun Phrase Translation
- Clause Level Restructuring

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#### Clause Structure

```
PPER-SB Ich
         werde will
PPER-DA Ihnen
VAFIN-HD
VP-OC
                                                                  MATN
         NP-OA
                  ART-OA
ADJ-NK
                                the
                                                                  CLAUSE
                          entsprechenden
                                            corresponding
                          Anmerkungen con
digen pass on
         VVFIN
                  KOUS-CP damit
                                  so that
                         SIIR-
                                                                 ORDINATE
                                                                 CLAUSE
                                                 ung vote
                          VVINF
```

- Syntax tree from German parser
  - statistical parser by Amit Dubay, trained on TIGER treebank

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# Clause Level Reordering

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```
PPER-SB
              Ich
     VAFIN-HD werde
                                             will
                                             you
               Ihnen
                                          - 4
                       die
                                             the
     NP-OA
               ART-OA
                                          - 5
               ADJ-NK
                       entsprechenden
                                              corresponding
               NN-NK
                       Anmerkungen
                                              comments
                                             pass on
     VVFIN
               aushaendigen
S-MO KOUS-CP
              damit
     PPER-SB
              Sie
                                             you
     PDS-OA
                                             that
               das
     ADJD-MO
               eventuel1
                                             perhaps
               APRD-MO
     PP-MO
               ART-DA
                        der
                                              the
               NN-NK
                        Abstimmung
                                              vote
     VVINF
                                           5
                                             include
               uebernehmen
                                             can
     VMFIN
               koennen
```

- Clause level reordering is a well defined task
  - label German constituents with their English order
  - done this for 300 sentences, two annotators, high agreement

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#### Syntax-Based Statistical Machine Translation

### Systematic Reordering German → English

- Many types of reorderings are systematic
  - move verb group together
  - subject verb object
  - move negation in front of verb

#### ⇒ Write rules by hand

- apply rules to test and training data
- train standard phrase-based SMT system

System	BLEU
baseline system	25.2%
with manual rules	26.8%

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### Other Syntax-Based Approaches

- ISI: extending work of Yamada/Knight
  - more complex rules
  - performance approaching phrase-based
- Prague: Translation via dependency structures
  - parallel Czech-English dependency treebank
  - tecto-grammatical translation model [EACL 2003]
- U.Alberta/Microsoft: treelet translation
  - translating from English into foreign languages
  - using dependency parser in English
  - project dependency tree into foreign language for training
  - map parts of the dependency tree ("treelets") into foreign languages

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# Syntax: Does it help?

- Not yet
  - best systems still phrase-based, treat words as tokens
- Well, maybe...
  - work on reordering German
  - automatically trained tree transfer systems promising
- Why not yet?
  - if real syntax, we need good parsers are they good enough?
  - syntactic annotations add a level of complexity
  - ightarrow difficult to handle, slow to train and decode
  - few researchers good at statistical modeling and understand syntactic theories

#### **Improved Translations**

• we must also this criticism should be taken seriously .

→ we must also take this criticism seriously

- i am with him that it is necessary, the institutional balance by means of a political revaluation of both the commission and the council to maintain.
- i agree with him in this, that it is necessary to maintain the institutional balance by means of a political revaluation of both the commission and the council.
- thirdly, we believe that the principle of differentiation of negotiations note.
- $\,\,\,\,\,\,\,\,\,\,\,\,\,\,\,\,\,\,\,$  thirdly , we maintain the principle of differentiation of negotiations .
- perhaps it would be a constructive dialog between the government and opposition parties, social representative a positive impetus in the right direction
- perhaps a constructive dialog between government and opposition parties and social representative could give a positive impetus in the right direction.

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### Other Syntax-Based Approaches (2)

- Reranking phrase-based SMT output with syntactic features
  - create n-best list with phrase-based system
  - POS tag and parse candidate translations
  - rerank with syntactic features
  - see [Koehn, 2003] and JHU Workshop [Och et al., 2003]
- JHU Summer workshop 2005
  - final presentations this week
  - tools for syntax-based SMT

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