

Syntactic analysis of natural languages

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PA153 Natural Language Processing

Syntactic analysis

■ What?

- reveal the structure of the sentence
- relationships among words, phrases

■ Why?

- basis for more informed language analysis
- more than keywords
- semantic and logical analysis, question answering, ...
- applications can benefit from syntactic information
- red brick house vs. red house brick vs. brick house red

■ Origins

- Noam Chomsky: Syntactic structures (1957)
- theory of formal languages

Automatic syntactic analysis of natural languages

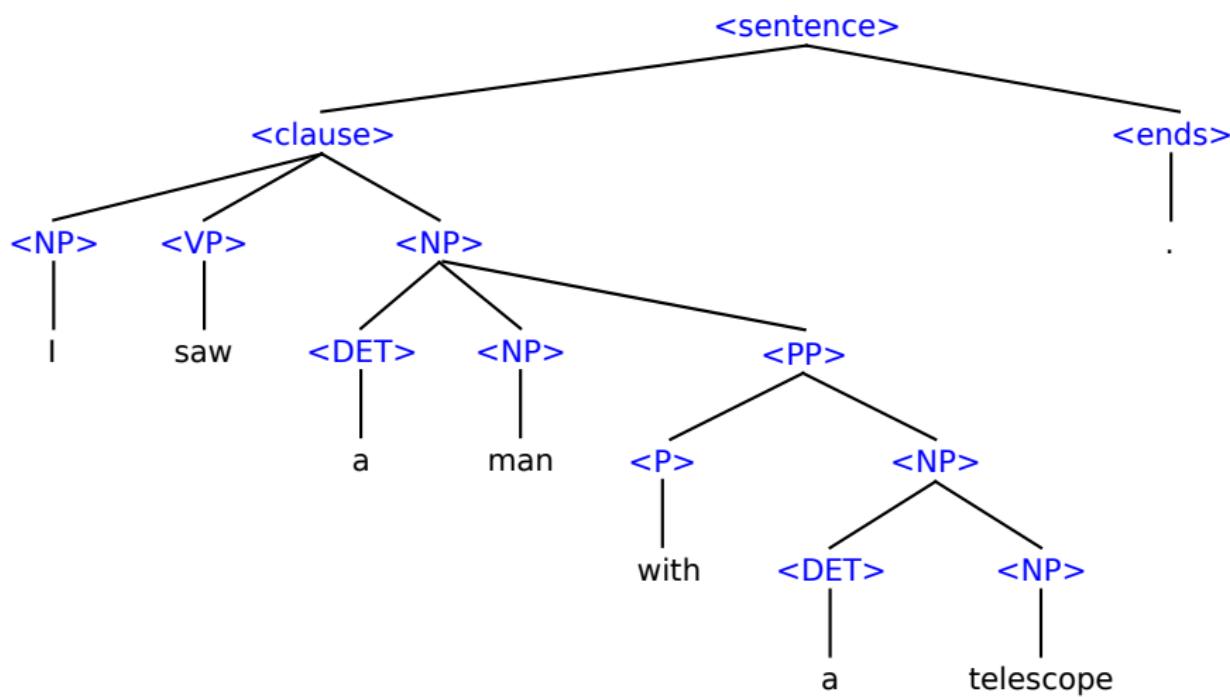
■ Preprocessing

- sentence boundary detection
- word segmentation
- morphological analysis and disambiguation
- (named entity MWE recognition, lexical semantics, ...)
- compatibility issues

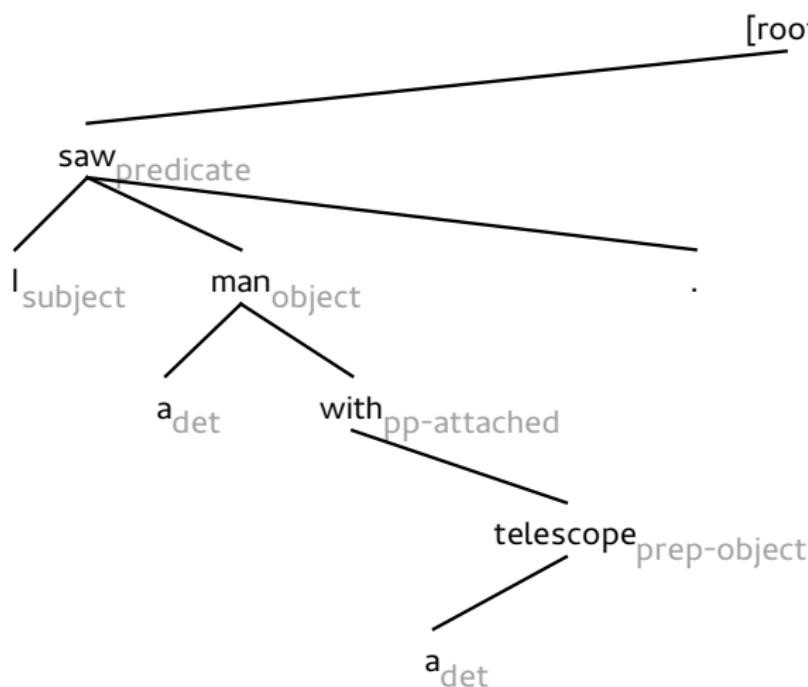
■ Encoding

- phrase structure formalism
- dependency formalism
- partial analysis
- advanced – CCG, LFG, HPSG, TAG

