Chart-based RRG parsing using an automatically extracted RRG grammar with features

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Design of RRG Grammars

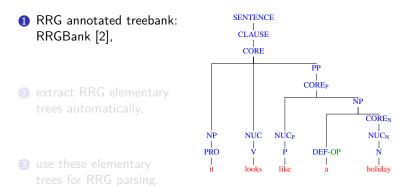
Automatic RRG Grammar Extraction

Parsing experiments

Issues

Summary & Outlook

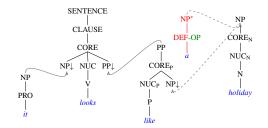
Introduction



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 RRG annotated treebank: RRGBank [2],

 extract RRG elementary trees automatically,

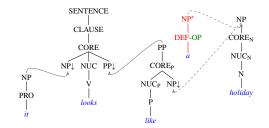


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3 use these elementary trees for RRG parsing.

Outline

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Automatic RRG Grammar Extraction

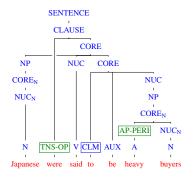
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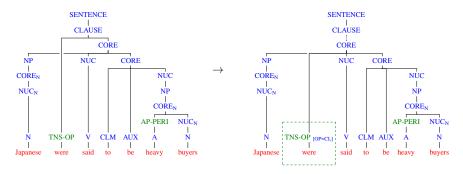
Summary & Outlook

RRGbank

- ★ Corpus of RRG annotated sentences [2] → automatically converted from Penn Treebank,
 - \rightarrow manually checked and validated;
- 395 gold sentences, 1090 silver annotated sentences;
- * RRGbank and RRG annotation tool: rrgbank.phil.hhu.de.

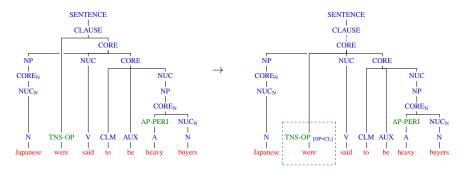


Removing crossing branches



 $\star\,$ we transform the RRG structures to remove crossing branches,

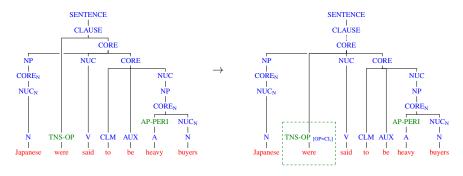
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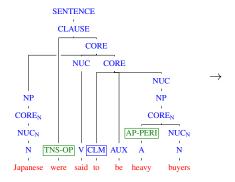
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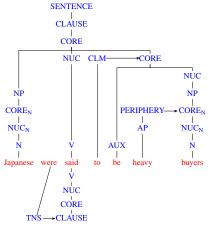
Removing crossing branches



- \star we transform the RRG structures to remove crossing branches,
- \star we mark the original position of the node with [OP=CL],
- \star original tree structure is easily recovered.

Operator projection and periphery can be recovered





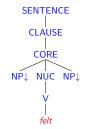
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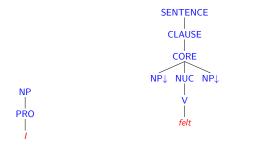
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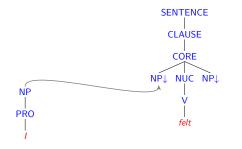
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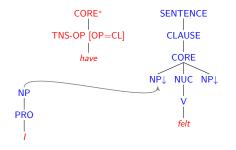
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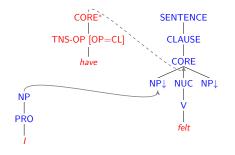
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- \star Three tree composition operations:
 - \rightarrow substitution (argument slot filling)
 - \rightarrow wrapping substitution (displaced argument slot filling)
 - \rightarrow sister adjunction (adding operators and periphery elements);
- $\star\,$ Such RRG grammars capture long-distance dependencies $\rightarrow\,$ for example, WH-movement.

