

Introduction
oooooooooo

Derivation Projection
oooooo

Experiments
ooooooo

Conclusions
oo

References

Cross-lingual CCG Induction

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Introduction
oooooooooo

Derivation Projection
oooooo

Experiments
ooooooo

Conclusions
oo

References

Outline

Introduction

Derivation Projection

Experiments

Conclusions

Introduction
●oooooooo

Derivation Projection
ooooo

Experiments
ooooooo

Conclusions
oo

References

Outline

Introduction

Derivation Projection

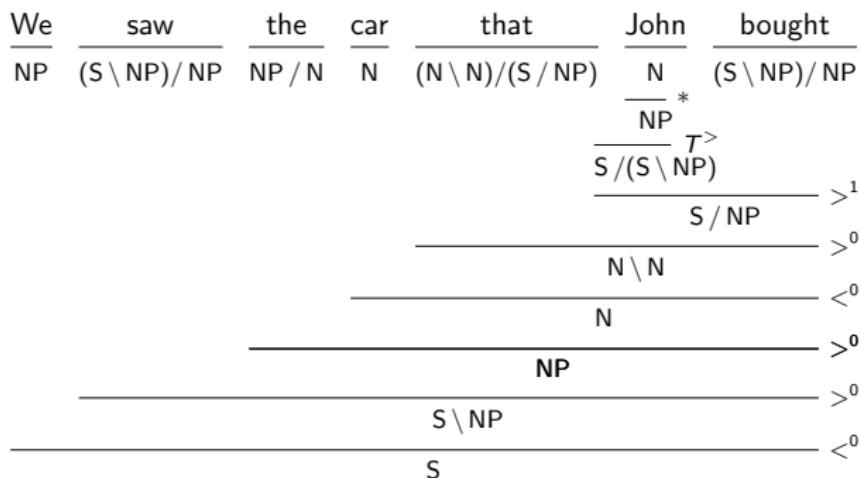
Experiments

Conclusions

Combinatory Categorial Grammar

$$\frac{\begin{array}{c} \text{We} & \text{sang} \\ \hline \text{NP} & \text{S} \setminus \text{NP} \end{array}}{\text{S}} <^0$$

Combinatory Categorial Grammar



Appeal

- coordination
- universal rules
- syntax-semantics interface

Most CCG Parsers

- trained on large treebanks, or
- hand-crafted



David Blackwell, CC-BY-NC

What about low-resource languages?

Unsupervised CCG Induction?

target-language text
+
magic
=

target-language CCG parser

(Bisk and Hockenmaier, 2013; Bisk et al., 2015)

Cross-lingual CCG Induction?

English CCG parser
+
parallel corpus
+
magic
=

target-language CCG parser

Cross-lingual CCG Induction via Derivation Projection

parallel corpus
+
English CCG derivations
+
word alignments
+
derivation projection
=

target-language CCG derivations
=

target-language training data

Introduction
oooooooooo

Derivation Projection
●oooo

Experiments
ooooooo

Conclusions
oo

References

Outline

Introduction

Derivation Projection

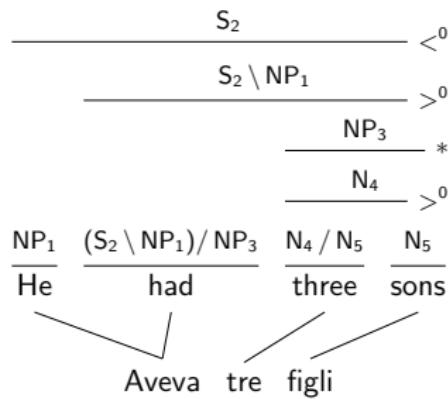
Experiments

Conclusions

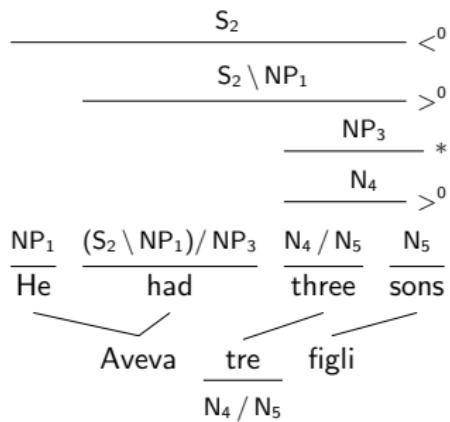
Derivation Projection

- project lexical categories along word alignments
- n:1 alignment → merge
- word order difference → flip slash

