IN2110: Methods in Language Technology Grammatical Structure Wrap-Up

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Language Technology Group (LTG)

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- ► Short recap:
 - Universal Dependencies
 - Transition-based dependency parsing



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- Syntactic structure in negation resolution



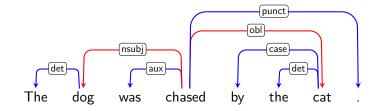
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- Syntactic structure in negation resolution
- Sample exam questions

Recap: 'Universal' Dependency Types

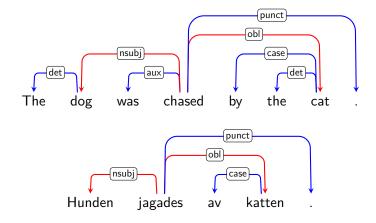


nsubj	nominal subject	She <u>arrived</u> .		
csubj	clausal subject	That she arrived surprised me.		
obj	(direct) object	My mother <u>called</u> me.		
iobj	indirect object	She <u>teaches</u> my daughter maths.		
ccomp	clausal complement	She <u>knew</u> that she arrived.		
xcomp	open clausal complement	She promised to sing.		
obl	oblique nominal	She <u>arrived</u> on Monday		
obl	oblique nominal	She <u>depends</u> on <u>me</u> .		
nmod	nominal modifier	the <u>office</u> of the chair is empty.		
amod	adjectival modifier	the fierce dog barks.		
acl	adjectival clause	the \underline{dog} that \underline{barks} arrived.		
conj	conjunct	Kim and Sandy arrived.		
сс	coordinating conjunction	Kim and Sandy arrived.		

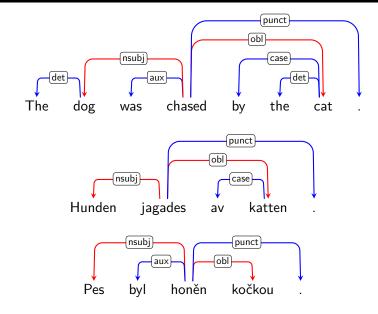




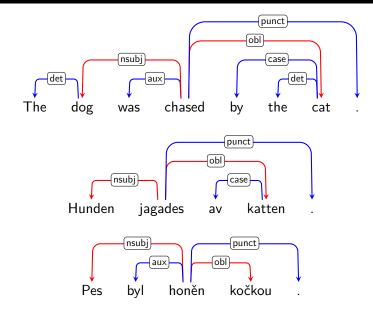








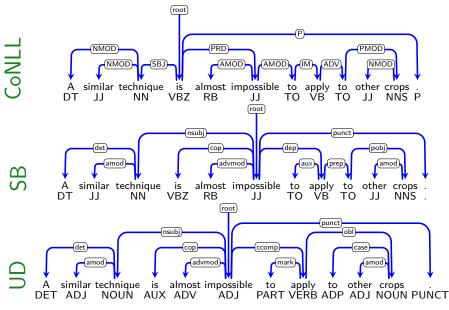




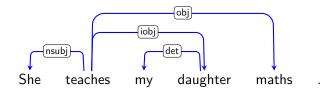
• Capitalize on content words, e.g. demote case-marking prepositions.

Functional vs. Content Heads

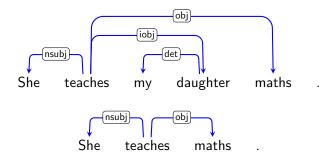




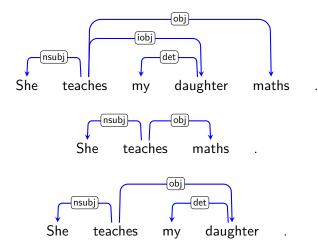




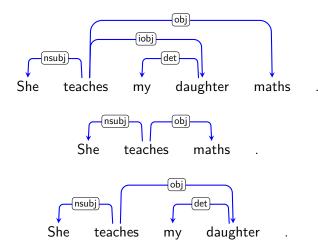












▶ UD guidelines: [...] if there is just one object, it should be labeled obj.

