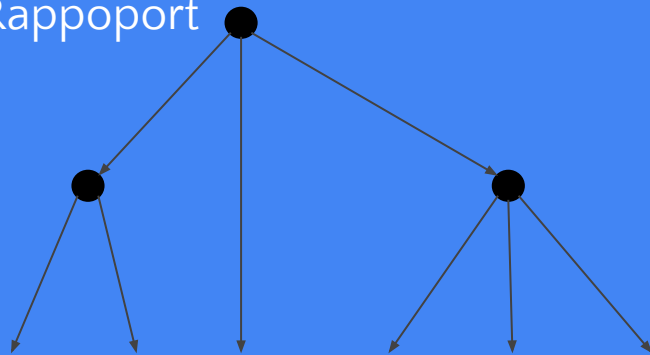


Broad-Coverage Semantic Parsing: A Transition-Based Approach

Daniel Hershcovich, Omri Abend and Ari Rappoport
Hebrew University of Jerusalem



Outline

- Background
- Conversion-Based Parsing
- Broad-Coverage Semantic Parsing

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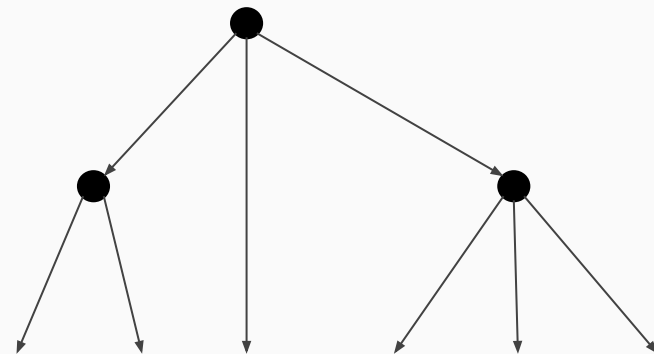
- Background
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- Broad-Coverage Semantic Parsing

Given a sequence of tokens $w = w_1, \dots, w_n$,

A (labeled) directed graph (V, E) where $\{w_i\} \subseteq V$ is a

grounded representation of w .

Examples:

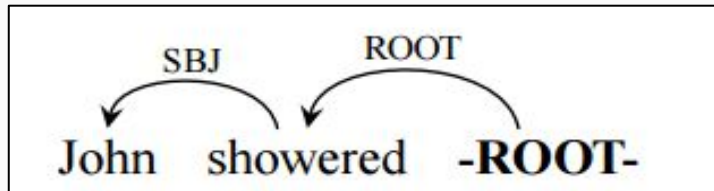
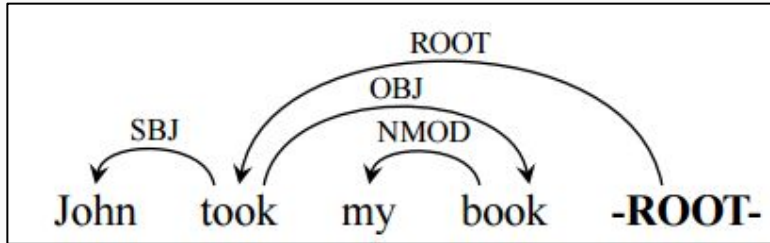
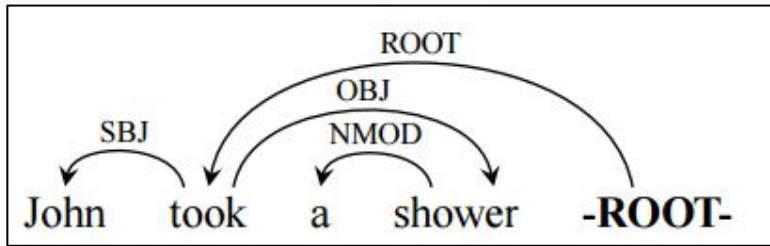


Dependency/constituency trees, UCCA...

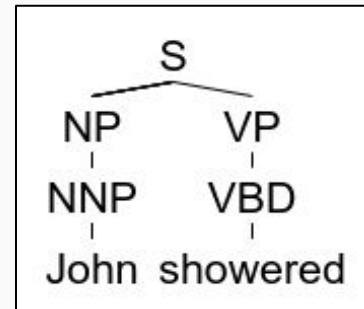
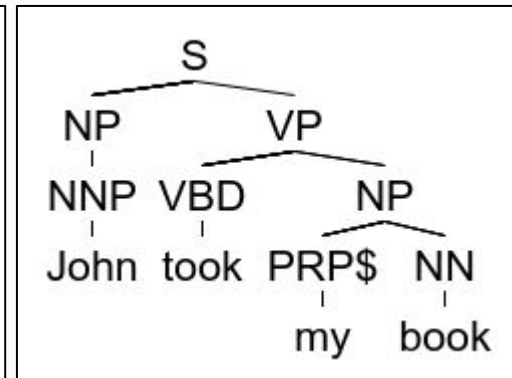
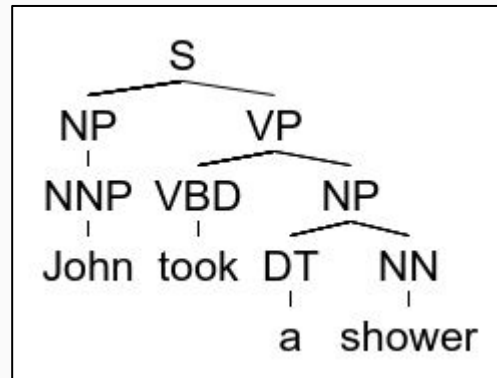
Why Semantic Representation?

