Empirical Methods in Natural Language Processing Lecture 18 Machine translation (V): Syntax-Based Models

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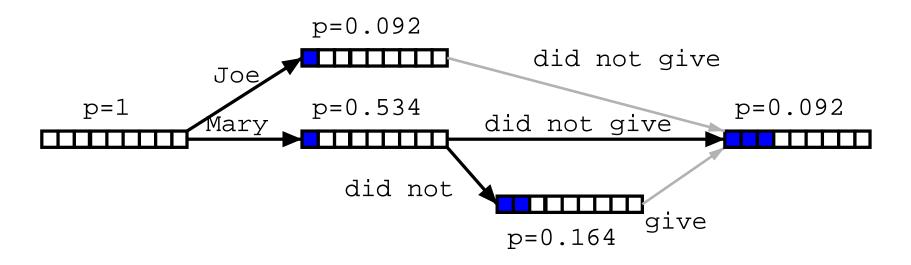
6 March 2008





Phrase-based SMT

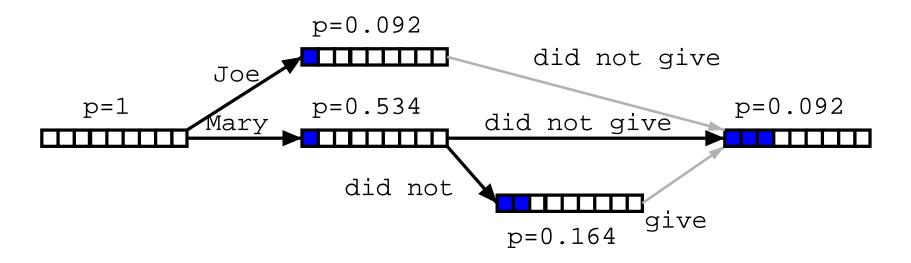
- Already works pretty well.
- Are there any problems that we need to solve here?





Phrase-based SMT

- *Computational*: computing all possible reorderings is NP-complete.
- *Linguistic*: language is not finite-state.





Syntax-based SMT

- What's going on here? A whole lot of things...
- Chiang (2005) makes a distinction between *formally* syntax-based and *linguistically* syntax-based.

		formally syntax-based
	phrase-based	hierarchical phrase-based
		[Chiang 2005]
linguistically	reordering + phrase-based	syntax-based SCFG
syntax-based	[Collins et al. 2005]	[Yamada & Knight 2002]



Linguistic Advantages of Syntax-Based Translation

- *Generalized* 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*
 - ensuring grammatical output

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Clause Level Restructuring [Collins et al.]

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Clause Level Restructuring [Collins et al.]

- 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*
- Details see [Collins, Kucerova and Koehn, ACL 2005]

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Clause Structure

S PPER-SB Ich Т VAFIN-HD werde will VP-OC PPER-DA Ihnen you MAIN NP-OA ART-OA die the CLAUSE 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 VP-OC das that SUB-ADJD-MO eventuell perhaps PP-MO ORDINATE APRD-MO bei in ART-DA der the CLAUSE NN-NK Abstimmung vote VVINF uebernehmen include VMFIN koennen can \$. .

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

informatics



Reordering When Translating

S	PPER-SB VAFIN-HD PPER-DA NP-OA VVFIN	Ihnen	die entsprechenden Anmerkungen digen	I will you the corresponding comments pass on
\$,		ausitaette	rigen	pass on
		damit		' an that
S-MO	KOUS-CP			so that
	PPER-SB	Sie		you
	PDS-OA	das		that
	ADJD-MO	eventue]	11	perhaps $X \setminus$
	PP-MO	APRD-MO	bei	in)
	II MO	ART-DA	der	the
		NN-NK	Abstimmung	vote
	VVINF	ueberneł	ımen	include /
	VMFIN	koennen		can
\$				•

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



Clause Level Reordering

S PPER-SB Ich 1 I 2 will VAFIN-HD werde 4 you PPER-DA Ihnen the NP-OA die ART-OA - 5 corresponding ADJ-NK entsprechenden comments NN-NK Anmerkungen VVFIN aushaendigen 3 pass on \$, so that S-MO KOUS-CP damit 2 you PPER-SB Sie 6 that PDS-OA das perhaps ADJD-MO 4 eventuell in PP-MO APRD-MO bei 7 the ART-DA der vote Abstimmung NN-NK 5 include VVINF uebernehmen 3 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



Systematic Reordering German \rightarrow English

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



Improved Translations

- we must also this criticism should be taken seriously .
- $\rightarrow\,$ 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 .
- $\rightarrow\,$ 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 .
- $\rightarrow\,$ 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 .
- \rightarrow perhaps a constructive dialog between government and opposition parties and social representative could give a positive impetus in the right direction .



