### **Syntax-Based Models**

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# what is syntax?

#### **Tree-Based Models**



- Traditional statistical models operate on sequences of words
- Many translation problems can be best explained by pointing to syntax
  - reordering, e.g., verb movement in German–English translation
  - long distance agreement (e.g., subject-verb) in output
- ⇒ Translation models based on tree representation of language
  - significant ongoing research
  - state-of-the art for some language pairs

#### **Phrase Structure Grammar**

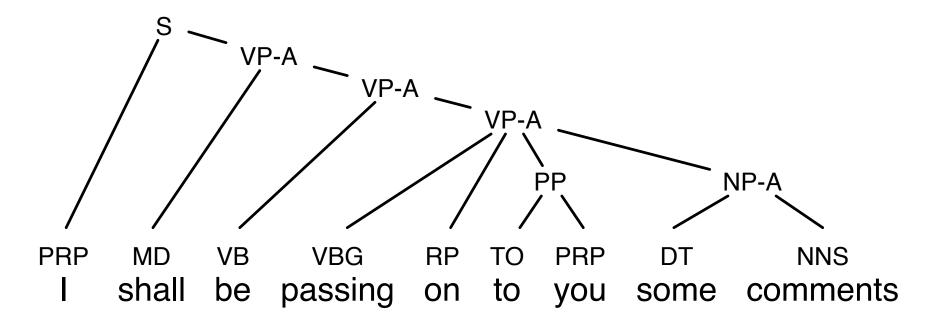


#### • Phrase structure

- noun phrases: the big man, a house, ...
- prepositional phrases: at 5 o'clock, in Edinburgh, ...
- verb phrases: going out of business, eat chicken, ...
- adjective phrases, ...
- Context-free Grammars (CFG)
  - non-terminal symbols: phrase structure labels, part-of-speech tags
  - terminal symbols: words
  - production rules: NT → [NT,T]+ example: NP → DET NN

#### **Phrase Structure Grammar**





Phrase structure grammar tree for an English sentence (as produced Collins' parser)

## **Synchronous Phrase Structure Grammar**



• English rule

$$NP \rightarrow DET JJ NN$$

• French rule

$$\mathsf{NP} \to \mathsf{DET} \; \mathsf{NN} \; \mathsf{JJ}$$

• Synchronous rule (indices indicate alignment):

$$NP \rightarrow DET_1 NN_2 JJ_3 \mid DET_1 JJ_3 NN_2$$

### **Synchronous Grammar Rules**



• Nonterminal rules

$$NP \rightarrow DET_1 NN_2 JJ_3 \mid DET_1 JJ_3 NN_2$$

• Terminal rules

$$N o maison \mid house$$
 NP  $o la maison bleue \mid the blue house$ 

Mixed rules

$$NP \rightarrow la \ mais on \ JJ_1 \mid \ the \ JJ_1 \ house$$

#### **Tree-Based Translation Model**



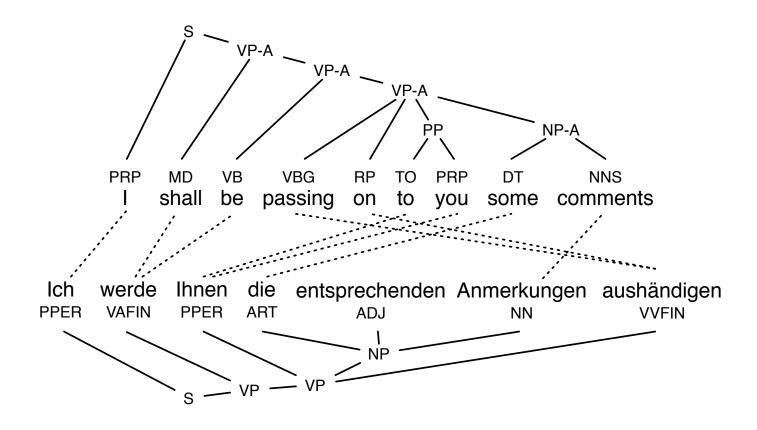
- Translation by parsing
  - synchronous grammar has to parse entire input sentence
  - output tree is generated at the same time
  - process is broken up into a number of rule applications
- Translation probability

$$SCORE(TREE, E, F) = \prod_{i} RULE_{i}$$

• Many ways to assign probabilities to rules

### **Aligned Tree Pair**



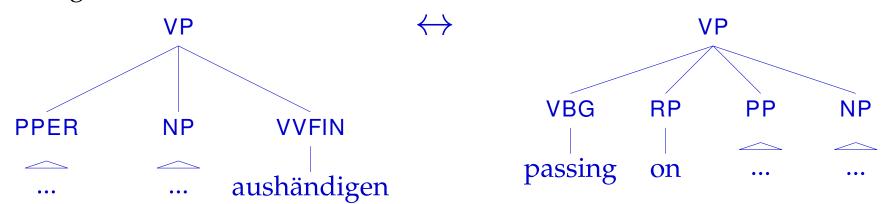


Phrase structure grammar trees with word alignment (German–English sentence pair.)

