

# Syntactic Formalisms for Parsing Natural Languages

Aleš Horák, Miloš Jakubíček, Vojtěch Kovář  
(based on slides by Juyeon Kang)

`ia161@nlp.fi.muni.cz`

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## Dependency Syntax and Parsing

# Outline

- 1 Motivation
- 2 Dependency Syntax
- 3 Dependency Parsing

# Motivation

- what you have seen as far: applying analysis of formal languages to a natural language – creating a phrase-structure derivation tree according to some grammar
- PS accounts for one important syntactic property: **constituency**
- is that all?
- but what about: discontinuous phrases, structure sharing

# Motivation

- another crucial syntactic phenomenon is **dependency**
- what is a dependency? "some relation between two words"
- what is the difference to phrase-structure?
- what does constituency express?
- what does dependency express?

## Dependency Syntax (Mel'chuk 1988)

A more formal account – what is a dependency? A relation!

### Dependency Relation

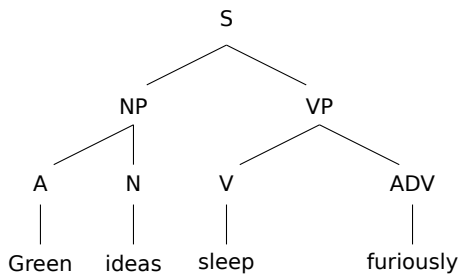
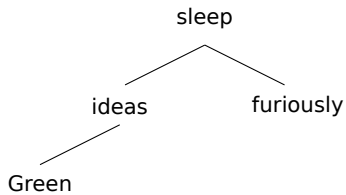
Let  $W$  be a set of all words within a sentence, then dependency relation  $\rightarrow$  is  $D \subseteq W \times W$  such that:

- $D$  is **anti-reflexive**:  $a \rightarrow b \Rightarrow a \neq b$
- $D$  is **anti-symmetric**:  $a \rightarrow b \wedge b \rightarrow a \Rightarrow a = b, \equiv$   
(anti-reflexivity)  $a \rightarrow b \Rightarrow b \nrightarrow a$
- $D$  is **anti-transitive**:  $a \rightarrow b \wedge b \rightarrow c \Rightarrow a \nrightarrow c$
- optionally:  $D$  is **labeled**: there is a mapping  $l : D \rightarrow L, L$  being the set of labels

# Dependency Representation

- $a \rightarrow b$ :  $a$  depends on  $b$ ,  $a$  is a dependent  $b$ ,  $b$  is the head of  $a$
- a **dependency graph**
- a **dependency tree**

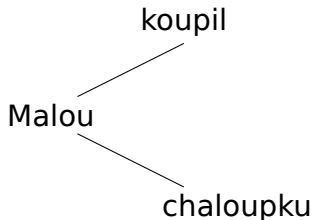
# Dependency Tree vs. PS Tree





## Non-projectivity

- a property of a dependency tree: a sentence is non-projective whenever drawing (projecting) a line from a node to the surface of the tree crosses an arc
- a lot of attention has been paid to this problem
- practical implications are rather limited (in most cases non-projectivity can be easily handled or avoided)
- hard cases:



# Czech Tradition of Dependency Syntax

- a long tradition of dependency syntax in the Prague linguistic school (Sgall, Hajičová, Panevová)
- Institute of Formal and Applied Linguistics at Charles University
- formalized as Functional Generative Description (FGD) of language
- Prague Dependency Treebank (PDT)

## Dependencies vs. PS

- is one of the formalisms clearly better than the other one?  
**No.**
  - dependencies:  $\oplus$  account for relational phenomena,  $\oplus$  simple
  - phrase-structure:  $\oplus$  account for constituency,  $\oplus$  easy chunking
- can we perform transformation from one of the formalism to the other one a vice versa? **Technically yes, but . . .**
  - It is not a problem to convert the structure between a dependency tree and a PS tree ...
  - ... but it is a problem to transform the information included
- $\Rightarrow$  both of the formalisms are convertible but not mutually equivalent

# Dependency Parsing

- rule-based vs. statistical
- transition-based ( $\rightarrow$  deterministic parsing)
- graph-based ( $\rightarrow$  spanning trees algorithms)
- various other approaches (ILP, PS conversion, . . . )
- very recent advances (vs. long studied PS parsing algorithms)