Other Linguistically Syntax-Based Approaches

- *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]
- Incorporate syntax into decoder [Tillman and Ney, 2003]
 - Add finite-state control structure to allow long-distance movement of verbs in German-English translation.



Formal Advantages of Syntax-Based Translation

- Foundation in *well-understood* models from formal language theory (theoretical computer science).
 - Maybe they have some use after all
- *Computational complexity* is (in principle) just as much as we need to model linguistic phenomena, and no more.
 - Polynomial even with full reordering.
 - Caveat: no easy trick to speed it up as with phrase-based models.
- *Apply advances* made algorithms for statistical parsing.
 - Earley, CKY, etc.



Synchronous Context-Free Grammars

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		[Chiang 2005]
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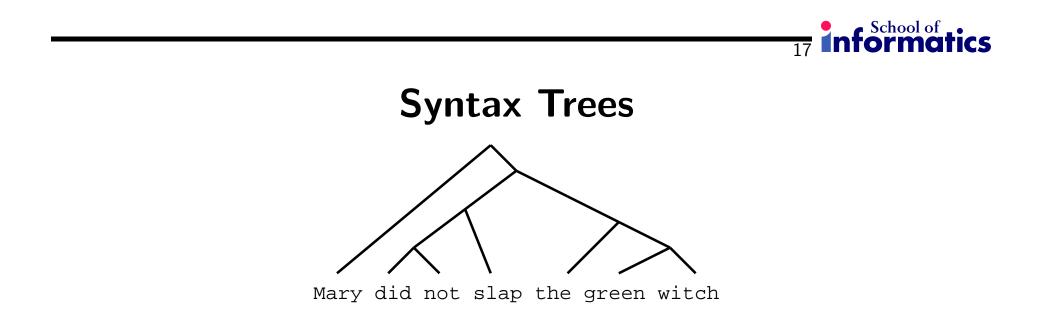
Synchronous Context-Free Grammars

- Finite-state transducers model *regular* language
- Regular tree transducers model *context-free* language
- Various guises of SCFG
 - Syntax-directed Transduction (Lewis and Stearns 1968)
 - Inversion Transduction Grammar (Wu 1995-1998)
 - Head Transducers (Alshawi et al. 2000)
 - Multitext Grammar (Melamed 2003)

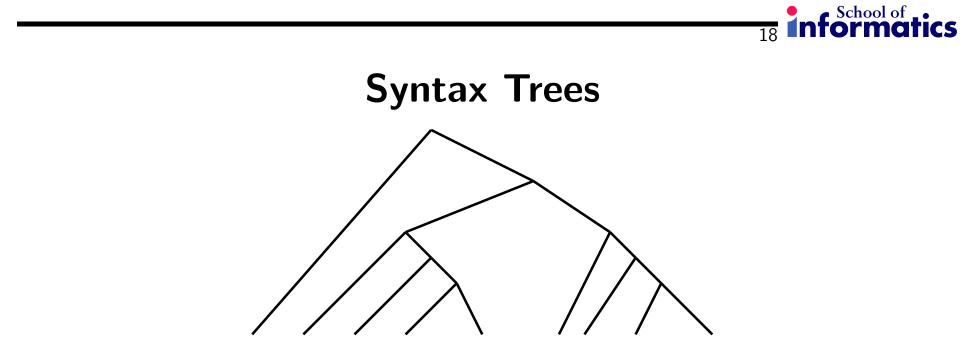


Inversion Transduction Grammars

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

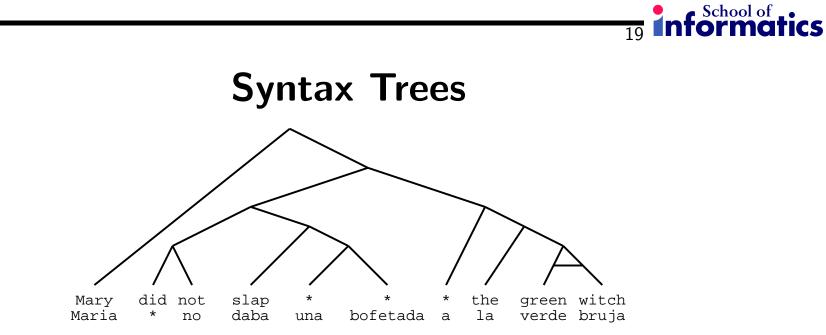


• English binary tree



Maria no daba una bofetada a la bruja verde

• Spanish binary tree



• Combined tree with reordering of Spanish



Chiang: Hierarchical Phrase-based 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