Syntactic representation is sensitive to formal variations.

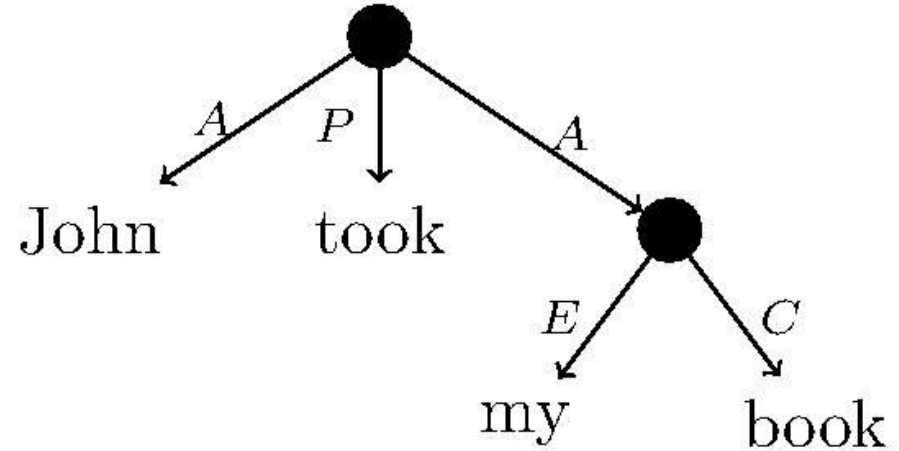
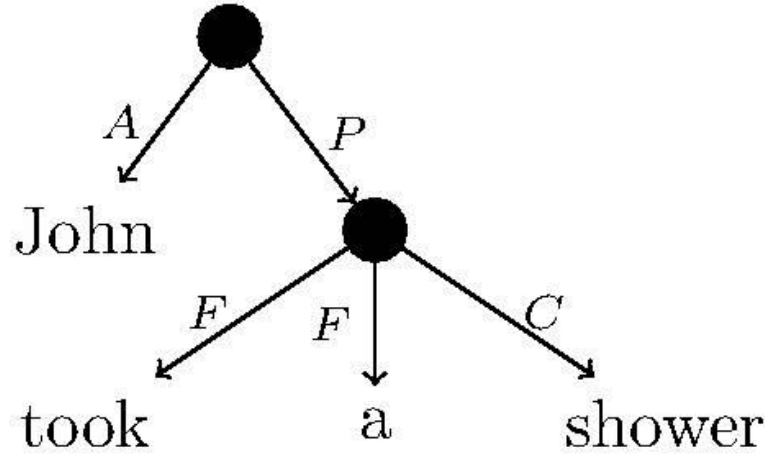
Dependency:



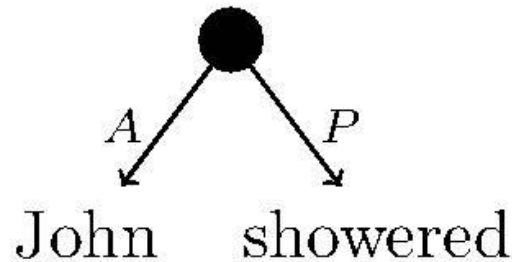
Constituency:



Universal Conceptual Cognitive Annotation (UCCA)



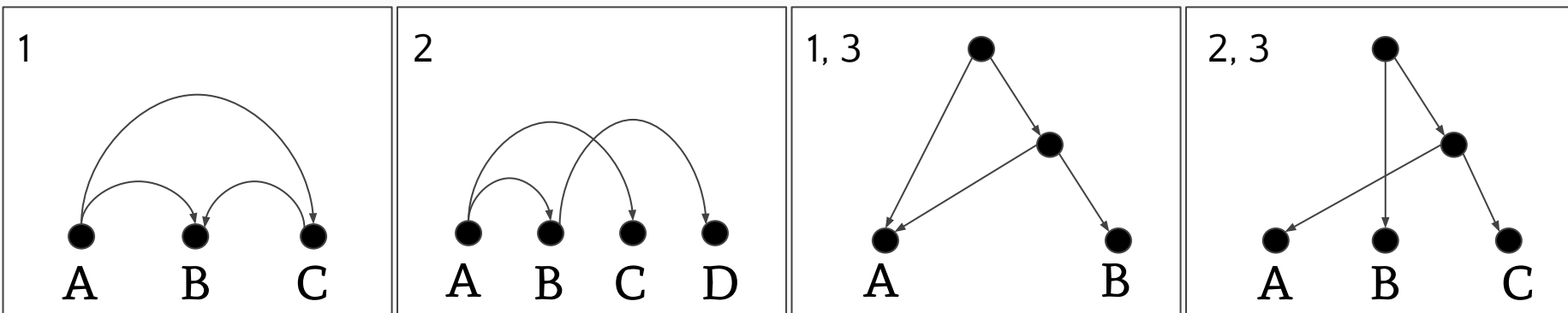
A participant
P process
F function
E elaborator
C center



