

Hieu Hoang

MT Marathon 2011 Trento

# **MIXED SYNTAX TRANSLATION**

# Contents

- What is a syntactic model?
- What's wrong with Syntax?
- Which syntax model to use?
- Why use syntactic models?
- Mixed-Syntax Model
  - Extraction
  - Decoding
  - Results
- Future Work

# What is a syntactic model?

- Hierarchical Phrase-Based Model
  - String-to-string
  - Non-terminals are unlabelled

$X \rightarrow \text{habe } X_1 \text{ gegessen} \ # \text{ have eaten } X_1$

- Tree-to-string Model
  - Source non-terminals are labelled
    - match input parse tree

$S \rightarrow \text{habe } NP_1 \text{ gegessen} \ # \text{ have eaten } NP_1$

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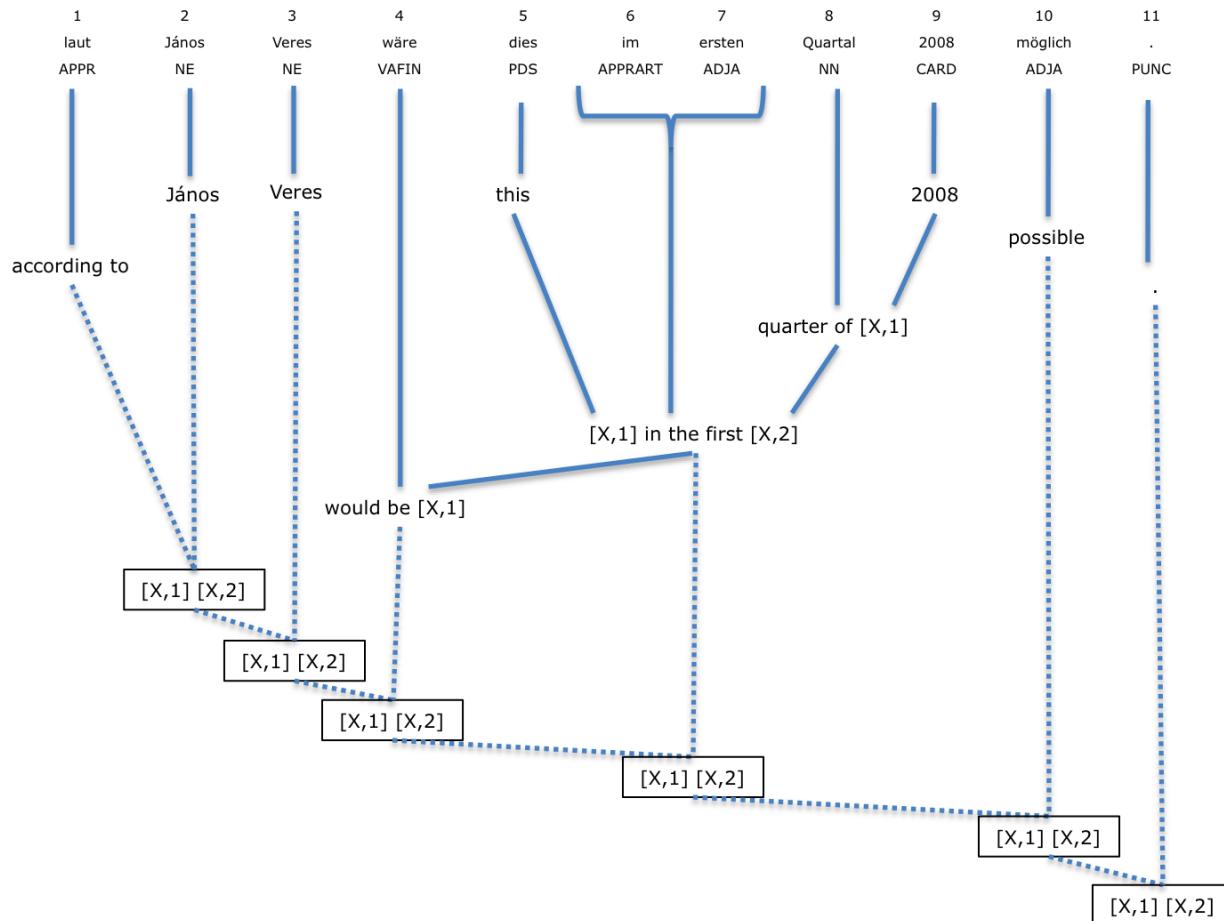
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# What's Wrong with Syntax?

	BLEU	METEOR
Tree-to-string	27.02	57.68
Tree-to-tree	22.23	54.05
Moses (phrase-based)	30.18	58.13

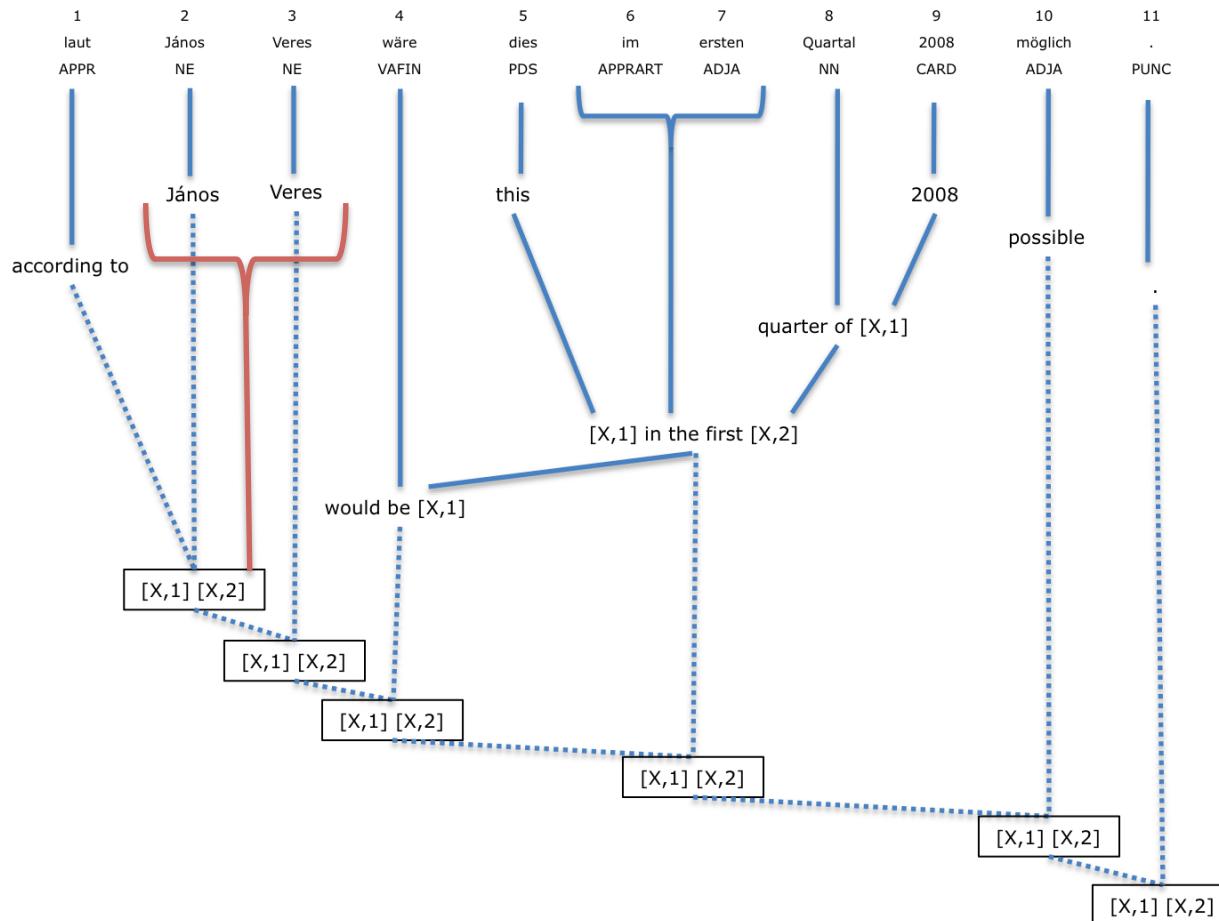
Evaluation of French-English MT System  
(Ambati and Lavie, 2009)