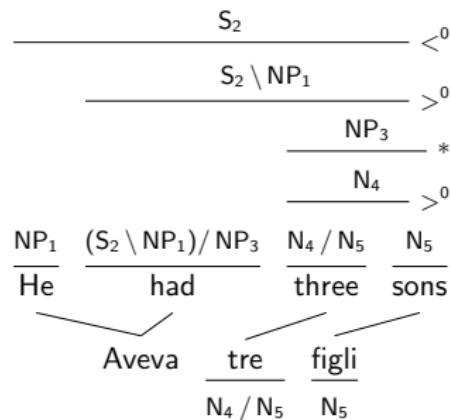
Example 1/3



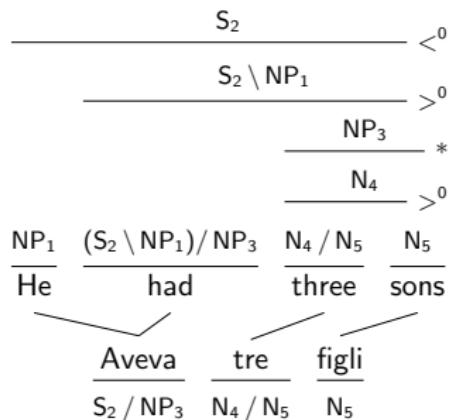
Example 1/3



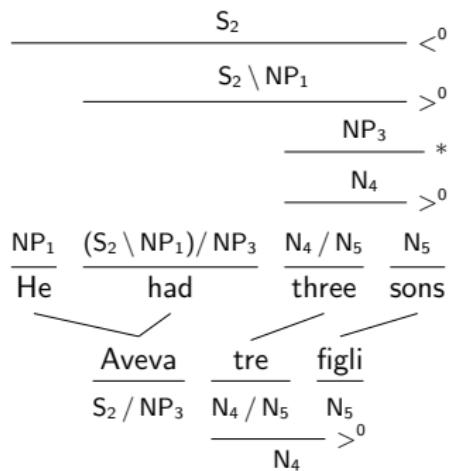
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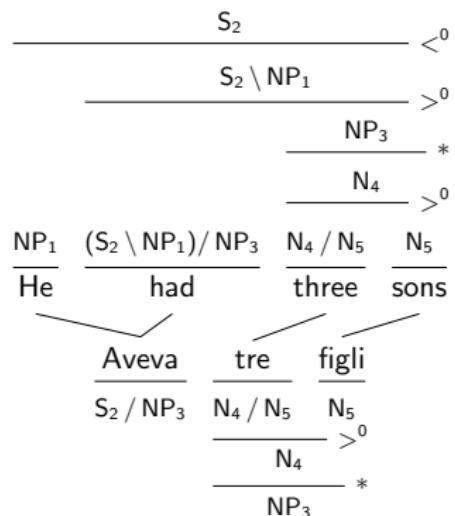
