
Syntax-Based Decoding

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syntax-based models

Synchronous Context Free Grammar Rules



- Nonterminal rules

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

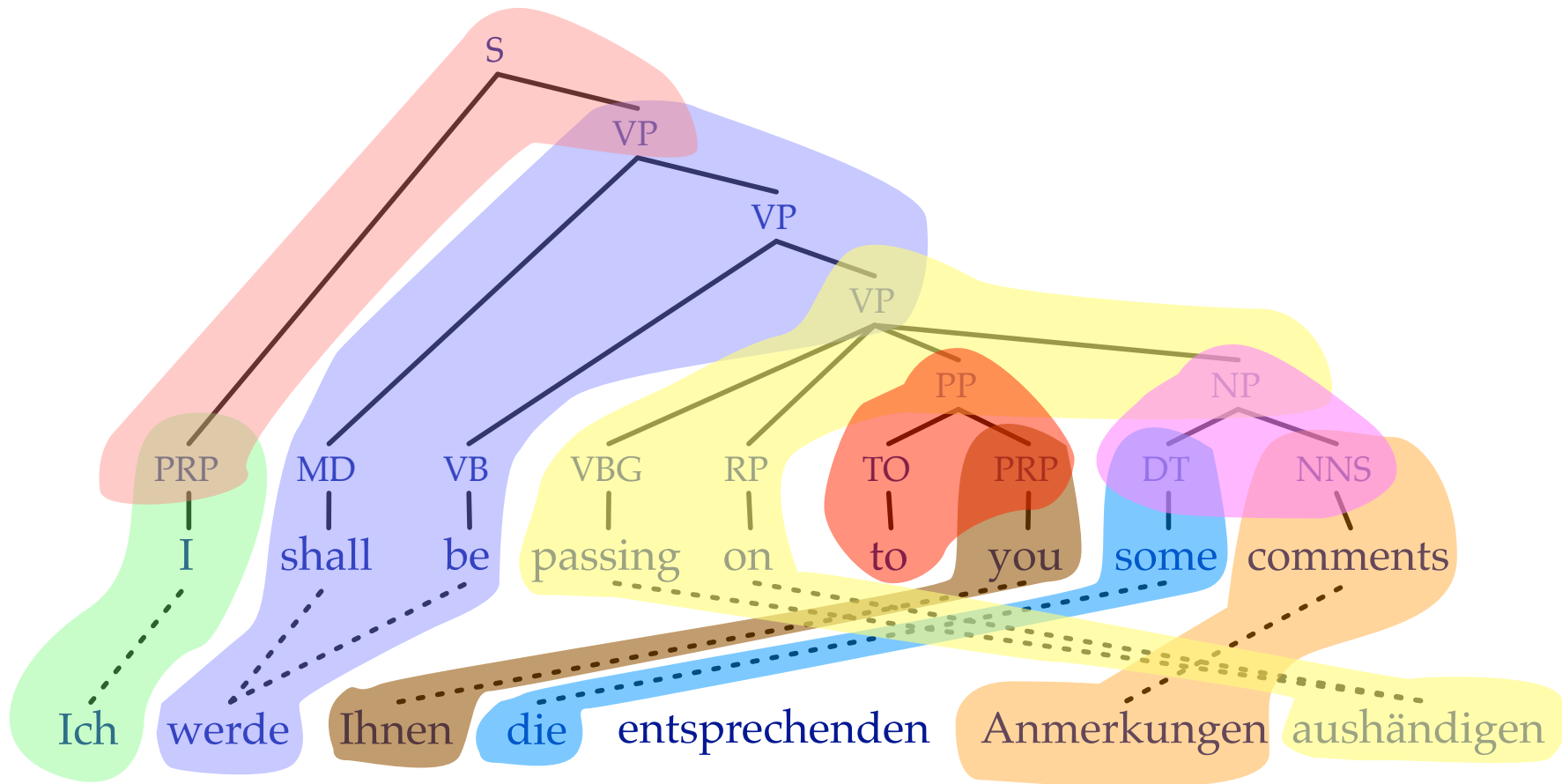
- Terminal rules

$$N \rightarrow \text{maison} \mid \text{house}$$
$$NP \rightarrow \text{la maison bleue} \mid \text{the blue house}$$

- Mixed rules

$$NP \rightarrow \text{la maison } JJ_1 \mid \text{the } JJ_1 \text{ house}$$

Extracting Minimal Rules



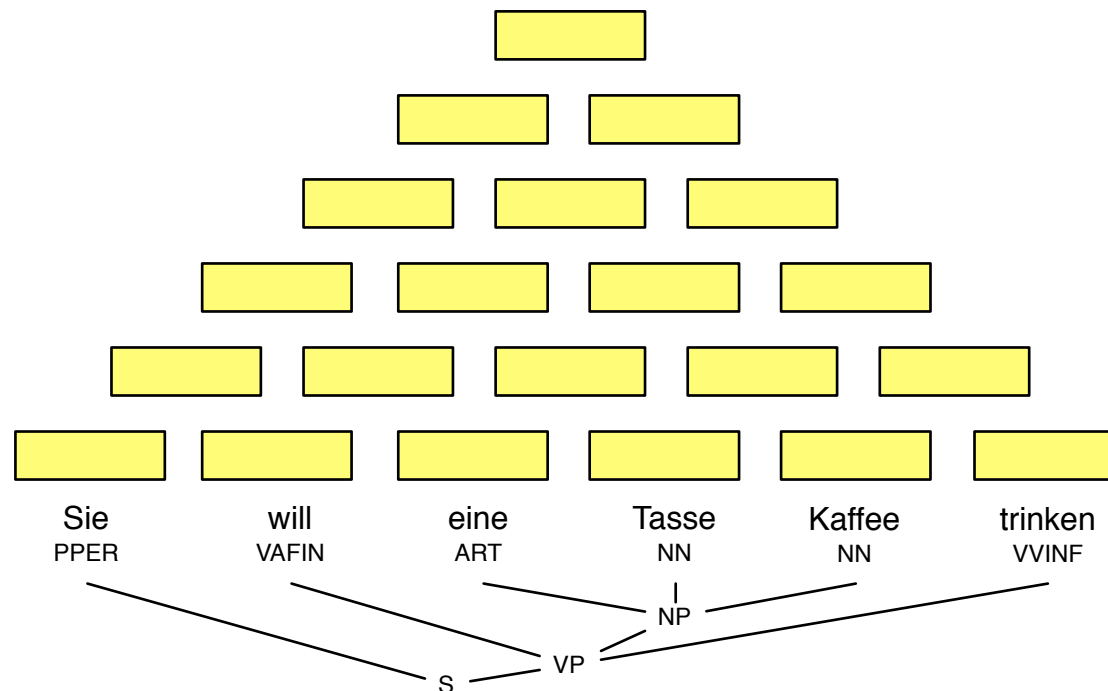
Extracted rule: $S \rightarrow X_1 X_2 \mid \text{PRP}_1 \text{VP}_2$
DONE — note: one rule per alignable constituent

decoding

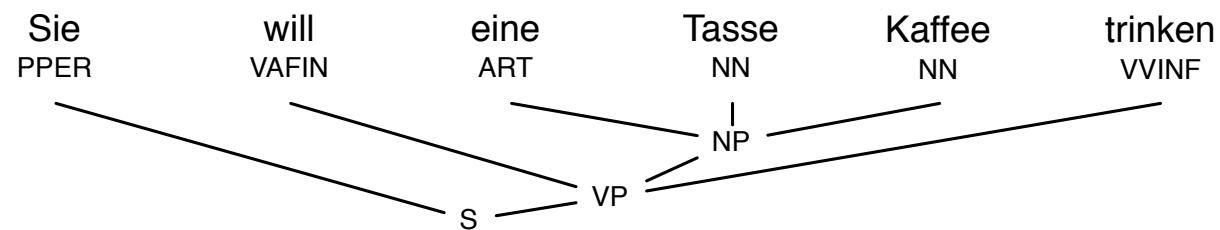
Syntactic Decoding

Inspired by monolingual syntactic chart parsing:

During decoding of the source sentence,
a chart with translations for the $O(n^2)$ spans has to be filled

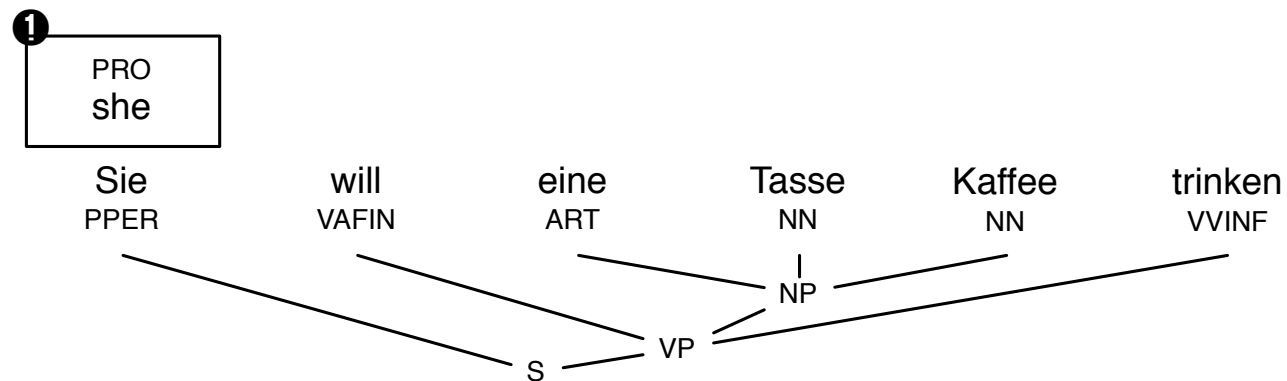


Syntax Decoding



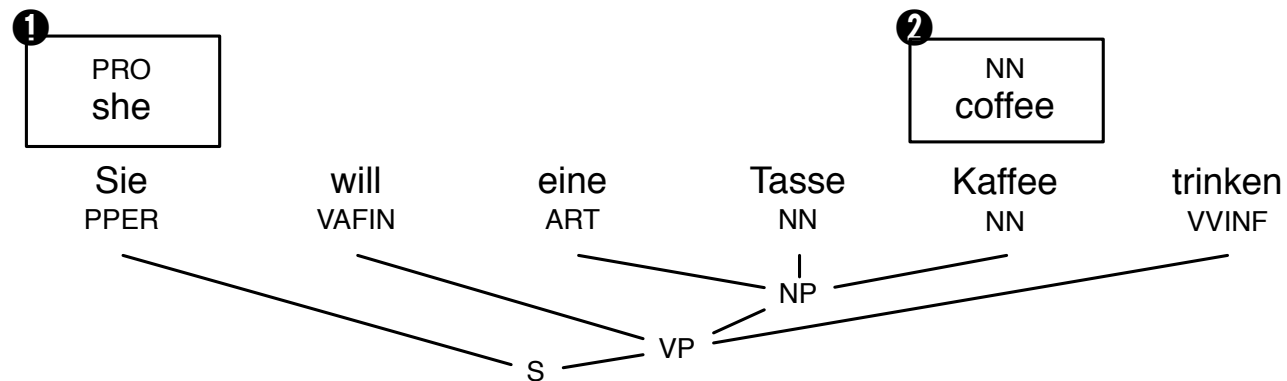
German input sentence with tree

Syntax Decoding



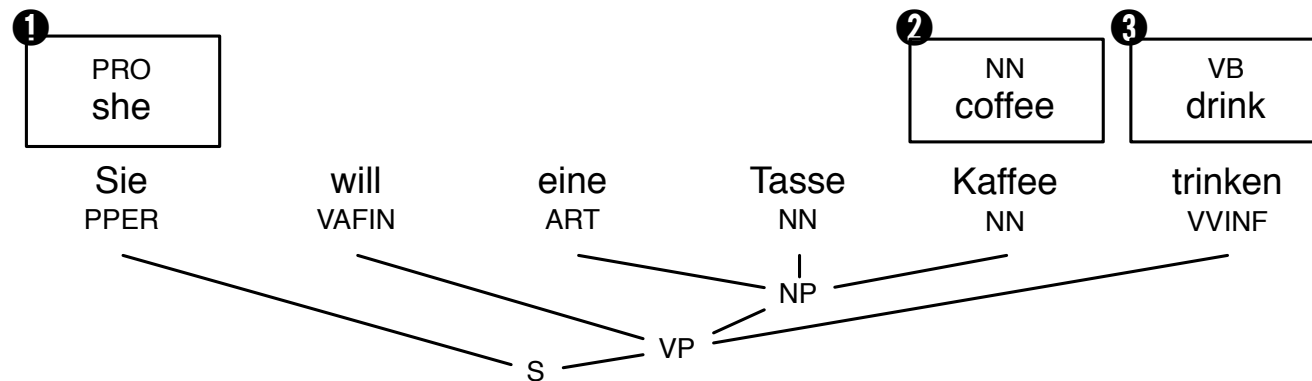
Purely lexical rule: filling a span with a translation (a constituent in the chart)

Syntax Decoding

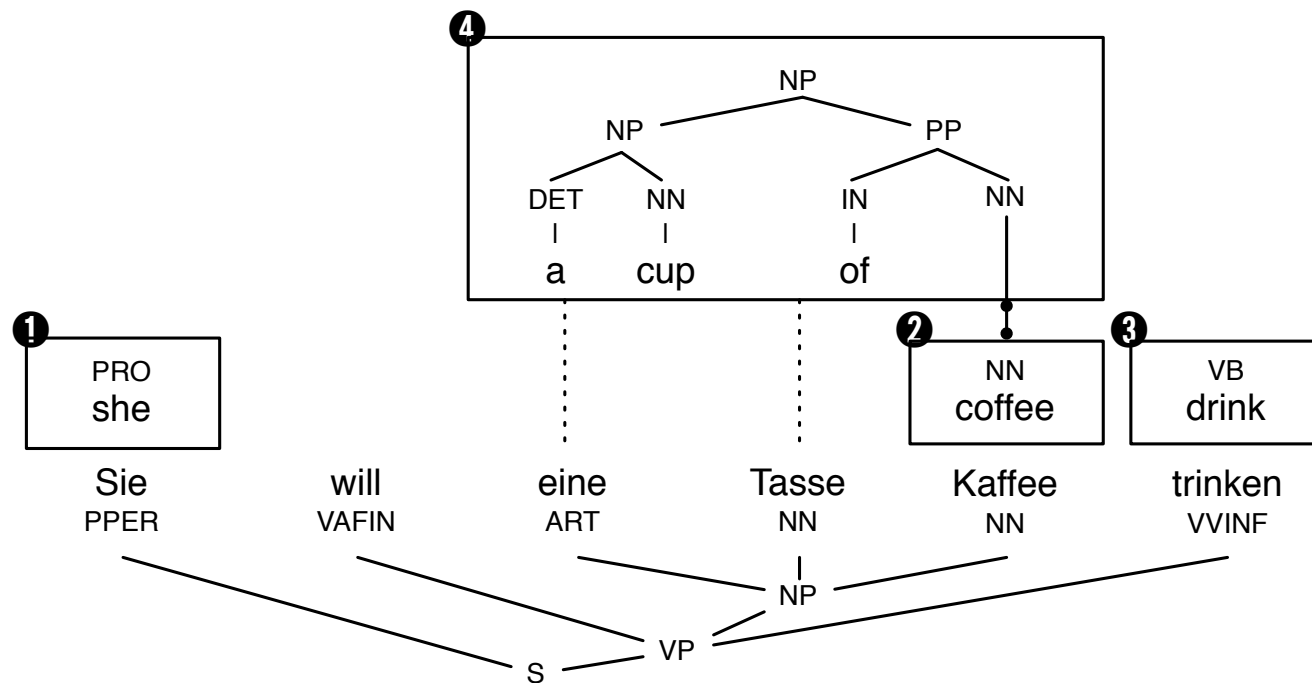


Purely lexical rule: filling a span with a translation (a constituent in the chart)