### **Reordering Rule**



• Subtree alignment



• Synchronous grammar rule

$$VP \rightarrow PPER_1 NP_2$$
 aushändigen | passing on  $PP_1 NP_2$ 

- Note:
  - one word aushändigen mapped to two words passing on ok
  - but: fully non-terminal rule not possible (one-to-one mapping constraint for nonterminals)

#### **Another Rule**



• Subtree alignment



• Synchronous grammar rule (stripping out English internal structure)

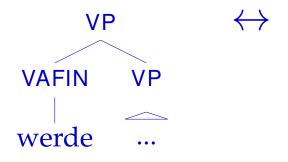
$$PRO/PP \rightarrow Ihnen \mid to you$$

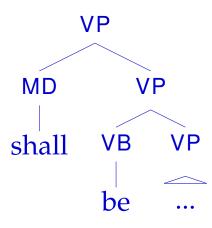
• Rule with internal structure

#### **Another Rule**



• Translation of German werde to English shall be





- Translation rule needs to include mapping of VP
- $\Rightarrow$  Complex rule

#### **Internal Structure**



• Stripping out internal structure

$$VP \rightarrow werde VP_1 \mid shall be VP_1$$

- ⇒ synchronous context free grammar
- Maintaining internal structure

$$VP \rightarrow egin{pmatrix} VAFIN & VP_1 & MD & VP & VP_1 & VB & VB & VP_1 & VB & VB & VB_1 & VB_1 & VB & VB_1 & VB$$

⇒ synchronous tree substitution grammar

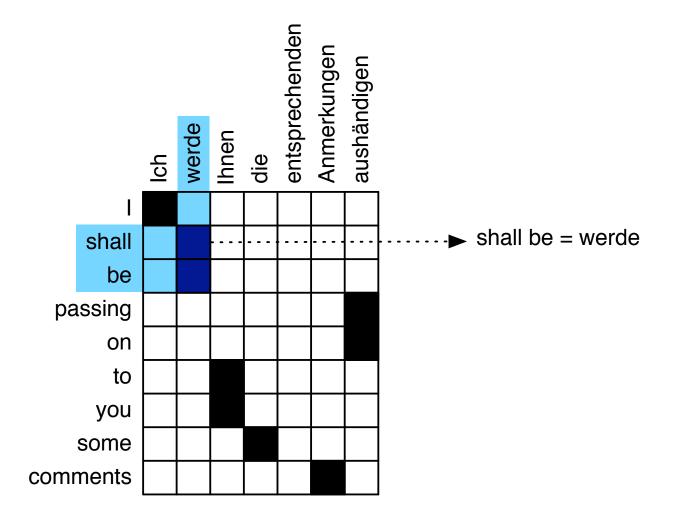
### **Learning Synchronous Grammars**



- Extracting rules from a word-aligned parallel corpus
- First: Hierarchical phrase-based model
  - only one non-terminal symbol X
  - no linguistic syntax, just a formally syntactic model
- Then: Synchronous phrase structure model
  - non-terminals for words and phrases: NP, VP, PP, ADJ, ...
  - corpus must also be parsed with syntactic parser

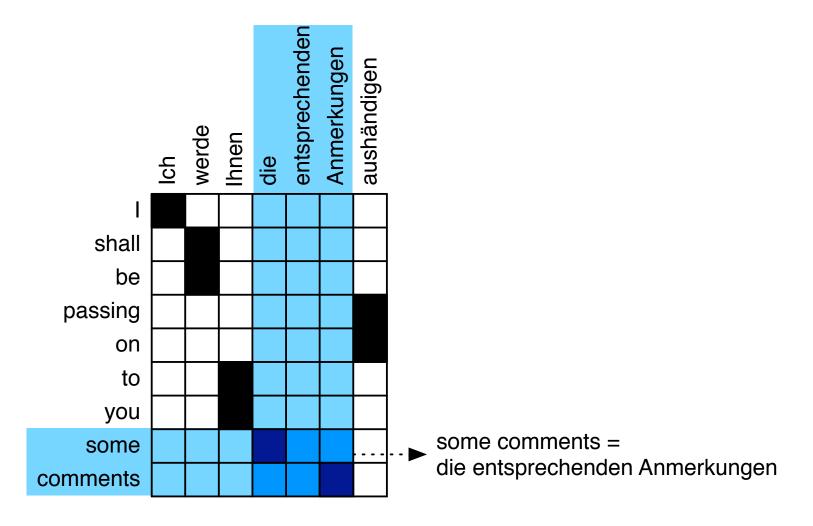
### **Extracting Phrase Translation Rules**