# Hierarchical Model



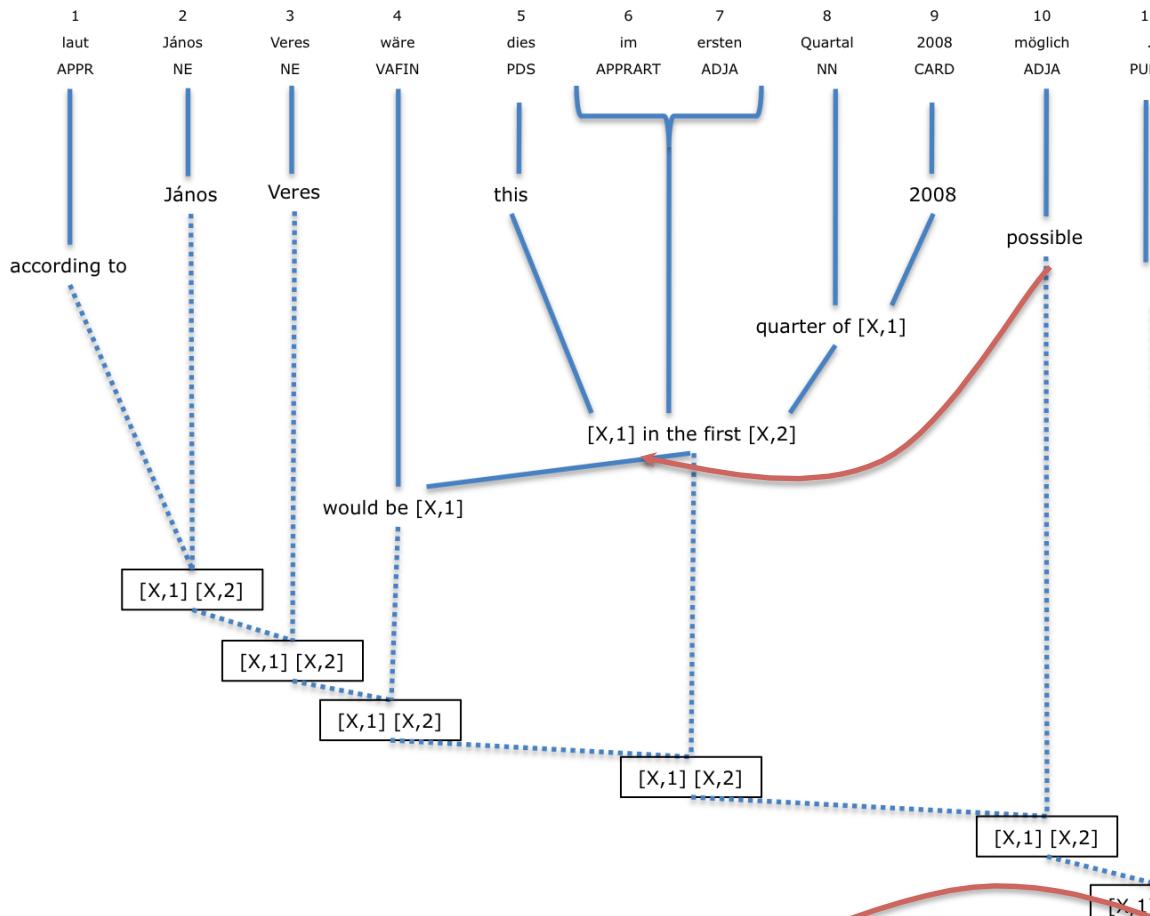
according to János Veres , this would be in the first quarter of 2008 **possible** .

# Hierarchical Model



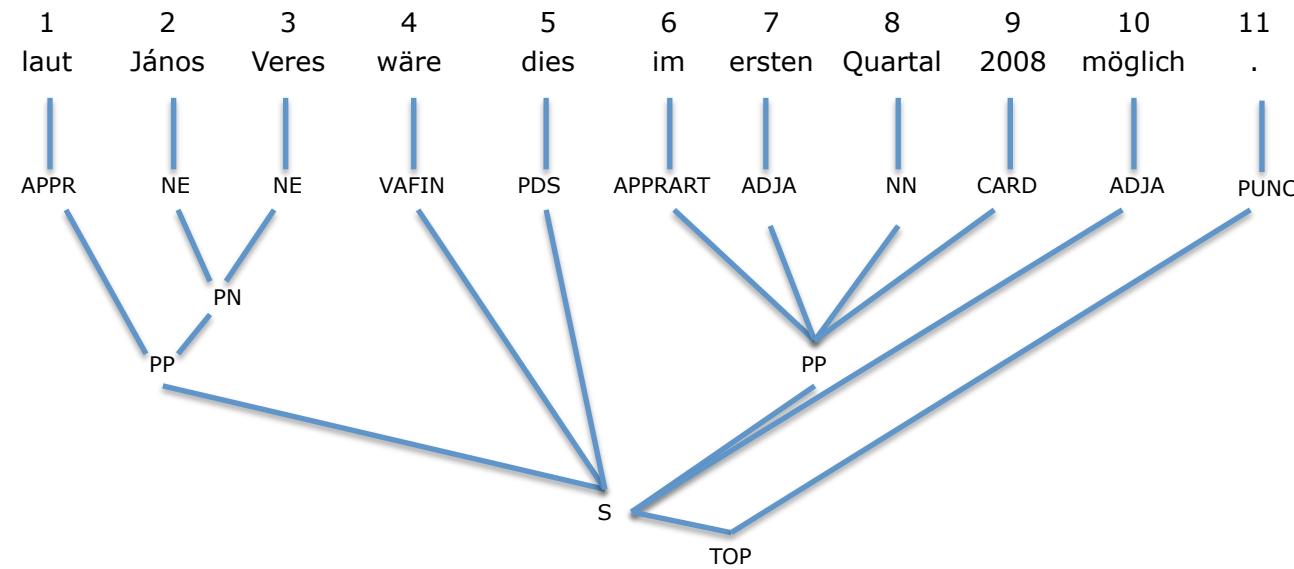
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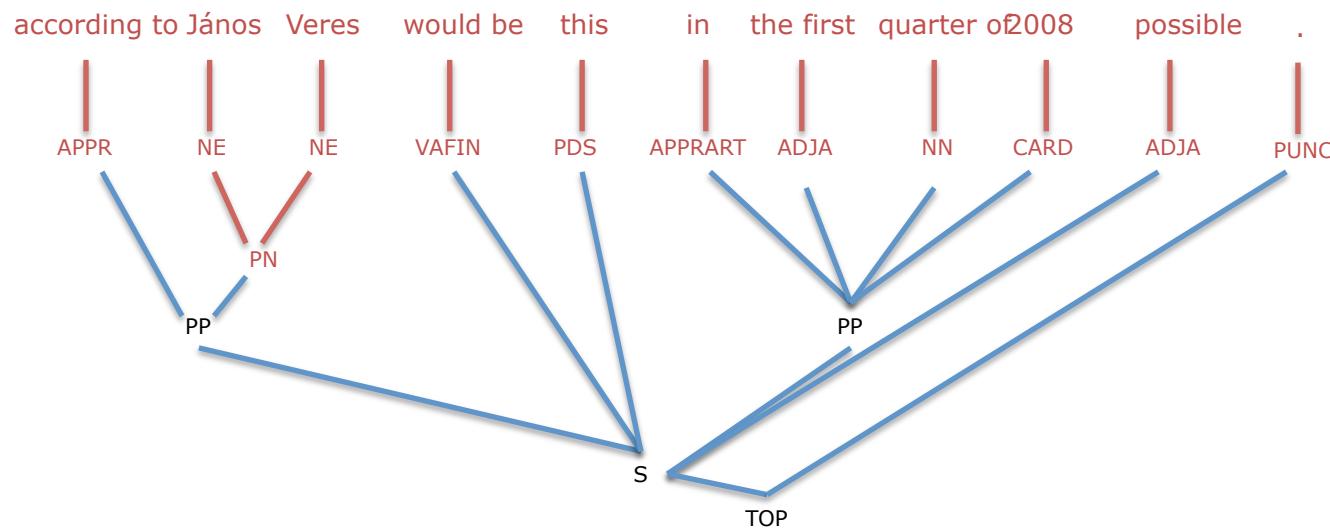