# Introduction to Dependency parsing

## ■ Motivation

- a. dependency-based syntactic representation seem to be useful in many applications of language technology: machine translation, information extraction
  - transparent encoding of predicate-argument structure
- b. dependency grammar is better suited than phrase structure grammar for language with free or flexible word order
  - analysis of diverse languages within a common framework
- c. leading to the development of accurate syntactic parsers for a number of languages
  - combination with machine learning from syntactically annotated corpora (e.g. treebank)

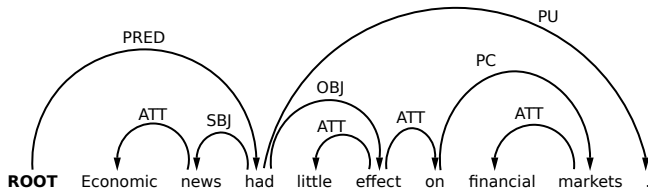
# Introduction to Dependency parsing

## ■ Dependency parsing

“Task of automatically analyzing the dependency structure of a given input sentence”

## ■ Dependency parser

“Task of producing a labeled dependency structure of the kind depicted in the follow figure, where the words of the sentence are connected by typed dependency relations”



# Definitions of dependency graphs and dependency parsing

**Dependency graphs:** syntactic structures over sentences

**Def. 1.:** A sentence is a sequence of tokens denoted by

$$S = w_0 w_1 \dots w_n$$

**Def. 2.:** Let  $R = \{r_1, \dots, r_m\}$  be a finite set of *possible dependency relation types* that can hold between any two words in a sentence. A relation type  $r \in R$  is additionally called an *arc label*.

# Definitions of dependency graphs and dependency parsing

**Dependency graphs:** syntactic structures over sentences

**Def. 3.:** A dependency graph  $G = (V, A)$  is a labeled directed graph, consists of nodes,  $V$ , and arcs,  $A$ , such that for sentence  $S = w_0w_1 \dots w_n$  and label set  $R$  the following holds:

- 1  $V \subseteq \{w_0w_1 \dots w_n\}$
- 2  $A \subseteq V \times R \times V$
- 3 if  $(w_i, r, w_j) \in A$  then  $(w_i, r', w_j) \notin A$  for all  $r' \neq r$



# Approach to dependency parsing

a. **data-driven**

it makes essential use of machine learning from linguistic data in order to parse new sentences

b. **grammar-based**

it relies on a formal grammar, defining a formal language, so that it makes sense to ask whether a given input is in the language defined by the grammar or not.

→ **Data-driven have attracted the most attention in recent years.**

## Data-driven approach

according to the *type of parsing model* adopted,  
*the algorithms used to learn the model from data*  
*the algorithms used to parse new sentences with the model*

a. **transition-based**

start by defining a transition system, or state machine, for mapping a sentence to its dependency graph.

b. **graph-based**

start by defining a space of candidate dependency graphs for a sentence.

# Data-driven approach

## a. transition-based

- **learning problem:** induce a model for predicting the next state transition, given the transition history
- **parsing problem:** construct the optimal transition sequence for the input sentence, given induced model

## b. graph-based

- **learning problem:** induce a model for assigning scores to the candidate dependency graphs for a sentence
- **parsing problem:** find the highest-scoring dependency graph for the input sentence, given induced model

# Transition-based Parsing

- Transition system consists of a set  $C$  of parser configurations and of a set  $D$  of transitions between configurations.
- **Main idea:** a sequence of valid transitions, starting in the *initial configuration* for a given sentence and ending in one of several *terminal configurations*, defines a valid dependency tree for the input sentence.

$$D_{1'm} = d_1(c_1), \dots, d_m(c_m)$$

# Transition-based Parsing

## ■ Definition

Score of  $D_{1'm}$  factors by configuration-transition pairs  $(c_i, d_i)$ :

$$s(D_{1'm}) = \sum_{i=1}^m s(c_i, d_i)$$

## ■ Learning

**Scoring function**  $s(c_i, d_i)$  for  $d_i(c_i) \in D_{1'm}$

## ■ Inference

Search for highest scoring sequence  $D_{1'm}^*$  given  $s(c_i, d_i)$

# Transition-based Parsing

## Inference for transition-based parsing

### ■ Common inference strategies:

- Deterministic [Yamada and Matsumoto 2003, Nivre et al. 2004]
- Beam search [Johansson and Nugues 2006, Titov and Henderson 2007]
- Complexity given by upper bound on transition sequence length