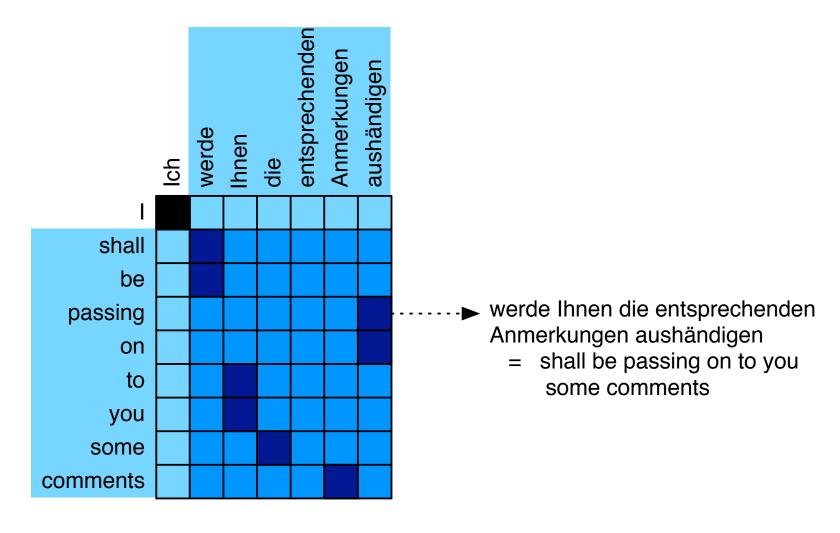
### **Extracting Phrase Translation Rules**



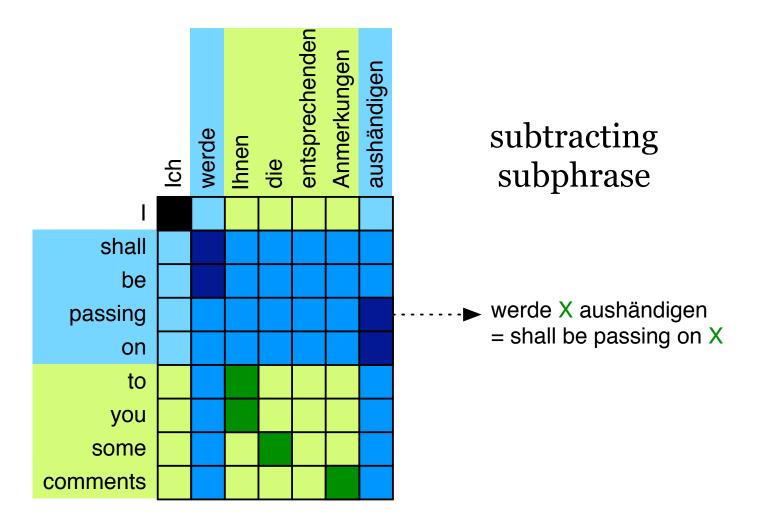


### **Extracting Phrase Translation Rules**





## Extracting Hierarchical Phrase Translation Rules



### **Formal Definition**



• Recall: consistent phrase pairs

$$(ar{e},ar{f})$$
 consistent with  $A\Leftrightarrow$  
$$\forall e_i\in ar{e}:(e_i,f_j)\in A \to f_j\in ar{f}$$
 and  $\forall f_j\in ar{f}:(e_i,f_j)\in A \to e_i\in ar{e}$  and  $\exists e_i\in ar{e},f_j\in ar{f}:(e_i,f_j)\in A$ 

• Let P be the set of all extracted phrase pairs  $(\bar{e}, \bar{f})$ 

### **Formal Definition**



• Extend recursively:

$$\begin{split} &\text{if } (\bar{e},\bar{f}) \in P \text{ AND } (\bar{e}_{\text{SUB}},\bar{f}_{\text{SUB}}) \in P \\ &\text{AND } \bar{e} = \bar{e}_{\text{PRE}} + \bar{e}_{\text{SUB}} + \bar{e}_{\text{POST}} \\ &\text{AND } \bar{f} = \bar{f}_{\text{PRE}} + \bar{f}_{\text{SUB}} + \bar{f}_{\text{POST}} \\ &\text{AND } \bar{e} \neq \bar{e}_{\text{SUB}} \text{ AND } \bar{f} \neq \bar{f}_{\text{SUB}} \\ &\text{add } (e_{\text{PRE}} + \mathbf{X} + e_{\text{POST}}, f_{\text{PRE}} + \mathbf{X} + f_{\text{POST}}) \text{ to } P \end{split}$$

(note: any of  $e_{PRE}$ ,  $e_{POST}$ ,  $f_{PRE}$ , or  $f_{POST}$  may be empty)

• Set of hierarchical phrase pairs is the closure under this extension mechanism

#### **Comments**



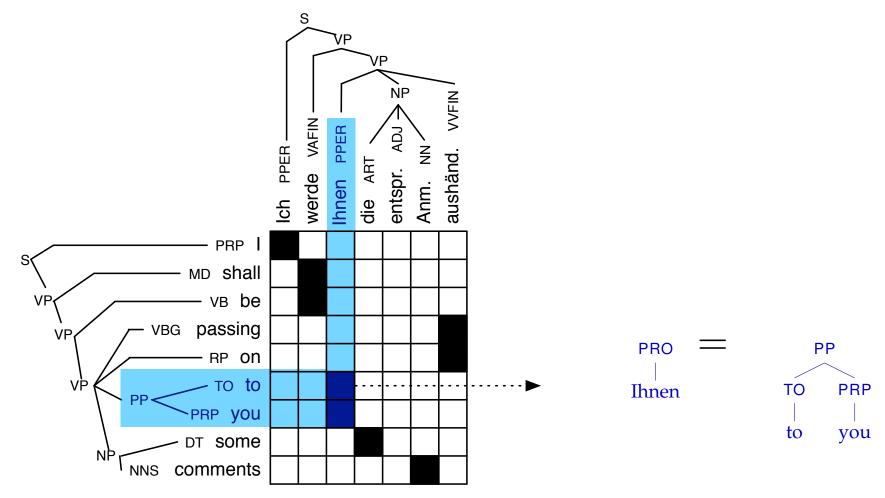
• Removal of multiple sub-phrases leads to rules with multiple non-terminals, such as:

$$Y \rightarrow X_1 X_2 \mid X_2 \text{ of } X_1$$

- Typical restrictions to limit complexity (Chiang, 2005)
  - at most 2 nonterminal symbols
  - at least 1 but at most 5 words per language
  - span at most 15 words (counting gaps)

### **Learning Syntactic Translation Rules**





## **Constraints on Syntactic Rules**

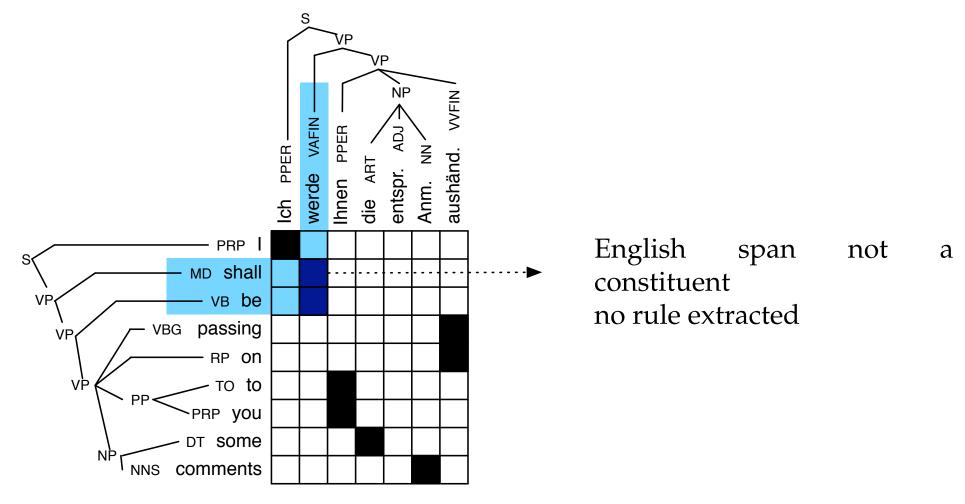


- Same word alignment constraints as hierarchical models
- Hierarchical: rule can cover any span
   syntactic rules must cover constituents in the tree
- Hierarchical: gaps may cover any span
   ⇔ gaps must cover constituents in the tree

• Much less rules are extracted (all things being equal)

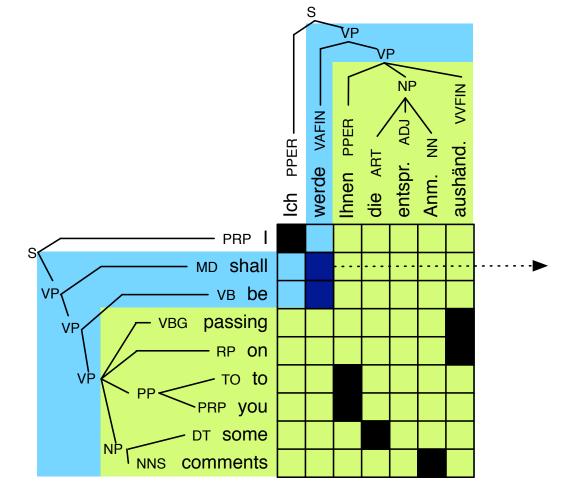
### **Impossible Rules**



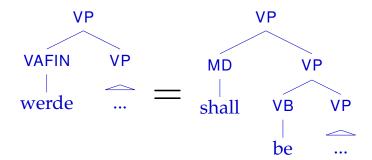


#### **Rules with Context**





Rule with this phrase pair requires syntactic context



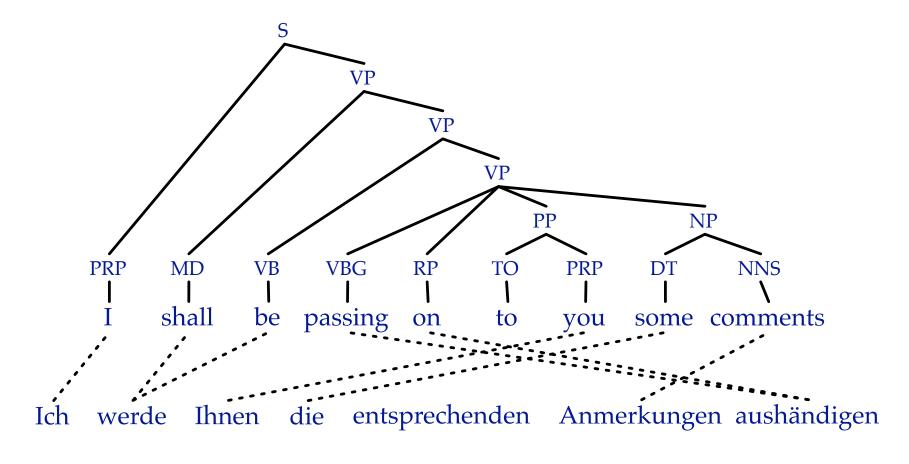
### **Too Many Rules Extractable**



- Huge number of rules can be extracted
   (every alignable node may or may not be part of a rule → exponential number of rules)
- Need to limit which rules to extract
- Option 1: similar restriction as for hierarchical model (maximum span size, maximum number of terminals and non-terminals, etc.)
- Option 2: only extract minimal rules ("GHKM" rules)