Syntax Decoding

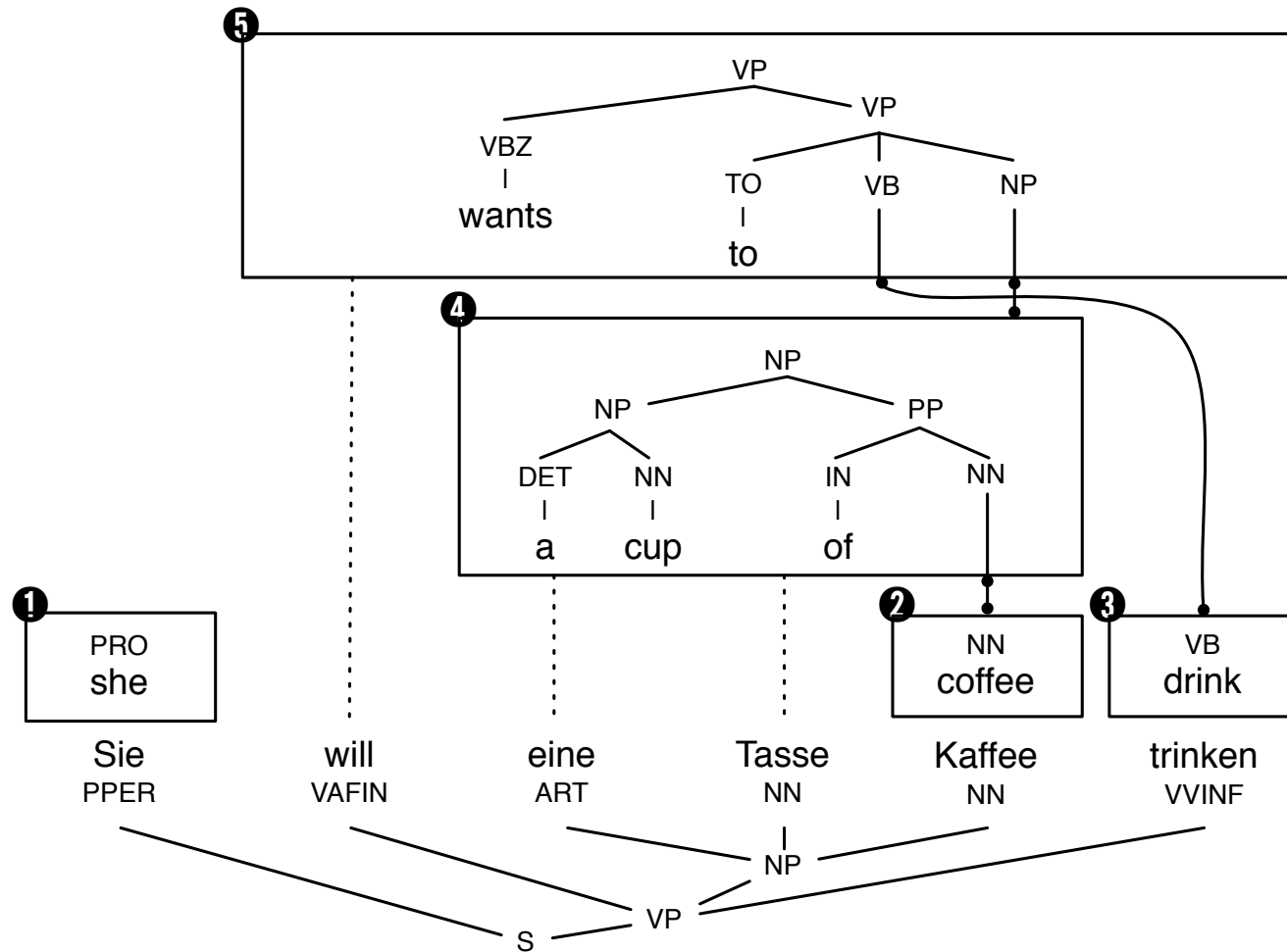


Purely lexical rule: filling a span with a translation (a constituent in the chart)

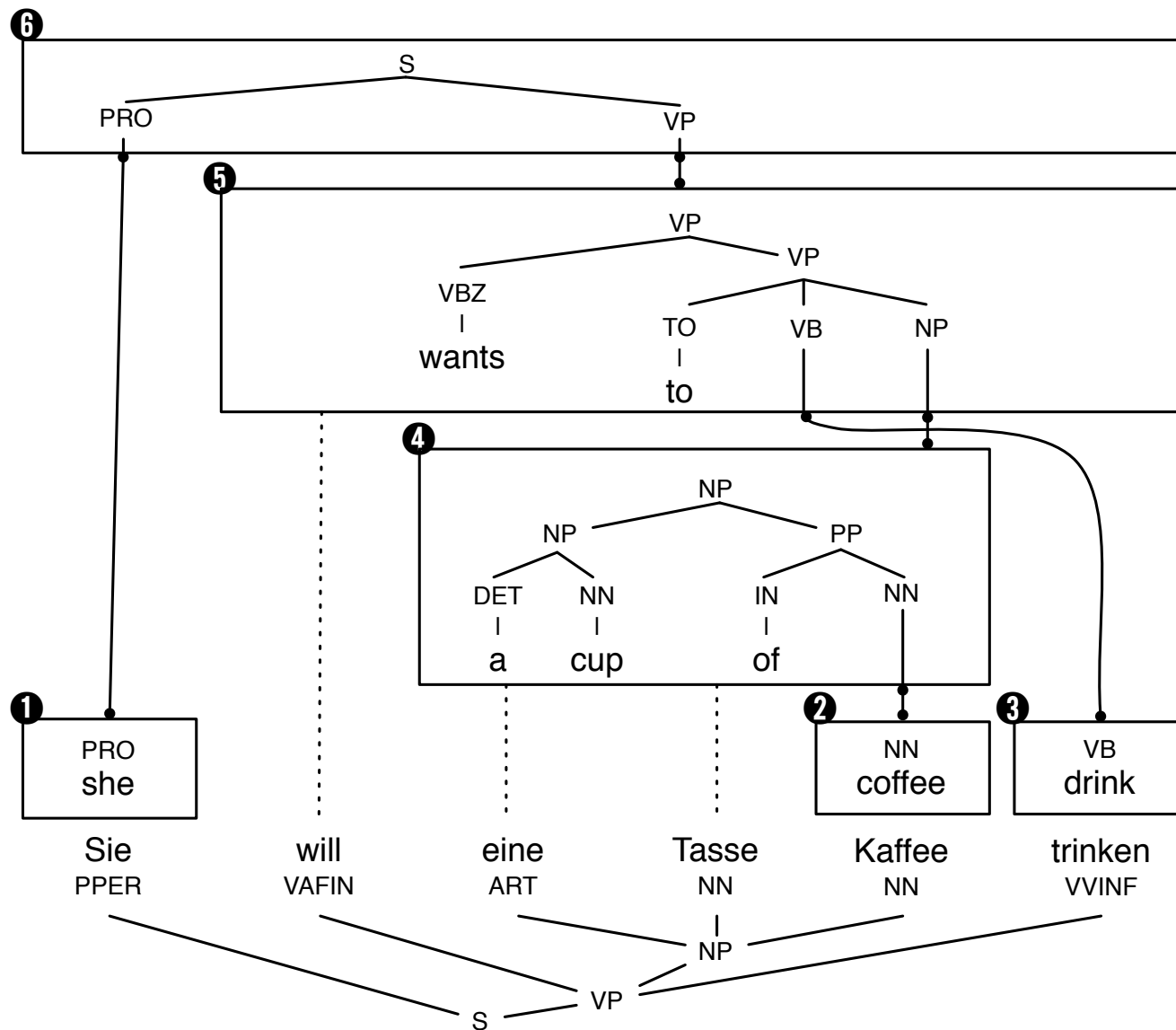


Complex rule: matching underlying constituent spans, and covering words

Syntax Decoding

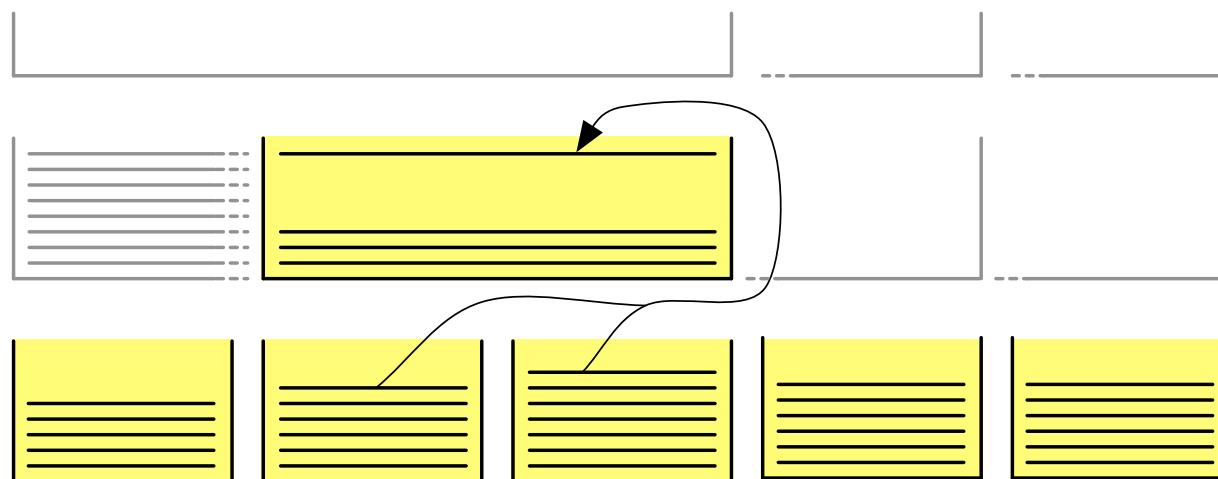


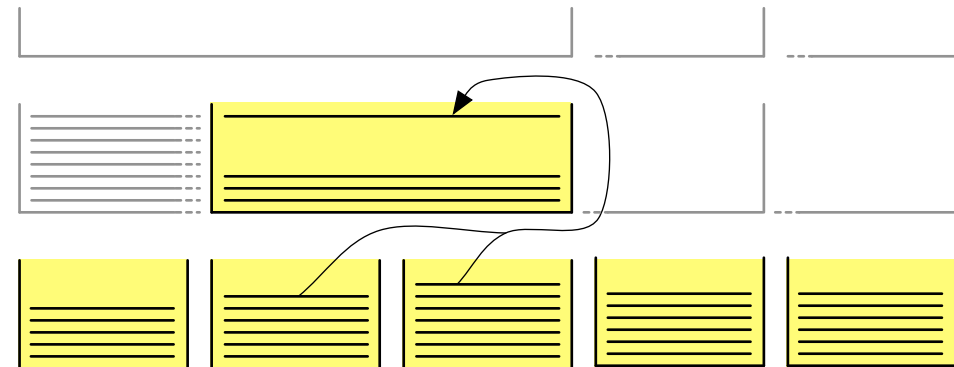
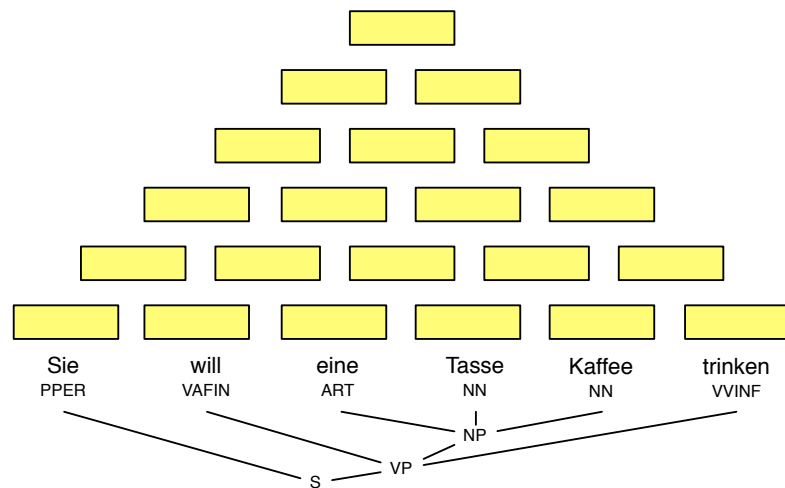
Complex rule with reordering



Bottom-Up Decoding

- For each span, a stack of (partial) translations is maintained
- Bottom-up: a higher stack is filled, once underlying stacks are complete





- Chart consists of cells that cover contiguous spans over the input sentence
- Each cell contains a set of hypotheses¹
- Hypothesis = translation of span with target-side constituent

¹In the book, they are called chart entries.

Naive Algorithm

Input: Foreign sentence $\mathbf{f} = f_1, \dots, f_{l_f}$, with syntax tree

Output: English translation \mathbf{e}

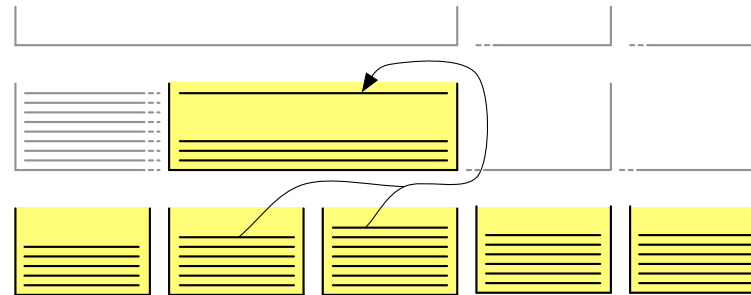
```
1: for all spans [start,end] (bottom up) do
2:   for all sequences  $s$  of hypotheses and words in span [start,end] do
3:     for all rules  $r$  do
4:       if rule  $r$  applies to chart sequence  $s$  then
5:         create new hypothesis  $c$ 
6:         add hypothesis  $c$  to chart
7:       end if
8:     end for
9:   end for
10: end for
11: return English translation  $\mathbf{e}$  from best hypothesis in span  $[0, l_f]$ 
```


- Number of hypotheses in each chart cell explodes
- Dynamic programming (recombination) not enough

⇒ need to discard bad hypotheses
e.g., keep 100 best only

- Different stacks for different output constituent labels?
- Cost estimates
 - translation model cost known
 - language model cost for internal words known
→ estimates for initial words
 - outside cost estimate?
(how useful will be a NP covering input words 3–5 later on?)

Naive Algorithm: Blow-ups



- Many subspan sequences

for all sequences s of hypotheses and words in span $[start, end]$

- Many rules

for all rules r

- Checking if a rule applies not trivial

rule r applies to chart sequence s

⇒ Unworkable

- Prefix tree data structure for rules
- Dotted rules
- Cube pruning

storing rules efficiently

Storing Rules

- First concern: do they apply to span?
→ have to match available hypotheses and input words

- Example rule

$NP \rightarrow X_1 \text{ des } X_2 \mid NP_1 \text{ of the } NN_2$

- Check for applicability
 - is there an initial sub-span that with a hypothesis with constituent label **NP**?
 - is it followed by a sub-span over the word **des**?
 - is it followed by a final sub-span with a hypothesis with label **NN**?
- Sequence of relevant information