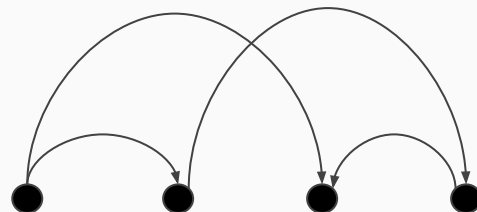
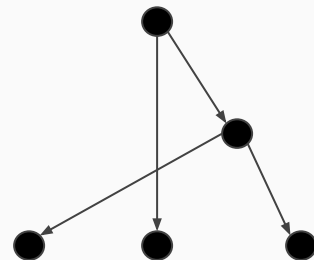
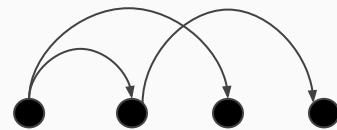
Abend and Rappoport (2013)

Properties required for full semantic coverage in grounded representations:

1. **Multiple parents (DAG).**
2. **Non-projectivity (discontinuity).**
3. **Non-terminal nodes.**



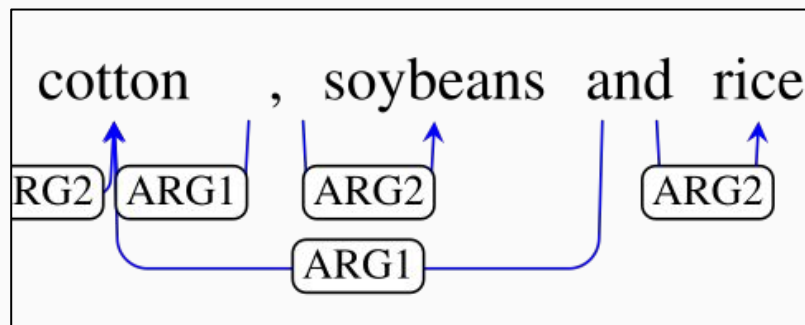
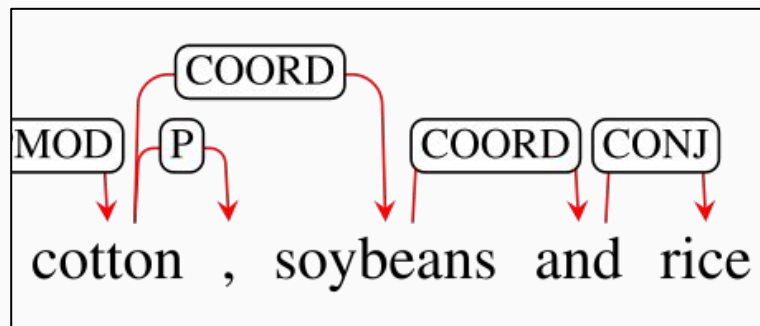
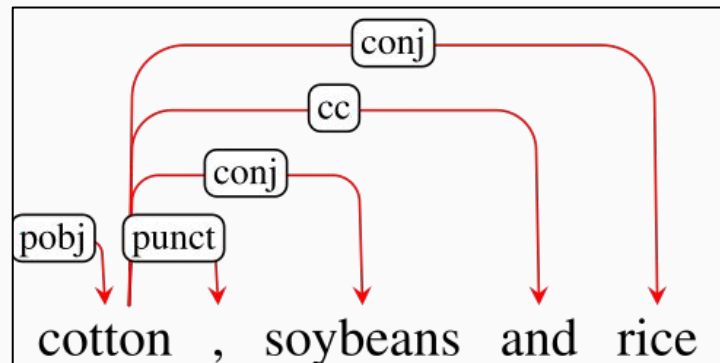
- Non-projective dependency parsing
- Discontinuous constituency parsing
- Semantic dependency parsing (SDP)