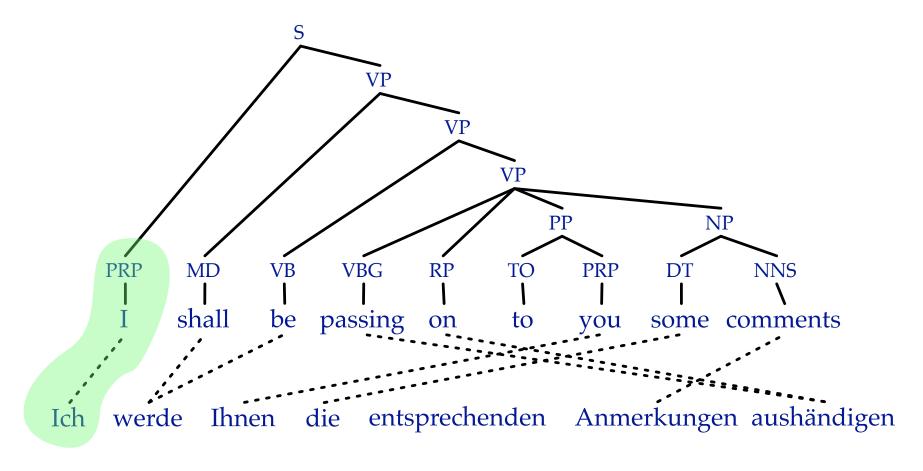
#### **Minimal Rules**





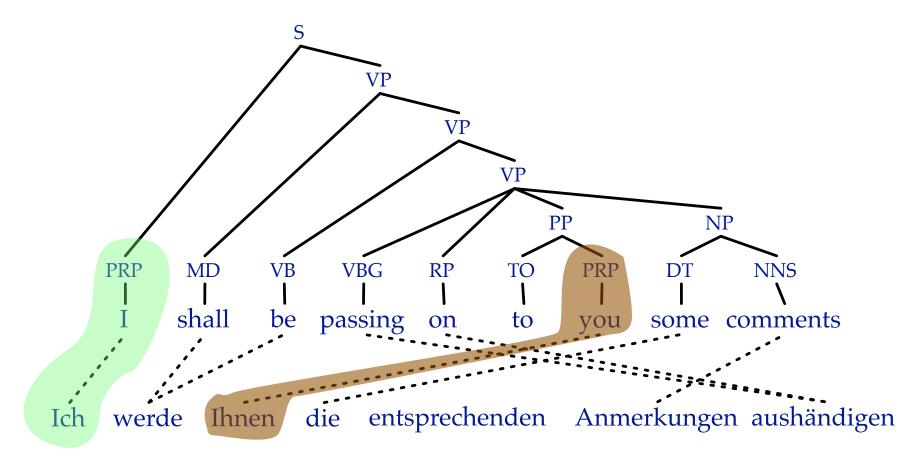
Extract: set of smallest rules required to explain the sentence pair





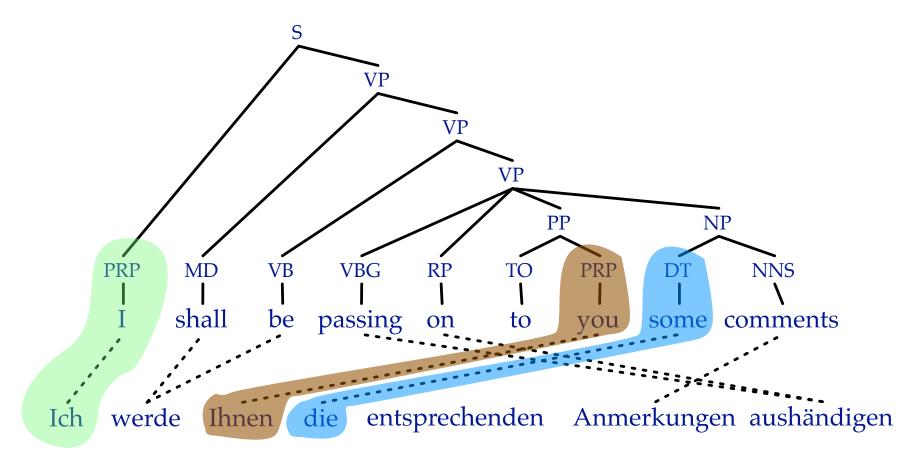
Extracted rule:  $PRP \rightarrow Ich \mid I$ 