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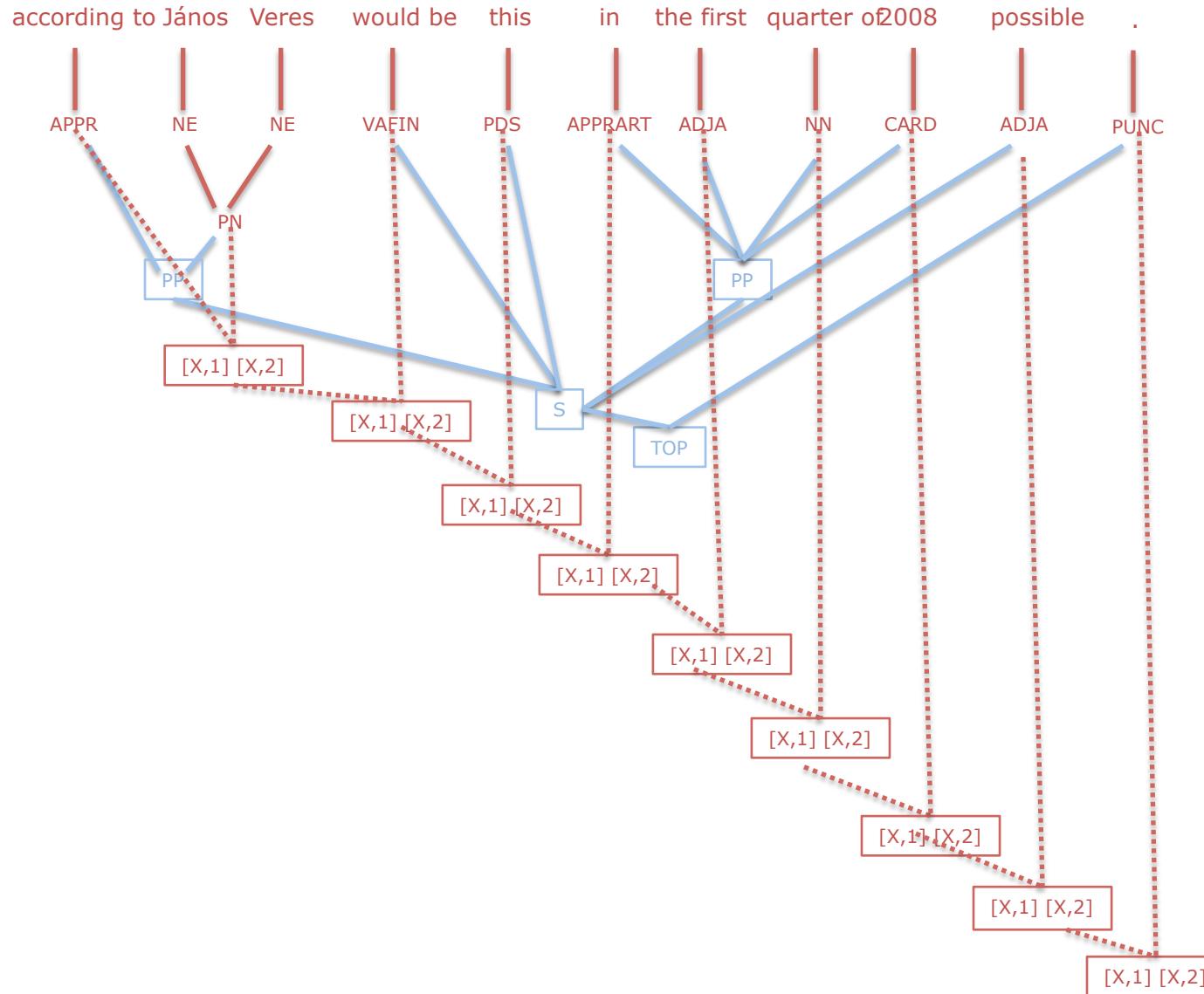
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# Other Syntactic Models

- Syntax-Augmented MT (SAMT)
  - Not constrained only to parse tree
  - (Zollmann and Venugopal, 2006)
- Binarization
  - Restructure and relabel parse tree
  - (Wang et al, 2010)
- Forest-based translation
  - Recover from parse errors
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- Soft constraint
  - Reward/Penalize derivations which follows parse structure
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  - rules covering more words than max-span limit

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  - 3+ non-terminals
  - consecutive non-terminals
  - non-lexicalized rules

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- Decrease decoding time
  - Derivation constrained by source parse tree
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  - rules covering more words than max-span limit
- Other rule-forms
  - 3+ non-terminals
  - consecutive non-terminals
  - non-lexicalized rules

$X \rightarrow S_1 O_2 V_3$

#  $S_1 V_3 O_2$

$X \rightarrow PRO_1 PRO_2$  aime bien #  $PRO_1$  like  $PRO_2$

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- Soft constraint
  - Reward/Penalize derivations which follows parse structure
  - (Chiang 2010)
- Ignore Syntax (occasionally)

# Mixed-Syntax Model

- Tree-to-string model
  - input is a parse tree
- Roles of non-terminals
  - Constrain derivation to parse constituents
  - State information
    - Consistent node label on target derivation
    - hypotheses with different head NT cannot be recombined

# Mixed-Syntax Model

- Tree-to-string model
  - input is a parse tree
- Roles of non-terminals
  - Constrain derivation to parse constituents
    - Can sometime have no constraints
  - State information
    - Consistent node label on target derivation
    - hypotheses with different head NT cannot be recombined
    - always X

# Mixed-Syntax Model

## Example Translation Rules

- Naïve syntax model

$\text{VP} \rightarrow \text{VVFIN}_1 \text{ zu } \text{VVINF}_2 \# \text{ to } \text{VVFIN}_2 \text{ VVINF}_1$

- Mixed-Syntax Model

$\text{VP} \rightarrow X_1 \text{ zu } \text{VVINF}_2 \# X \rightarrow \text{to } X_2 X_1$

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# Extraction

- Allow rules
  - Max 3 non-terminals
  - Adjacent non-terminals
    - At least 1 NT must be syntactic
  - Non-lexicalized rules

# Example Rules Extracted

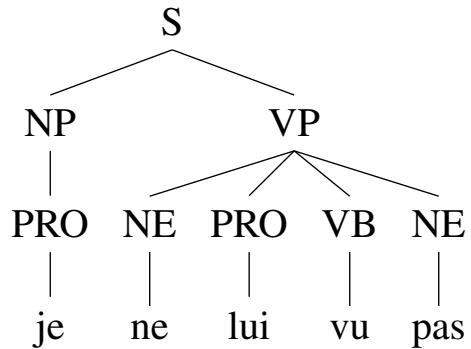
Rule	Factional Count	$p(t   s)$
<u>Syntactic Rules</u>		
$VP \rightarrow NP_1 VVINF_2 \# X \rightarrow X_2 X_1$	167.3	68%
<u>Mixed Rules</u>		
$VP \rightarrow X_1 VZ_2 \# X \rightarrow X_2 X_1$	63.3	64%
$VP \rightarrow X_1 zu VVINF_2 \# X \rightarrow to X_2 X_1$	39.9	56%
$TOP \rightarrow NP_1 X_2 \# X \rightarrow X_1 X_2$	43.1	92%

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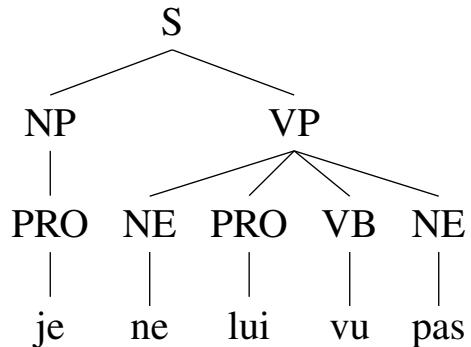
# Synchronous CFG

Input:



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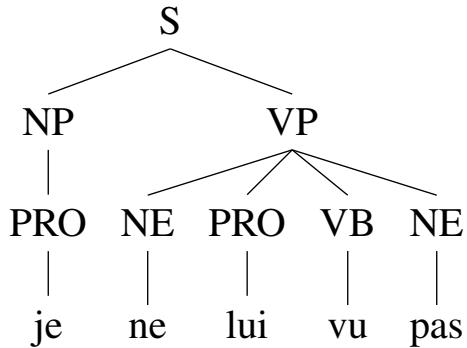


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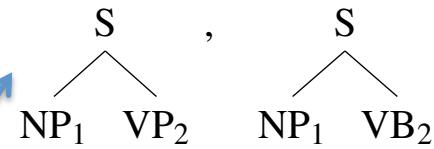
- |                                    |                        |
|------------------------------------|------------------------|
| $S \rightarrow NP_1 VP_2$          | # $NP_1 VP_2$          |
| $NP \rightarrow je$                | # I                    |
| $PRO \rightarrow lui$              | # him                  |
| $VB \rightarrow vu$                | # see                  |
| $VP \rightarrow ne PRO_1 VB_2 pas$ | # did not $VB_2 PRO_1$ |

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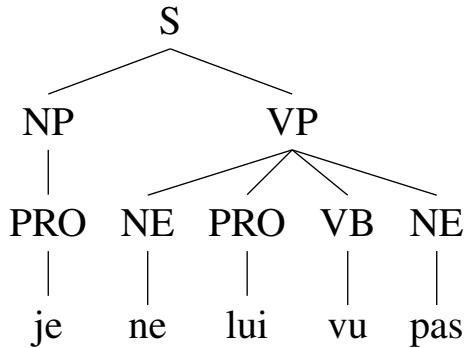


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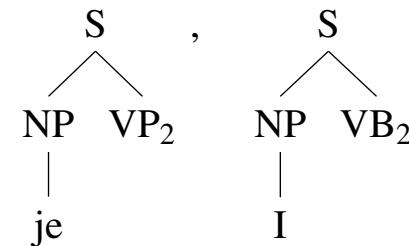
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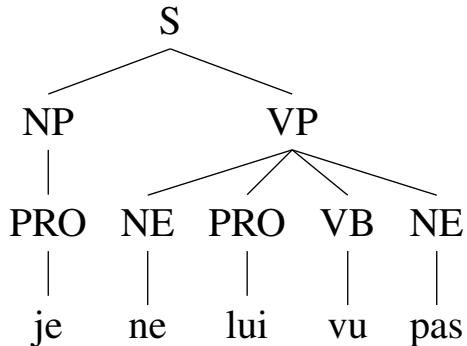