Phrase structure formalism – example



Dependency formalism – example



Dependency vs. phrase-structure

■ Non-projectivity

- disconnected phrases
- not natural in the phrase structure notation
- 20% of Czech sentences are reported to contain a non-projective dependency

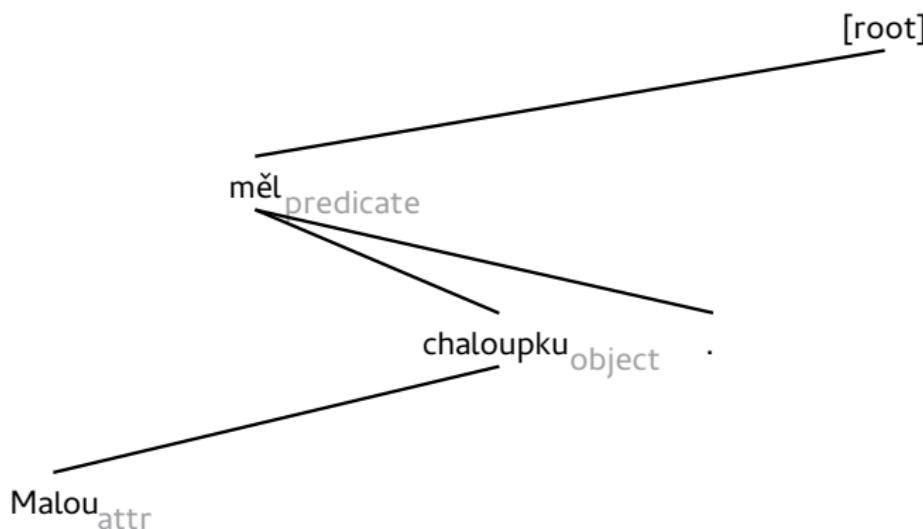
■ Phrase structure – more fine-grained analysis

- (new (queen of beauty))
- (new generation)(of fighters)

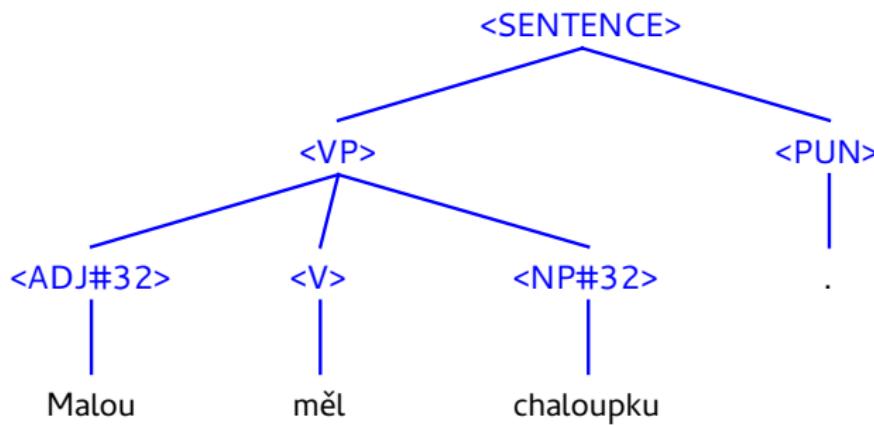
■ Coordinations and other “flat” phenomena