NP • des • NN • NP₁ of the NN₂

Rule Applicability Check

Trying to cover a span of six words with given rule

NP • des • NN → NP: NP of the NN



das Haus des Architekten Frank Gehry

Rule Applicability Check

First: check for hypotheses with output constituent label **NP**

NP • des • NN → NP: NP of the NN

das

Haus

des

Architekten

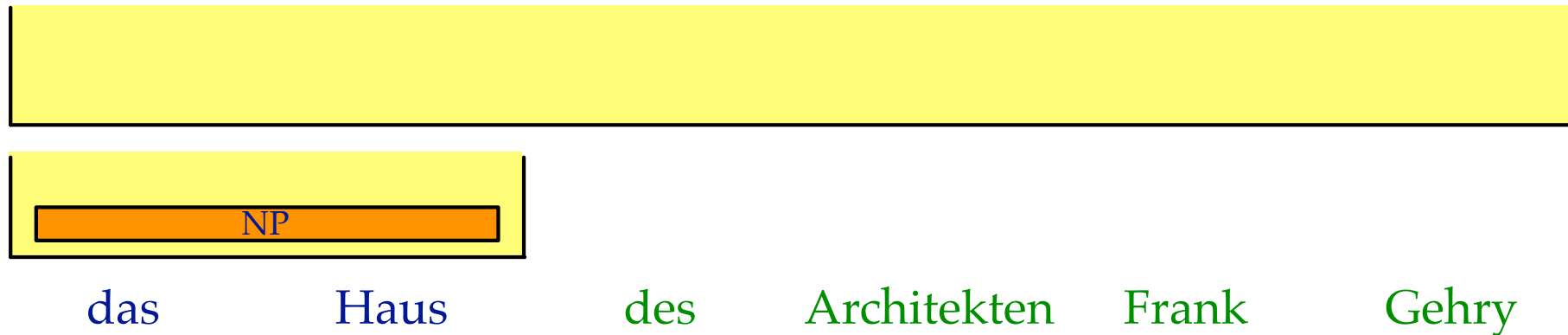
Frank

Gehry

Rule Applicability Check

Found NP hypothesis in cell, matched first symbol of rule

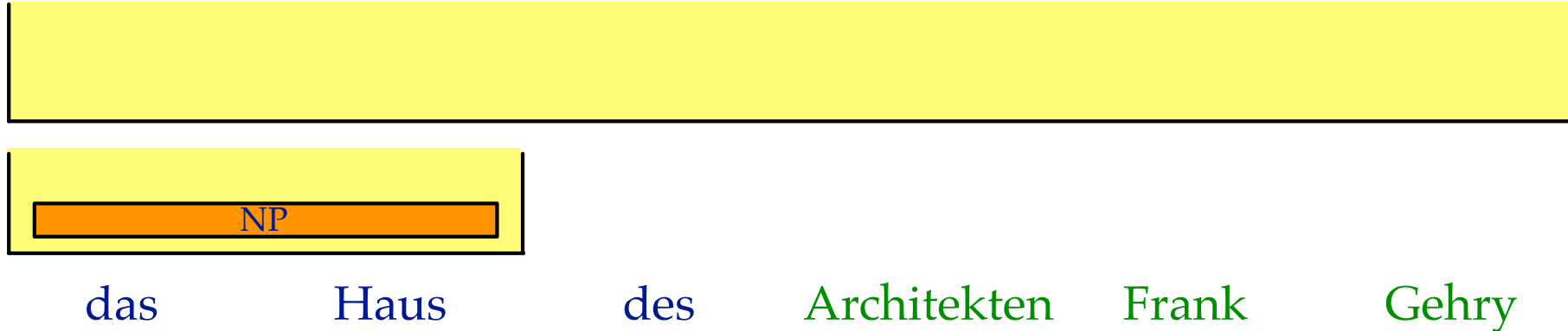
NP • des • NN → NP: NP of the NN



Rule Applicability Check

Matched word **des**, matched second symbol of rule

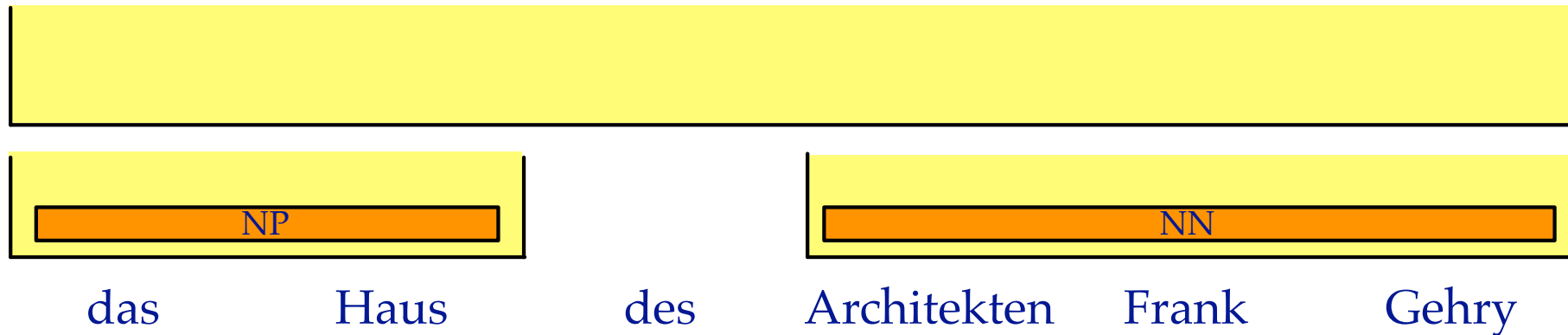
NP • **des** • NN → NP: NP of the NN



Rule Applicability Check

Found a **NN** hypothesis in cell, matched last symbol of rule

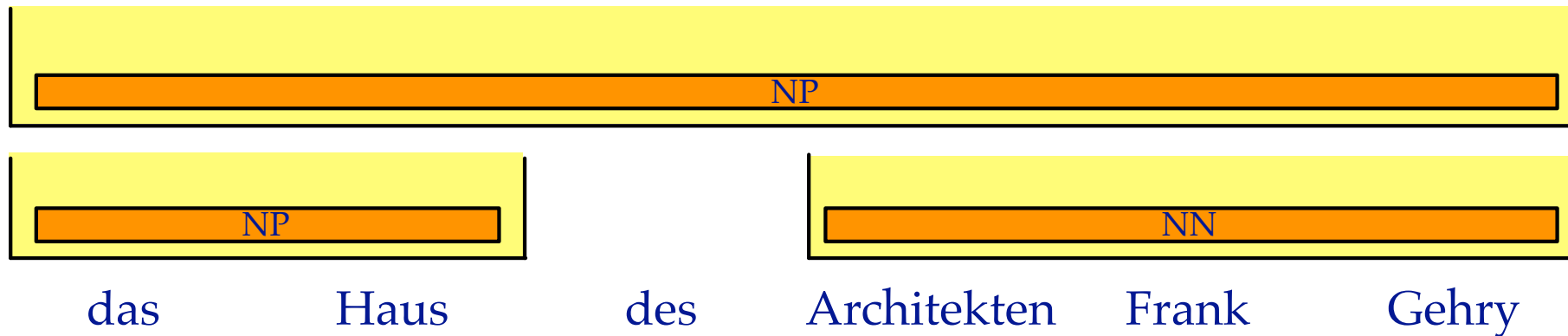
NP • **des** • **NN** → **NP: NP of the NN**



Rule Applicability Check

Matched entire rule \rightarrow apply to create a NP hypothesis

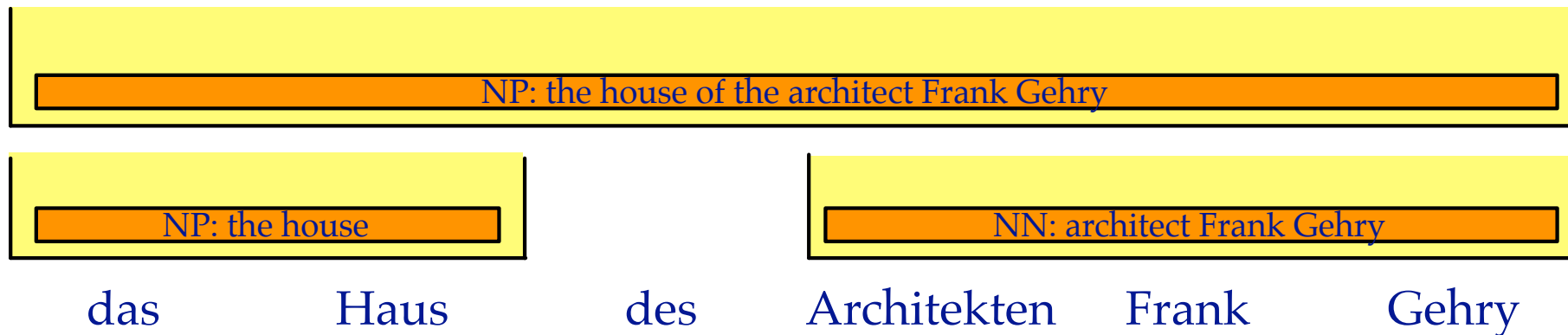
NP • des • NN \rightarrow NP: NP of the NN



Rule Applicability Check

Look up output words to create new hypothesis
(note: there may be many matching underlying NP and NN hypotheses)

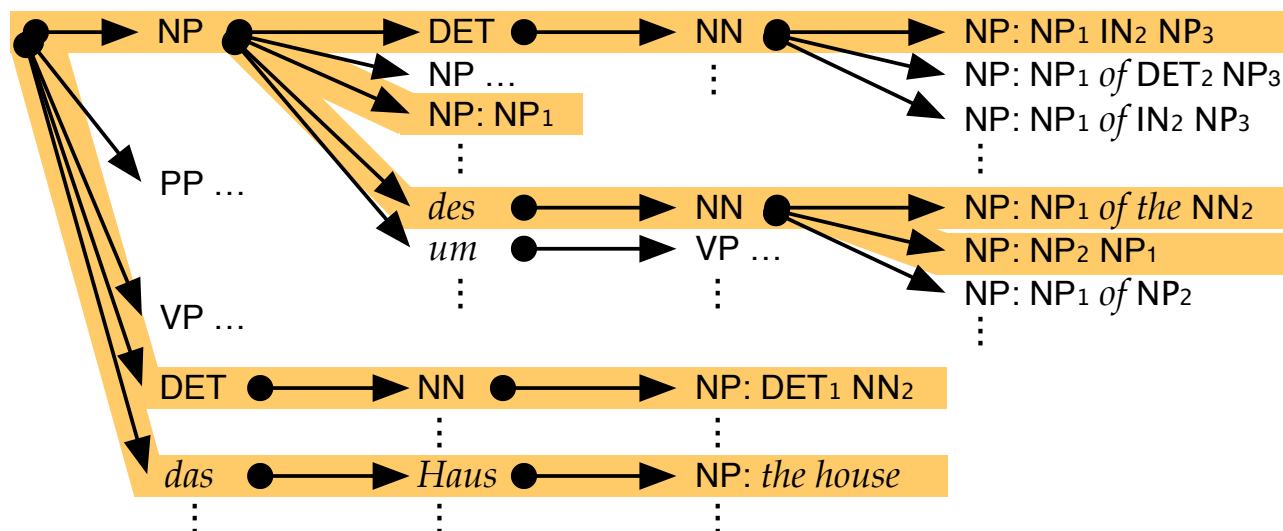
NP • des • NN → NP: NP of the NN



Checking Rules vs. Finding Rules

- What we showed:
 - given a rule
 - check if and how it can be applied
- But there are too many rules (millions) to check them all
- Instead:
 - given the underlying chart cells and input words
 - find which rules apply

Prefix Tree for Rules



Highlighted Rules

$NP \rightarrow NP_1 \text{ DET}_2 \text{ NN}_3 \mid NP_1 \text{ IN}_2 \text{ NN}_3$
 $NP \rightarrow NP_1 \mid NP_1$
 $NP \rightarrow NP_1 \text{ des } \text{NN}_2 \mid NP_1 \text{ of the } \text{NN}_2$
 $NP \rightarrow NP_1 \text{ des } \text{NN}_2 \mid NP_2 \text{ NP}_1$
 $NP \rightarrow \text{DET}_1 \text{ NN}_2 \mid \text{DET}_1 \text{ NN}_2$
 $NP \rightarrow \text{das Haus} \mid \text{the house}$

dotted rules

Dotted Rules: Key Insight

- If we can apply a rule like

$$p \rightarrow A B C \mid x$$

to a span

- Then we could have applied a rule like

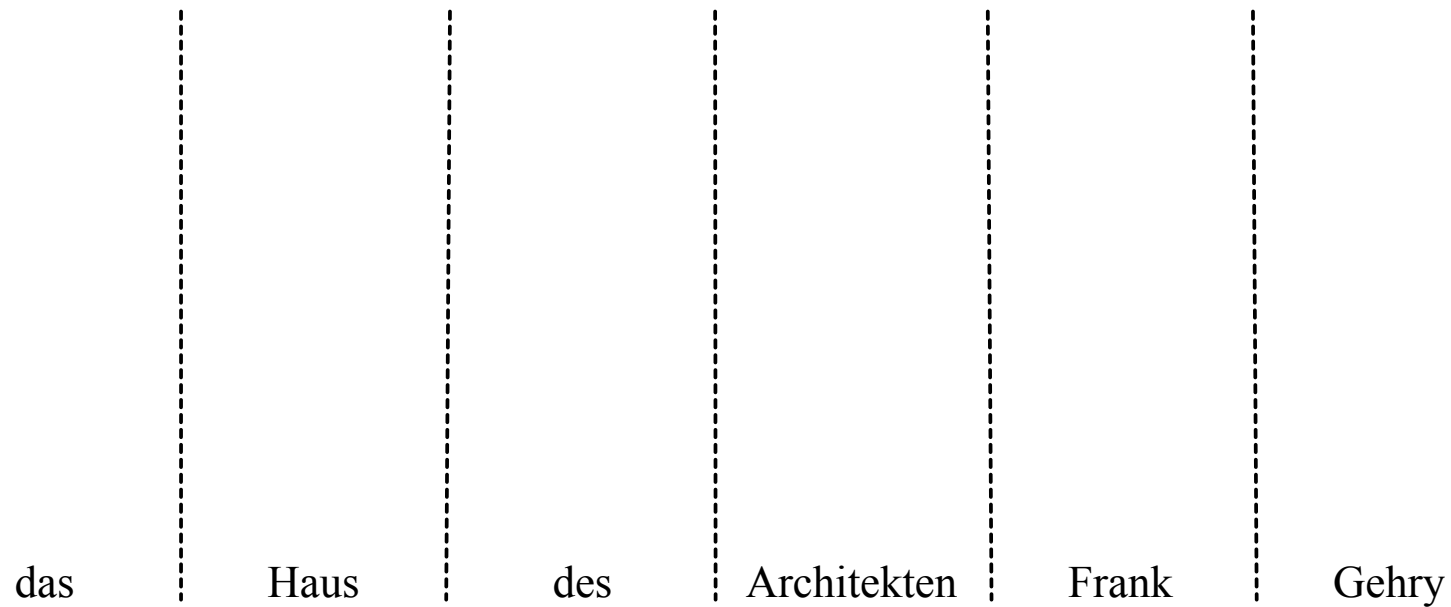
$$q \rightarrow A B \mid y$$

to a sub-span with the same starting word

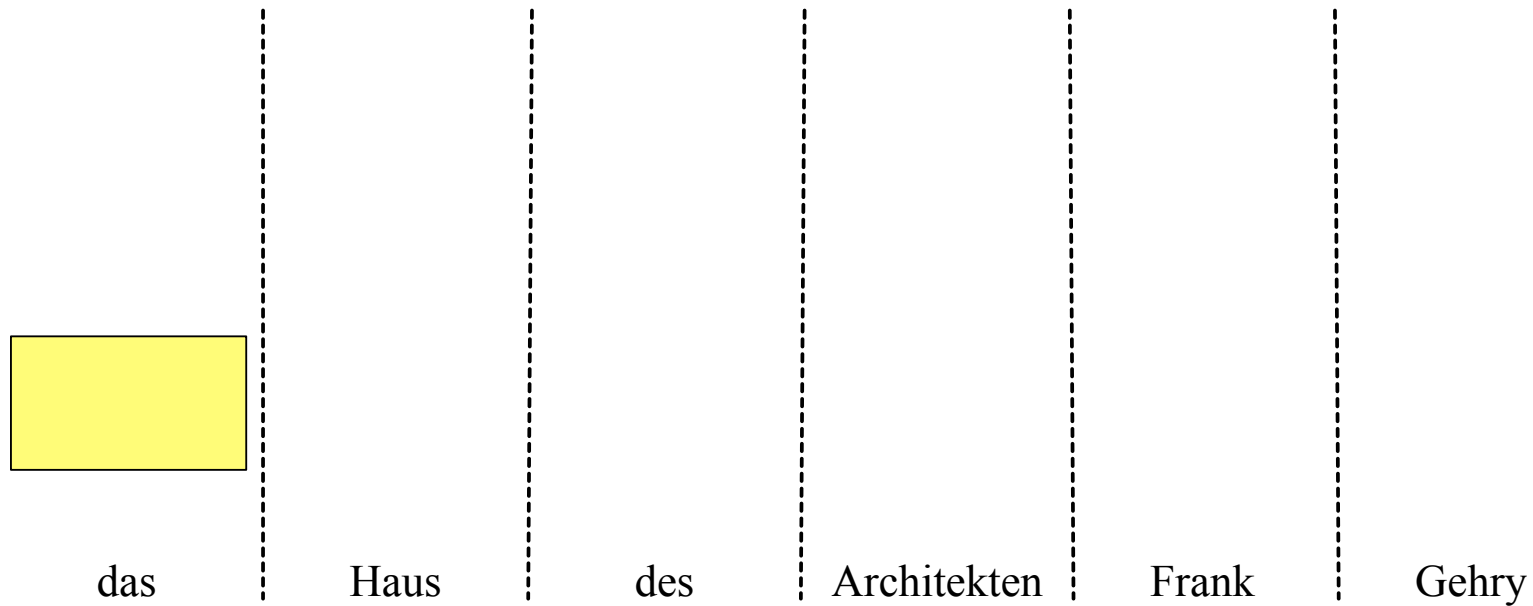
\Rightarrow We can re-use rule lookup by storing $A B \bullet$ (dotted rule)