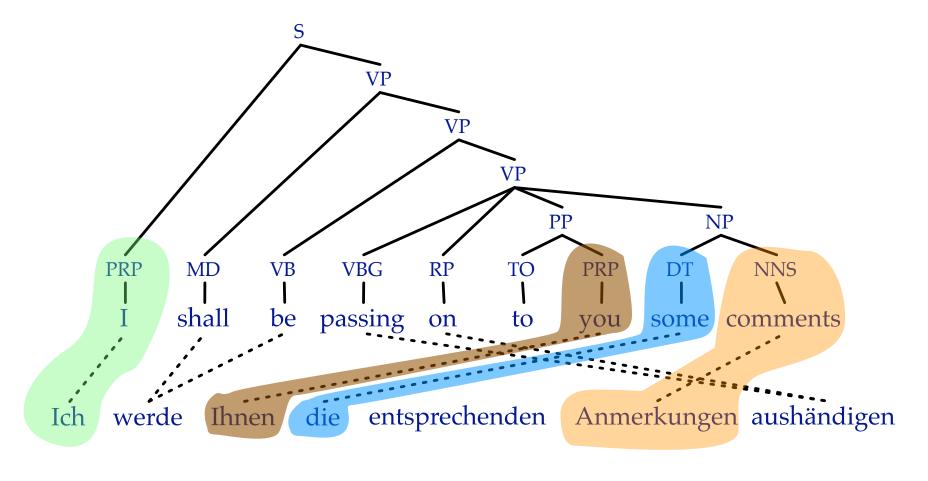
Extracted rule:  $PRP \rightarrow Ihnen \mid you$ 