UD: The Big Picture



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	Nominals	Clauses	Modifier words	Function Words	
Core arguments	<u>nsubj</u> obj	<u>csubj</u> ccomp			
	iobj	xcomp			
Non-core dependents	<u>obl</u>	<u>advcl</u>	advmod*	aux	
	<u>vocative</u> <u>expl</u> <u>dislocated</u>		<u>discourse</u>	<u>cop</u> mark	
Nominal dependents	nmod	acl	amod	<u>det</u>	
	<u>appos</u> nummod			<u>clf</u> <u>case</u>	
Coordination	MWE	Loose	Special	Other	
<u>conj</u>	fixed	list	orphan	punct	
<u>cc</u>	<u>flat</u> compound	<u>parataxis</u>	goeswith reparandum	<u>root</u> dep	



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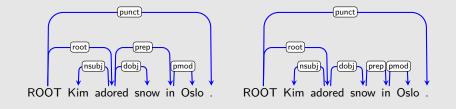
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LAS: Labeled Attachment Score

▶ In addition to the head, is the dependency type (edge label) correct?

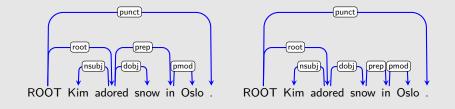
Exercise (5): Dependency Evaluation





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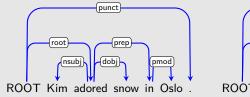


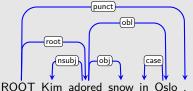


(5) What are the LAS and UAS scores for the two trees? Gold standard on the left, system prediction on the right.

Exercise (6): More Dependency Evaluation

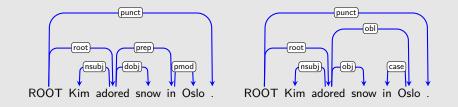






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Recap: Dependency Parsing Summary

Transition-Based Dependency Parsing

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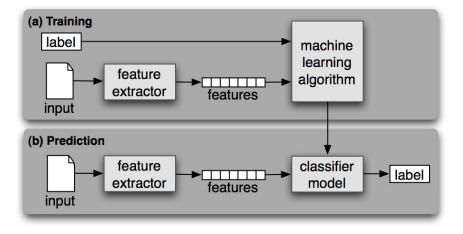


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- ► Multi-stratal (multi-layer) representations: MTT, FGD, enhanced UD.

The Transition Oracle as a Classifier





Recap: Adapting Shift-Reduce Parsing



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- Shift ('read') tokens from input buffer, one at a time, left-to-right;
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- ► At REDUCE, token should be fully processed (head and dependents).
- ► LEFT-ARC must respect single-head constraint and unique root node.

Recent Advances in Dependency Parsing

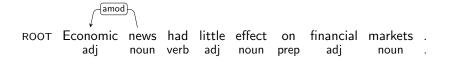
Tutorial, EACL, April 27th, 2014

Ryan McDonald¹ Joakim Nivre²

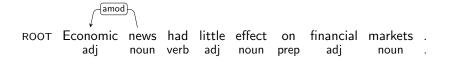
¹Google Inc., USA/UK E-mail: ryanmcd@google.com

²Uppsala University, Sweden E-mail: joakim.nivre@lingfil.uu.se

[ROOT]_S [news, had, little, effect, on, financial, markets, .]_B



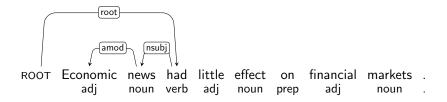
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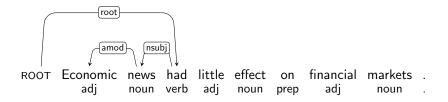
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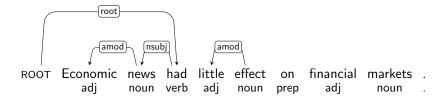
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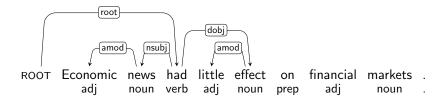
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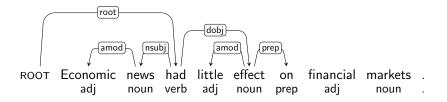
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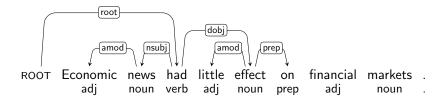
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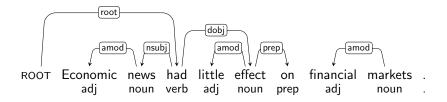
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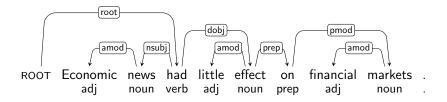
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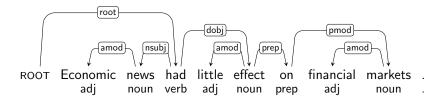
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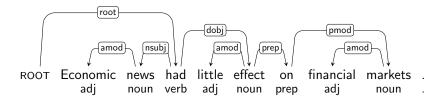
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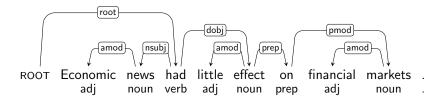
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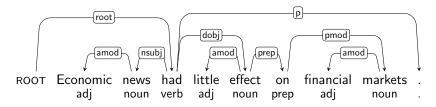
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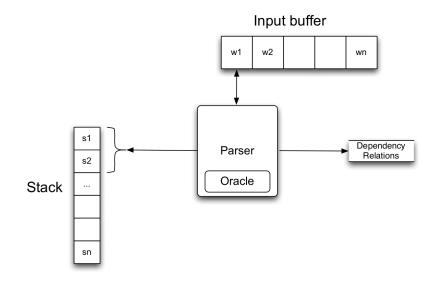


