

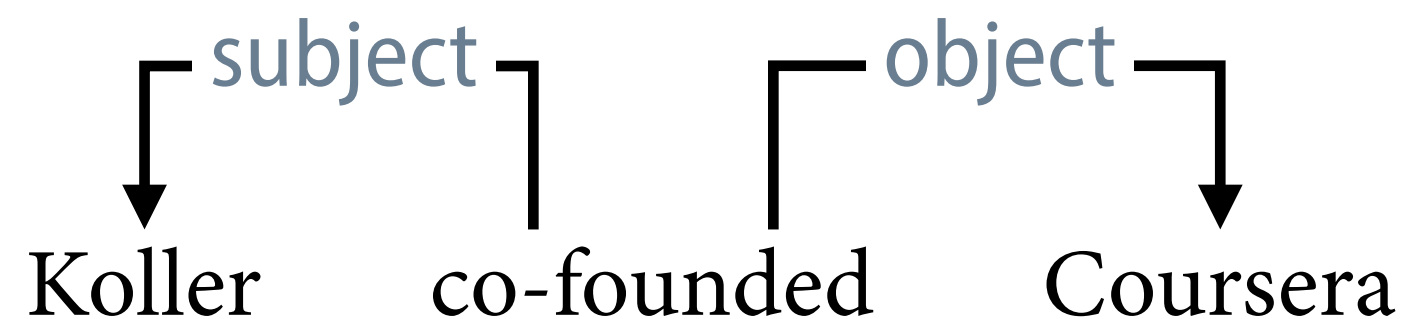
Introduction to dependency parsing

Marco Kuhlmann

Department of Computer and Information Science
Linköping University

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](https://wals.info/) (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		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		Sino-Tibetan
▶		Classical Chinese	1	74K		Sino-Tibetan
▶		Coptic	1	40K		Afro-Asiatic, Egyptian
▶		Croatian	1	199K		IE, Slavic
▶		Czech	5	2,222K		IE, Slavic
▶		Danish	2	100K		IE, Germanic
▶		Dutch	2	306K		IE, Germanic
▶		English	9	620K		IE, Germanic
▶		Erzya	1	15K		Uralic, Mordvin
▶		Estonian	2	465K		Uralic, Finnic
▶		Faroese	1	10K	W	IE, Germanic
▶		Finnish	3	377K		Uralic, Finnic
▶		French	8	1,157K		IE, Romance
▶		Galician	2	164K		IE, Romance
▶		German	4	3,753K		IE, Germanic
▶		Gothic	1	55K		IE, Germanic
▶		Greek	1	63K		IE, Greek
▶		Hebrew	1	161K		Afro-Asiatic, Semitic

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](#)).

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 tree

- 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