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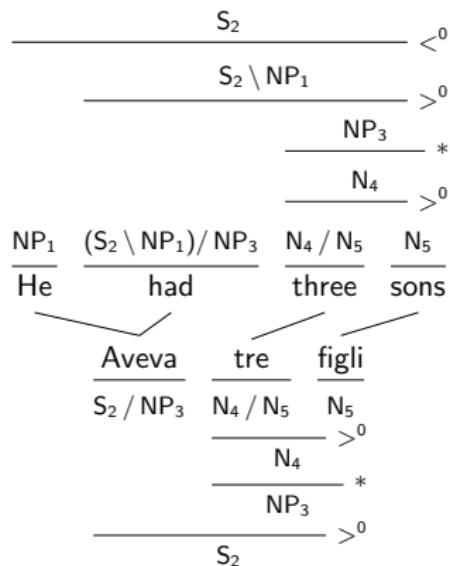
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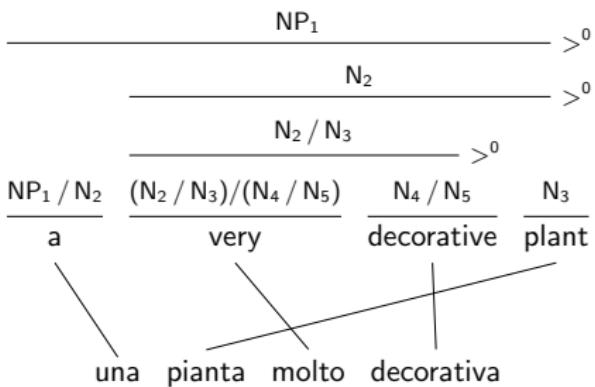
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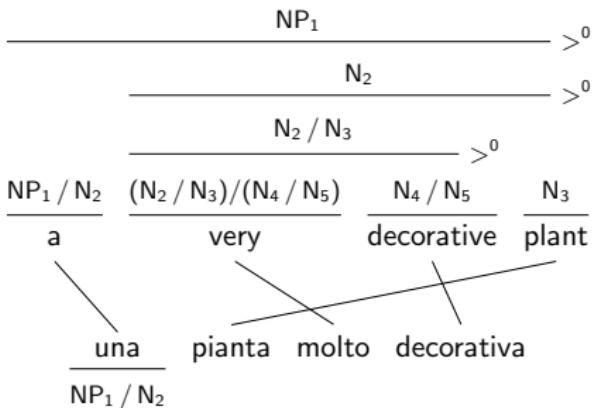
Example 1/3