Finding Applicable Rules in Prefix Tree

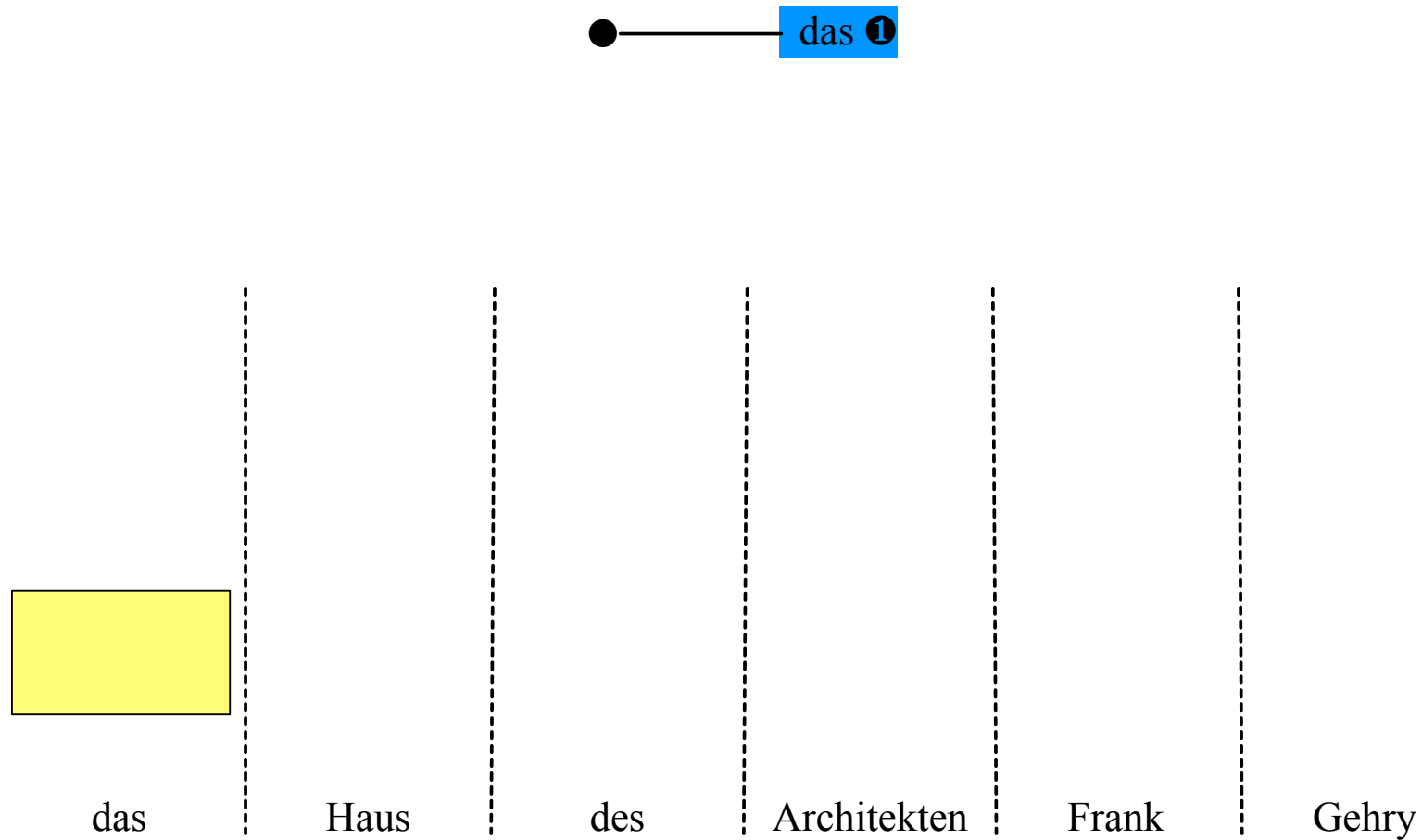
32



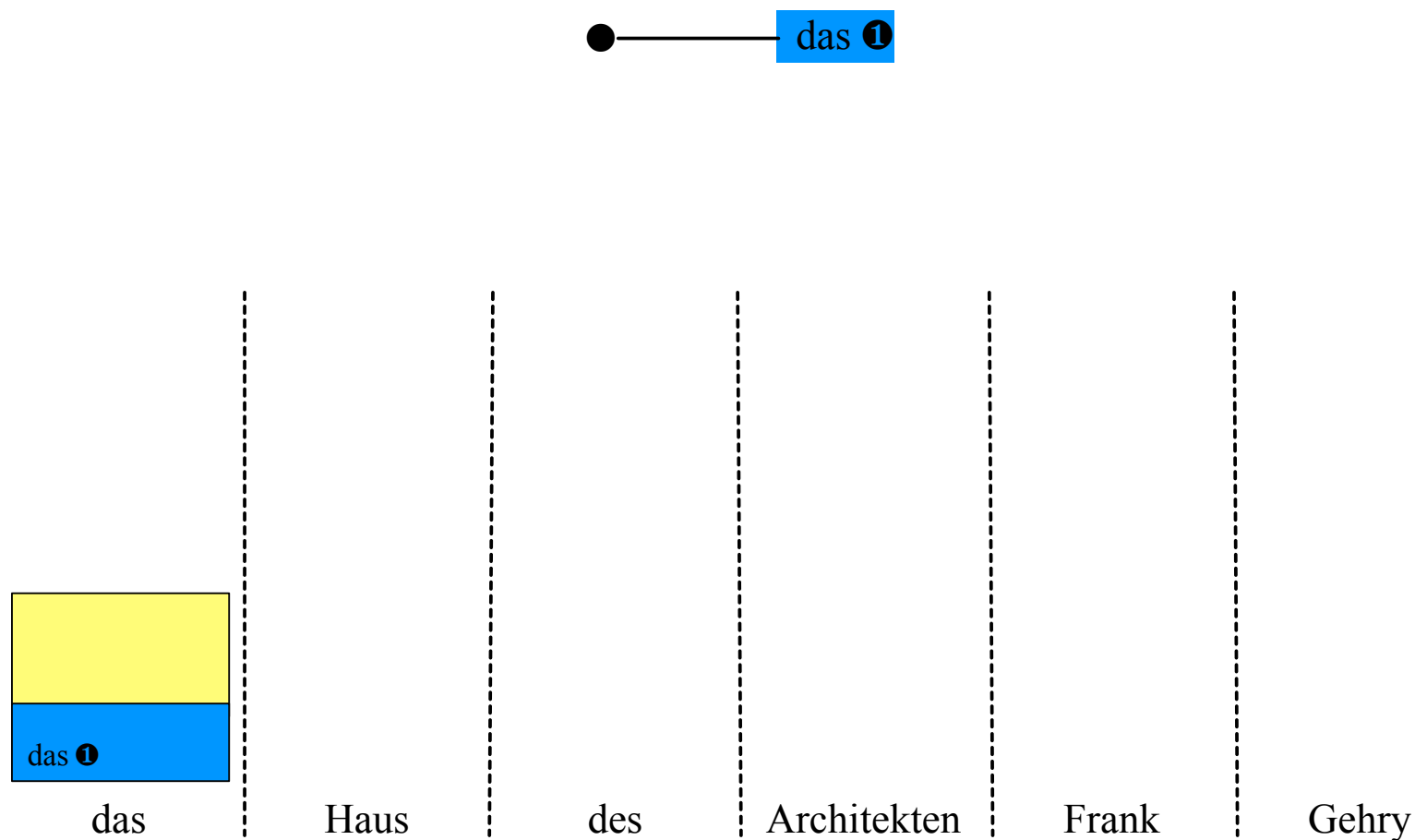
Covering the First Cell



Looking up Rules in the Prefix Tree

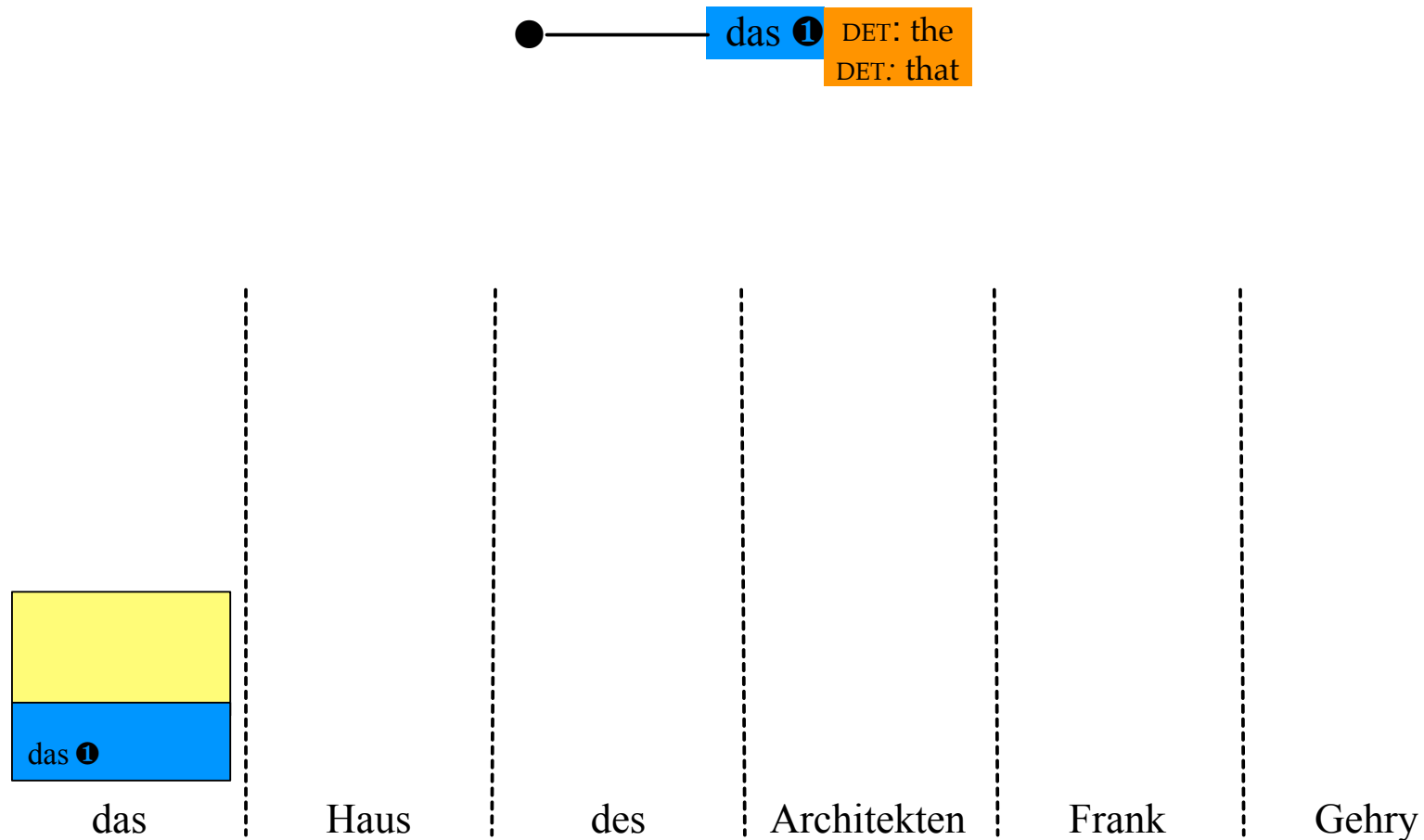


Taking Note of the Dotted Rule

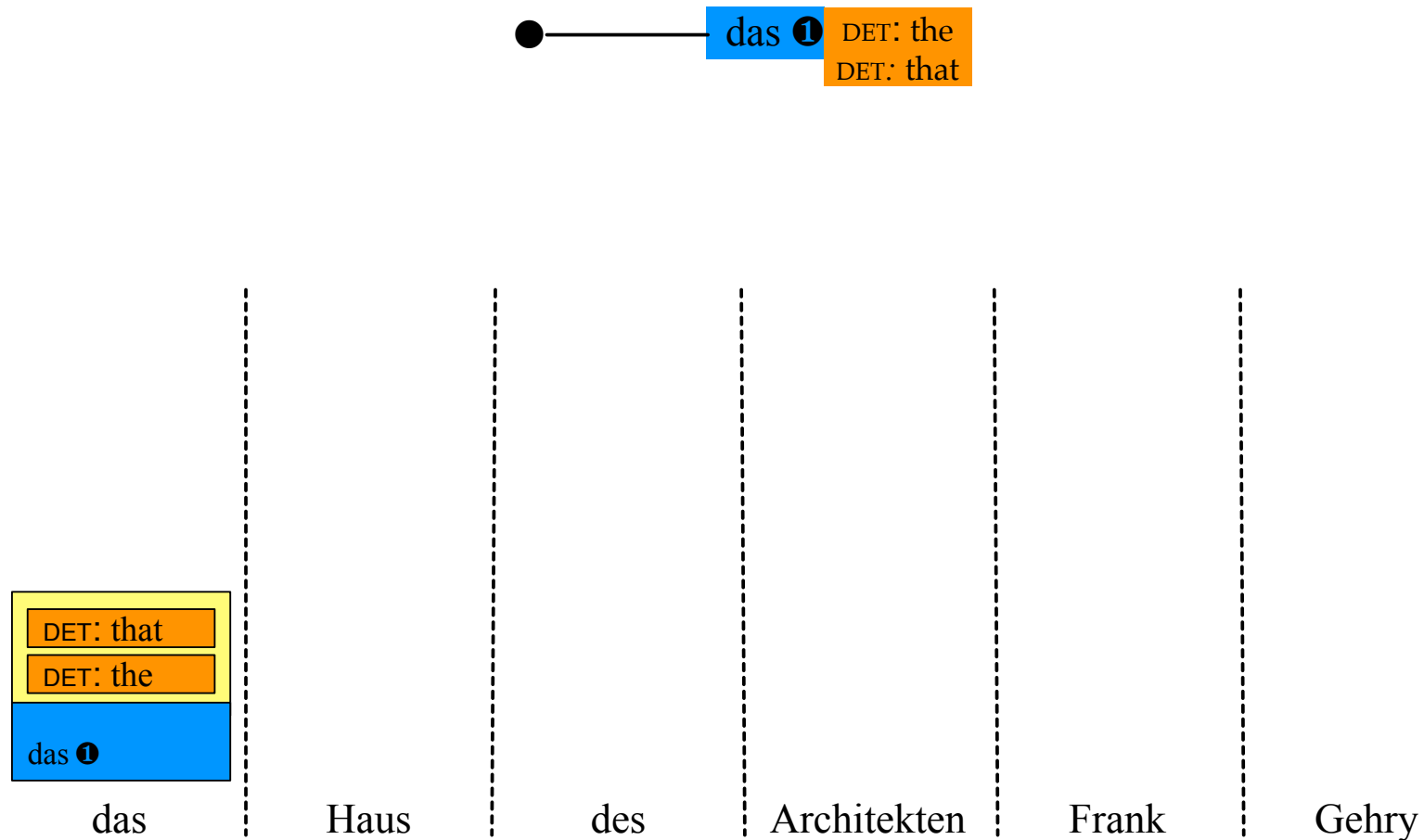


Checking if Dotted Rule has Translations

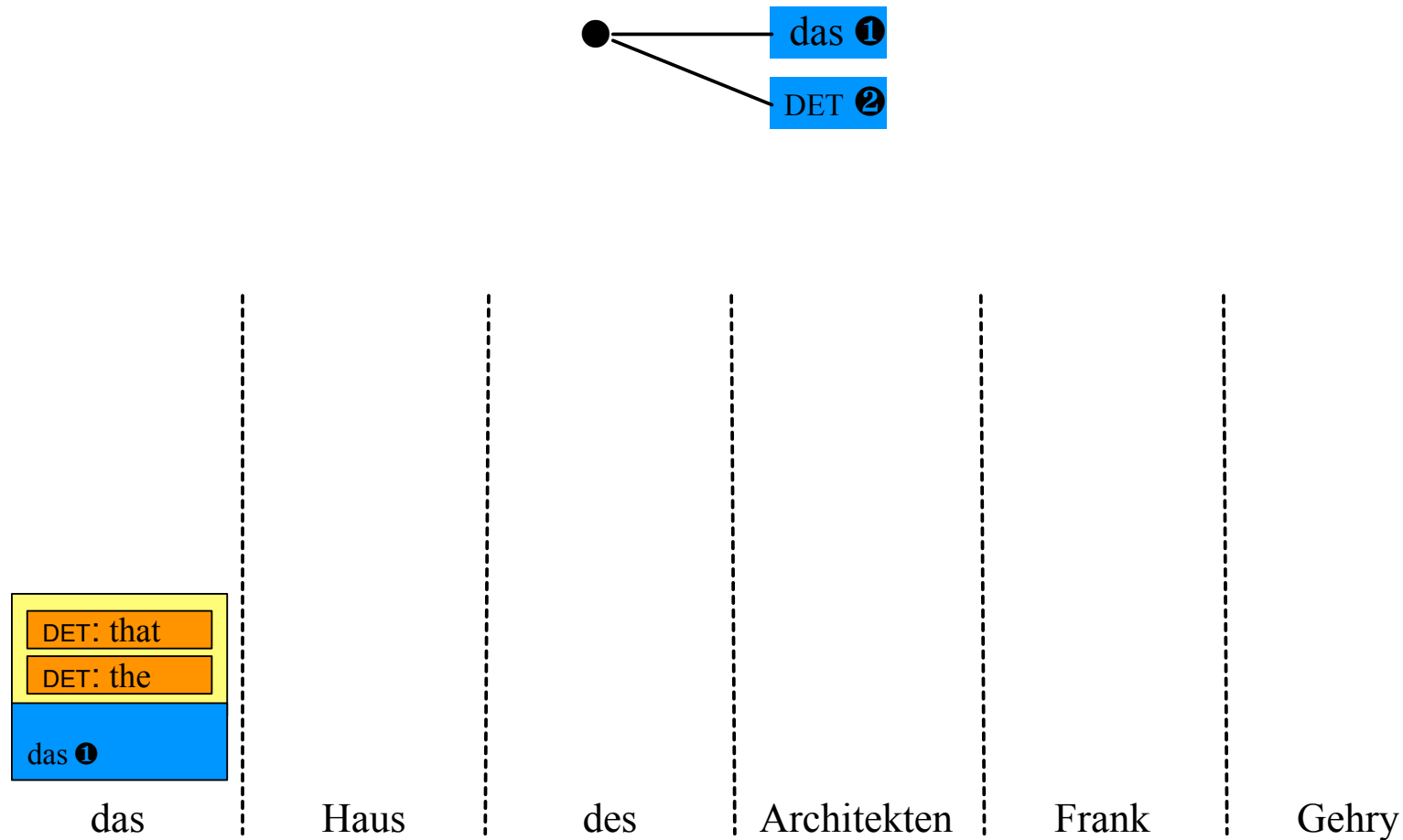
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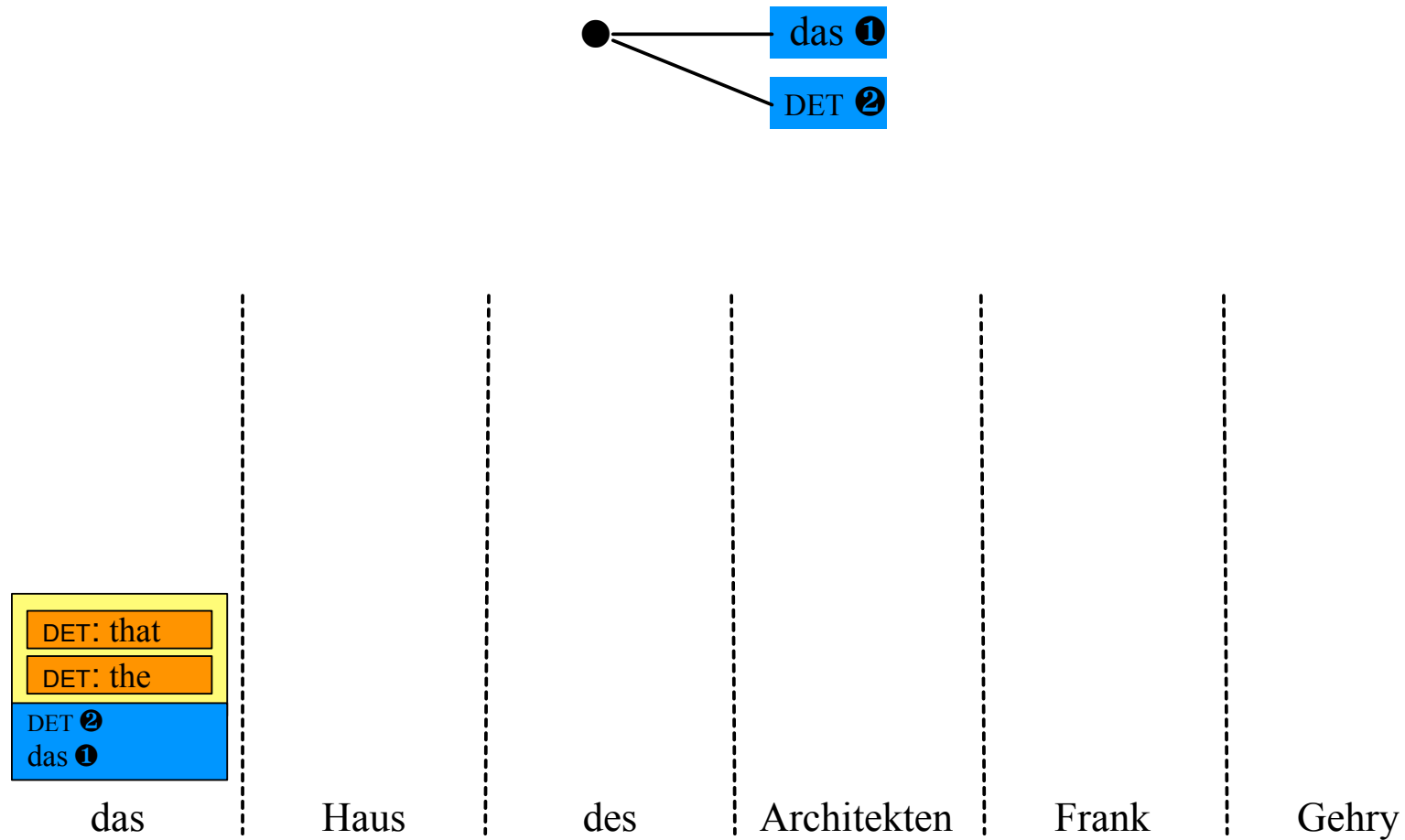
Applying the Translation Rules