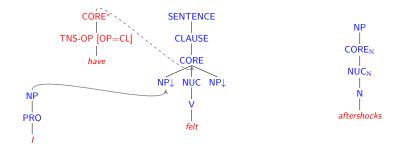


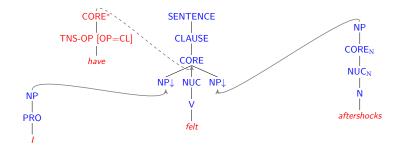


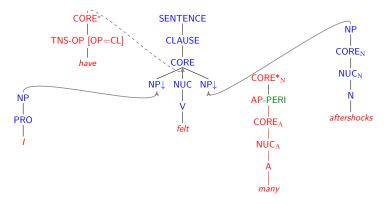




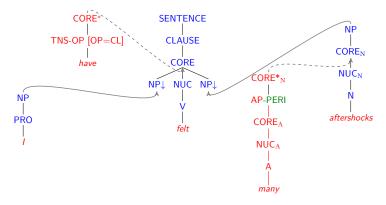




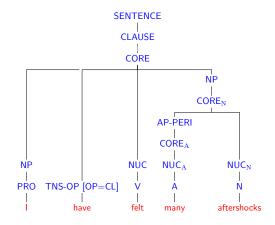




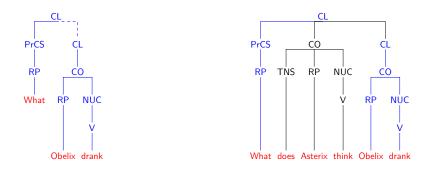
Sentence: I have felt many aftershocks



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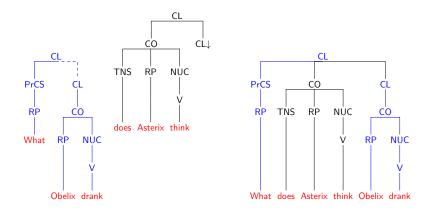


Combination operations: Wrapping substitution

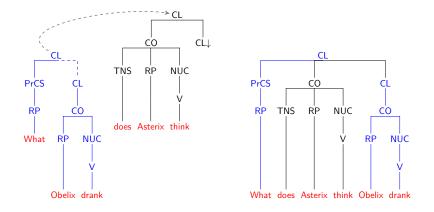


Sentence: What does Asterix think Obelix drank

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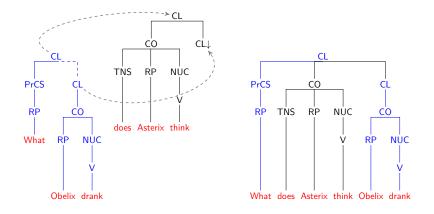


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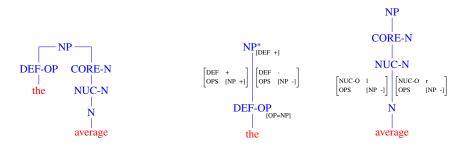


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Elementary trees with features

RRGbank

Extracted Elementary Trees



 \star elementary trees are enhanced with features

Elementary trees with features

RRGbank

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 $\star\,$ elementary trees are enhanced with features $\rightarrow\,$ edge features

Elementary trees with features

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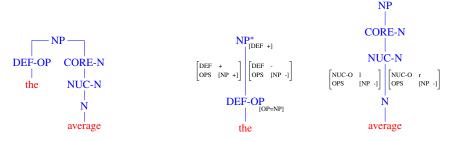


- \star elementary trees are enhanced with features
 - \rightarrow edge features
 - \rightarrow node features

Edge features

RRGbank

Extracted Elementary Trees



Left and right edge feature structures:

Edge features

RRGbank

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- \star unify adjacent structures in the derived tree
- ★ model ordering constraints
- \star percolate upwards until phrasal nodes

Node features

RRGbank

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***** One feature structure per node:

- unify during tree composition
- store syntactic or syn-sem interface information.