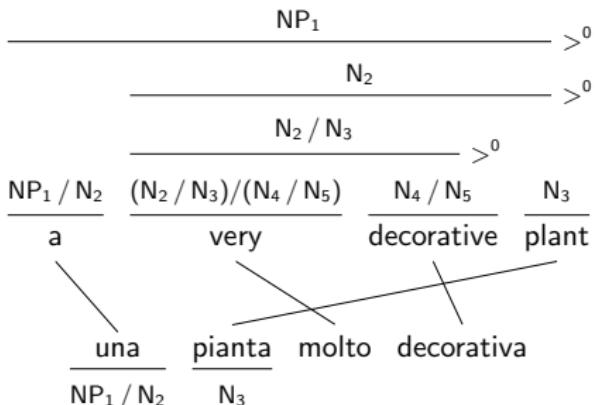
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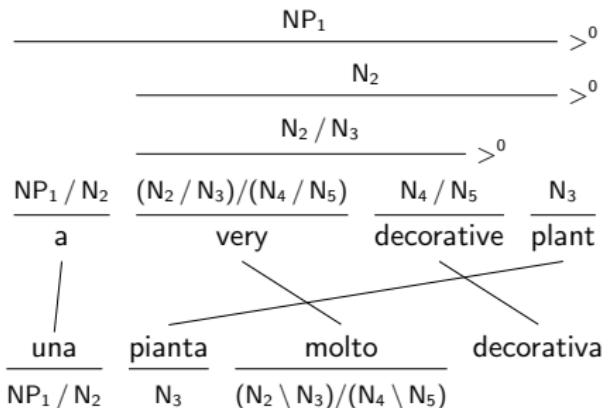
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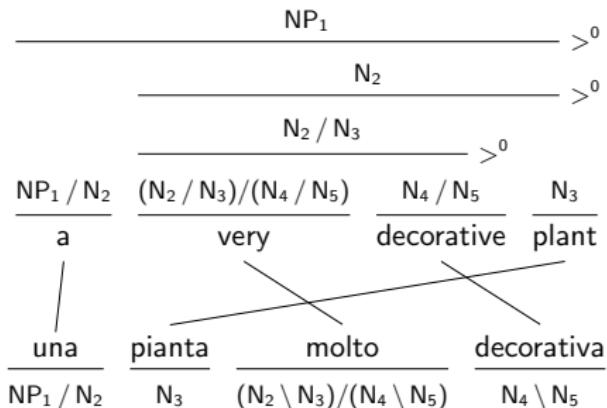
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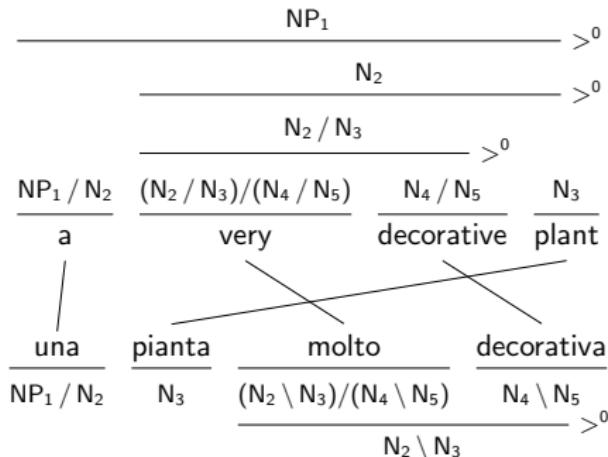
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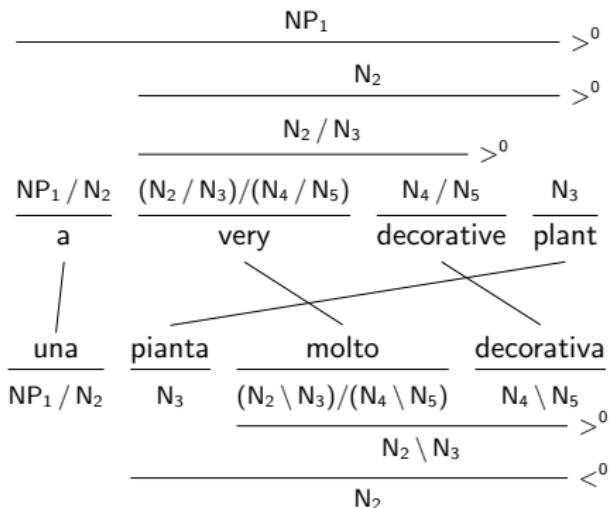
Example 2/3