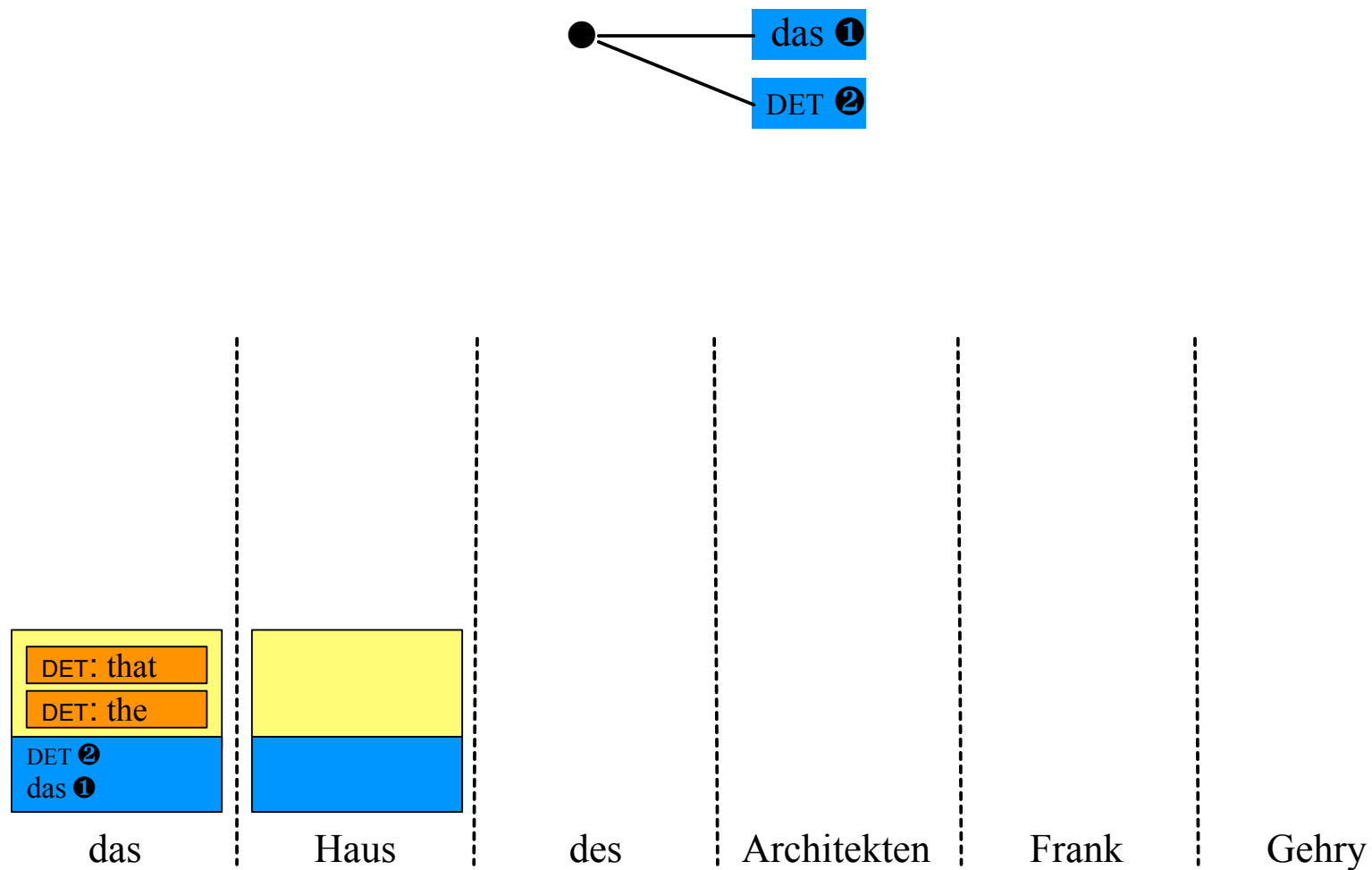
Looking up Constituent Label in Prefix Tree 38



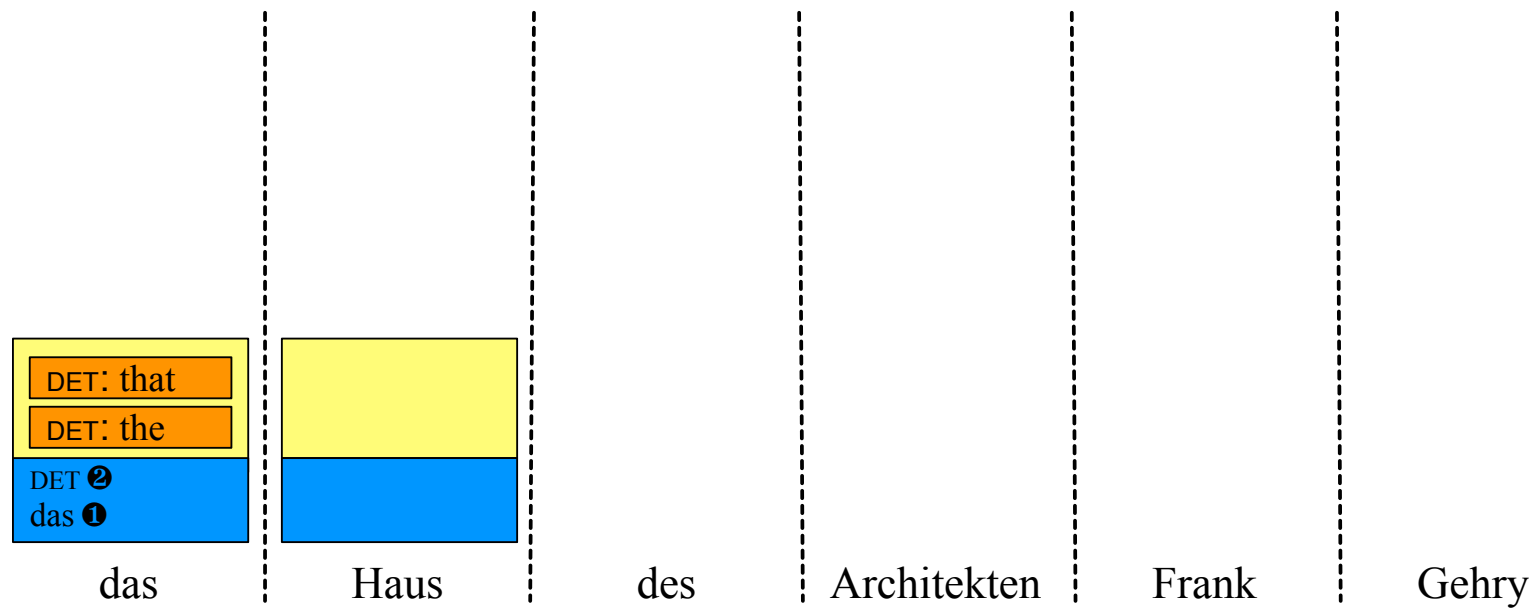
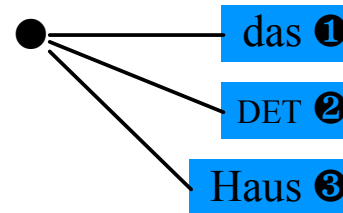
Add to Span's List of Dotted Rules



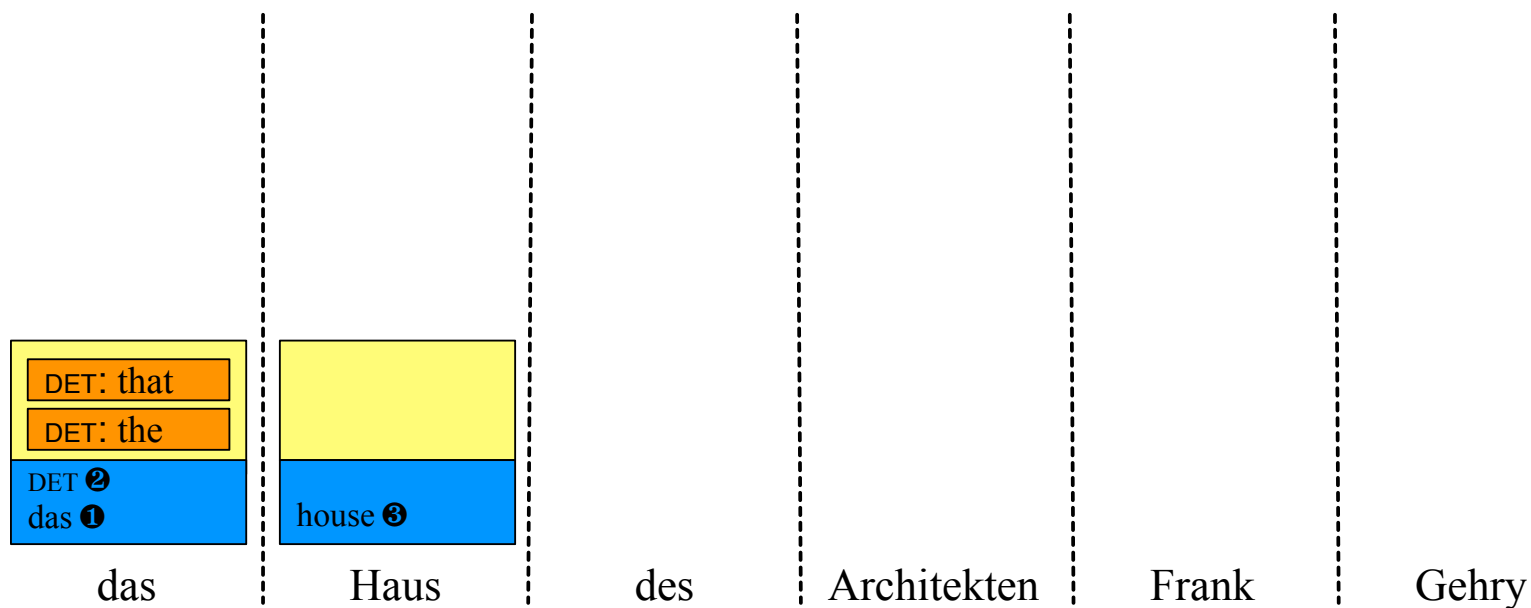
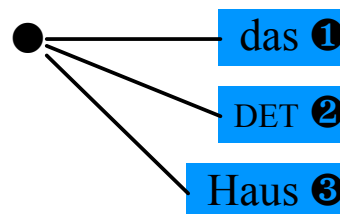
Moving on to the Next Cell