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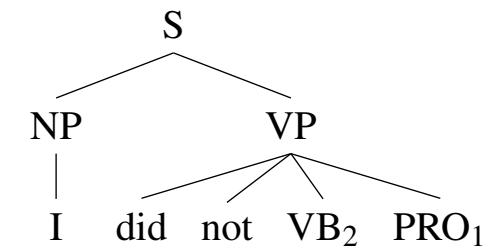
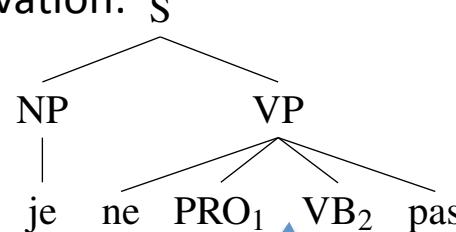
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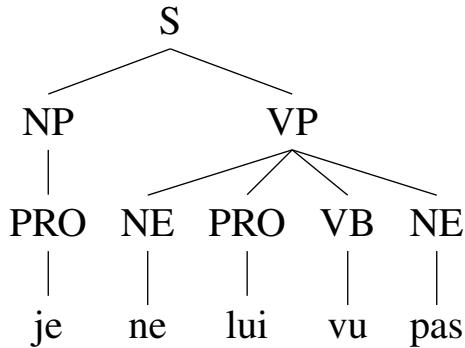
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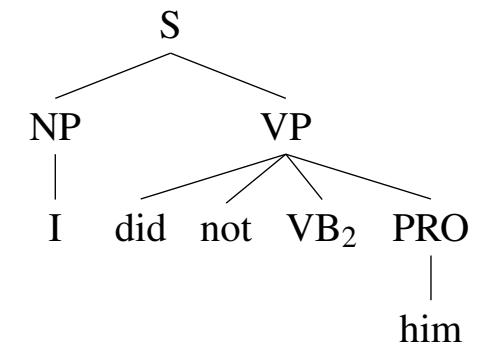
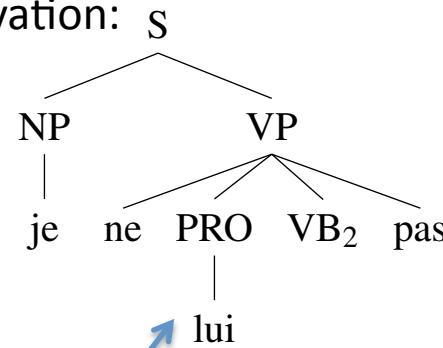


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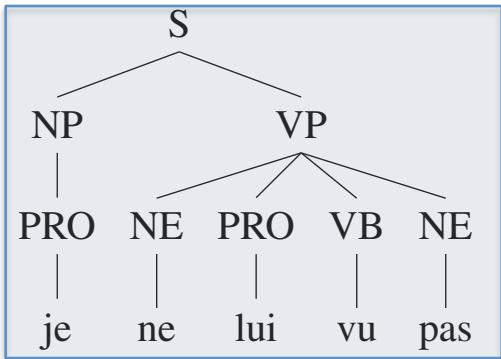


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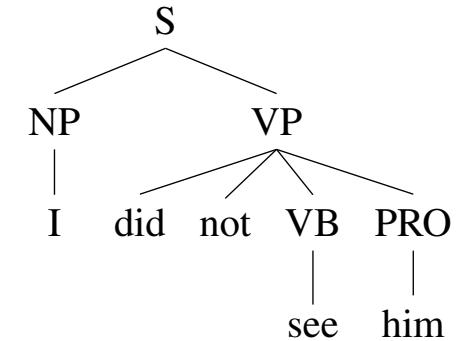
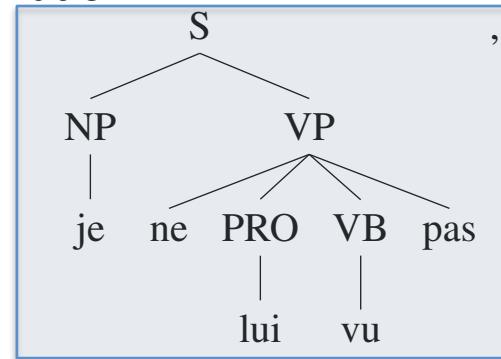
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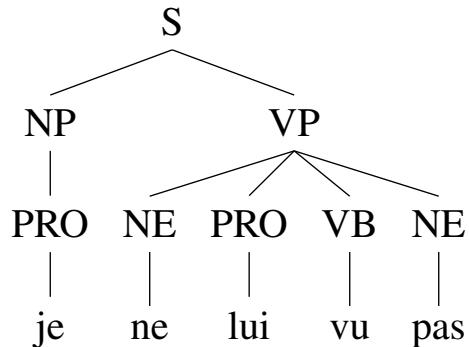
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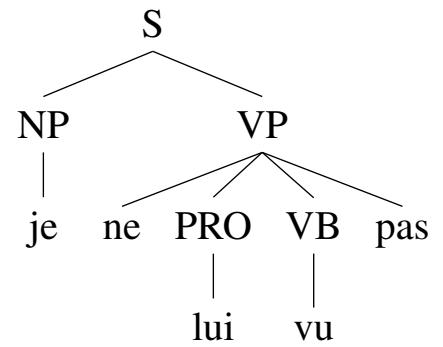
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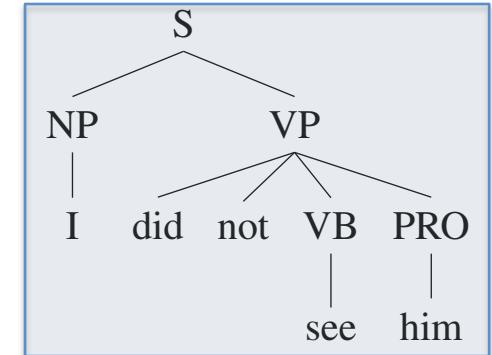
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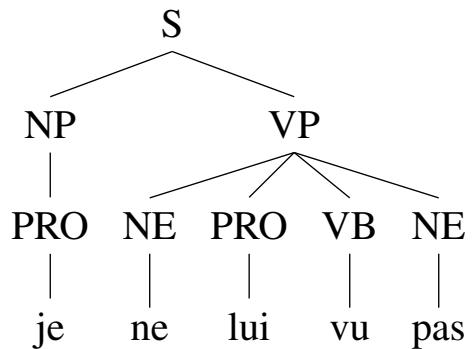
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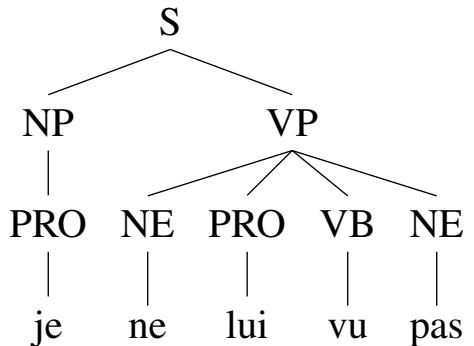
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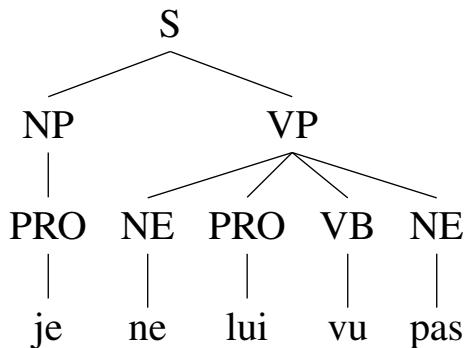
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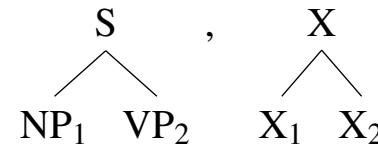
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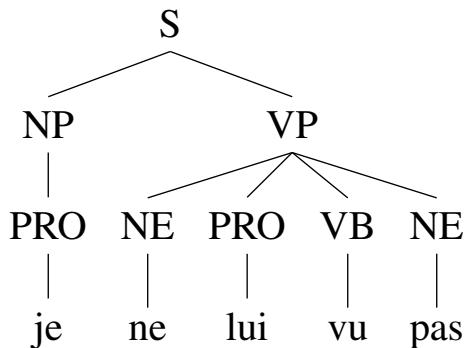
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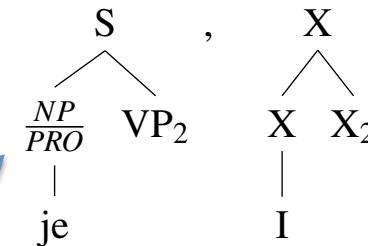
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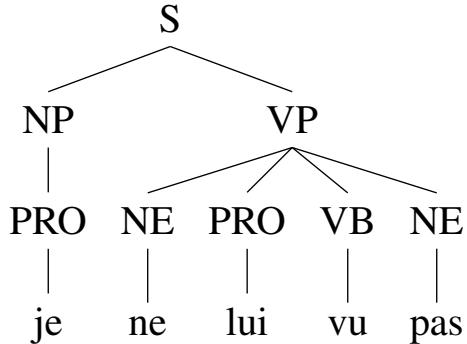