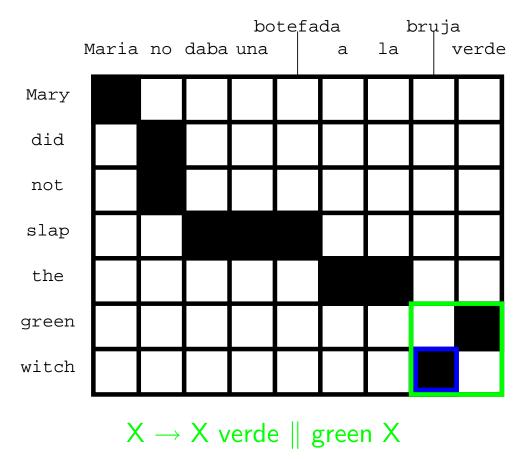


Types of Rules

- *Word* translation
 - $X \rightarrow$ maison \parallel house
- *Phrasal* translation
 - $X \rightarrow daba una bofetada | slap$
- Mixed non-terminal / terminal *hierarchial phrases*
 - $X \rightarrow X_1$ bleue \parallel blue X_1
 - $X \rightarrow$ ne X_1 pas \parallel not X_1
 - $X \to X_1 X_2 \parallel X_2 \text{ of } X_1$
- Technical rules
 - $S \to S_1 X_2 \parallel S_1 X_2$
 - $S \to X_1 \parallel X_1$

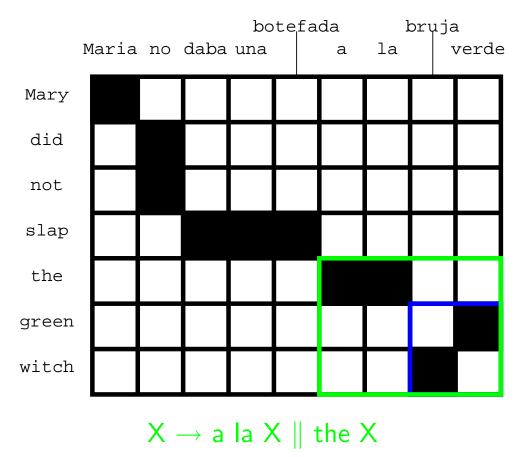


Learning Hierarchical Rules





Learning Hierarchical Rules





Details of Chiang's Model

- Too many rules
 - \rightarrow *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
 - m-gram language model integration increases complexity by $O(n^{2m})$



Language is not Context-Free!

- Maybe it's *mildly context-sensitive*?
 - Synchronous Tree-Adjoining Grammar [Shieber 1992, others]
 - Generalized Multitext Grammar [Melamed 2004]
- Various transducer formalisms [Knight & Graehl 2005] for overview.



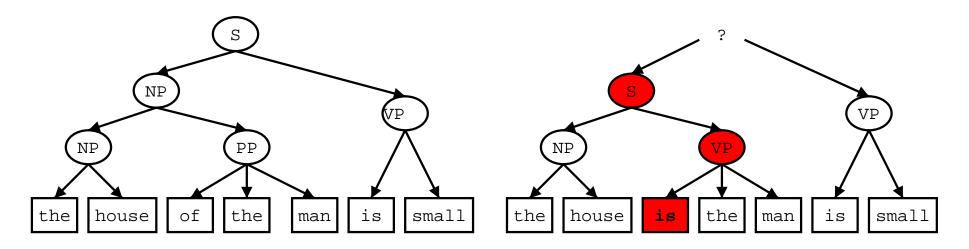
Syntactic Language Model

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Syntactic Language Model

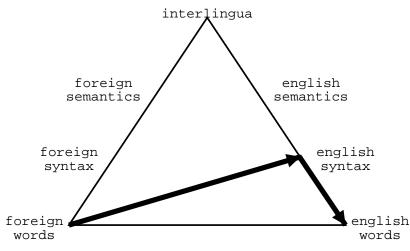
- Good syntax tree \rightarrow good English
- Allows for *long distance constraints*



