

Natural Language Processing

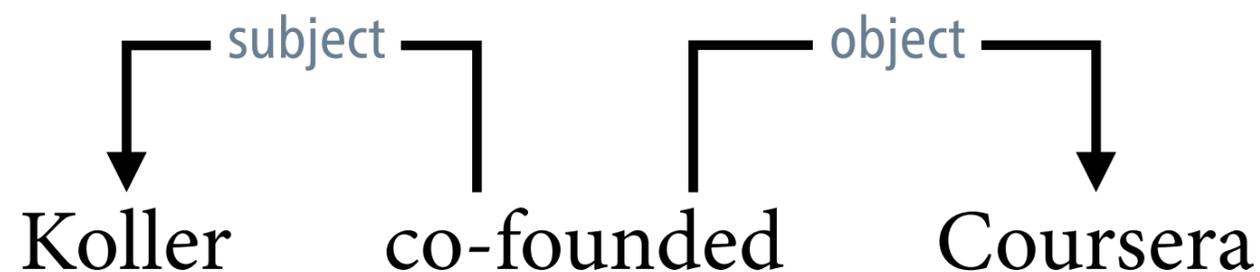
# Introduction to dependency parsing

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# Dependency parsing

- **Syntactic parsing** is the task of mapping a sentence to a formal representation of its syntactic structure.
- We focus on representations in the form of **dependency trees**.



- A syntactic dependency is an asymmetric relation between a **head** and a **dependent**.

## Current UD Languages

Information about language families (and genera for families with multiple branches) is mostly taken from [WALS Online](http://WALS Online) (IE = Indo-European).

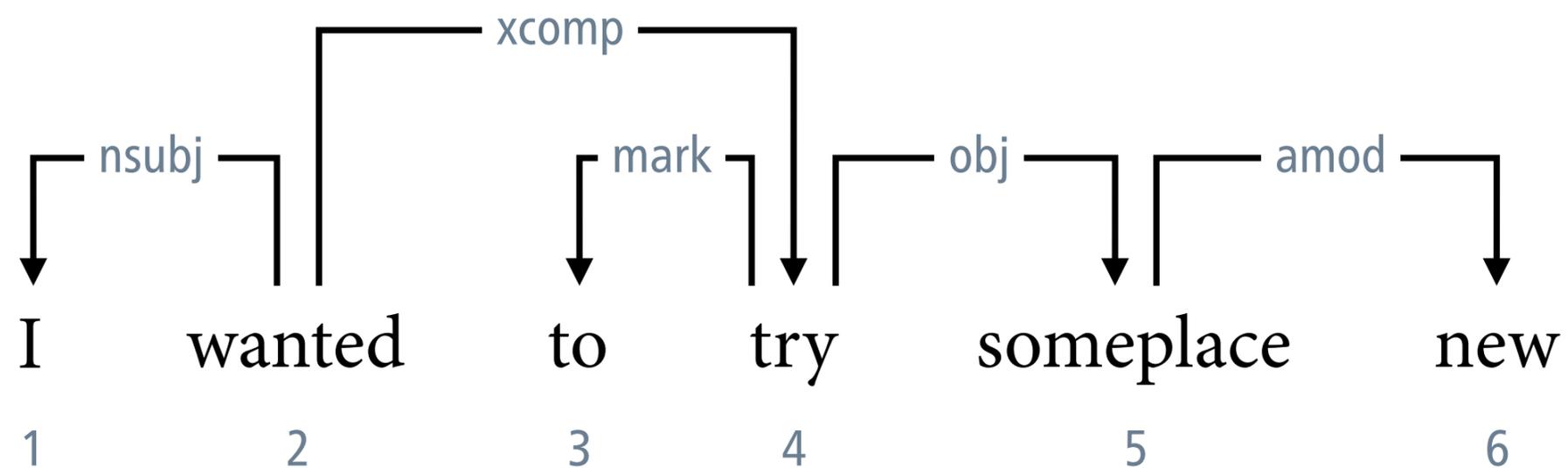
	Afrikaans	1	49K		IE, Germanic
	Akkadian	1	1K		Afro-Asiatic, Semitic
	Albanian	1	<1K	W	IE, Albanian
	Amharic	1	10K	  	Afro-Asiatic, Semitic
	Ancient Greek	2	416K	 	IE, Greek
	Arabic	3	1,042K	 W	Afro-Asiatic, Semitic
	Armenian	1	52K	  	IE, Armenian
	Assyrian	1	<1K	 	Afro-Asiatic, Semitic
	Bambara	1	13K	 	Mande
	Basque	1	121K		Basque
	Belarusian	1	13K	  	IE, Slavic
	Bhojpuri	2	4K	 	IE, Indic
	Breton	1	10K	   	IE, Celtic
	Bulgarian	1	156K	 	IE, Slavic
	Buryat	1	10K	 	Mongolic
	Cantonese	1	13K		Sino-Tibetan
	Catalan	1	531K		IE, Romance
	Chinese	5	285K	   W	Sino-Tibetan
	Classical Chinese	1	74K		Sino-Tibetan
	Coptic	1	40K	  	Afro-Asiatic, Egyptian
	Croatian	1	199K	 W	IE, Slavic
	Czech	5	2,222K	    W	IE, Slavic
	Danish	2	100K	  	IE, Germanic
	Dutch	2	306K	 W	IE, Germanic
	English	9	620K	                            	