- not natural in the dependency notation
- problem for dependency analysis

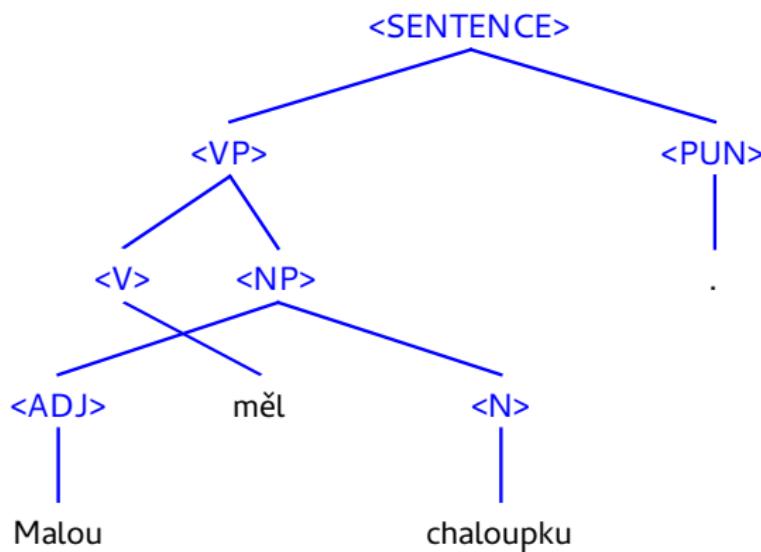
Non-projectivity – example



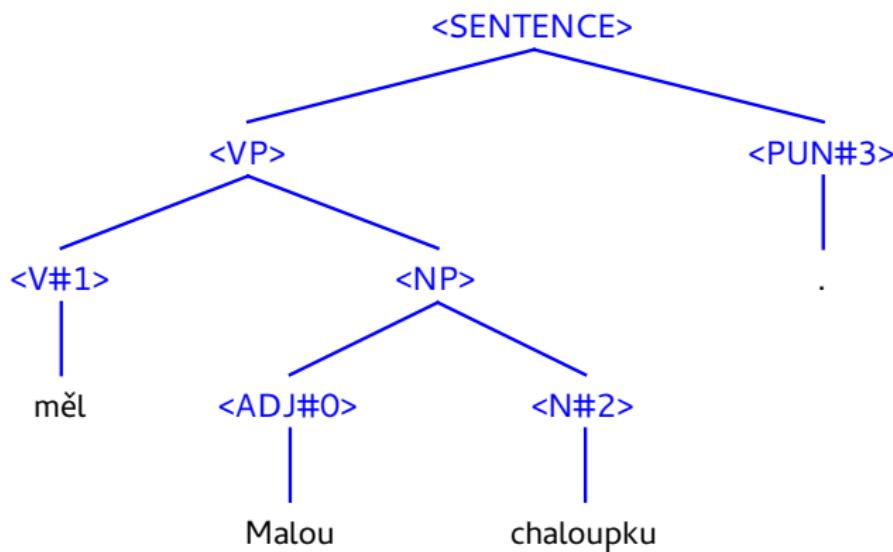
Non-projectivity in phrase structure formalism



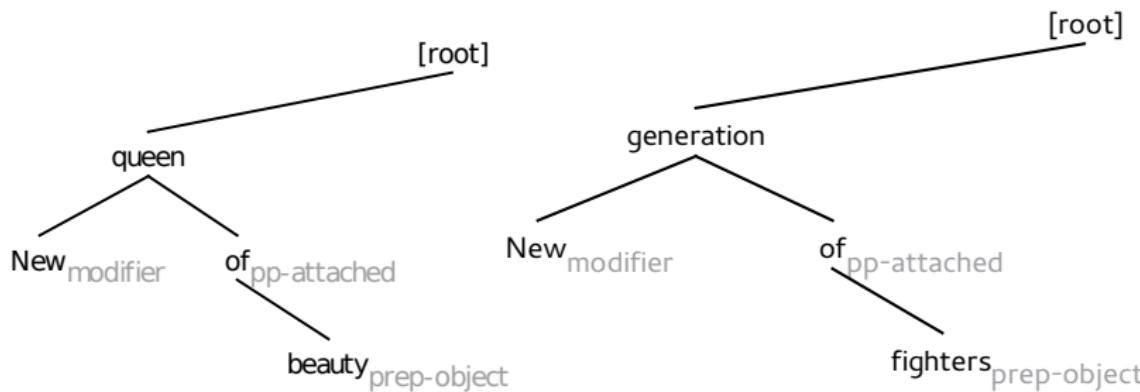
Non-projectivity in phrase structure formalism



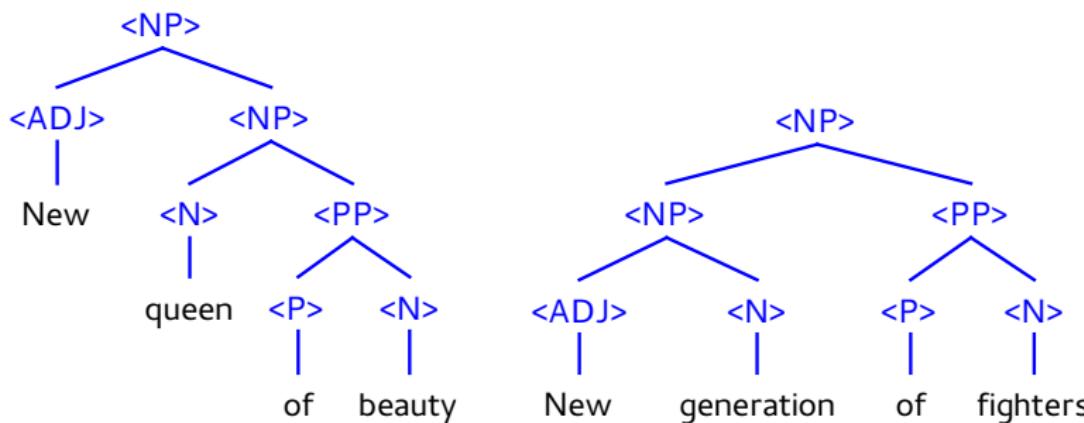
Non-projectivity in phrase structure formalism