Looking up Rules in the Prefix Tree

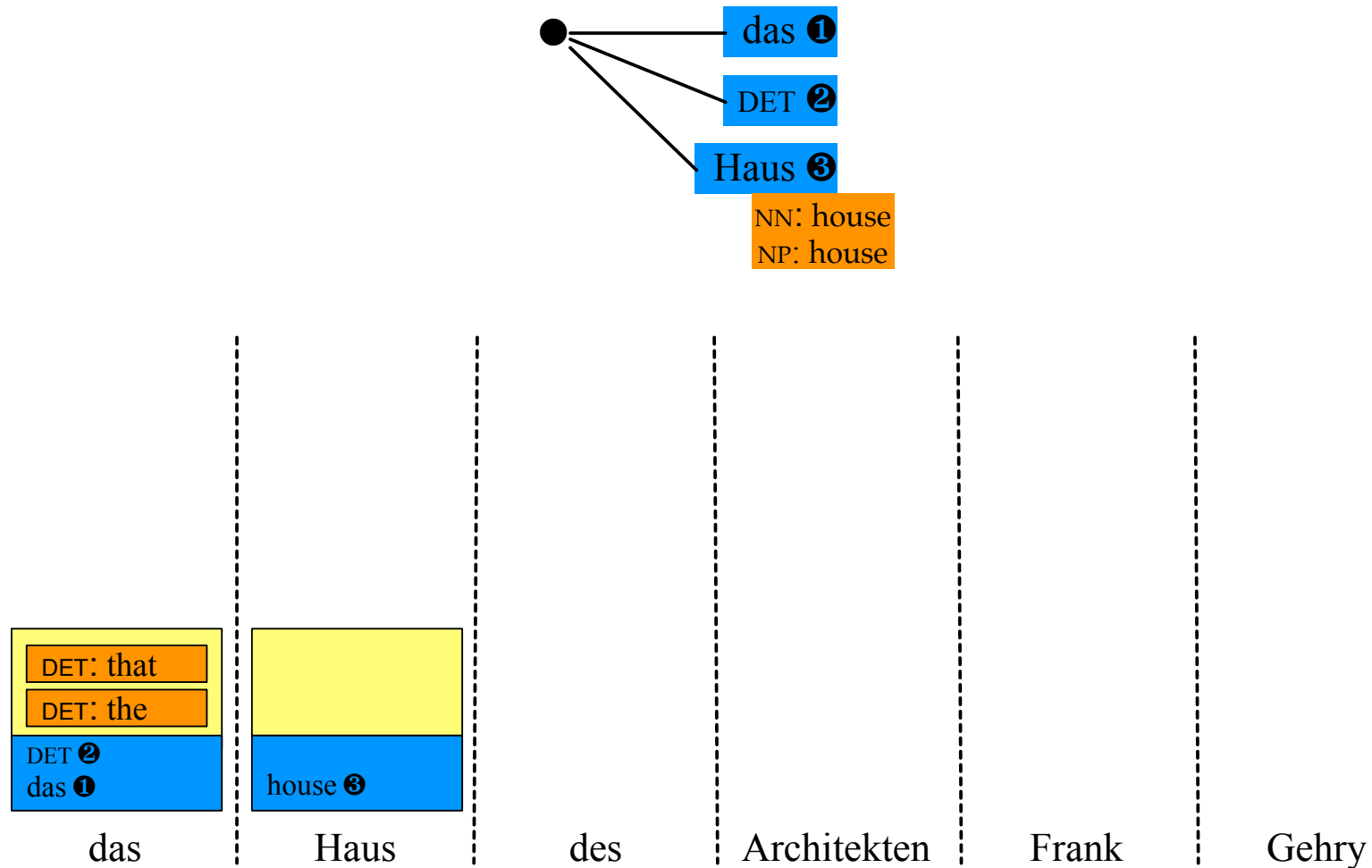


Taking Note of the Dotted Rule

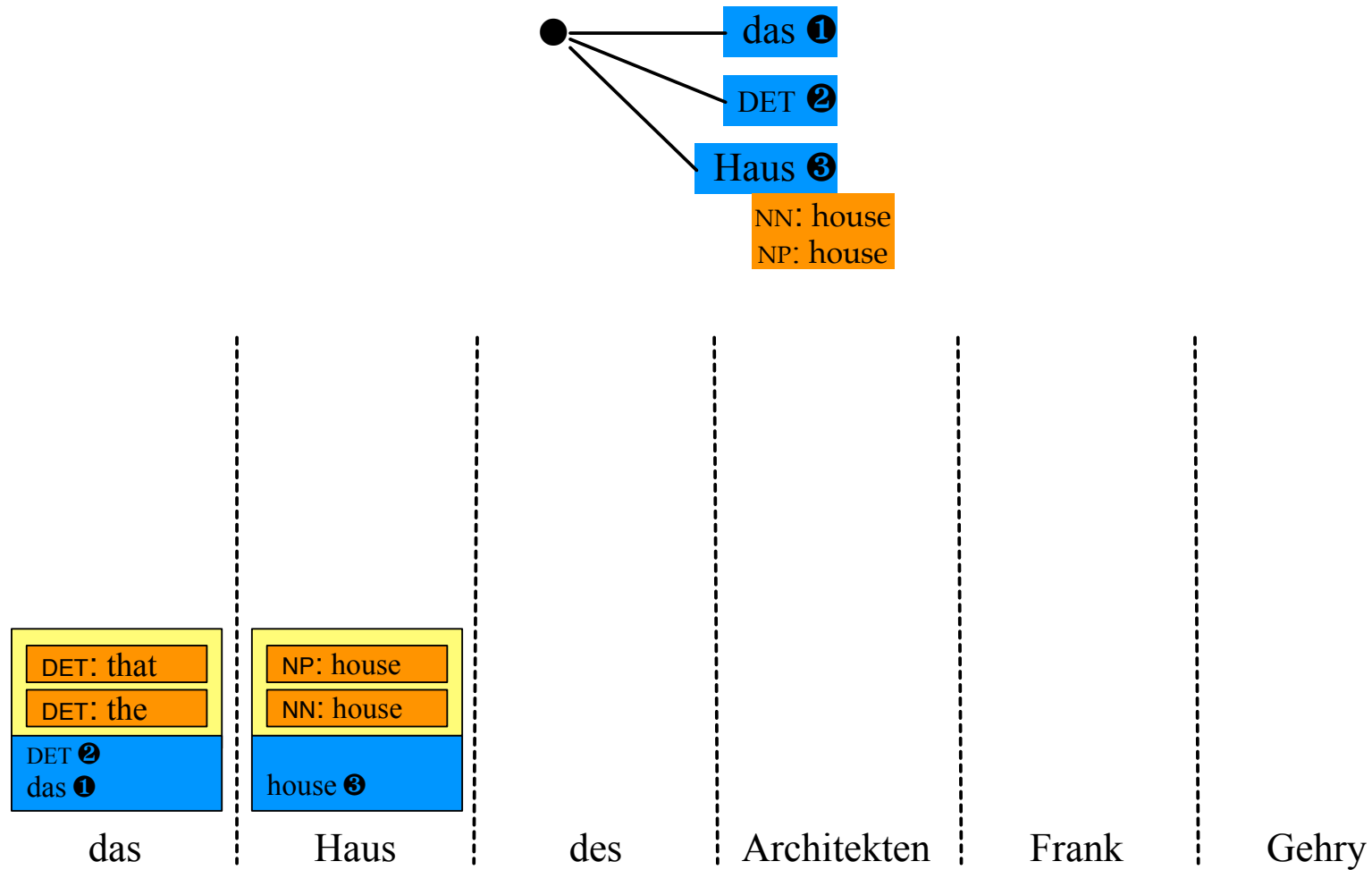


Checking if Dotted Rule has Translations

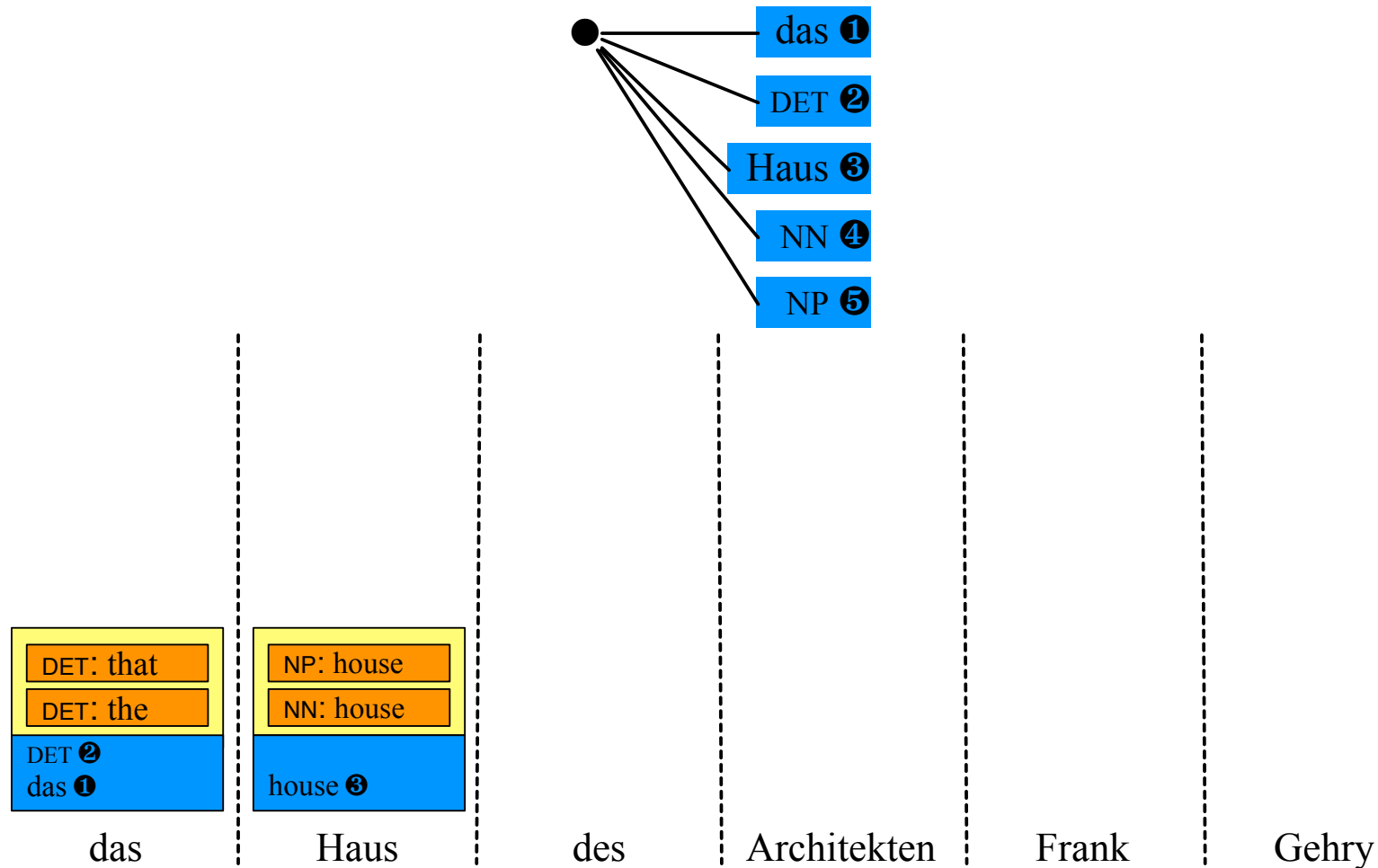
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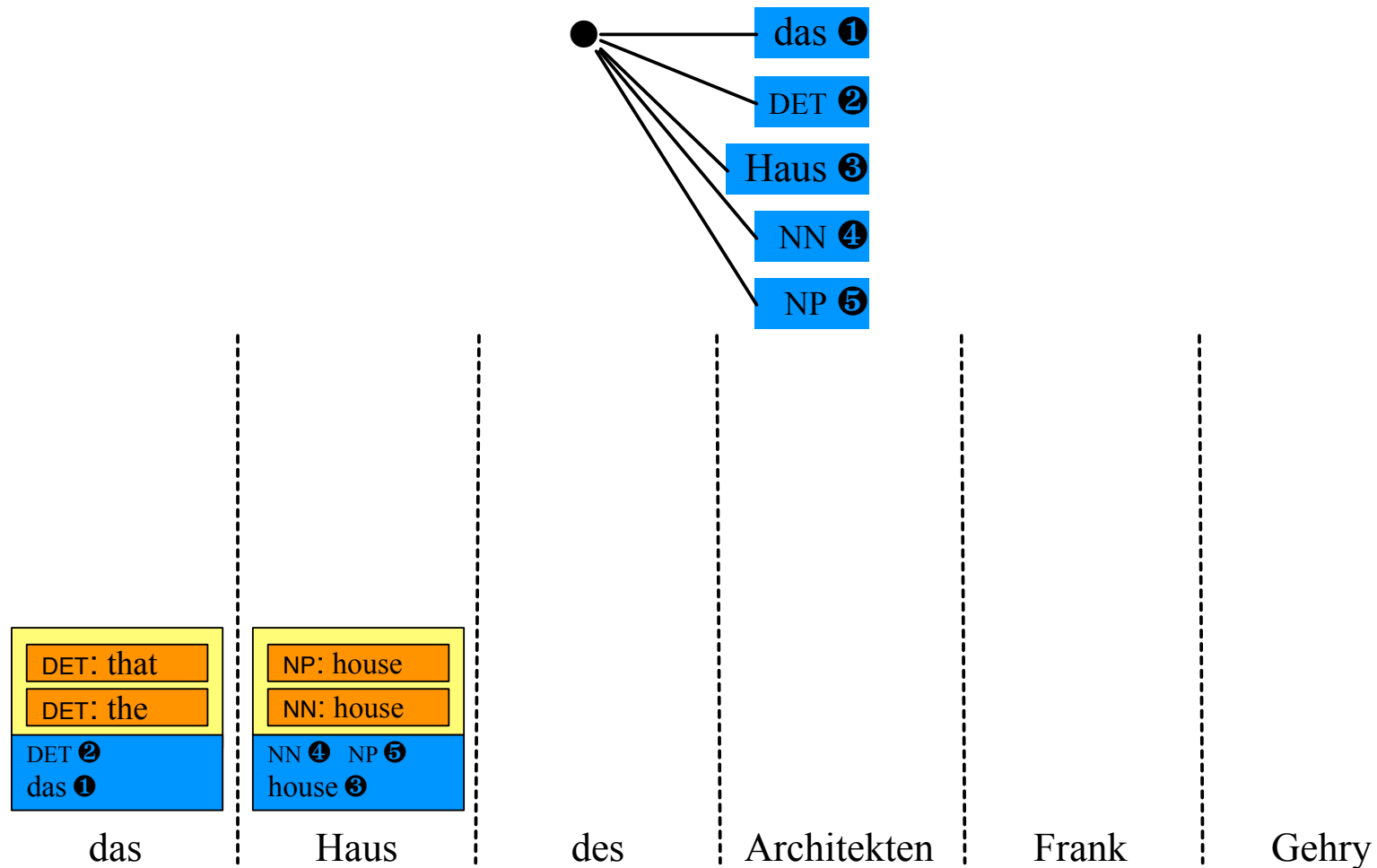
Applying the Translation Rules



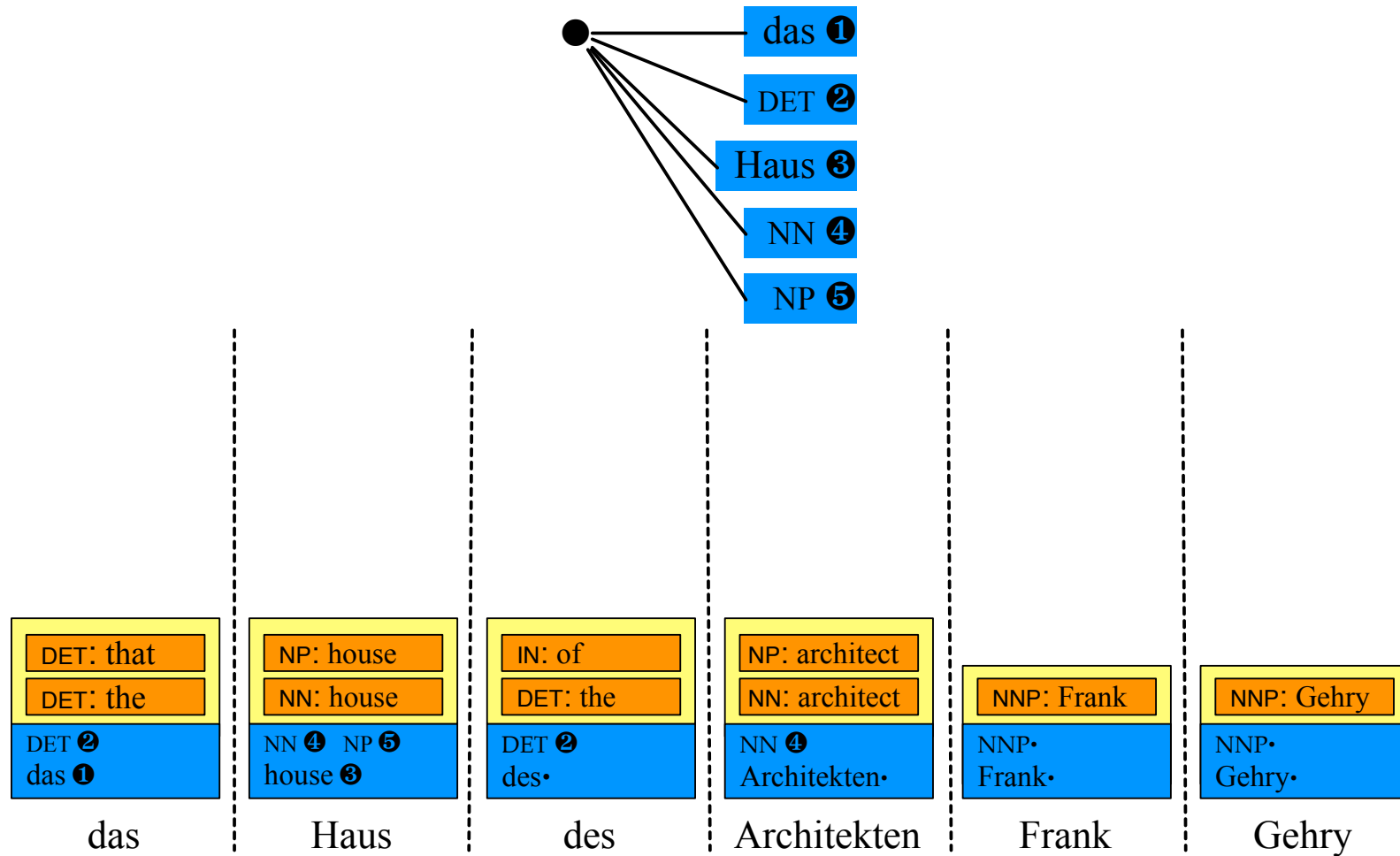
Looking up Constituent Label in Prefix Tree 45