# Dependency trees

- A **dependency tree** for a sentence  $x$  is a digraph  $G = (V, A)$  where  $V = \{1, \dots, |x|\}$  and where there exists a  $r \in V$  such that every  $v \in V$  is reachable from  $r$  via exactly one directed path.
- The vertex  $r$  is called the **root** of  $G$ .
- The arcs of a dependency tree may be labelled to indicate the type of the syntactic relation that holds between the two elements.

Universal Dependencies v2 uses 37 universal syntactic relations ([list](#)).

# Representation of dependency trees



word position	1	2	3	4	5	6
head position	2	0	4	2	4	5
dependency relation	nsubj	root	mark	xcomp	obj	amod

# Two parsing paradigms

- **Graph-based dependency parsing**

Cast parsing as a combinatorial optimisation problem over a (possibly restricted) set of dependency trees.

- **Transition-based dependency parsing**

Cast parsing as a sequence of local classification problems: at each point in time, predict one of several parser actions.

# Graph-based dependency parsing

- Given a sentence  $x$  and a set  $Y(x)$  of candidate dependency trees for  $x$ , we want to find a highest-scoring tree  $\hat{y} \in Y(x)$ :

$$\hat{y} = \arg \max_{y \in Y(x)} \text{score}(x, y)$$

- The computational complexity of this problem depends on the choice of the set  $Y(x)$  and the scoring function.

# The arc-factored model

- Under the **arc-factored model**, the score of a dependency tree is expressed as the sum of the scores of its arcs:

$$\hat{y} = \arg \max_{y \in Y(x)} \sum_{a \in y} \text{score}(x, a) \text{ ————— head-dependent arc}$$

- The score of a single arc can be computed by means of a neural network that receives the head and the dependent as input.

for example, a simple linear layer:  $\text{score}(x, h \rightarrow d) = [\mathbf{h} ; \mathbf{d}] \cdot \mathbf{w} + b$

# Computational complexity

- Under the arc-factored model, the highest-scoring dependency tree can be found in  $O(n^3)$  time ( $n =$  sentence length).

Chu–Liu/Edmonds algorithm; McDonald et al. (2005)

- Even seemingly minor extensions of the arc-factored model entail intractable parsing.

McDonald and Satta (2007)

- For some of these extensions, polynomial-time parsing is possible for restricted classes of dependency trees.

# Transition-based dependency parsing

- We cast parsing as a sequence of local classification problems such that solving these problems builds a dependency tree.
- In most approaches, the number of classifications required for this is linear in the length of the sentence.

# Transition-based dependency parsing

- The parser starts in the **initial configuration**.

empty dependency graph

- It then calls a classifier, which predicts the **transition** that the parser should make to move to a next configuration.

extend the partial dependency tree

- This process is repeated until the parser reaches a **terminal configuration**.

complete dependency tree

# Training transition-based dependency parsers

- To train a transition-based dependency parser, we need a treebank with gold-standard dependency trees.
- In addition to that, we need an algorithm that tells us the gold-standard transition sequence for a tree in that treebank.
- Such an algorithm is conventionally called an **oracle**.

# Comparison of the two parsing paradigms

## **Graph-based parsing**

slow (in practice, cubic in the length of the sentence)

restricted feature models  
(in practice, arc-factored)

features and weights directly  
defined on target structures

## **Transition-based parsing**

fast (quasi-linear in the length of the sentence)

rich feature models  
defined on configurations

indirection – features and  
weights defined on transitions