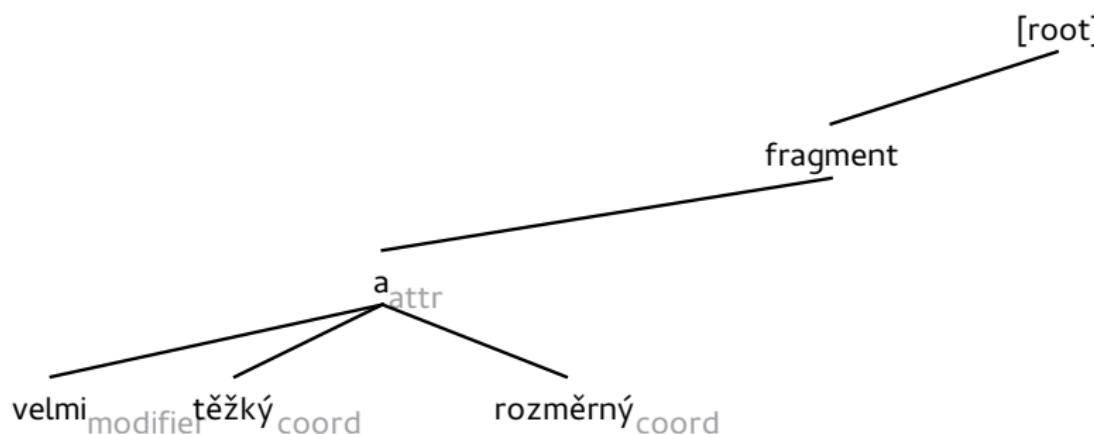
Phrase structure expressivity



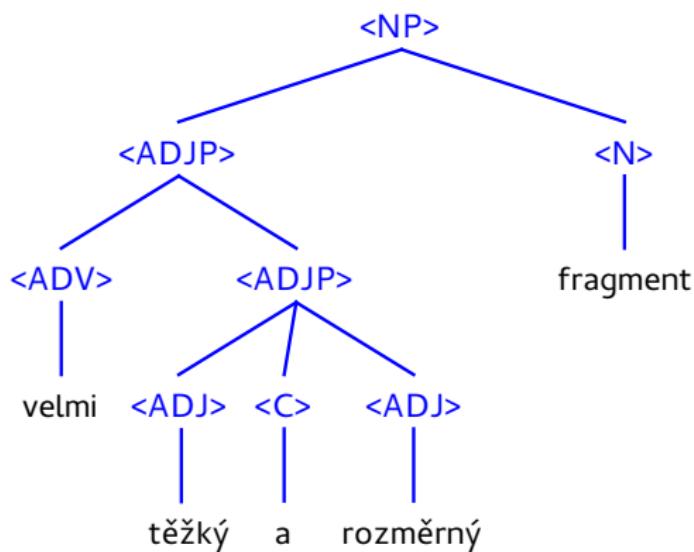
Phrase structure expressivity



Coordinations – dependency structure



Coordinations – phrase structure



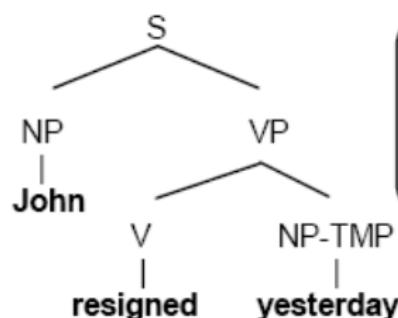
CCG: Combinatory Categorial Grammar

the : NP/N *dog* : N *John* : NP *bit* : (S\NP)/NP

$$\frac{\frac{\frac{the}{NP/N} \quad \frac{dog}{N}}{NP} > \quad \frac{\frac{bit}{(S\setminus NP)/NP} \quad \frac{John}{NP}}{S\setminus NP} >}{S} <$$

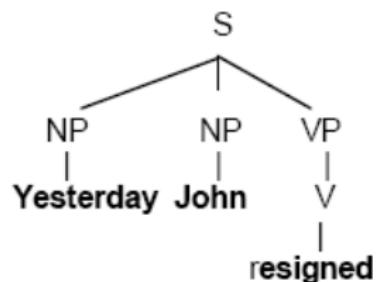
LFG: Lexical Functional Grammar

C-structure:



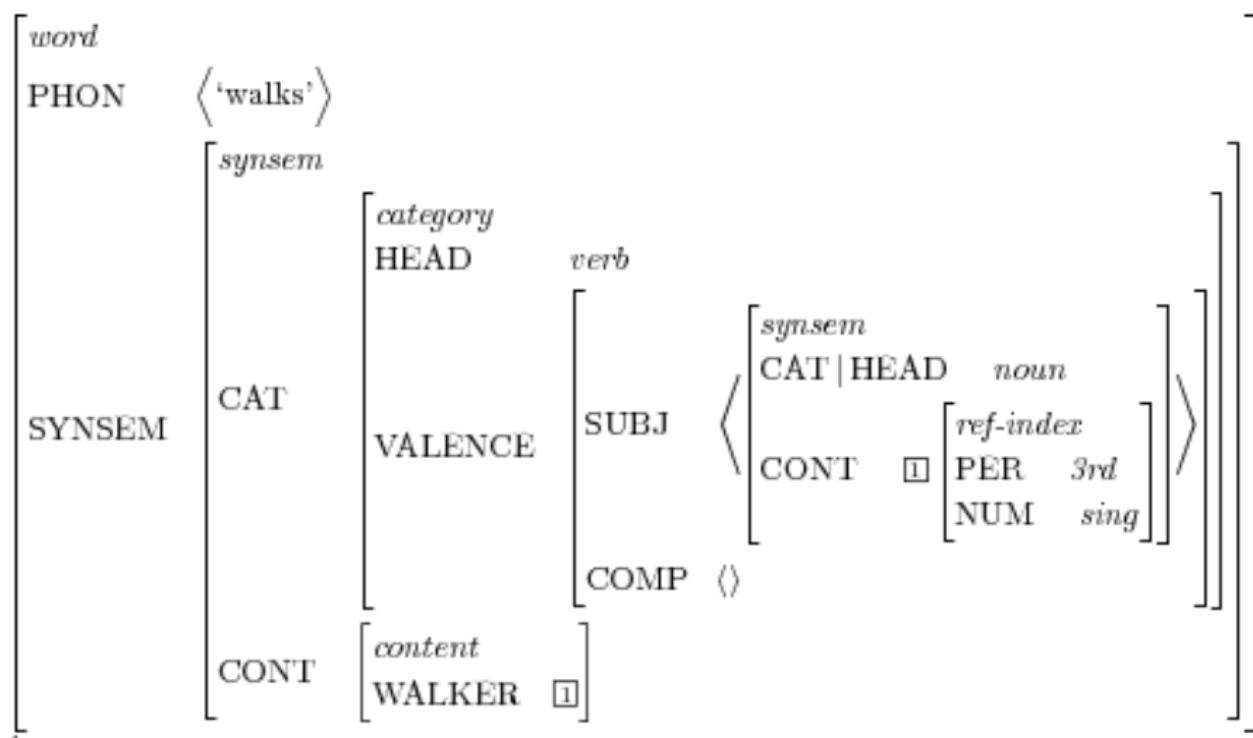
F-structure:

SUBJ	PRED john
NUM	sg
PERS	3
PRED	resign
TENSE	past
ADJ	{[PRED yesterday]}

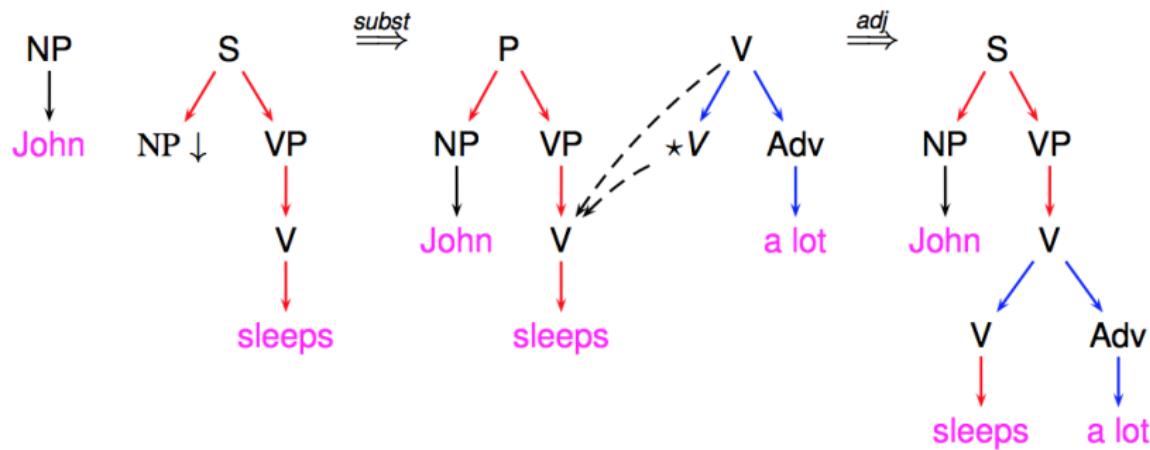


SUBJ	PRED john
NUM	sg
PERS	3
PRED	resign
TENSE	past
ADJ	{[PRED yesterday]}

HPSG: Head-driven Phrase Structure Grammar



TAG: Tree Adjoining Grammar



Parsing methods

■ Rule-based

- RASP, synt, SET, Žabokrtský, Dis/VaDis
- manually created grammar
- CFG (CKY parser, chart parser), dependency grammar, Prolog DCG, ...

■ Statistical

- MaltParser, MST Parser, Stanford parser, ...
- grammars created from annotated data by statistical methods
- direct guessing the tree shape

Parsing evaluation

■ Treebanks

- corpora manually annotated for syntactic structure
- Penn Treebank, Prague Dependency Treebank (PDT)

■ Tree similarity metrics

- PARSEVAL: precision, recall, F-score over phrases
- Leaf-ancestor assessment: edit distance over root-leaf paths
- dependency precision
- labelled or unlabelled
- best results: 85–93 percent

Problems

■ Central problem

- massive ambiguity
- “I saw a man with a telescope”
- “A plane fell into a field next to a forest.”
- problems with evaluation

■ Is the task well-defined?

- inter-annotator agreement rarely reported
- in case of PDT around 90%
- Sampson showed that above 95% is unreachable
- → current parsers are very good
- however, rather low usage in applications

Problems (II)

■ Low usage

- compared to e.g. morphological tagging
- no use in Google, Seznam, Facebook, ...
- Wikipedia page for information extraction does not even mention parsing or syntax
- neither does a Czech question answering system (Konopík, Rohlík)
- ACL anthology: 7,232 matches for word “parser”, 133 matches for using parsers (Jakubíček)

■ Are the results useless?