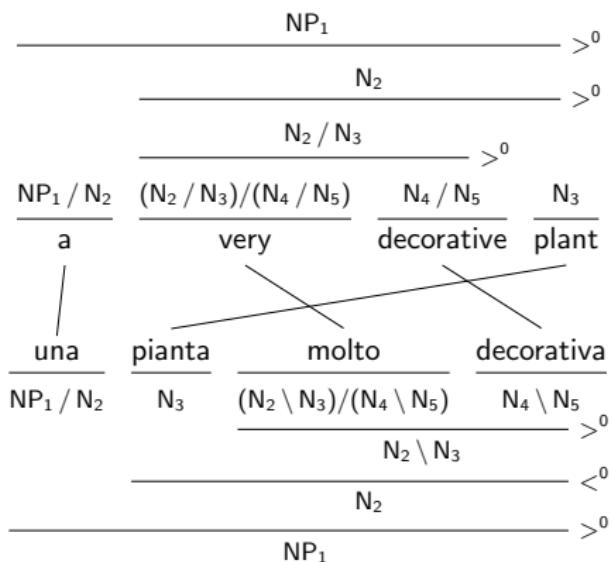
Example 2/3



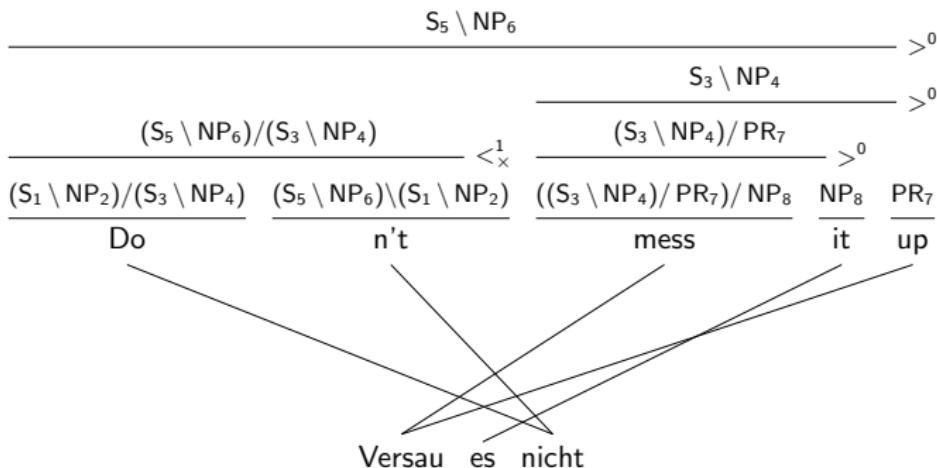
Example 2/3



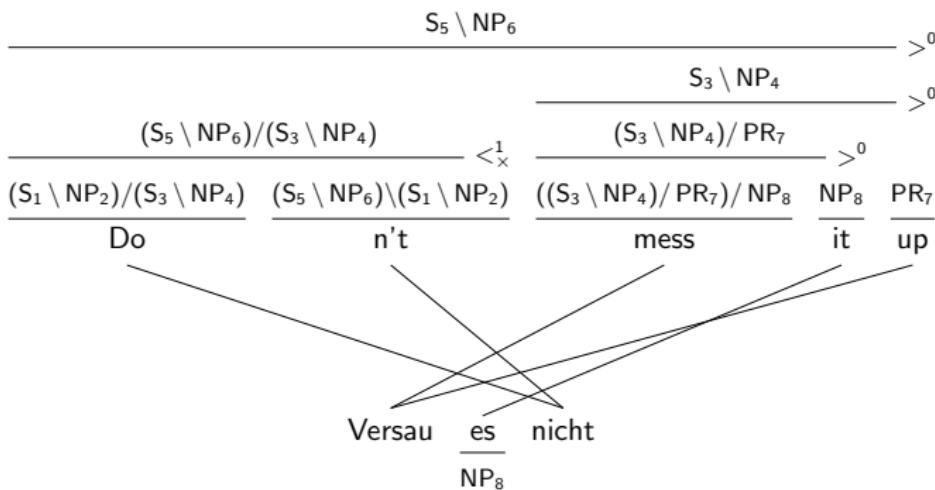
Example 2/3



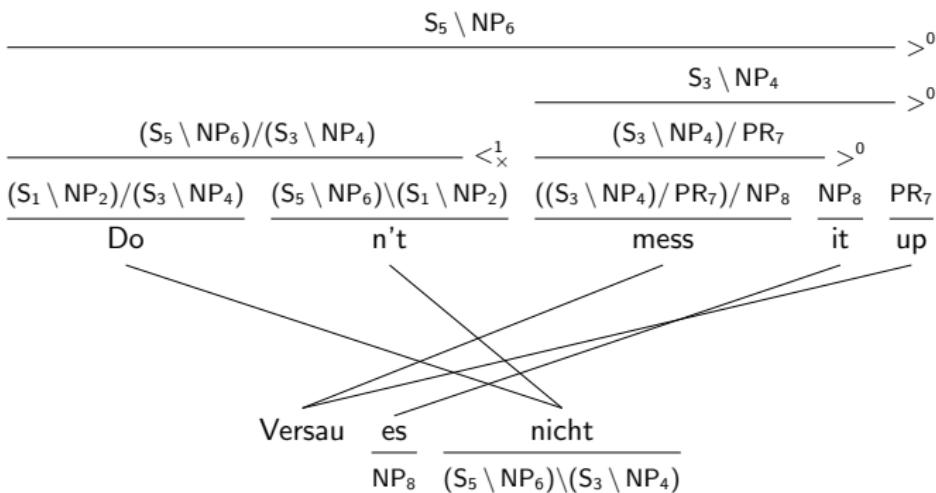
Example 3/3



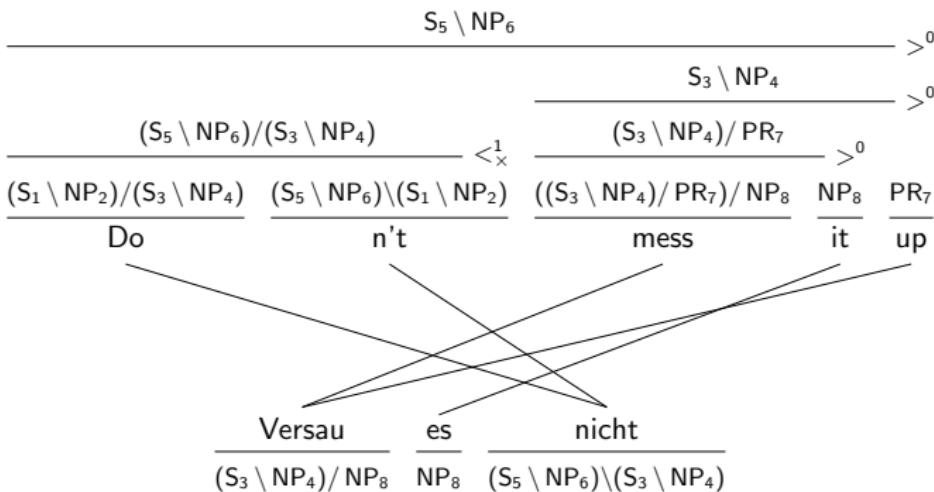
Example 3/3



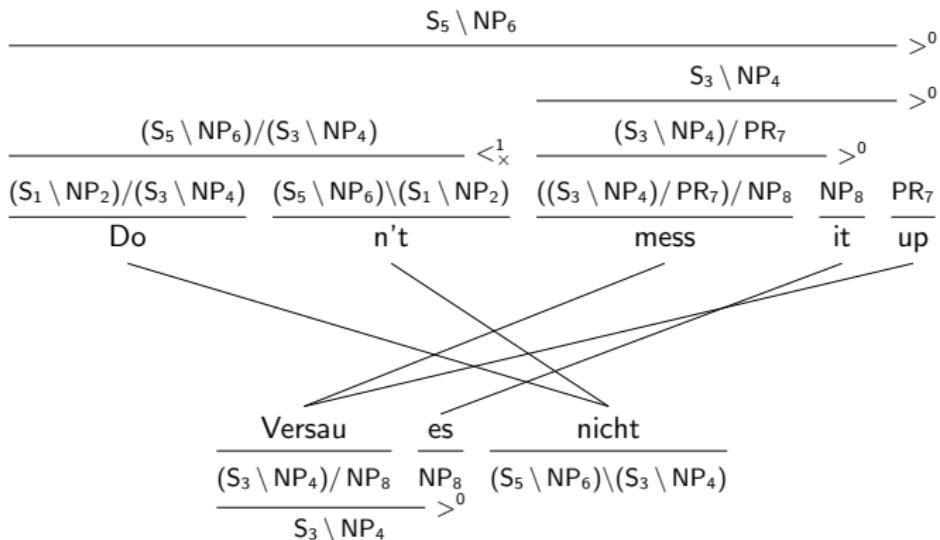
Example 3/3



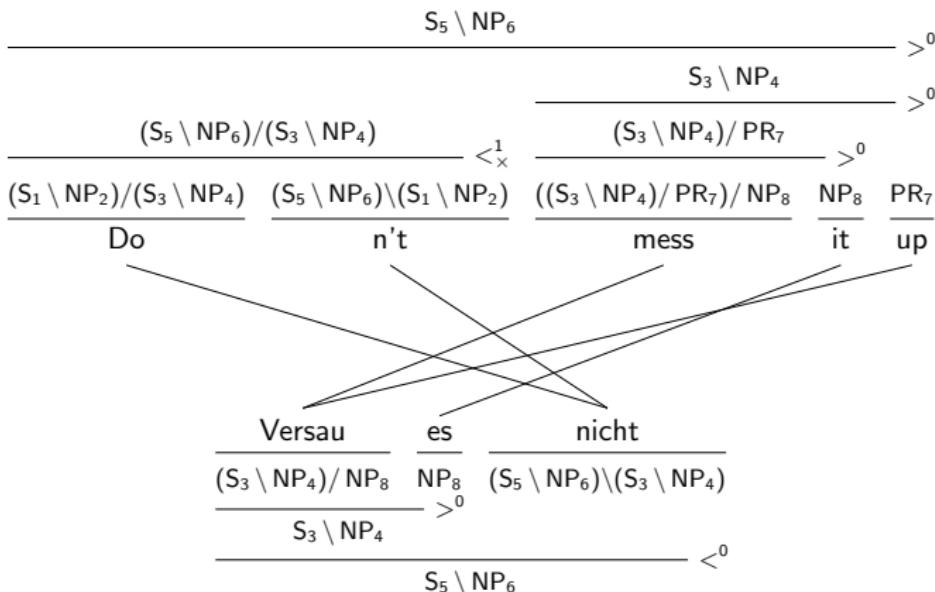
Example 3/3



Example 3/3



Example 3/3



Introduction
oooooooooo

Derivation Projection
ooooo

Experiments
●oooooo

Conclusions
oo

References

Outline

Introduction

Derivation Projection

Experiments

Conclusions

Training

- **Parallel corpus:** tatoeba.org
- **English parser:** EasyCCG trained on CCGBank
- **Word alignments:** GIZA++
- **Target-language parser:** EasyCCG trained on projected derivations