\star Unification successful \rightarrow accept parse tree

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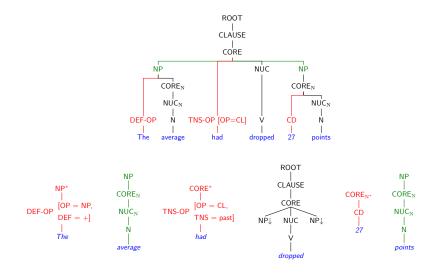
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Summary & Outlook

RRG Grammar extraction algorithm (1)

- ★ Elementary tree extraction inspired by Xia [6] algorithm for induction of Tree-Adjoining Grammars.
- * Top-down extraction of elementary trees.
- \star Heuristics from head-modifier percolation tables.
- $\star\,$ We use RRG structures from RRGbank for automatic grammar induction
 - \rightarrow rrgbank.phil.hhu.de.

RRG Grammar extraction algorithm (2)



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- $\star\,$ suitable for automatically extracted RRG grammars.

Extracted RRG grammar

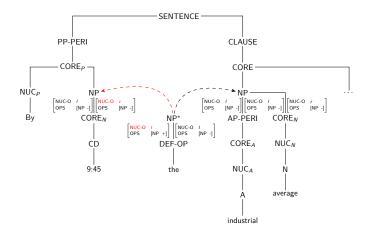
- removed punctuation
- exhaustive parsing (i.e. not probabilistic, overgenerating a lot)
- 2 versions:
 - no features
 - 2 edge features for operators model adjunction constraints
- do feature structures eliminate parse trees that contradict linguistic intuitions?

Parsing experiments

| | Gold Grammar | | Silver Grammar | |
|-----------------------|--------------|-----------|----------------|-----------|
| | w feats | w/o feats | w feats | w/o feats |
| Sentences | 395 | | 1480 | |
| avg. sentence length | 6.1 | | 8.0 | |
| token-supertag pairs | 1526 | 1497 | 6288 | 6044 |
| avg. number of parses | 6.9 | 12.7 | 1166 | 2939 |
| savings | 45.1% | | 39.7% | |

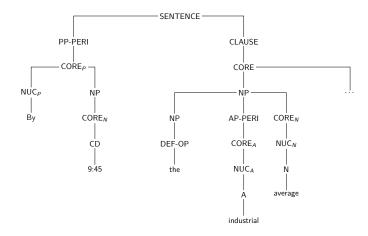
features decrease number of results by $\approx 45\%$

Features restrict adjunction of operators and periphery (1)



Sentence: By 9:45, the industrial average had dropped 27 points.

Features restrict adjunction of operators and periphery (2)



Sentence: By 9:45, the industrial average had dropped 27 points.

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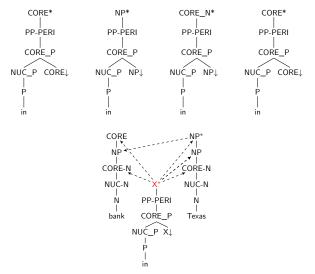
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Attachment ambiguities



syntactic information needed that might not be in RRGBank

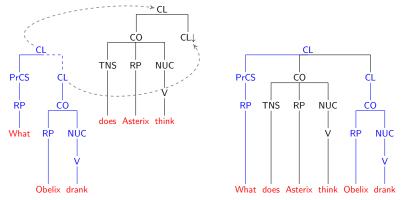
Size of the grammar

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the number of parses per sentence increases with the size of the grammar

Issues

Extraction of wrapping substitution trees



- discontinuous constituents are marked with traces in PTB; *
- no special marking in RRGBank; *
- * transfer traces from PTB to RRG trees in RRGbank?

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Summary

- * automatically extracted elementary trees from RRGBank;
- \star experiments with exhaustive parsing of 395/1480 sentences;
- \star parsing w/o edge features \rightarrow too many results;
- \star some edge features already rule out bad results.

Future Work

- \star adding more edge features = better results;
- ★ extract wrapping elementary trees;
- \star 100s of results \rightarrow not satisfying;
- * ambiguity and annotation/extraction mistakes have bad consequences;
- * use unlexicalized elementary trees (= supertags);
- * probabilistic grammar and parsing \rightarrow A* parsing algorithm ParTAGe by Waszczuk (2017) [5];
- ★ Web GUI.

Thank you!

THANK YOU VERY MUCH FOR YOUR ATTENTION!

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