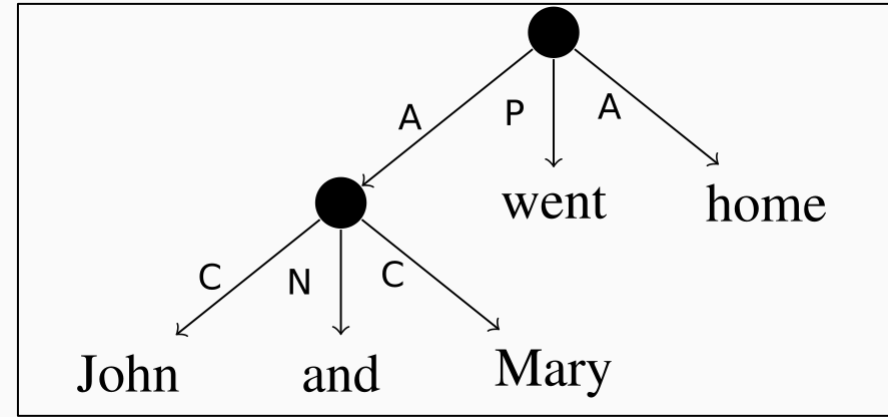
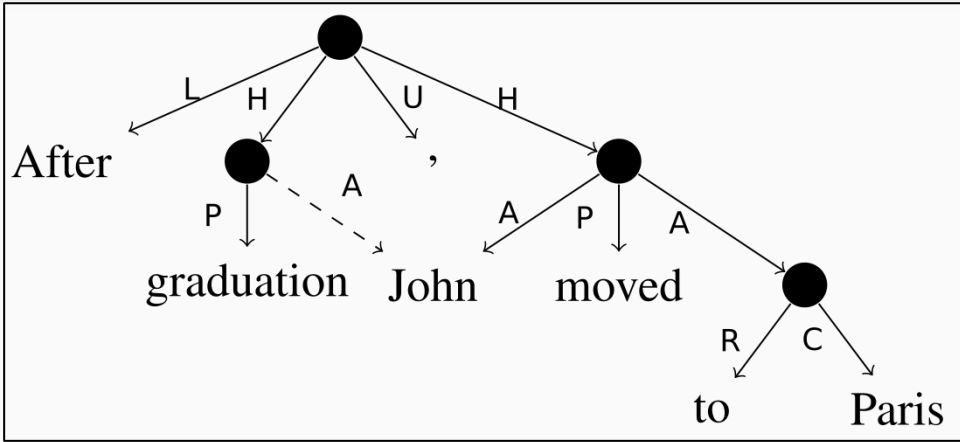
Why Non-Terminals?

Some frequent constructions do not have one clear head,
e.g. coordination, some multi-word expressions, compounds.

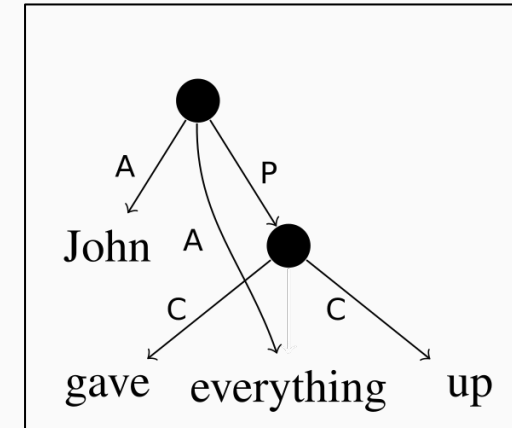
Coordination in SDP:



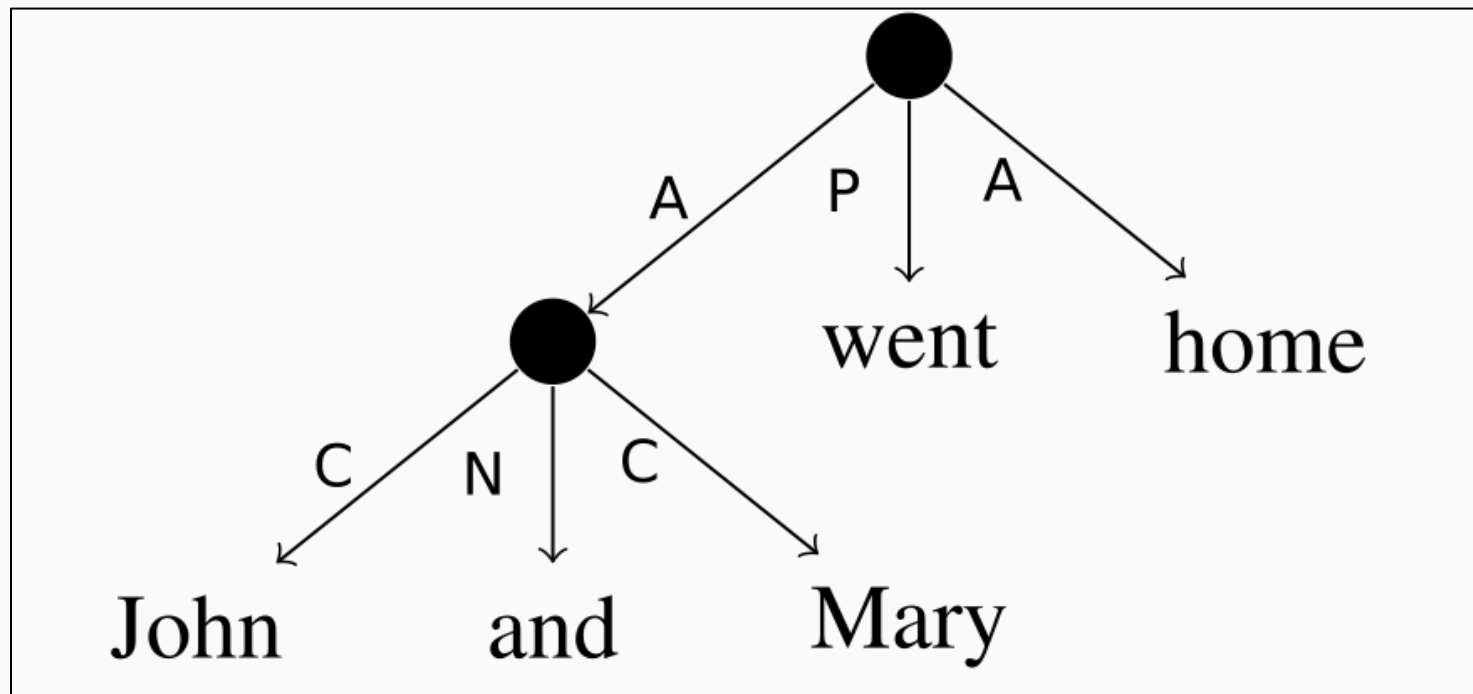
Structural Properties in UCCA



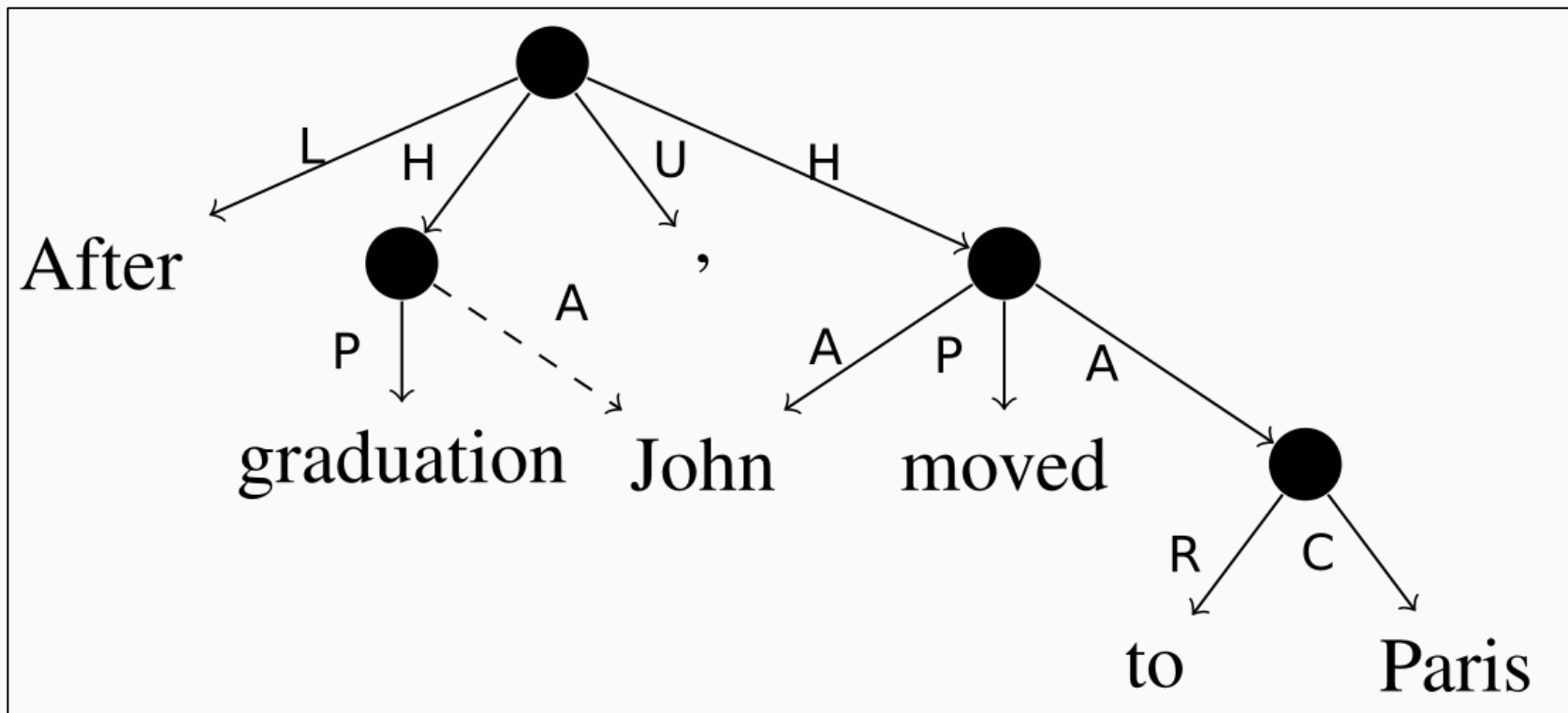
H	parallel scene	D	adverbial	N	connector
A	participant	E	elaborator	R	relator
P	process	S	state	C	center
G	ground	L	linker	F	function