• Left translation preferred by syntactic LM



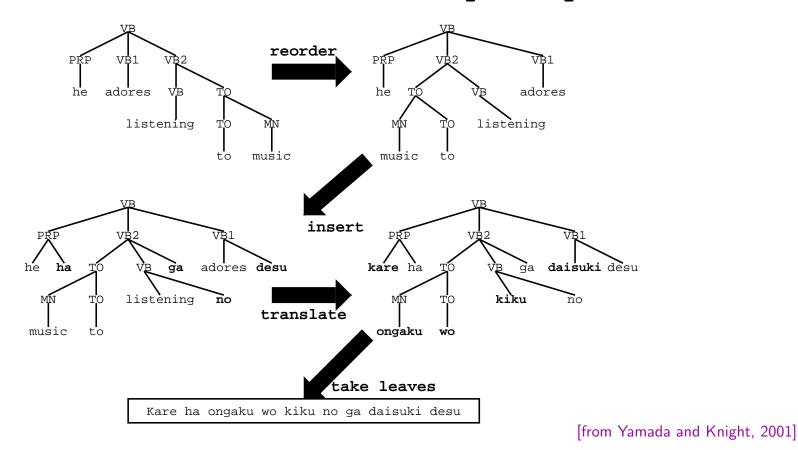
String to Tree Translation



- 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



Yamada and Knight [2001]





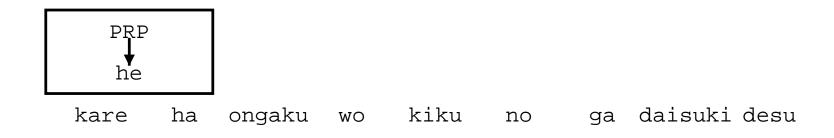
Reordering Table

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	το νβ	0.893
TO NN	TO NN	0.251
TO NN	ΝΝ ΤΟ	0.749