### ■ Transition system

- Projective  $O(n)$  [Yamada and Matsumoto 2003, Nivre 2003]
- Limited non-projective  $O(n)$  [Attardi 2006, Nivre 2007]
- Unrestricted non-projective  $O(n^2)$  [Nivre 2008, Nivre 2009]

# Transition-based Parsing - Nivre algorithm

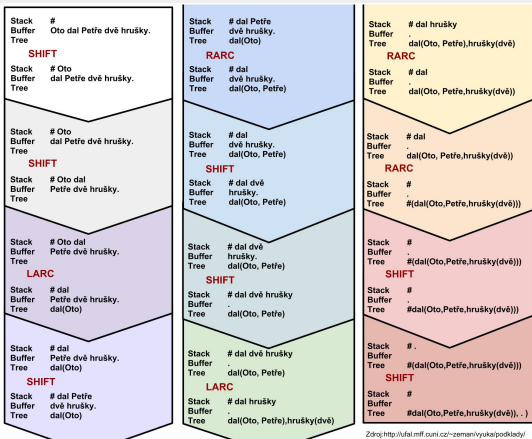
Prechody (*transitions*) z jedného stavu (konfigurácie) do druhého.

**Konfigurácia:**

- Vstupní buffer (slova vo vete zľava doprava)
- Zásobník
- Výstupní strom (slova, závislosti a značky závislosti)

Prechody:

- **Reduce:** uvoľnenie vrchného slovo na zásobníku
- **Shift:** presunutie slovo z bufferu na zásobník
- **Left-arc (LARC):** ľavá závislosť medzi dvoma hornými slovami v zásobníku
- **Right-arc (RARC):** pravá závislosť medzi dvoma hornými slovami v zásobníku



Zdroj: <http://lfa1.mff.cuni.cz/~zemani/vyuka/podklady/2017/2017-03-26-2017-03-27.pdf>

# Transition-based Parsing

## Learning for transition-based parsing

### ■ Typical scoring function:

- $s(c_i, d_i) = w \cdot f(c_i, d_i)$  where  $f(c_i, d_i)$  is a feature vector over configuration  $c_i$  and transition  $d_i$  and  $w$  is a weight vector  
[ $w_i =$  weight of feature  $f_i(c_i, d_i)$ ]

### ■ Transition system

- Projective  $O(n)$  [Yamada and Matsumoto 2003, Nivre 2003]
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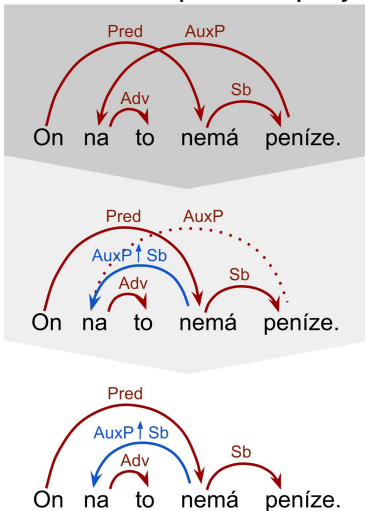
### ■ Problem

- Learning is local but features are based on the global history



# Transition-based Parsing

Projectivization to pseudo-projectivity:



# Graph-based Parsing

- For an input sentence  $S$  we define a graph  $G_S = (V_S, A_S)$  where  
 $V_S = \{w_0, w_1, \dots, w_n\}$  and  
 $A_S = \{(w_i, w_j, l) \mid w_i, w_j \in V \text{ and } l \in L\}$

- Score of a dependency tree  $T$  factors by subgraphs  $G_S, \dots, G_i$ :

$$s(T) = \sum_{i=1}^m s(G_i)$$

- Learning: **Scoring function**  $s(G_i)$  for a subgraph  $G_i \in T$
- Inference: Search for maximum spanning tree scoring sequence  $T^*$  of  $G_S$  given  $s(G_i)$