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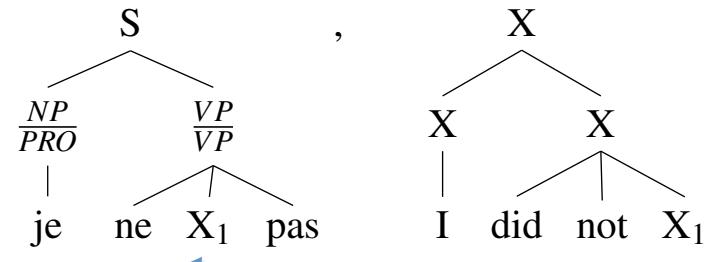
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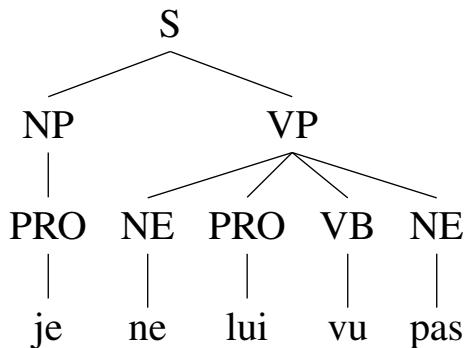
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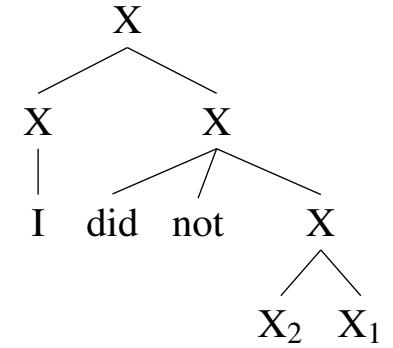
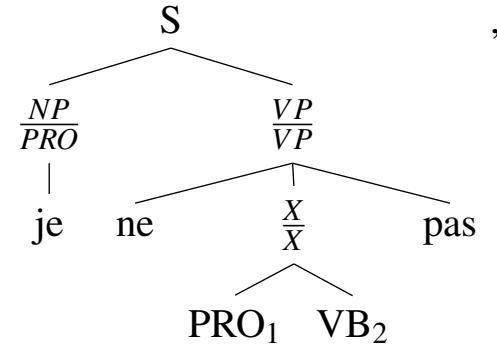
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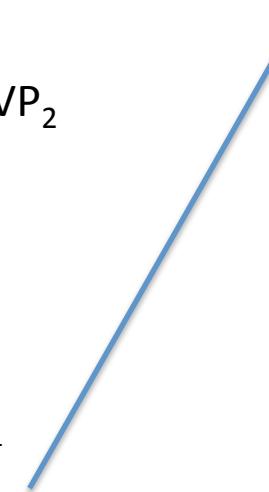
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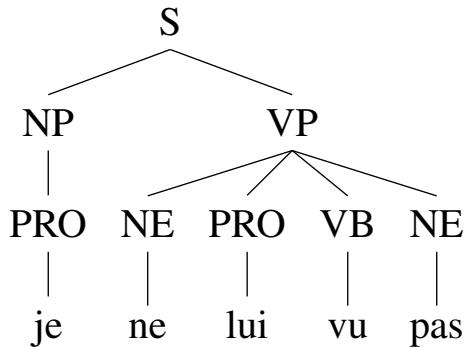
$$VP \rightarrow ne X_1 pas \quad \# did not X_1$$

$$X \rightarrow PRO_1 VB_2 \quad \# X \rightarrow X_2 X_1$$

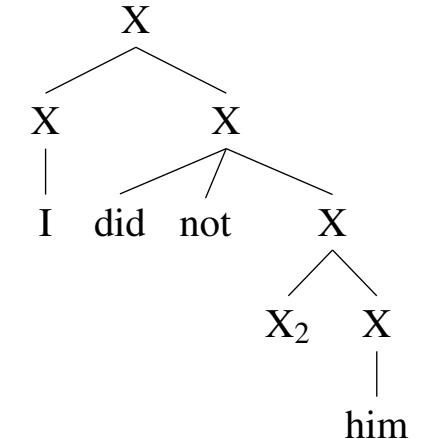
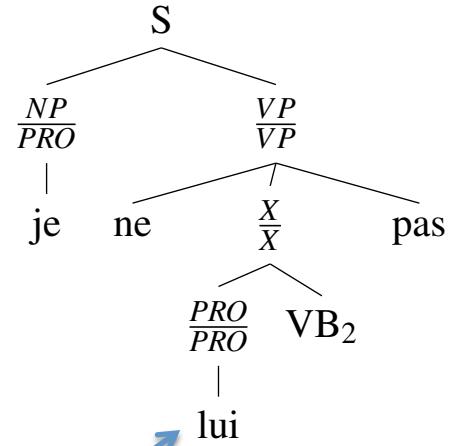


# Mixed-Syntax Model

Input:



Derivation:



Rules:

$$S \rightarrow NP_1 VP_2 \quad \# X \rightarrow NP_1 VP_2$$

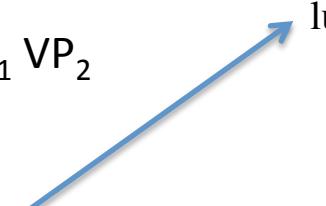
$$PRO \rightarrow je \quad \# X \rightarrow I$$

$$PRO \rightarrow lui \quad \# X \rightarrow him$$

$$VB \rightarrow vu \quad \# X \rightarrow see$$

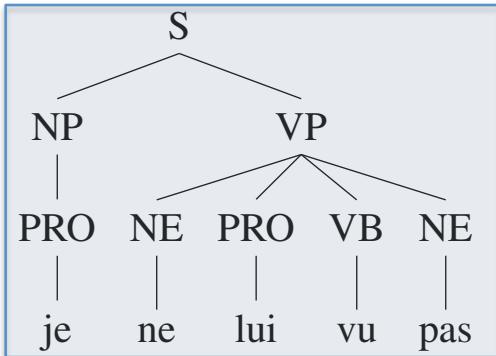
$$VP \rightarrow ne X_1 pas \quad \# did not X_1$$

$$X \rightarrow PRO_1 VB_2 \quad \# X \rightarrow X_2 X_1$$

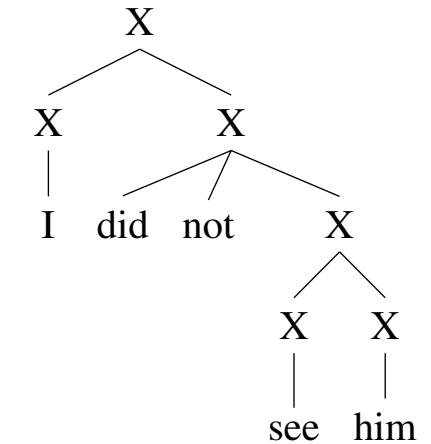
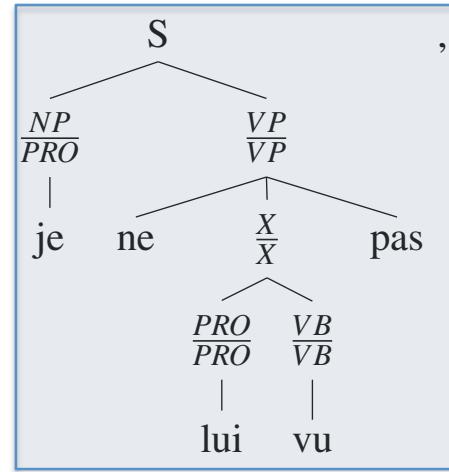


# Mixed-Syntax Model

Input:



Derivation:



Rules:

$$S \rightarrow NP_1 VP_2 \quad \# X \rightarrow NP_1 VP_2$$

$$PRO \rightarrow je \quad \# X \rightarrow I$$

$$PRO \rightarrow lui \quad \# X \rightarrow him$$

$$VB \rightarrow vu \quad \# X \rightarrow see$$

$$VP \rightarrow ne X_1 pas \quad \# did not X_1$$

$$X \rightarrow PRO_1 VB_2 \quad \# X \rightarrow X_2 X_1$$

# Contents

- What's Wrong with Syntax?
- Which syntax model to use?
- Why use syntactic models?
- Mixed-Syntax Model
  - Extraction
  - Decoding
  - Results
- Future Work

# Experiment

German-English

Corpus

		German	English
Train	Sentences	82,306	
	Words	2,034,373	1,965,325
Tune	Sentences	2000	
Test	Sentences	1026	