Problems (III)

■ Application-sparse output

- trees do not provide all the information needed
- but at the same time they do contain noise

■ Application-free evaluation

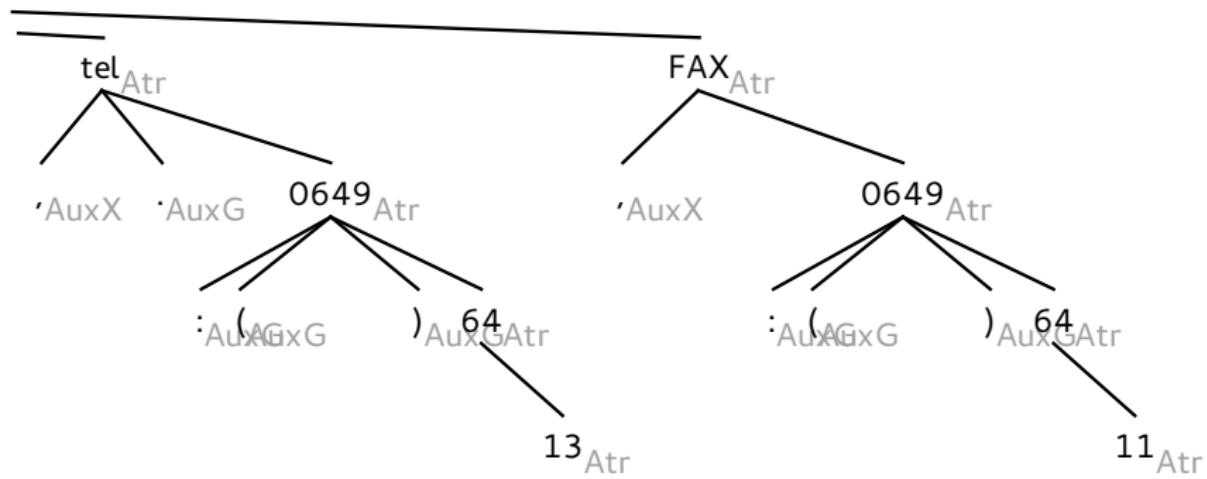
- tree similarity metrics do not correlate well with accuracy of the end applications
- as illustrated by Myiao, Google research, our collocation extraction research

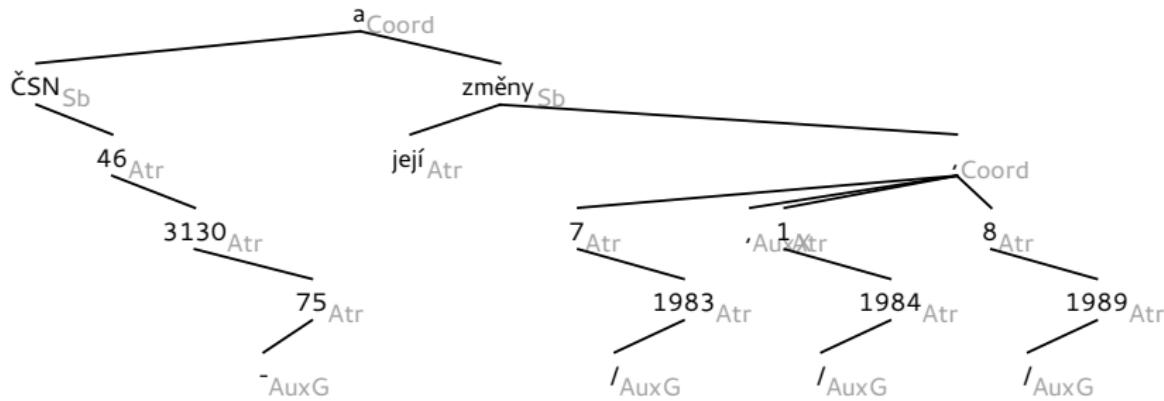
■ Technical aspects

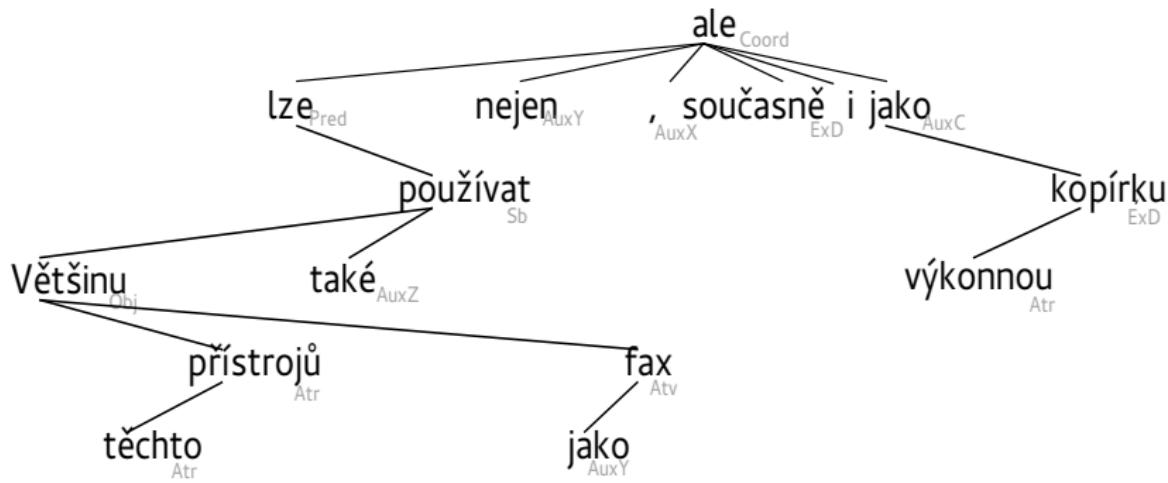
- parsers hard to run, output not readable