Extracted rule:  $DT \rightarrow die \mid some$ 

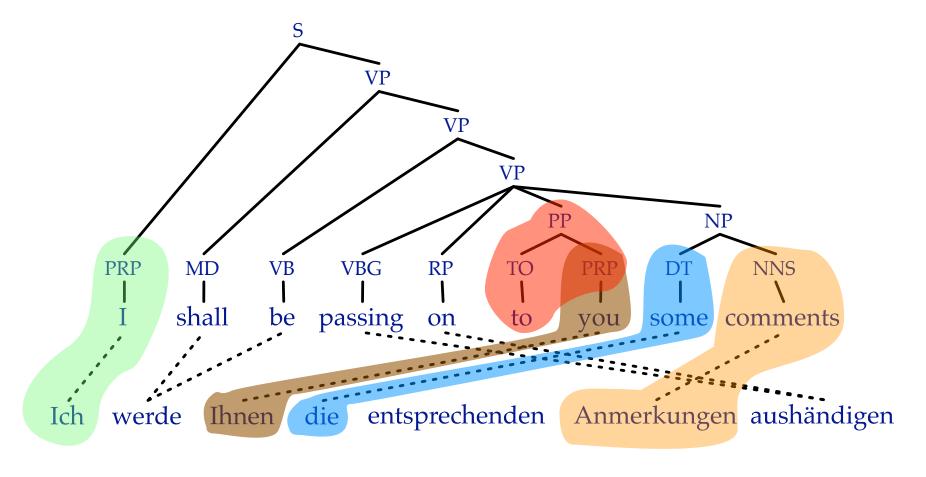




Extracted rule: NNS → Anmerkungen | comments

#### **Insertion Rule**

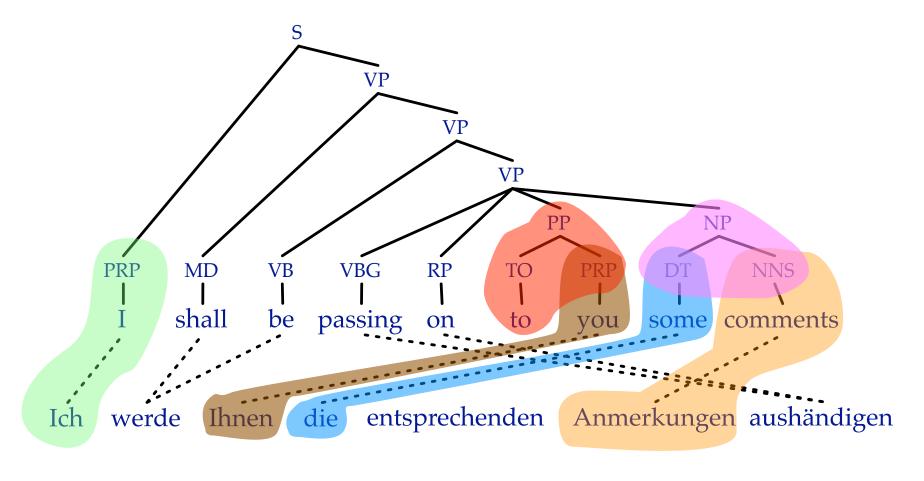




Extracted rule:  $PP \rightarrow X \mid to PRP$ 

#### **Non-Lexical Rule**

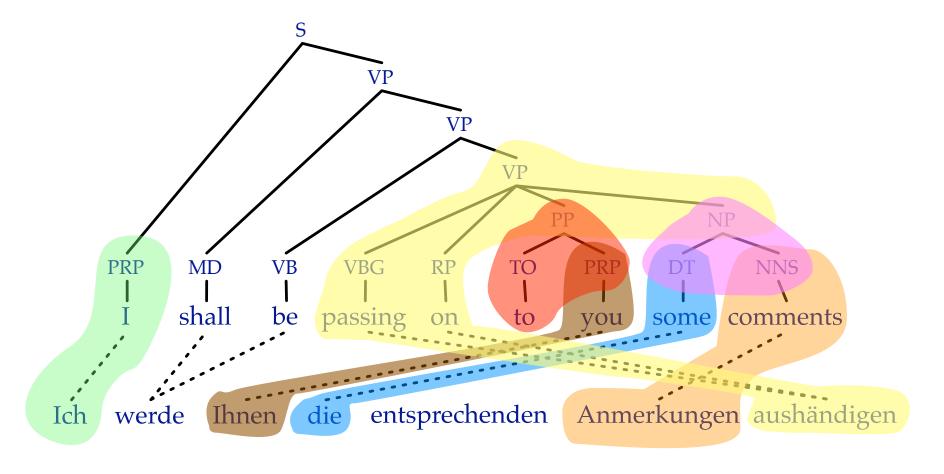




Extracted rule: NP  $\rightarrow$  X<sub>1</sub> X<sub>2</sub> | DT<sub>1</sub> NNS<sub>2</sub>

### **Lexical Rule with Syntactic Context**

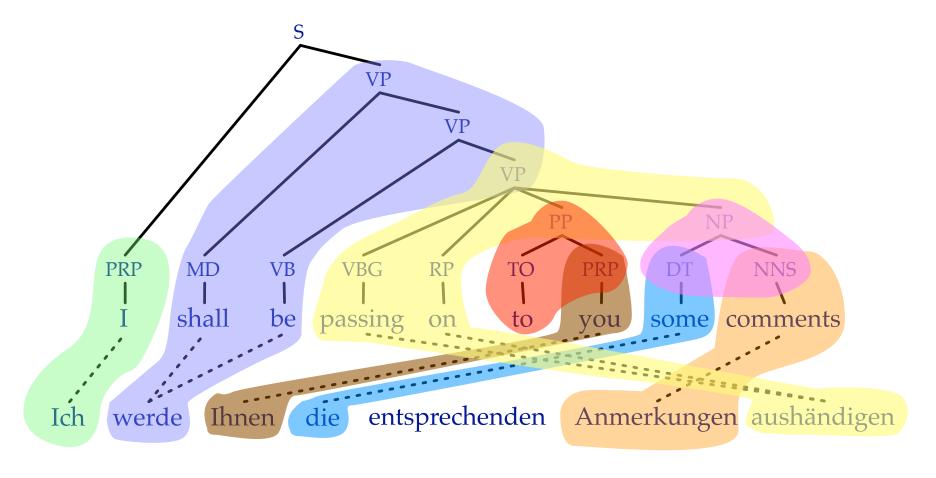




Extracted rule:  $VP \rightarrow X_1 X_2$  aushändigen | passing on  $PP_1 NP_2$ 

### **Lexical Rule with Syntactic Context**

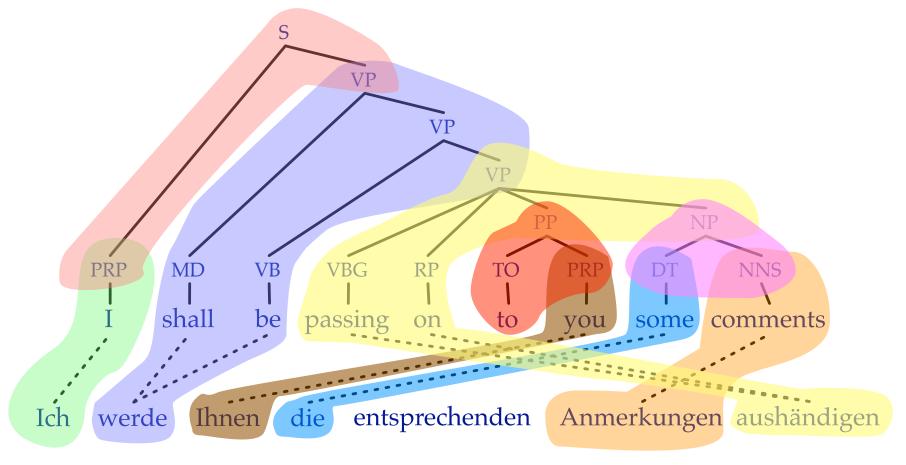




Extracted rule:  $VP \rightarrow werde \ X \mid shall \ be \ VP \ (ignoring internal structure)$ 