Trained: News Commentary 2009

Tuned: held out set

Tested: *nc test2007*

# Results

## Using constituent parse

German-English

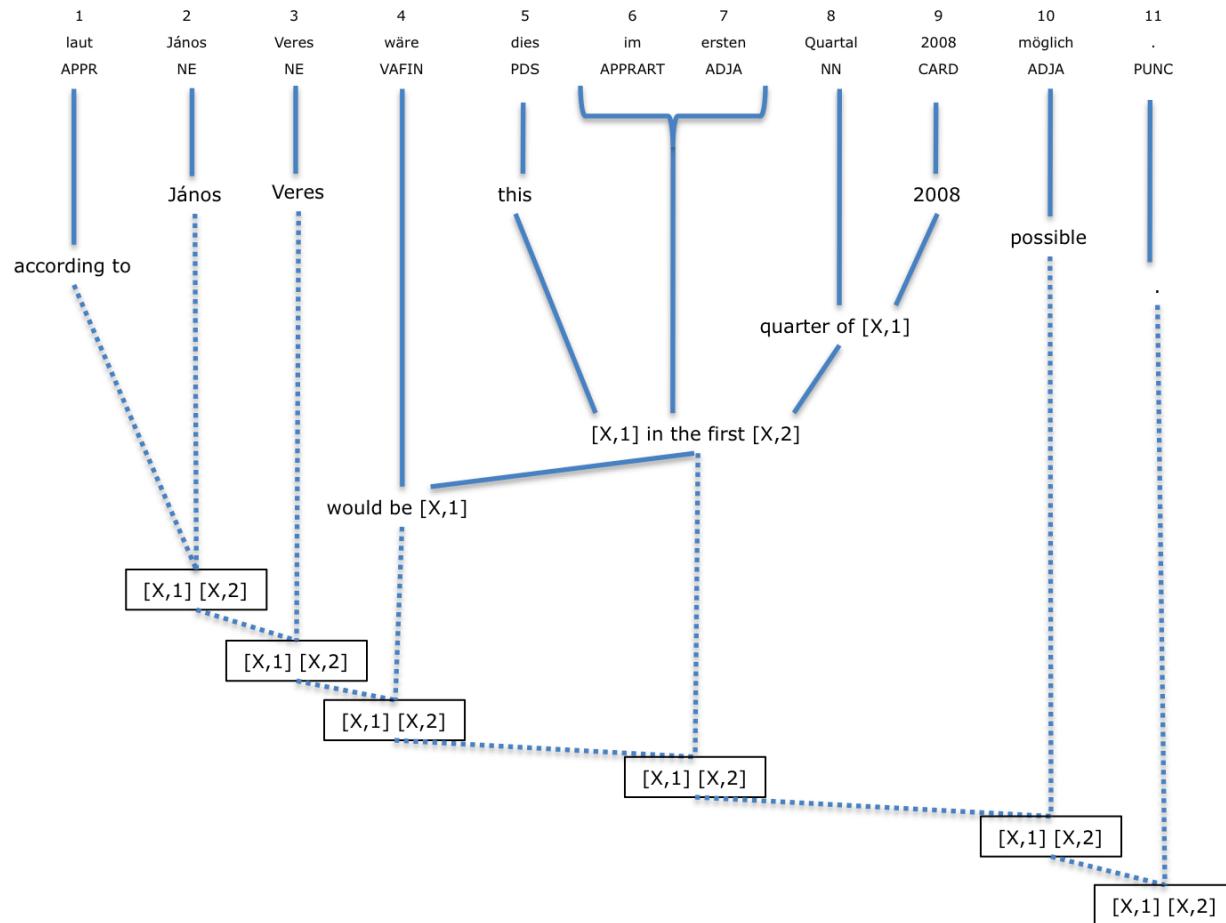
Model	# rules	%BLEU
Hierarchical	61.2m	15.9
Tree-to-String	4.7m	14.9
Mixed Syntax	128.7m	16.7

English-German

Model	# rules	%BLEU
Hierarchical	84.6m	10.2
Mixed Syntax	175.0m	10.6

# Example

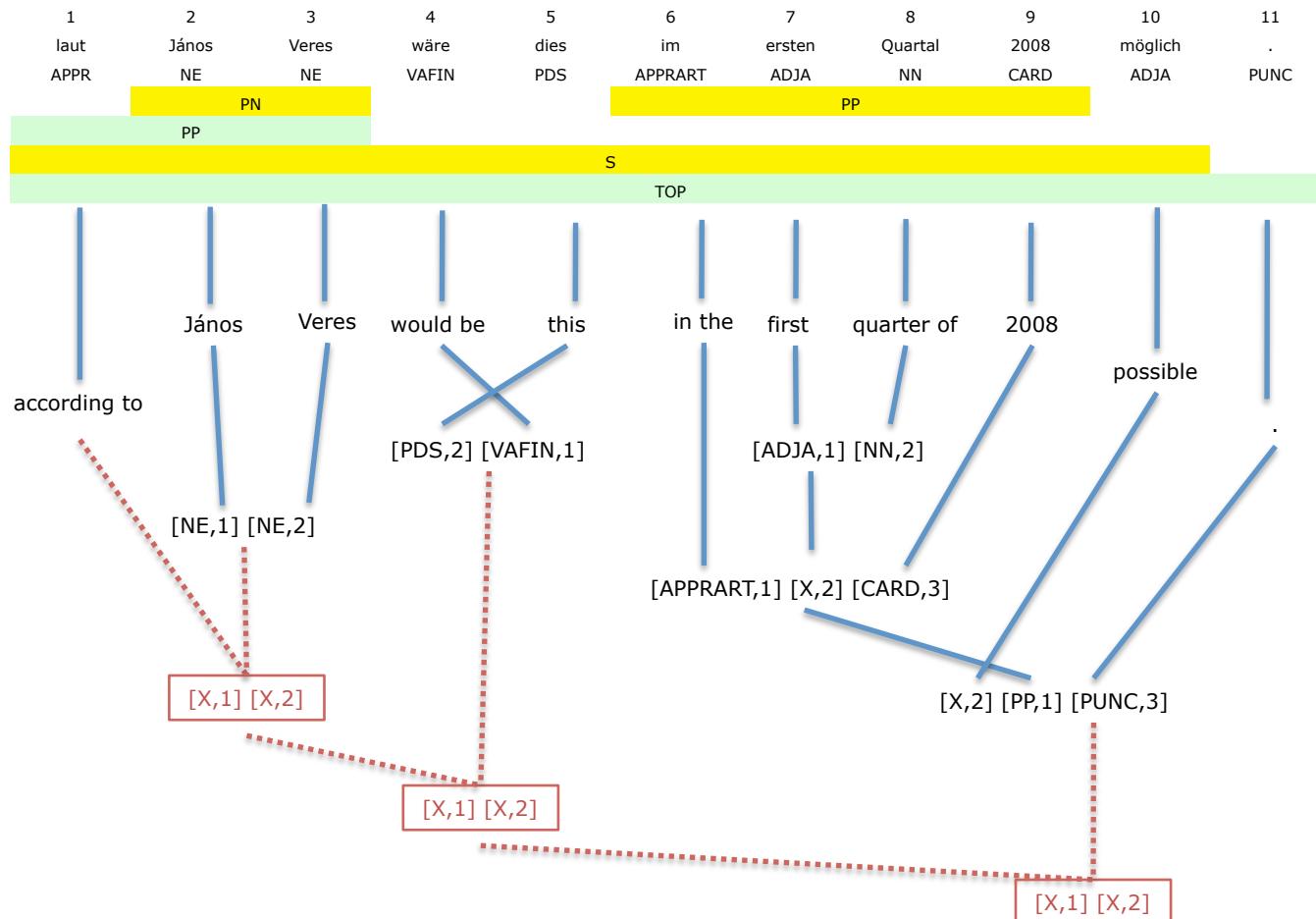
## Hierarchical Model



according to János Veres , this would be in the first quarter of 2008 **möglich** .