# Graph-based Parsing

## Learning graph-based models

### ■ Typical scoring function:

- $s(G_i) = w \cdot f(G_i)$  where  $f(G_i)$  is a high-dimensional feature vector over subgraphs and  $w$  is a weight vector  
[ $w_j =$  weight of feature  $f_j(G_i)$ ]

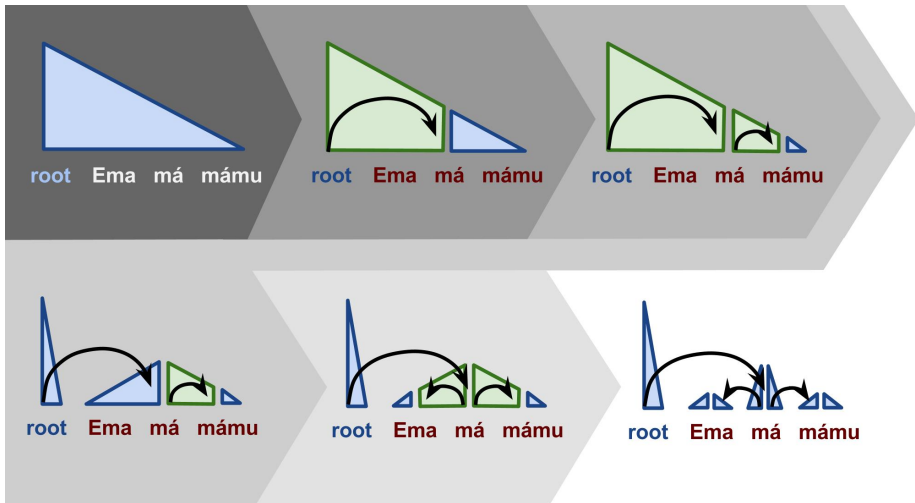
### ■ Structured learning [McDonald et al. 2005a, Smith and Johnson 2007]:

- Learn weights that maximize the score of the correct dependency tree for every sentence in the training set

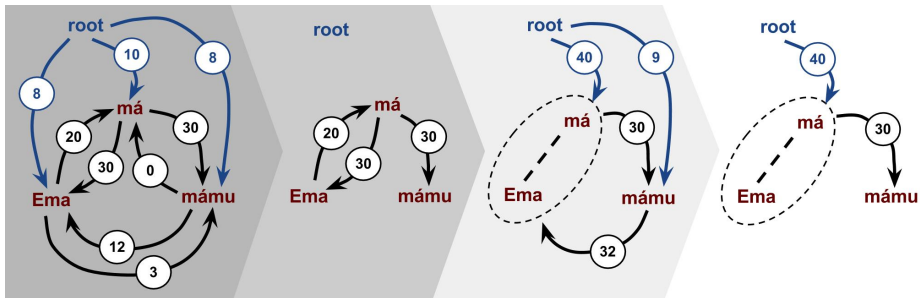
### ■ Problem

- Learning is global (trees) but features are local (subgraphs)

# Graph-based Parsing - Eisner algorithm



# Graph-based Parsing - Chu-Liu-Edmonds algorithm



# Grammar-based approach

## a. **context-free dependency parsing**

exploits a mapping from dependency structures to CFG structure representations and reuses parsing algorithms originally developed for CFG → chart parsing algorithms

## b. **constraint-based dependency parsing**

- parsing viewed as a constraint satisfaction problem
- grammar defined as a set of constraints on well-formed dependency graphs
- finding a dependency graph for a sentence that satisfies all the constraints of the grammar (having the best score)

# Grammar-based approach

## a. context-free dependency parsing

**Advantage:** Well-studied parsing algorithms such as CKY, Earley's algorithm can be used for dependency parsing as well.

→ need to convert dependency grammars into efficiently parsable context-free grammars; (e.g. bilingal CFG, Eisner and Smith, 2005)

## b. constraint-based dependency parsing

defines the problem as constraint satisfaction

- Weighted constraint dependency grammar (WCDG, Foth and Menzel, 2005)
- Transformation-based CDG

# Conclusions

- 1 Dependency syntax vs. constituency (phrase-structure) syntax
- 2 Non-projectivity
- 3 Graph-based and Transition-based methods