Coordination represented by one parent node

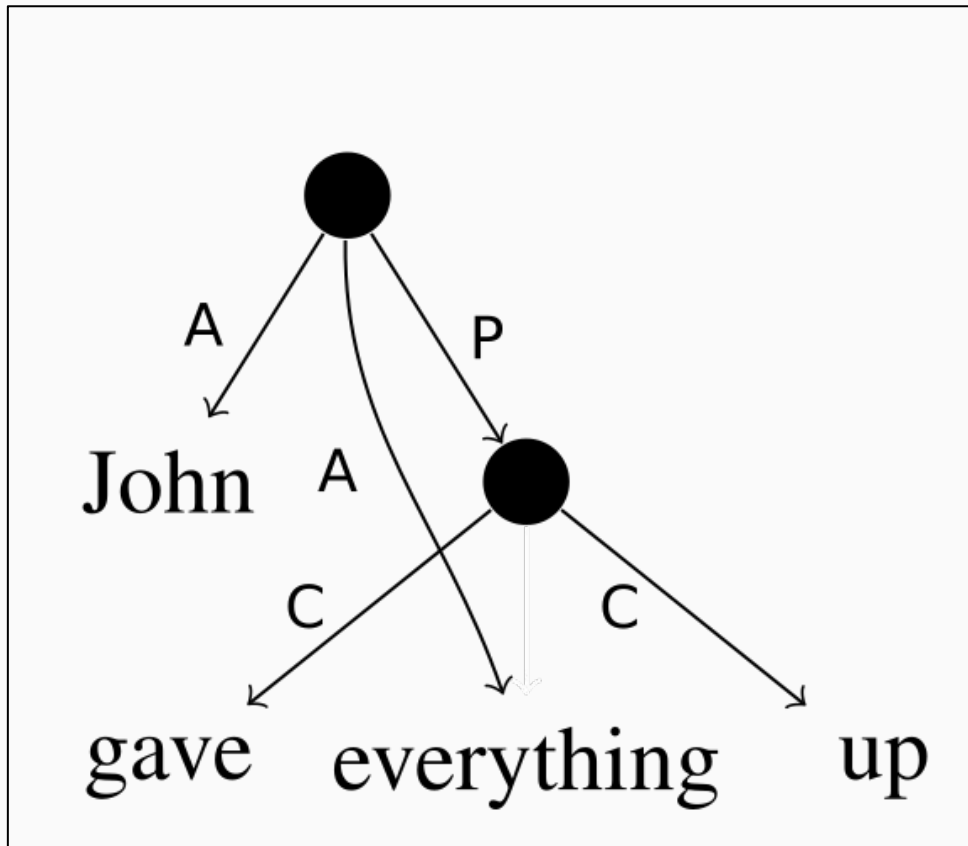


Multiple Parents

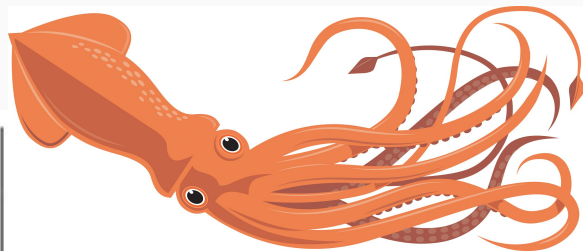


----> Remote edges denote implicit relations

Multi-word expression
annotated as one unit



160K tokens from English Wikipedia
+25K tokens from *Twenty Thousand
Leagues Under the Sea*