#### Non-Lexical Rule



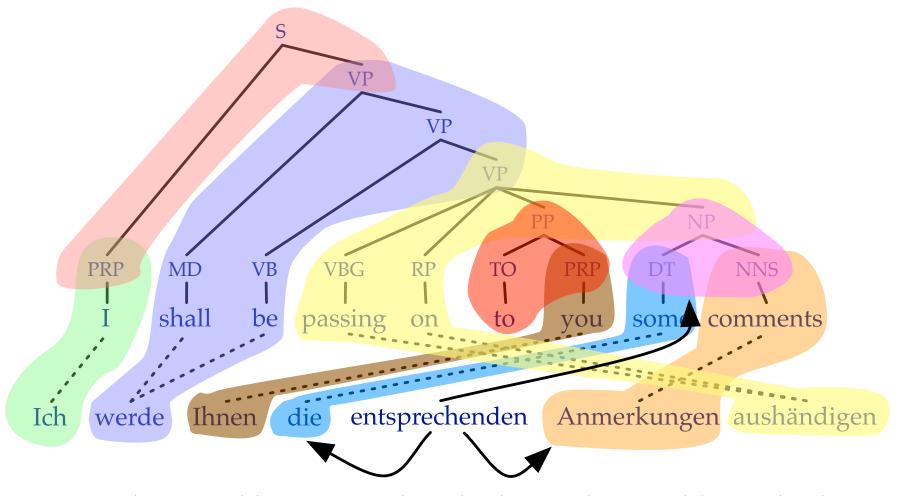


Extracted rule:  $S \rightarrow X_1 X_2 \mid PRP_1 VP_2$ 

DONE — note: one rule per alignable constituent

### **Unaligned Source Words**





Attach to neighboring words or higher nodes  $\rightarrow$  additional rules

#### **Too Few Phrasal Rules?**



- Lexical rules will be 1-to-1 mappings (unless word alignment requires otherwise)
- But: phrasal rules very beneficial in phrase-based models
- Solutions
  - combine rules that contain a maximum number of symbols (as in hierarchical models, recall: "Option 1")
  - compose minimal rules to cover a maximum number of non-leaf nodes

### **Composed Rules**



• Current rules

$$X_1 X_2 = NP$$

$$DT_1 NNS_1$$



• Composed rule



(1 non-leaf node: NP)

### **Composed Rules**



• Minimal rule:

 $X_1 \ X_2 \ aushändigen = VP$ PRP PRP PP1 NP2

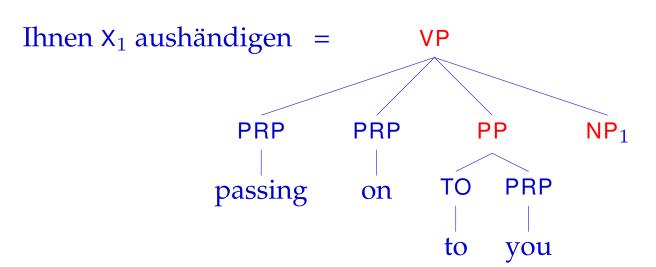
passing on

3 non-leaf nodes:

VP, PP, NP

• Composed rule:

3 non-leaf nodes: **VP**, **PP** and **NP** 



### **Relaxing Tree Constraints**



• Impossible rule

- Create new non-terminal label: MD+VB
- $\Rightarrow$  New rule

### **Zollmann Venugopal Relaxation**



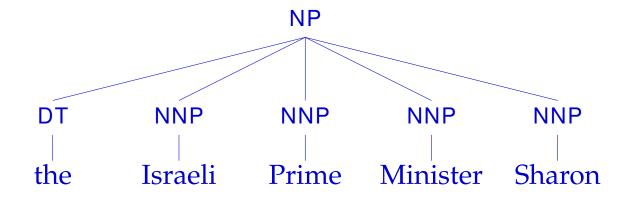
- If span consists of two constituents, join them: X+Y
- If span conststs of three constituents, join them: X+Y+Z
- If span covers constituents with the same parent x and include
  - every but the first child Y, label as X\Y
  - every but the last child Y, label as X/Y
- For all other cases, label as FAIL

⇒ More rules can be extracted, but number of non-terminals blows up

### **Special Problem: Flat Structures**



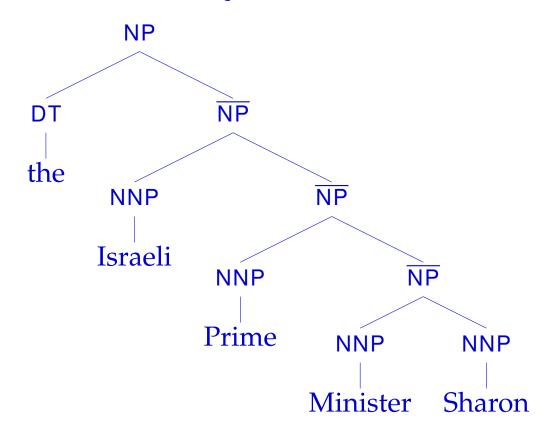
• Flat structures severely limit rule extraction



• Can only extract rules for individual words or entire phrase

### Relaxation by Tree Binarization





More rules can be extracted Left-binarization or right-binarization?

## **Scoring Translation Rules**



- Extract all rules from corpus
- Score based on counts
  - joint rule probability:  $p(LHS, RHS_f, RHS_e)$
  - rule application probability:  $p(RHS_f, RHS_e|LHS)$
  - direct translation probability:  $p(RHS_e|RHS_f, LHS)$
  - noisy channel translation probability:  $p(RHS_f|RHS_e, LHS)$
  - lexical translation probability:  $\prod_{e_i \in RHS_e} p(e_i | RHS_f, a)$