Decoding as Parsing

• Chart Parsing

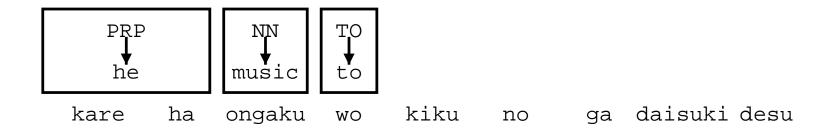


- Pick Japanese *words*
- Translate into *tree stumps*

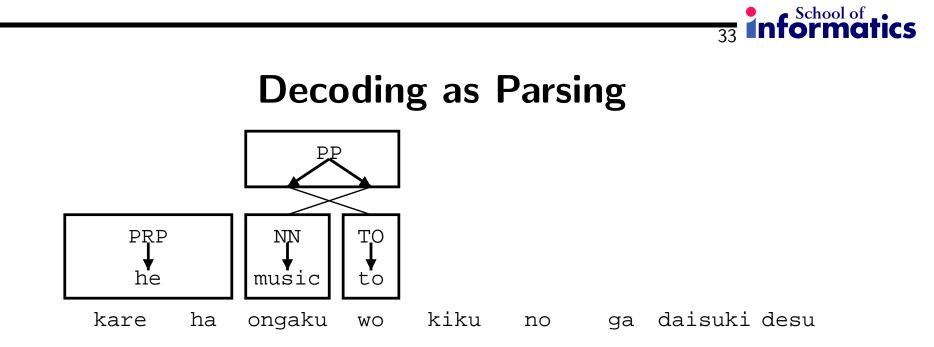


Decoding as Parsing

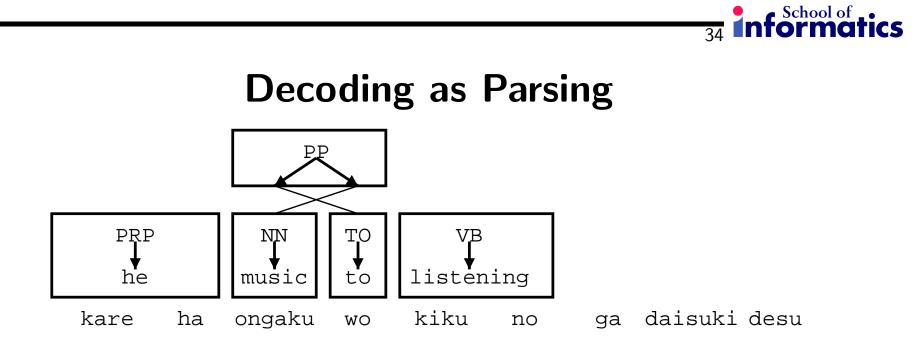
• Chart Parsing



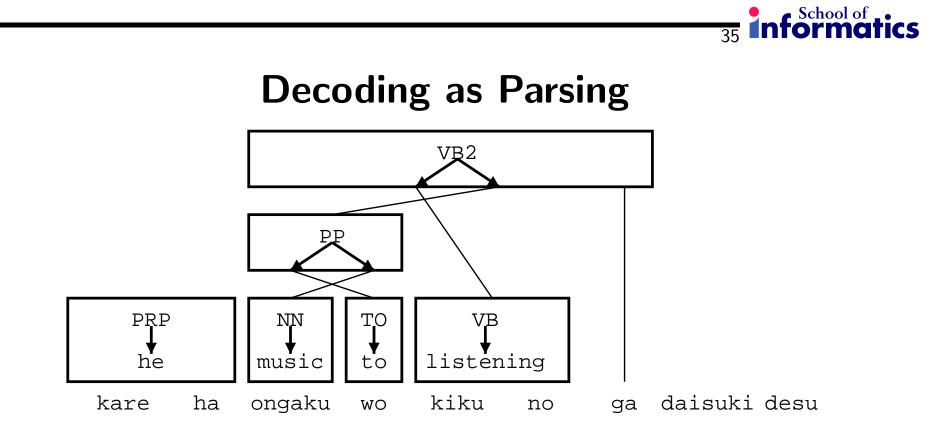
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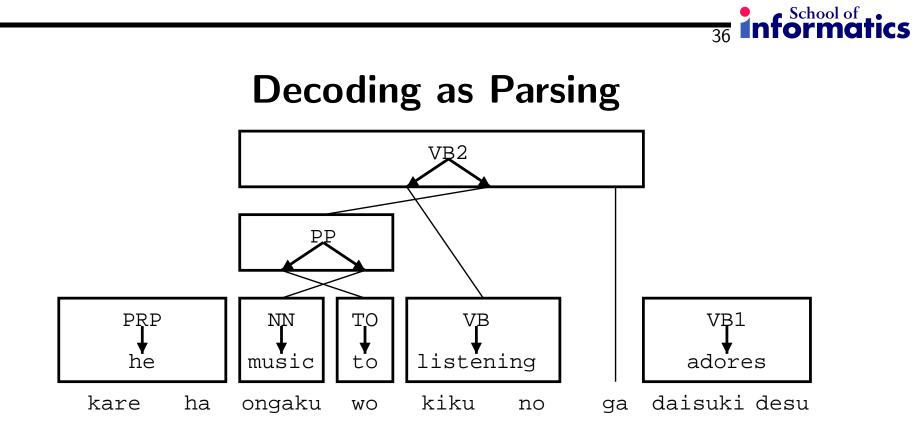


• Adding some *more entries*...

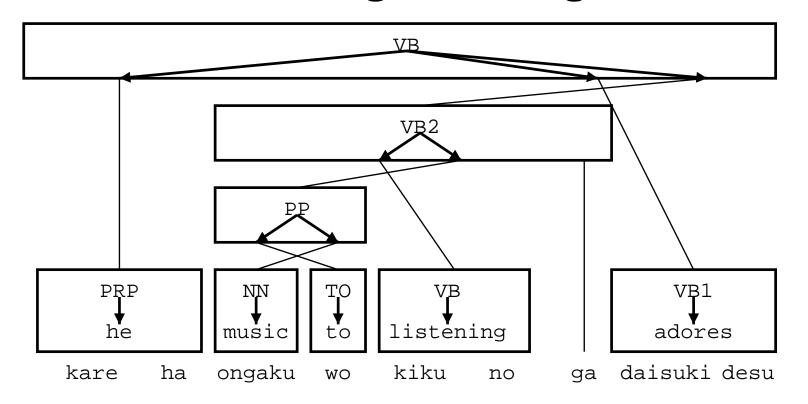


• Combine entries





Decoding as Parsing



• *Finished* when all foreign words covered

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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
 - \rightarrow unity of training and decoding as in IBM models



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**



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



Syntax: Does it help?

- Getting there
 - for some languages competitive with best phrase-based systems
- Some evidence
 - work on reordering German
 - ISI: better for short sentences Chinese-English
 - 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*
 - $\rightarrow\,$ difficult to handle, slow to train and decode
 - few researchers good at statistical modeling and syntactic theories