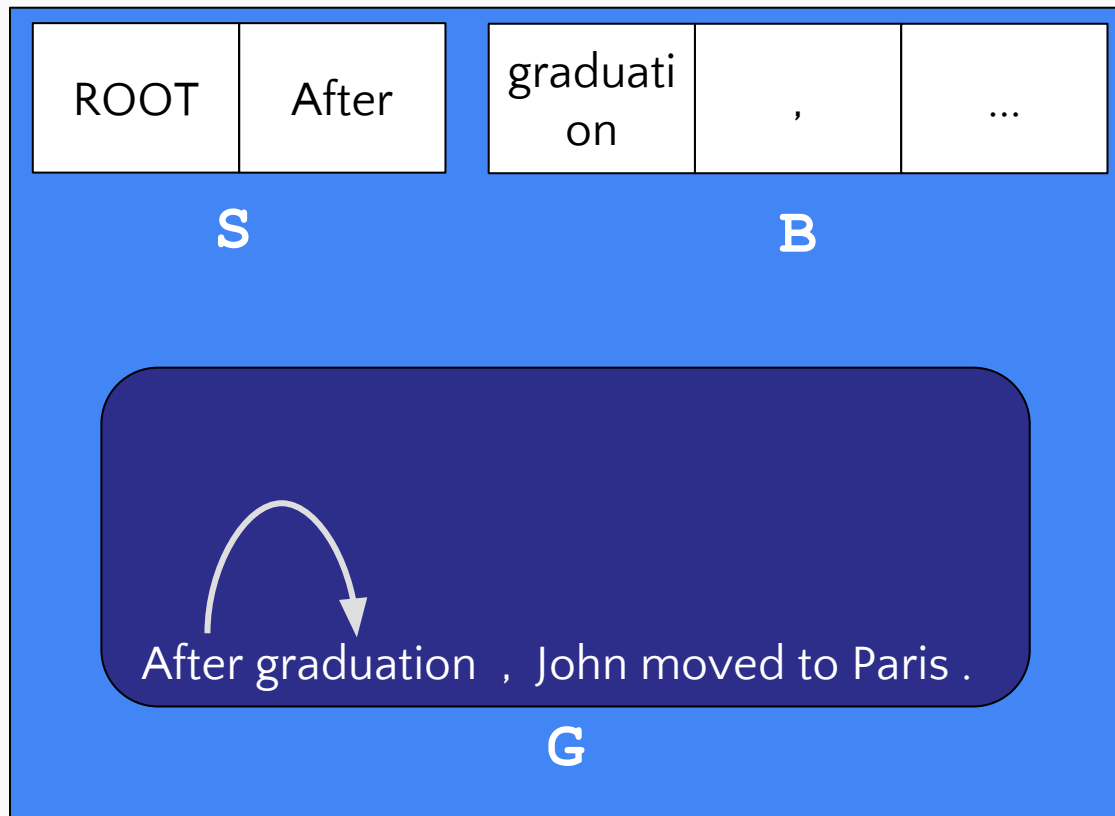


	Train	Dev	Test	
# passages	281	35	43	154
# sentences	4021	537	608	522
# nodes	277,587	40,700	45,047	29,965
% terminal	42.41	42.8	42.66	41.23
% non-term.	57.59	57.20	57.34	58.77
% discount.	0.52	0.55	0.47	0.79
% > 1 parent	2.29	1.89	2.21	1.98

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Transition-Based Dependency Parsing



B:

Buffer of nodes to process, initialized to the list of tokens.

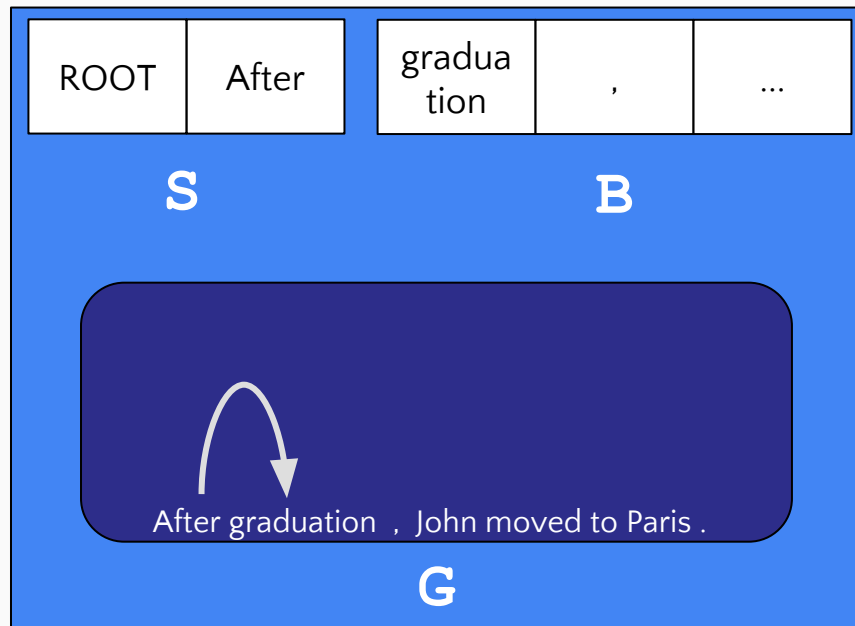
S:

Stack of partially processed nodes, initially just the root.

G:

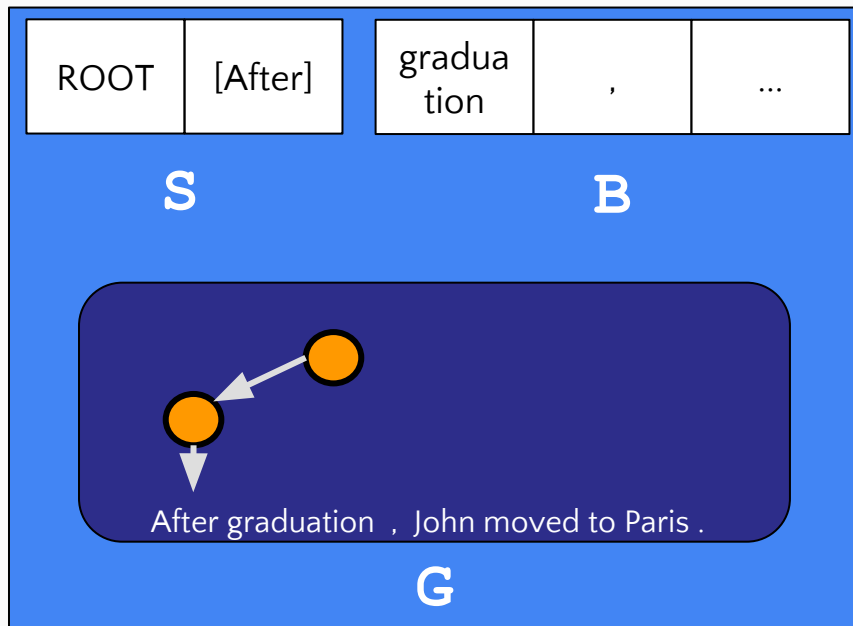
Graph of constructed edges.