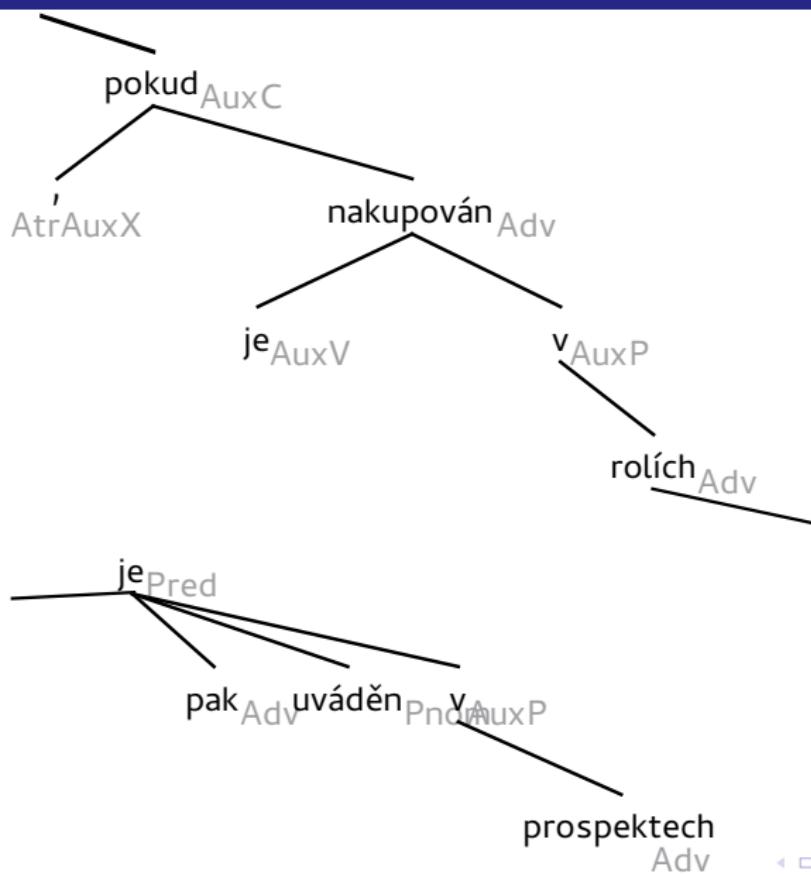
Treebank problems

- Apart from evaluation problems, treebanks are
 - expensive
 - old
 - domain-specific
 - unambiguous
- Treebank formalisms enforce
 - annotation manuals containing hundreds of pages
 - senseless annotations and garbage









Proposed solution: You aren't gonna need it

■ Rapid application development

- „worse is better”
- „keep it simple stupid” (KISS)
- „you aren't gonna need it” (YAGNI)
- completeness, consistency, correctness, simplicity

■ Implications

- start from applications
- strong emphasis on interaction with applications
- do not develop/implement theory that is not immediately needed
- simple, imperfect parsers, possibly task-specific
- rule based first, until we find what we actually need
- **extrinsic evaluations**

Sketch grammar: A shallow approach to syntax

■ Designed for collocation extraction

- Kilgarriff and Rychlý, The Sketch Engine
- based on Corpus Query Language
- results of queries scored statistically
- → pragmatic partial syntactic analysis

■ Extensions

- multi-word sketches
- bilingual word sketches
- terminology extraction
- bilingual terminology extraction

Word Sketch – original

goal

<u>object_of</u>	<u>58924</u>	<u>3.0</u>	<u>subject_of</u>	<u>25451</u>	<u>2.3</u>	<u>modifier</u>
score	<u>8390</u>	11.18	score	<u>903</u>	8.45	
achieve	<u>9422</u>	9.70				
concede	<u>141</u>	9.37	concede	<u>204</u>	7.5	
accomplish	<u>585</u>	7.9	gape	<u>76</u>	6.5	
reach	<u>1924</u>	7.57	kick	<u>76</u>	5.27	
net	<u>337</u>	7.4	orientate	<u>34</u>	5.03	
pursue	<u>648</u>	7.35	rule	<u>61</u>	5.02	
grab	<u>406</u>	7.33	come	<u>1316</u>	4.96	
attain	<u>400</u>	7.32	cap	<u>20</u>	4.32	
pull	<u>501</u>	6.69	beat	<u>53</u>	4.18	

choice is the key reform to **achieve** this **goal**, is that s
you are going to do the tasks to **achieve** these **goals**. For exa
various recommendations on how to **achieve** this **goal**. The loc
in Union, and help ensure this work **achieves** its **goals**. To help
ment departments, in a fun environment to **achieve** a **goal** for charit
term
strong opposition of the old spiritual forces could the **goal** be **achie
ve**.
actors may intend to use IO tools to **achieve** specific **goals**. Recent
winning
more environmentally friendly ways of **achieving** target **goals**. In the ci
primaria
reas of developing countries. **Achieving** these avowed **goals** will rema
seconding
that local solutions are key to **achieving** global **goals** à should
strateg
collective resources to identify and **achieve** system-wide **goals**. One ad
committ
e for plugging particular gaps or **achieving** some local **goals**, for getti
oment of an interoperable Federal PKI. To **achieve** the **goal** of an inte
realisti
achiev
t. an outcome or a clear confirmation that the learning **goal** was **ach
ieve**.