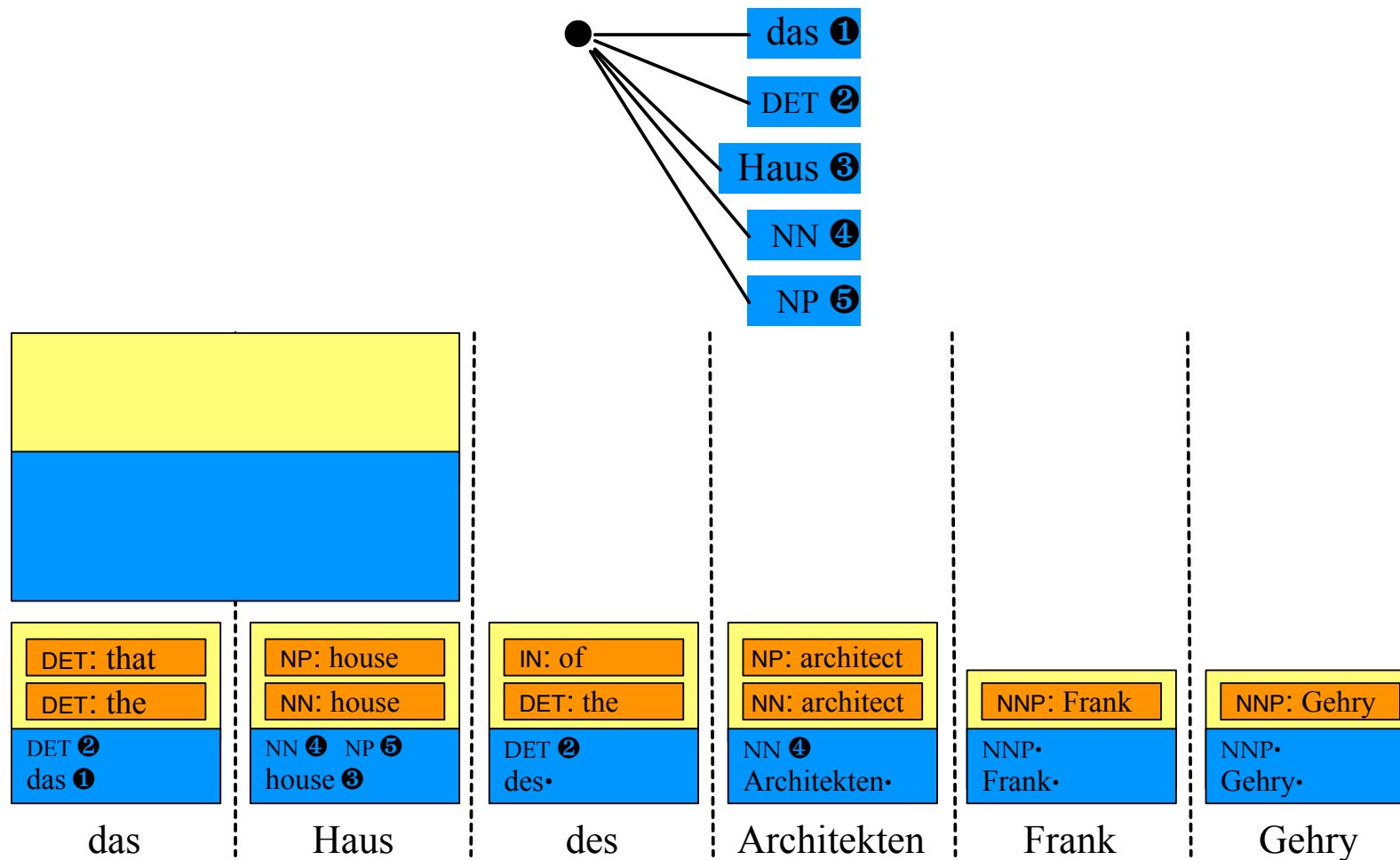
Add to Span's List of Dotted Rules



More of the Same

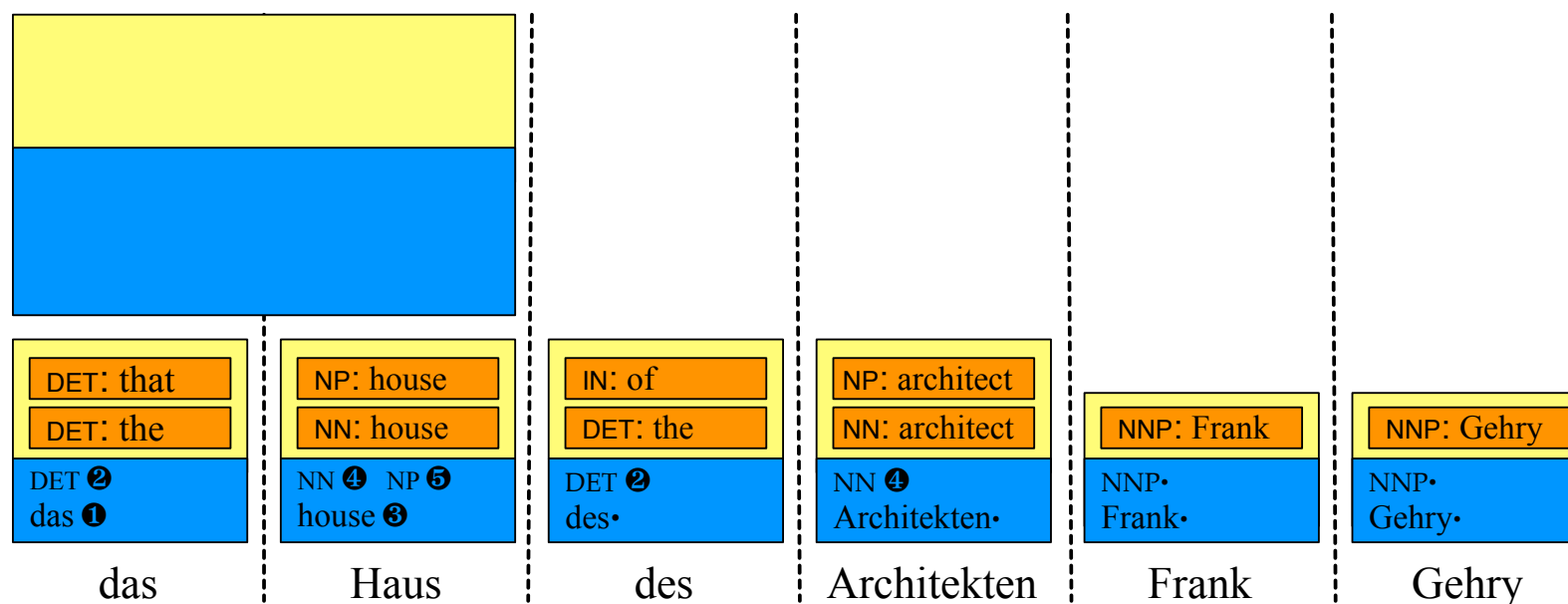


Moving on to the Next Cell

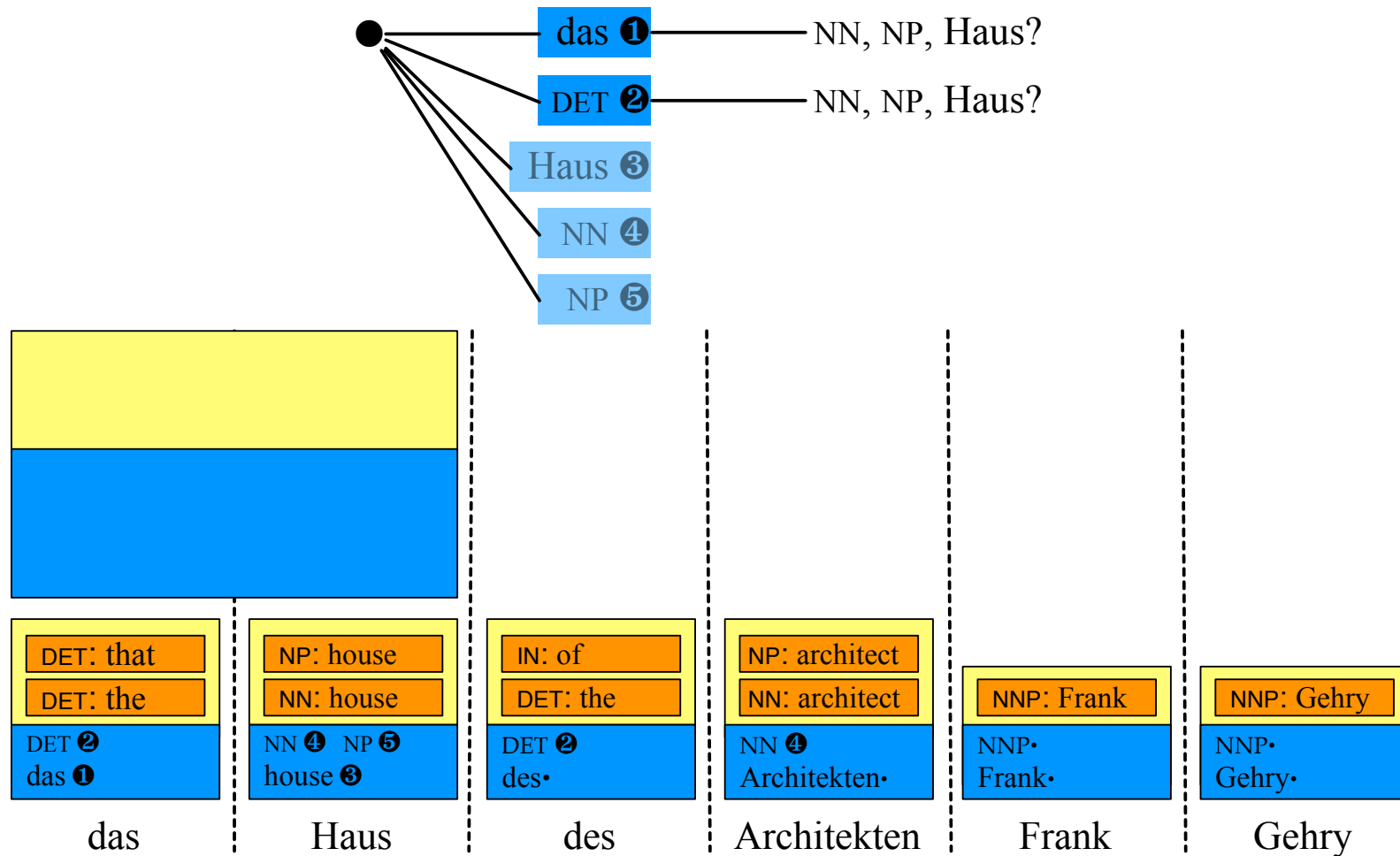


Covering a Longer Span

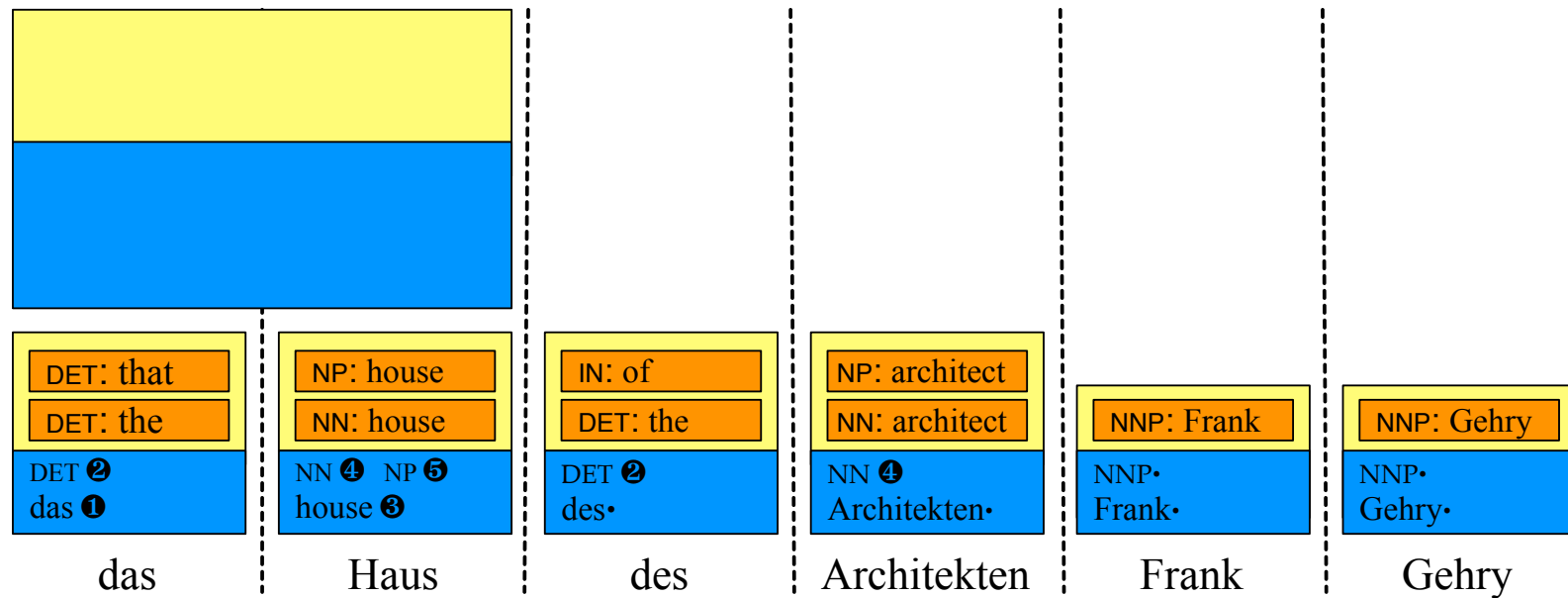
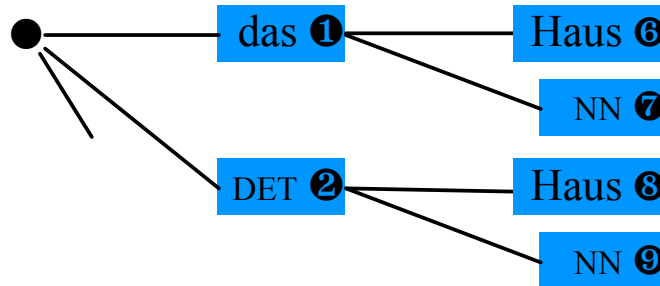
Cannot consume multiple words at once
All rules are extensions of existing dotted rules
Here: only extensions of span over **das** possible



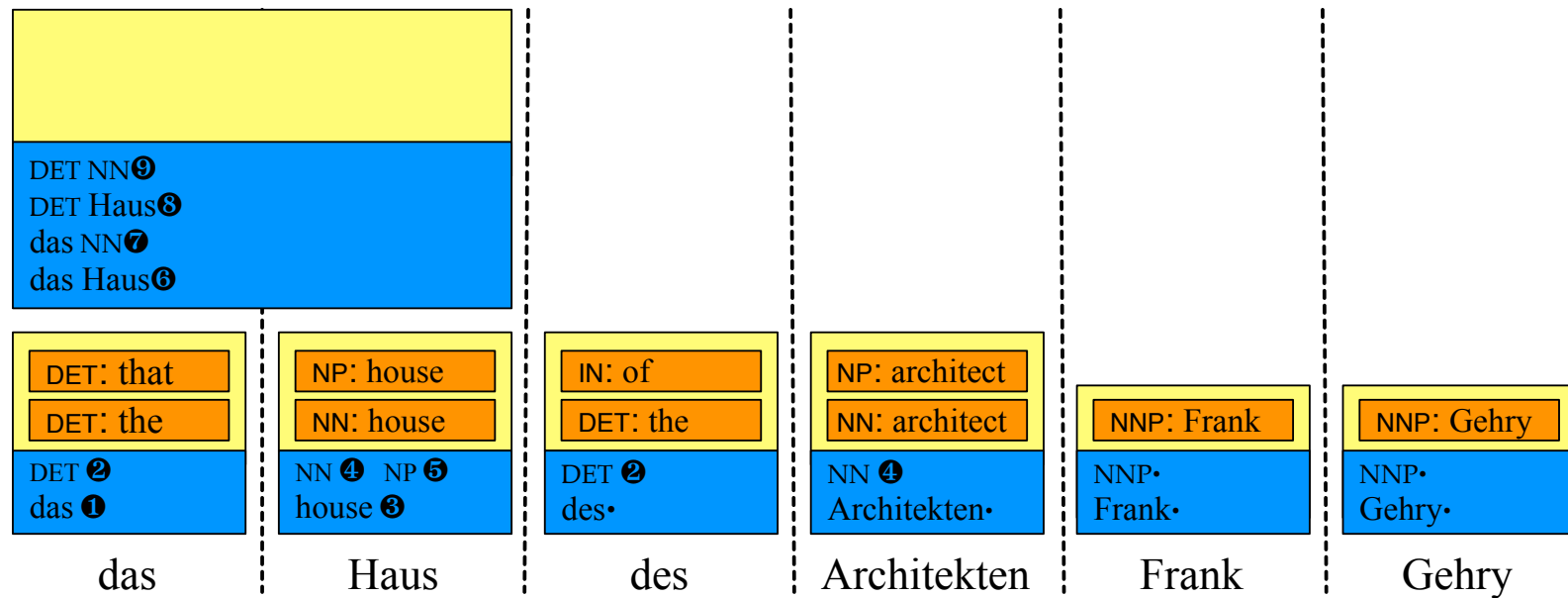
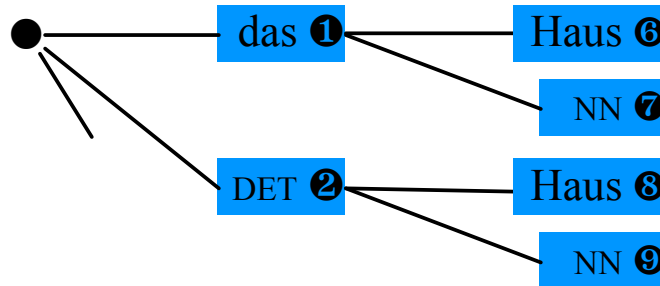
Extensions of Span over das



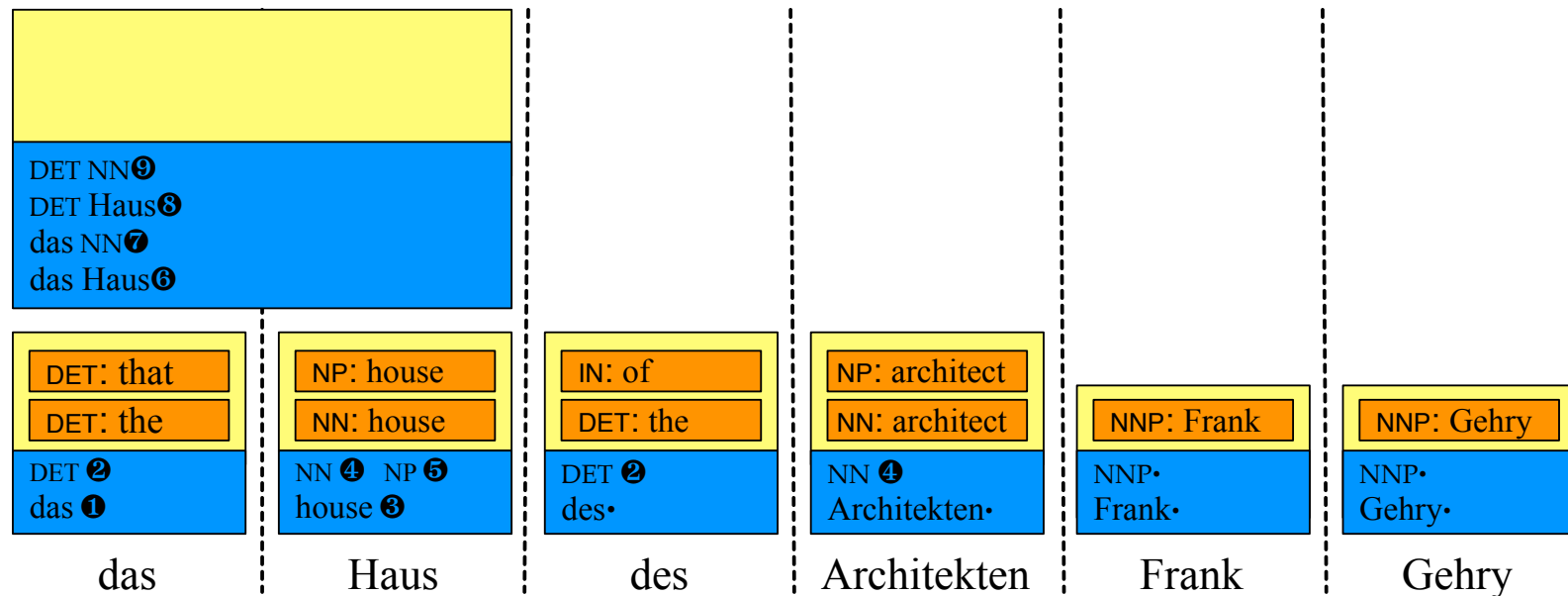
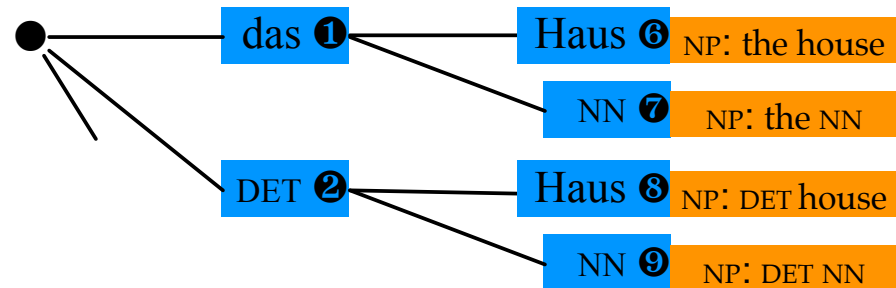
Looking up Rules in the Prefix Tree