Classifier selects next transition given current state



Shift
Reduce
Right-Edge _x
Left-Edge _x
Swap

Transitions to create new nodes



Shift
Reduce _{<i>x</i>}
Unary _{<i>x</i>}
Finish
Swap

1. Convert UCCA to dependency and constituency trees.
2. Apply existing transition-based parsers.
3. Convert back to UCCA.

Dependency parsers:

MaltParser (Nivre 2003), Stack-LSTM Parser (Dyer et al. 2015)

Constituency parser:

UPARSE (Maier 2015): discontinuous constituency parser

Scores on the *Wiki* test set:

	Primary			Remote		
	LP	LR	LF	LP	LR	LF
Constituency Tree Conversion						
UPARSE	64	67.3	65.4	—	0	0
Upper Bound	100	100	100	—	0	0
Dependency Tree Conversion						
Malt _{arc-standard}	63.4	57.3	60.1	—	0	0
Malt _{arc-eager}	63.9	57.9	60.5	—	0	0
LSTM	73.2	66.2	69.2	—	0	0
Upper Bound	93.8	83.7	88.4	—	0	0

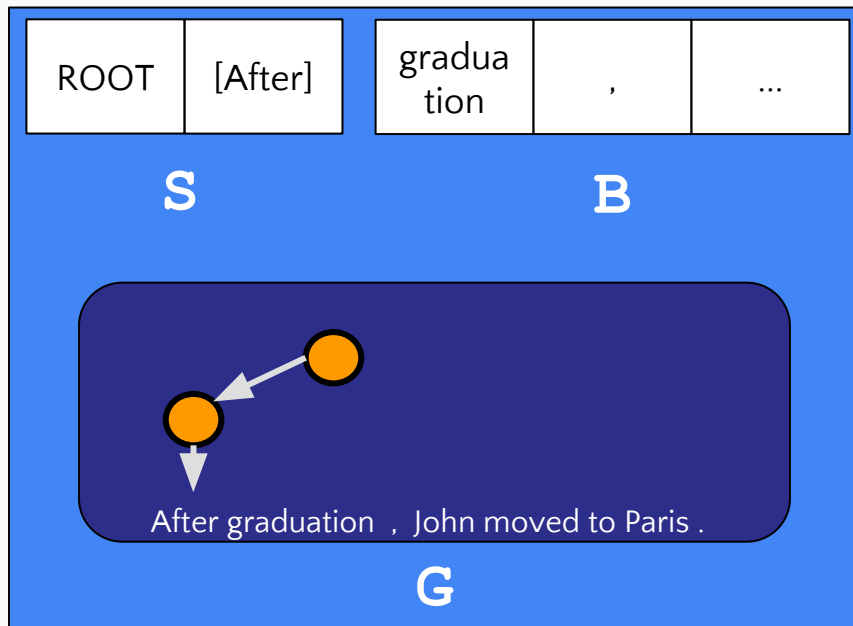
Upper bound is due to lossy conversion algorithms.

- MaltParser: perceptron/SVM
- UPARSE: perceptron
- Stack-LSTM parser: recurrent neural network
+ continuous features

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Discontinuous DAG parser



Perceptron

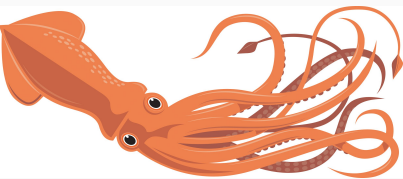


Shift
Reduce
Node _{<i>x</i>}
Left-Edge _{<i>x</i>}
Right-Edge _{<i>x</i>}
Left-Remote _{<i>x</i>}
Right-Remote _{<i>x</i>}
Swap
Finish

Scores on the *Wiki* test set and the *20K leagues* set:

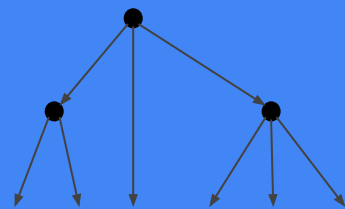


	Primary			Remote		
	LP	LR	LF	LP	LR	LF
BSP	62.4	56	59	15.3	11.8	13.3
BSP _{Tree}	63.8	56.5	59.9	—	0	0
Out-of-domain						
BSP	60.6	53.9	57.1	20.2	10.3	13.6
BSP _{Tree}	60.2	52.8	56.2	—	0	0



(BSP_{Tree} trained on converted trees without remote edges)

Conclusion



- The structural desiderata of grounded semantic parsing is not supported by today's parsers
- We present a transition-based system that does
- Encouraging results with UCCA suggest that NN-based classification may be helpful for better performance

Future Work

- Neural network for BSP classifier
- Improved conversions
- Beam search
- More languages, e.g. German

Thank you



References

- Angelina Ivanova et al. Who did what to whom?: A contrastive study of syntacto-semantic dependencies. 2012. In *Proc. of LAW*.
- Stephan Oepen et al. SemEval 2015 Task 18: Broad-Coverage Semantic Dependency Parsing. 2015. In *Proc. of SemEval*.
- Omri Abend and Ari Rappoport. Universal Conceptual Cognitive Annotation (UCCA). In *Proc. of ACL*.