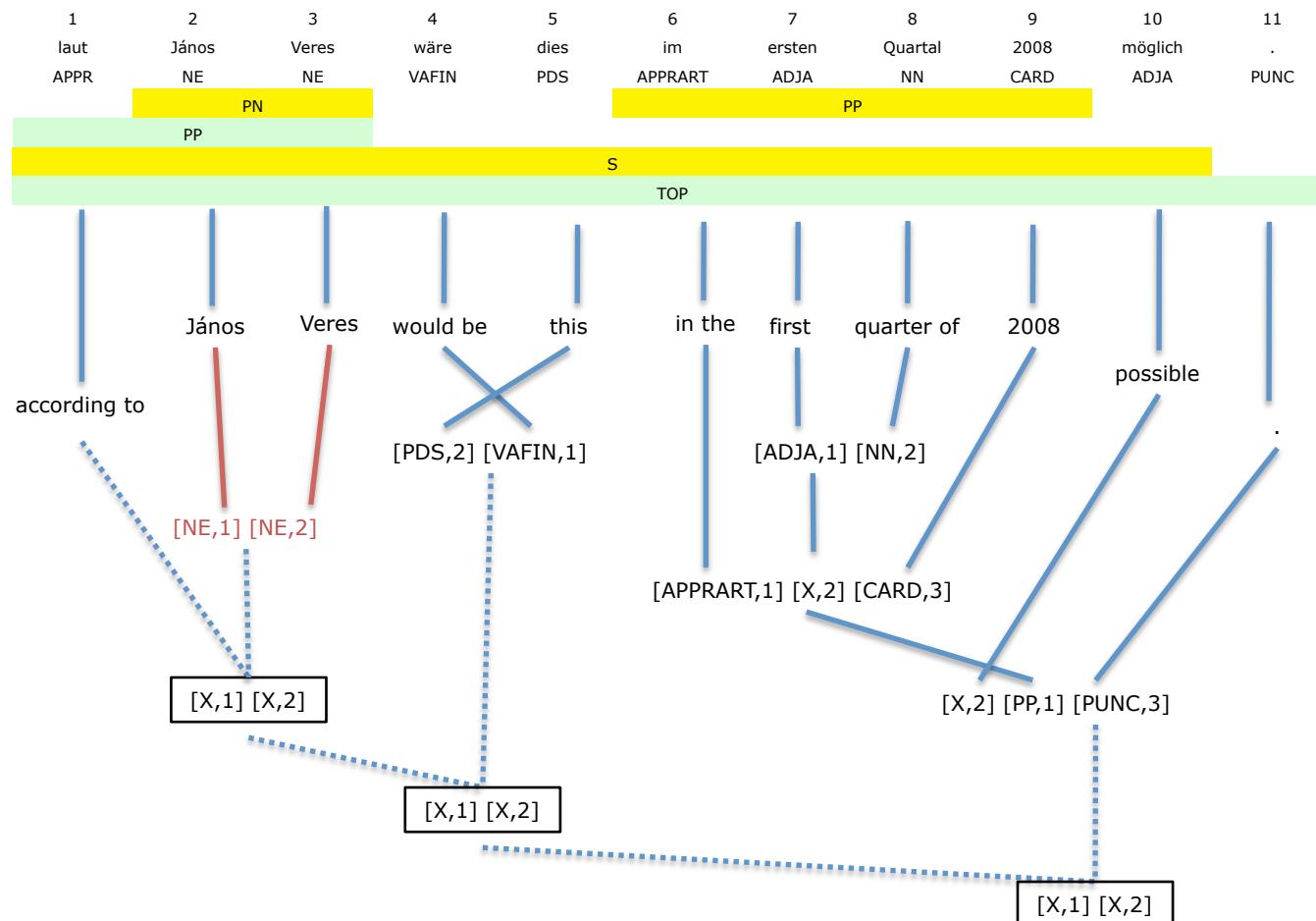
# Example

## Mixed-Syntax Model



# Example

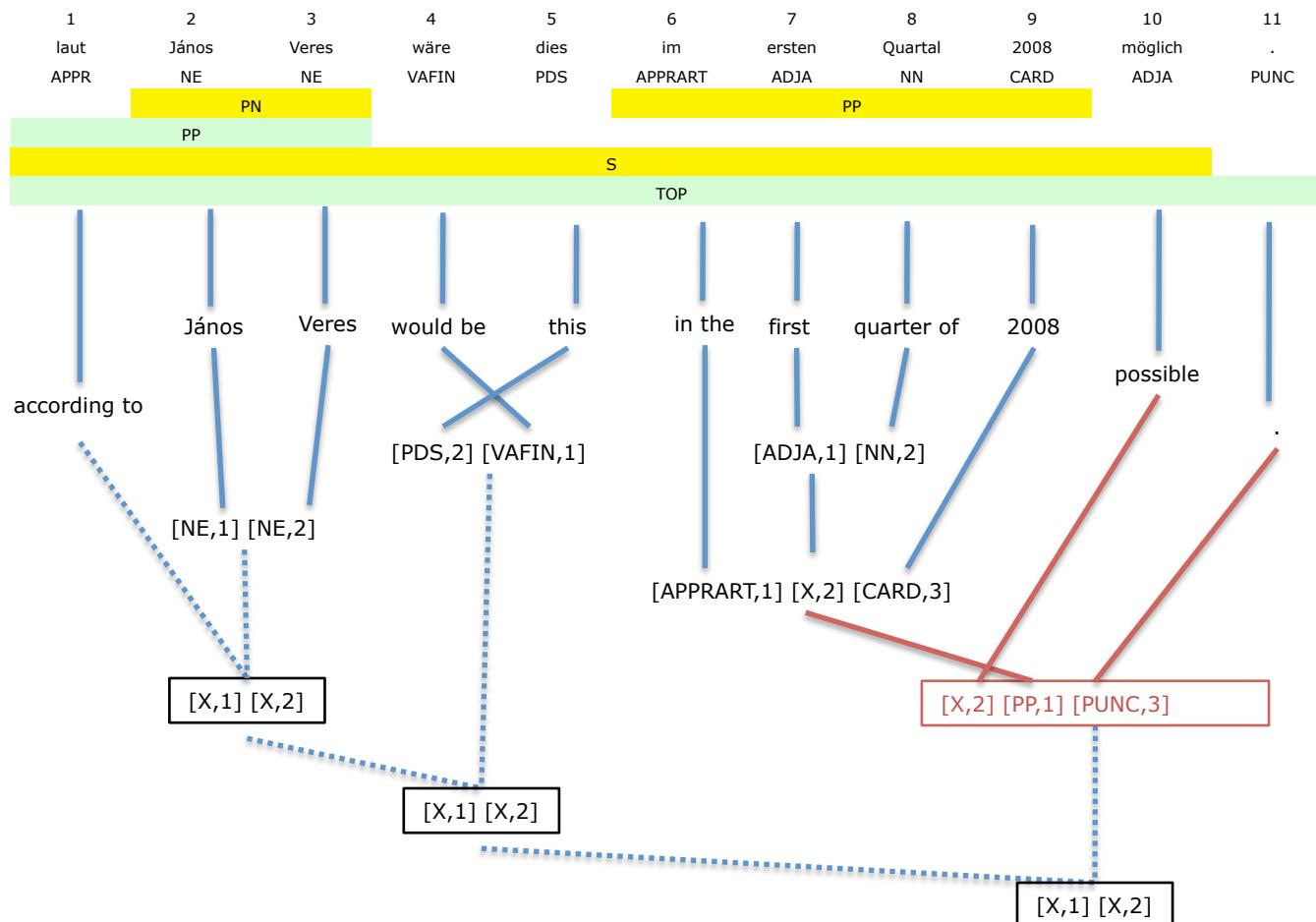
## Mixed Syntax



according to János Veres this would be **possible** in the first quarter of 2008.

# Example

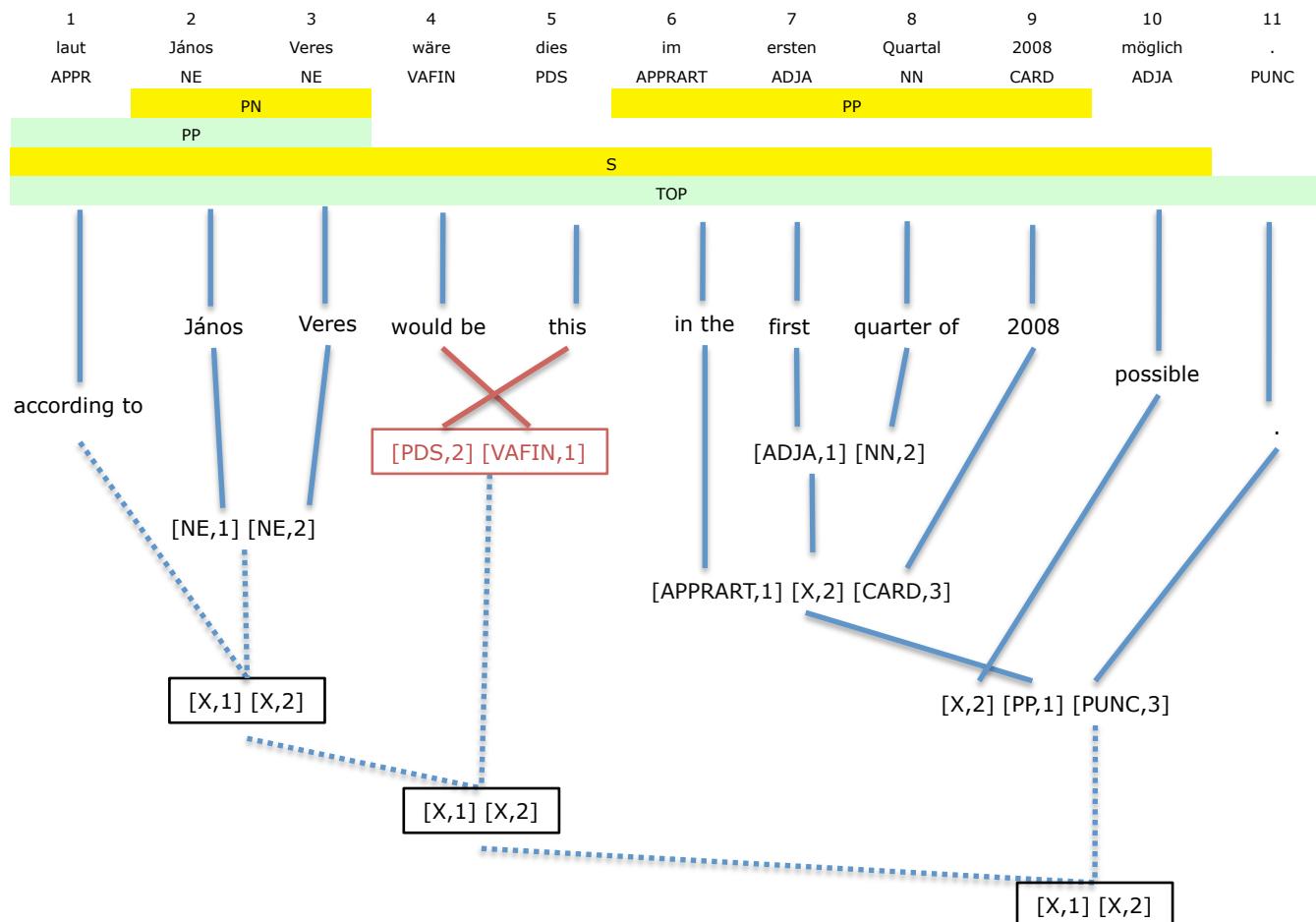
## Mixed Syntax



according to János Veres this would be **possible** in the first quarter of 2008.