Evaluation

- PASCAL challenge on unsupervised grammar induction:
Arabic, Czech, Danish, Basque, Dutch, Portuguese, Slovenian,
Swedish
- unlabeled dependency f-score

Training (cont.)

	ara	ces	dan	eus	nld	por	slv	swe
sentence pairs	20K	11K	21K	2K	44K	161K	835	24K
projected	7K	4K	11K	590	18K	50K	364	12K

Baselines

- **BH13:** CCG induction from raw text + POS tags
(Bisk and Hockenmaier, 2013)
- **BCH15:** CCG induction from raw text
(Bisk et al., 2015)

Results

Language	ara	ces	dan	eus	nld	por	slv	swe
<i>Monolingual training on PASCAL</i>								
Train tokens	5K	436K	25K	81K	79K	159K	54K	62K
BH13	.651	.507	.585	.450	.544	.629	.464	.669
BCH15	.437	.324	.377	.352	.438	.516	.236	.529
<i>Cross-lingual training on Tatoeba</i>								
Train tokens	20K	11K	21K	2K	44K	161K	835	24K
this work	.468	.449	.630	.290	.614	.678	.350	.637
	 2	 2	 1		 1	 1	 2	 2

Induced Lexicons

		eng	deu	ita	nld
SOV	$(S \setminus NP) \setminus NP$	-	+	-	+
right adj	$N \setminus N, (N \setminus N)/(N \setminus N)$	-	-	+	-
pro-drop	$S, S / NP$	-	-	+	-

Introduction
oooooooooo

Derivation Projection
oooooo

Experiments
ooooooo

Conclusions
●○

References

Outline

Introduction

Derivation Projection

Experiments

Conclusions

Take-home Message

- CCG derivations can be automatically projected along word alignments
- Cross-lingual supervision helps CCG induction
- Induces linguistically plausible lexicons
- Used for bootstrapping the Parallel Meaning Bank:
<https://pmb.let.rug.nl>

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