Arc-Standard Transition System [Nivre 2004]

Configuration:	(S, B, A) $[S = Stack, B = Buffer, A = Arcs]$
Initial:	$([], [0, 1, \ldots, n], \{\})$
Terminal:	([0], [], A)
Shift:	$(S, i B, A) \Rightarrow (S i, B, A)$
Right-Arc(k):	$(S i j,B,A) \Rightarrow (S i,B,A \cup \{(i,j,k)\})$
Left-Arc(k):	$(S i j,B,A) \Rightarrow (S j,B,A \cup \{(j,i,k)\}) i \neq 0$

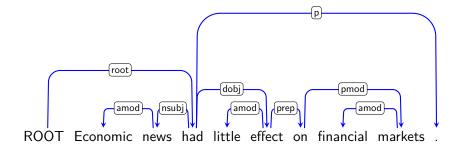
Arc-Standard More Like 'Classic' Shift-Reduce





Using the Arc-Standard Transition System





In Comparison



SHIFT SHIFT SHIFT LEFT-ARC(AMOD) LEFT-ARC(NSUBJ) SHIFT SHIFT LEFT-ARC(AMOD) SHIFT SHIFT SHIFT SHIFT LEFT-ARC(AMOD) RIGHT-ARC(PMOD) RIGHT-ARC(PREP) RIGHT-ARC(DOBJ) RIGHT-ARC(P)SHIFT RIGHT-ARC(ROOT)

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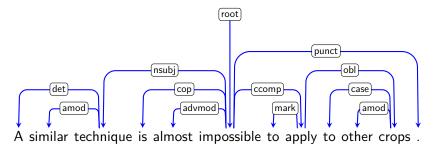


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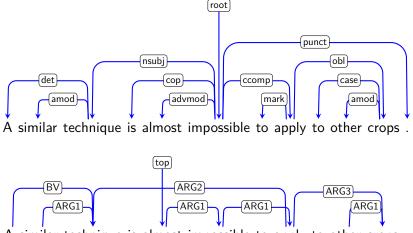
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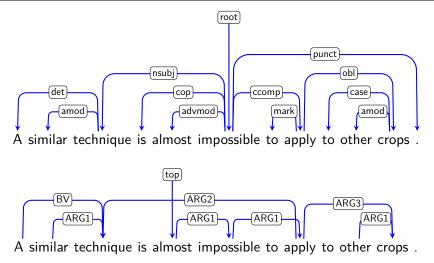




A similar technique is almost impossible to apply to other crops .

Outlook: Semantics as Dependency Graphs





- DELPH-IN MRS Dependencies: General graph (beyond rooted trees).
- Argument sharing requires re-entrancy, e.g. control or relative clauses.

Looking Back: How We had Motivated Syntactic Structure

Formal grammars describe a language, providing key notions of:

Wellformedness

- ► Kim was happy because _____ passed the exam.
- ► Kim was happy because _____ final grade was an A.
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- Kim gave Sandy the book.
- ► Kim gave the book to Sandy.
- Sandy was given the book by Kim.

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Ambiguity

- Kim ate sushi with chopsticks.
- Have her report on my desk by Friday!