# Example

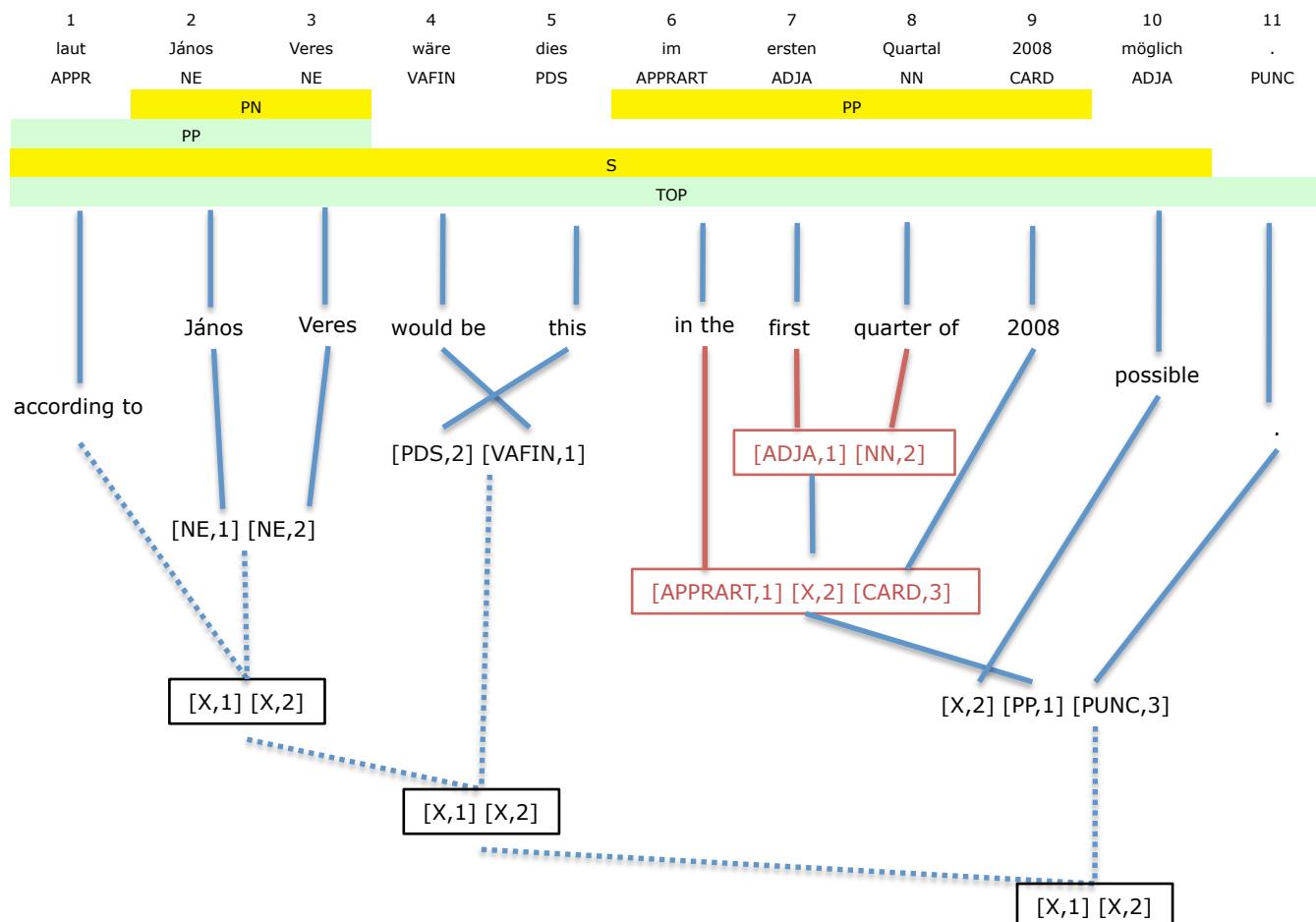
## Mixed Syntax



according to János Veres this would be **possible** in the first quarter of 2008.

# Example

## Mixed Syntax



# Chunk Tags

- Advantages of Shallow Tags
  - Don't need Treebank
  - More reliable
- Disadvantages
  - Not a tree structure
    - We don't rely on tree structure

# Results

## Shallow Tags

German-English

Model	# rules	%BLEU
Hierarchical	64.3m	16.3
Mixed Syntax	254.5m	16.8

# Larger Training Corpus

## German-English

### Corpus

		German	English	Corpus
Train	Sentences	1,446,224		Europarl v5
	Words	37,420,876	39,464,626	
Tune	Sentences	1910		dev2006
Test (in-domain) (out-of-domain)	Sentences	1920 1042		nc test2007 v2 devtest2006

# Larger Training Corpus

## German-English

Model	# rules	In-domain (BLEU)	Out-of-domain (BLEU)
Hierarchical	500m	22.1	16.5
Mixed Syntax (original)	2664m	21.6	16.3
Mixed Syntax (new extraction)	1435m	22.7	17.8

# Contents

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# Create your own label

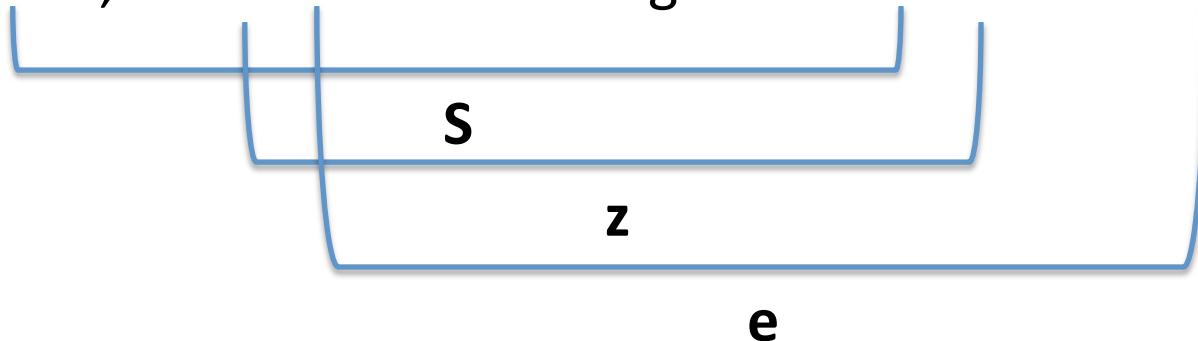
## Dumb labels

ich bitte Sie , sich zu einer Schweigeminute zu erheben .

# Create your own label

## Dumb labels

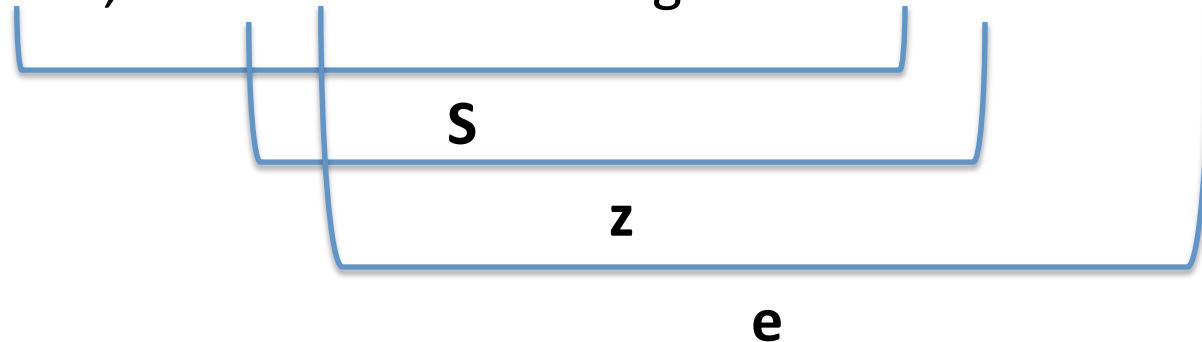
ich bitte Sie , sich zu einer Schweigeminute zu erheben .



# Create your own labels

## Dumb labels

ich bitte Sie , sich zu einer Schweigeminute zu erheben .



Model	In-domain (BLEU)	Out-of-domain (BLEU)
Hierarchical	22.1	16.5
Dumb Labels	22.0	16.3

# Create your own labels

## Labels motivated by reordering

Labelling patterns:

1. VMFIN...VVINF EOS
2. VVINF und ... VVINF
3. VAFIN ... (VVPP or VVINF) EOS
4. , PRELS ... VVINF EOS
5. EOS ... zu VVINF

Example:

ich bitte Sie , sich zu einer Schweigeminute zu erheben .

**label 5**

... werde ich dem Vorschlag von Herrn Evans folgen .

**label 3**

# Create your own labels

## Labels motivated by reordering

Model	In-domain (BLEU)	Out-of-domain (BLEU)
Hierarchical	22.1	16.5
Dumb Labels	22.0	16.3
Reordering Labels	22.1	16.9

# Conclusion

- Mixed-Syntax Model
  - SCFG-based decoding
  - Hierarchical phrase-based v. tree-to-string
  - Generality v. specificity
- Syntax Models
  - Many variations
  - Won't automatically make MT better
  - Question
    - which syntactic information?
    - how do we use it?
    - why use it?

**END**