Sketch grammar example

*DUAL

=subject/subject_of

```
2: [tag="N.*"]  [tag="RB.?"]{0,3}  [lemma="be"]?  
          [tag="RB.?"]{0,2}  1: ["V.[^N]?"]
```

Terminology extraction

Term	Frequency	Freq/mill	Score
carbon dioxide	<u>373</u>	3864.3	37.5
global warming	<u>317</u>	3284.1	30.8
water vapor	<u>71</u>	735.6	8.3
greenhouse effect	<u>69</u>	714.8	8.1
greenhouse gas	<u>71</u>	735.6	8.0
climate change	<u>78</u>	808.1	7.6
industrial ecology	<u>27</u>	279.7	3.8
fossil fuel	<u>26</u>	269.4	3.6
surface temperature	<u>20</u>	207.2	3.1
carbon cycle	<u>19</u>	196.8	3.0

Sketch grammar for terminology extraction

=terms

*COLLOC "%(2.1c)_%(1.1c)"

2: [tag=="NN" | tag=="JJ" | tag=="VVG"] 1: [tag=="NN"]

*COLLOC "%(3.1c)_%(2.1c)_%(1.1c)"

3: [tag=="NN" | tag=="JJ" | tag=="VVG"]

2: [tag=="NN" | tag=="JJ" | tag=="VVG"]

1: [tag=="NN"]

SET – a light-weight parsing system

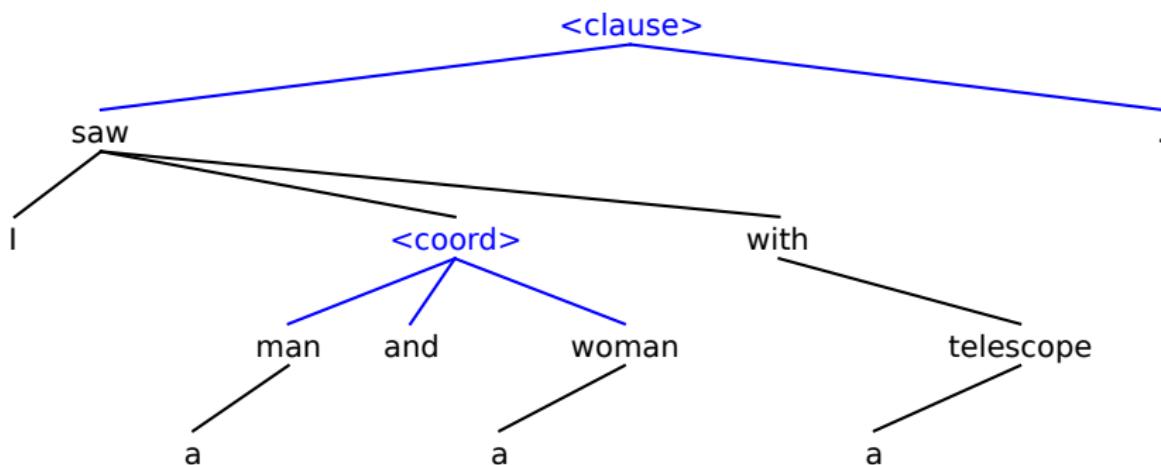
- Hybrid trees

- combination of dependency and phrase structure formalisms
- readability, natural analysis

- Pattern matching grammar

- similar to CQL
- manually created and ranked rules
- rules → matches → sorting → best tree

Hybrid tree



SET rule example

```
TMPL: (tag k5) ... $AND ... (tag k5)
      MARK 0 2 4 <coord> PROB 500 HEAD 2
$AND(word): , a ani nebo
```

Synt – a traditional CFG+ parser

- CFG backbone + contextual actions

- manually created CFG grammars for Czech, Slovak, English
- statistical ranking of rules
- chart parser + extensions

Conclusions

- There are many ways to approach syntactic analysis
 - none of them became dominant in practice (yet?)
- Basic formalisms
 - dependencies
 - phrase structure
- Manual as well as statistical approaches

Links

www.diotavelli.net/people/void/demos/cky.html
en.wikipedia.org/wiki/Definite_clause_grammar
en.wikipedia.org/wiki/Combinatory_categorical_grammar
en.wikipedia.org/wiki/Head-driven_phrase_structure_grammar
nlp.fi.muni.cz/projekty/wwwsynt
nlp.fi.muni.cz/projekty/wwwsynt/query.cgi
nlp.fi.muni.cz/trac/set
nlp.fi.muni.cz/projekty/set/wwwset.cgi/first_page
ufal.mff.cuni.cz/pdt2.0/index-cz.html