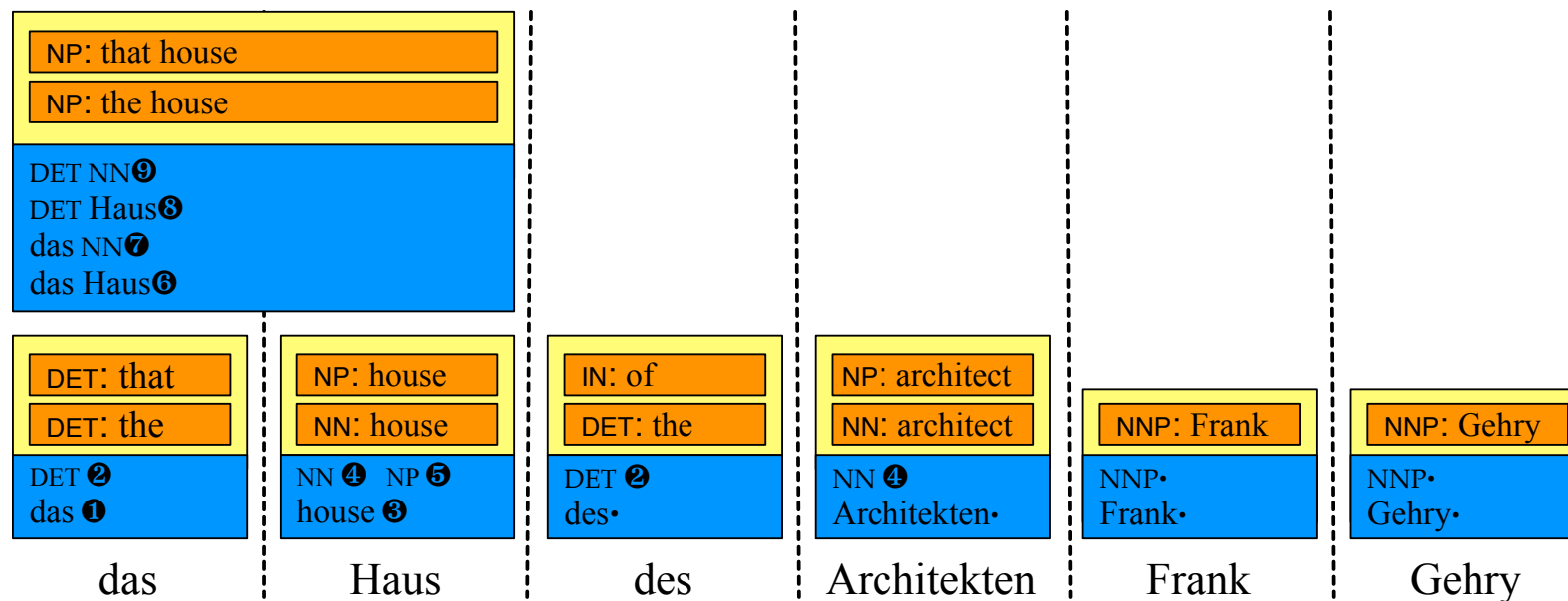
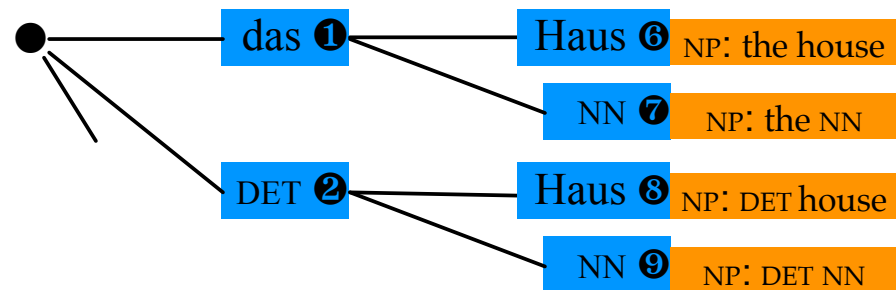
Taking Note of the Dotted Rule



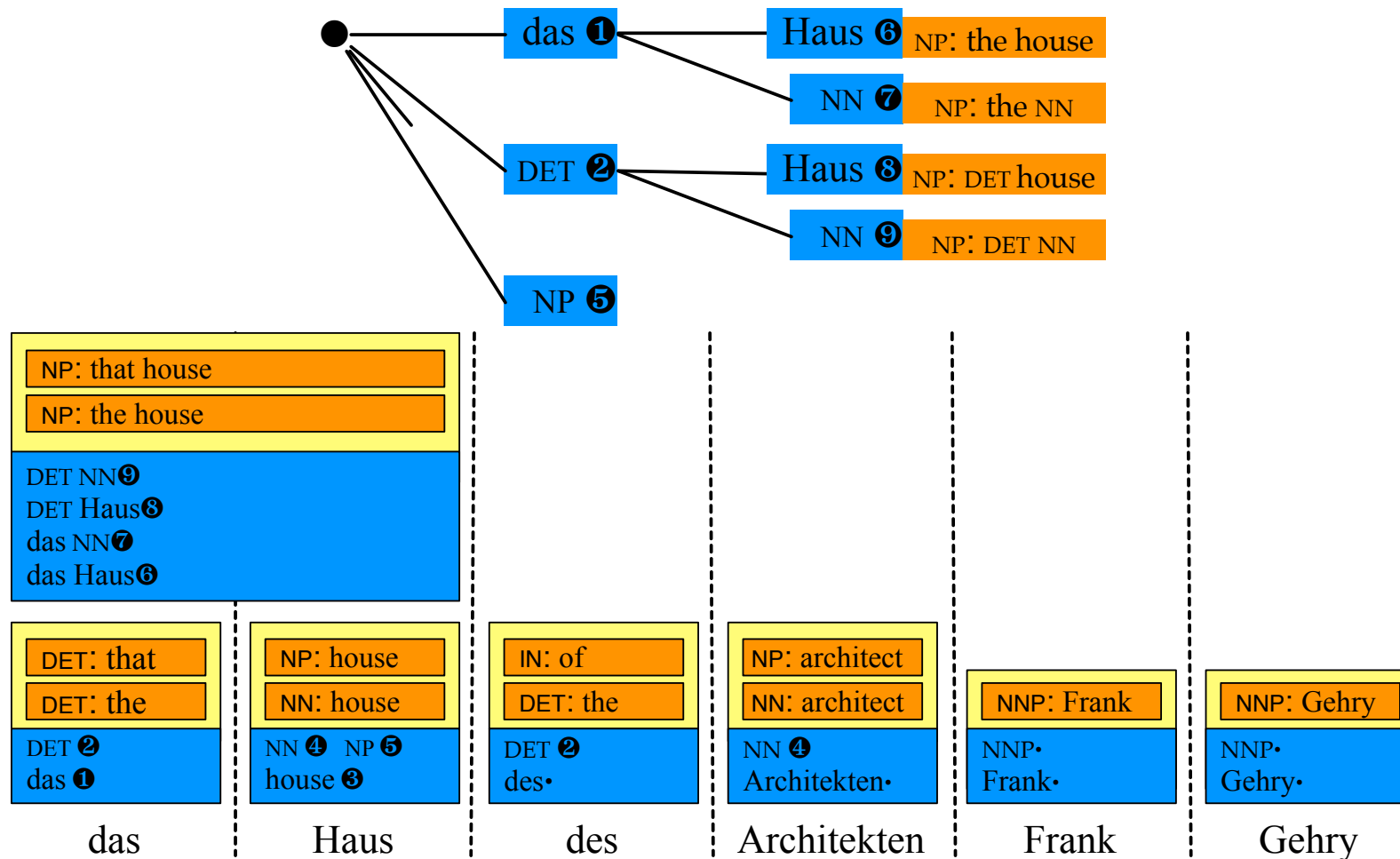
Checking if Dotted Rules have Translations 53



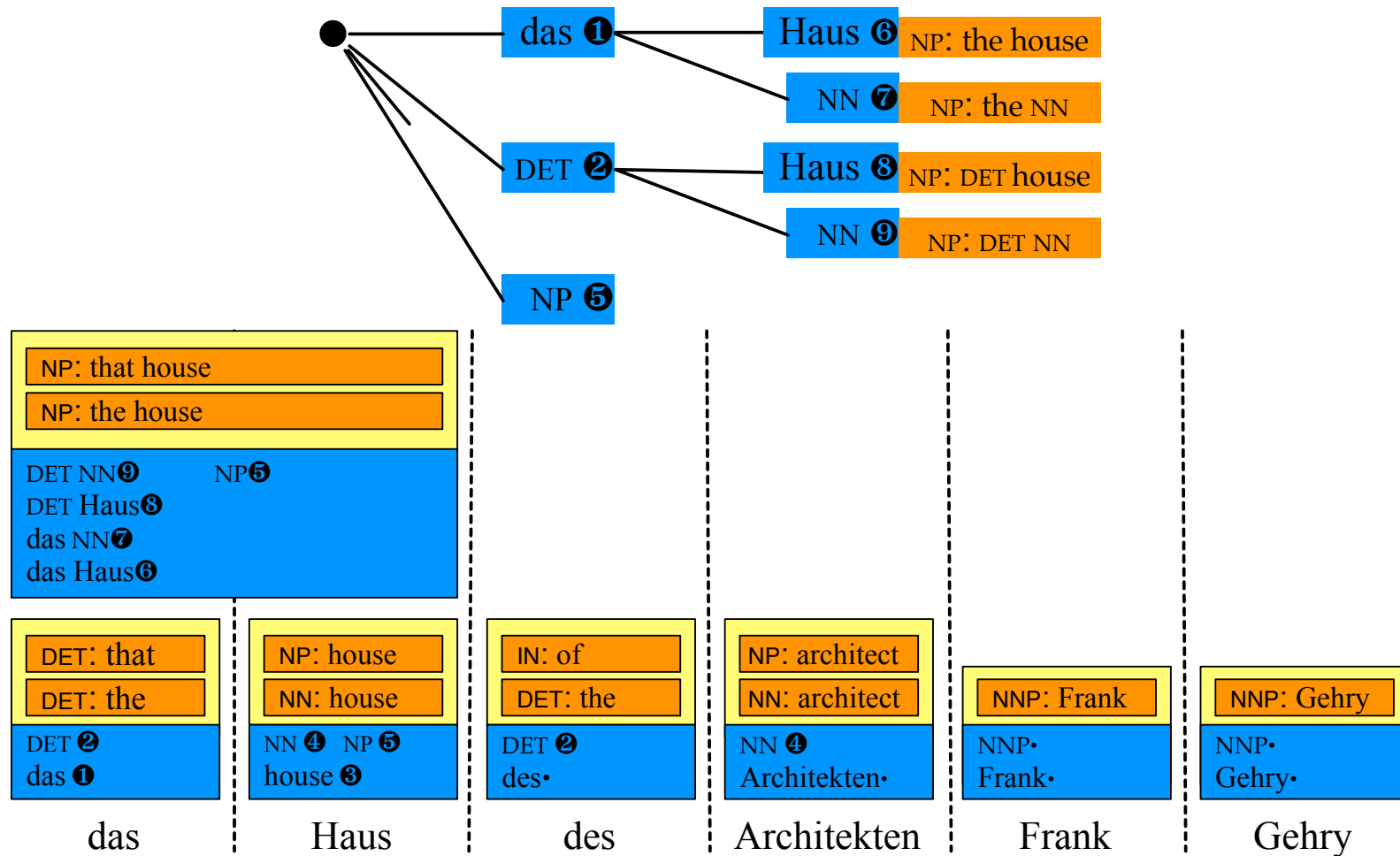
Applying the Translation Rules



Looking up Constituent Label in Prefix Tree 55



Add to Span's List of Dotted Rules



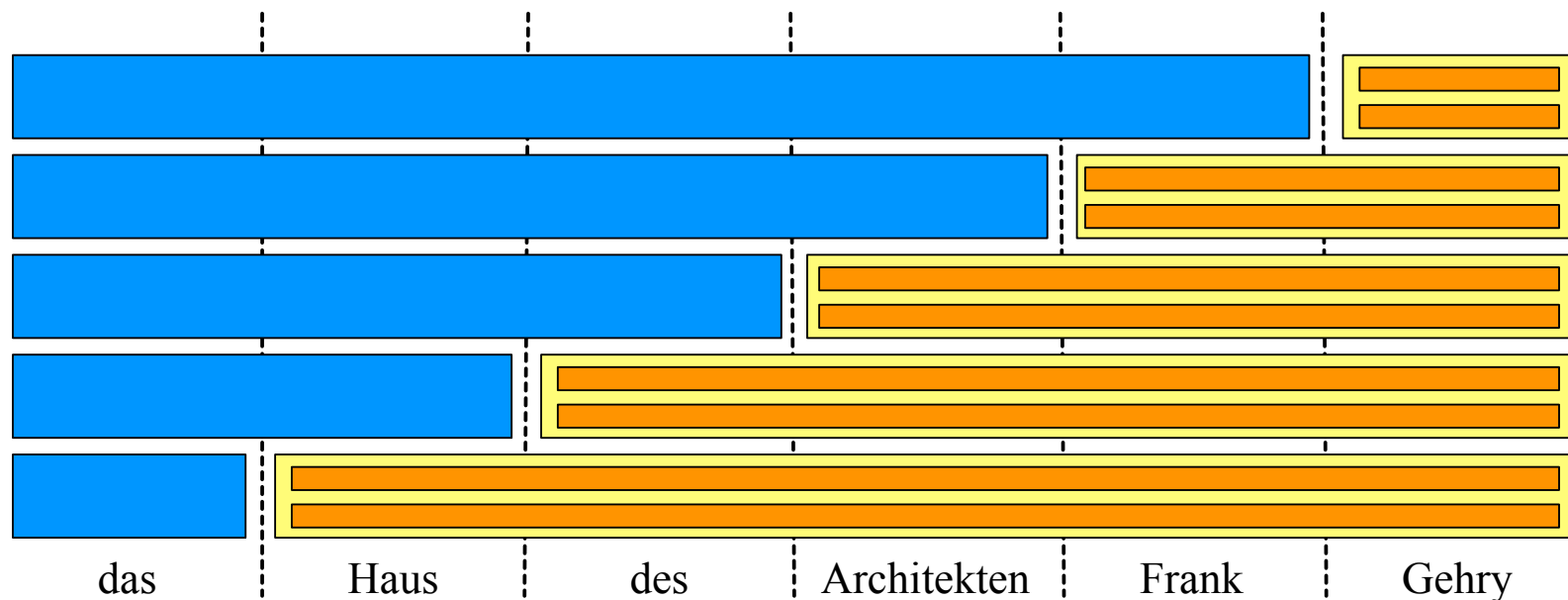
Even Larger Spans

Extend lists of dotted rules with cell constituent labels

span's dotted rule list (with same start)

plus neighboring

span's constituent labels of hypotheses (with same end)



- Complexity $O(rn^3)$ with sentence length n and size of dotted rule list r
 - may introduce maximum size for spans that do not start at beginning
 - may limit size of dotted rule list (very arbitrary)
- Does the list of dotted rules explode?
- Yes, if there are many rules with neighboring target-side non-terminals
 - such rules apply in many places
 - rules with words are much more restricted

- Some rules may apply in too many ways
- Neighboring input non-terminals

$NP \rightarrow X_1 X_2 \mid NP_2 \text{ to } NP_1$

- non-terminals may match many different pairs of spans
 - especially a problem for hierarchical models (no constituent label restrictions)
 - may be okay for syntax-models
- Three neighboring input non-terminals

$VP \rightarrow \text{trifft } X_1 X_2 X_3 \text{ heute} \mid \text{meets } NP_1 \text{ today } PP_2 PP_3$

- will get out of hand even for syntax models

- Basic idea: bottom up chart parsing
- Prefix structure for easy rule access
- Caching rule matching with dotted rules
- Coming up...
 - cube pruning for syntax-based decoding
 - recombination and state
 - scope3 pruning
 - recursive cky+
 - coarse-to-fine