

Syntactic Analysis

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Operational Programme Research,
Development and Education

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Faculty of Mathematics and Physics
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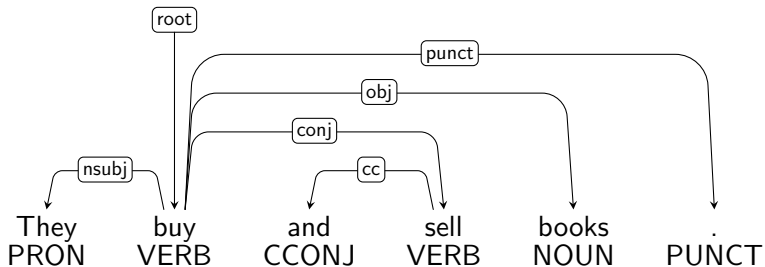


unless otherwise stated

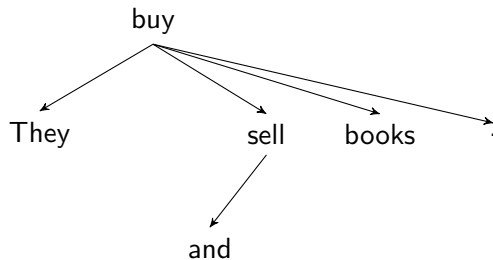
Syntactic Annotation

ID	FORM	LEMMA	POS	FEATS	HEAD	DEPREL
1	They	they	PRON	Case...	2	nsubj
2	buy	buy	VERB	Mood...	0	root
3	and	and	CCONJ	_	4	cc
4	sell	sell	VERB	Mood...	2	conj
5	books	book	NOUN	Number...	2	obj
6	.	.	PUNCT	_	2	punct

Dependency Tree



Dependency Tree



They buy and sell books.

Syntactic Annotation

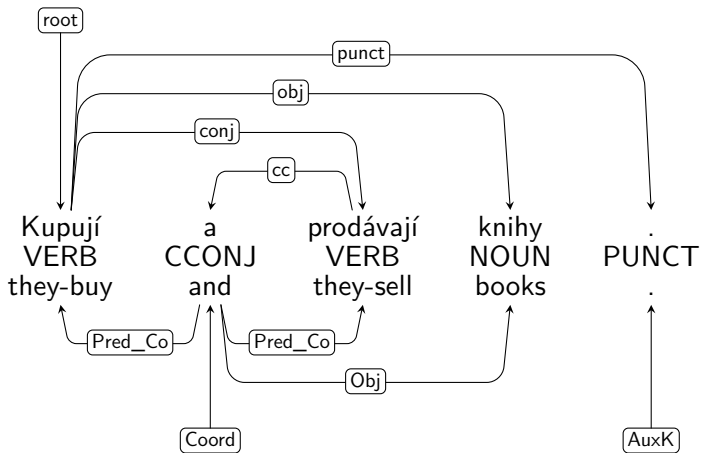
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2	a	a	CCONJ	_	3	cc
3	prodávají	prodávat	VERB	Mood...	1	conj
4	knihy	kniha	NOUN	Case...	1	obj
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ID	FORM	LEMMA	XPOS	HEAD	DEPREL
1	Kupují	kupovat	VB-P---3P-AA---	2	Pred_Co
2	a	a	J^-----	0	Coord
3	prodávají	prodávat	VB-P---3P-AA---	2	Pred_Co
4	knihy	kniha	NNFP4-----A----	2	Obj
5	.	.	Z:-----	0	AuxK

Dependency Tree



Constituents vs. Dependencies

What Is (Surface) Syntax?

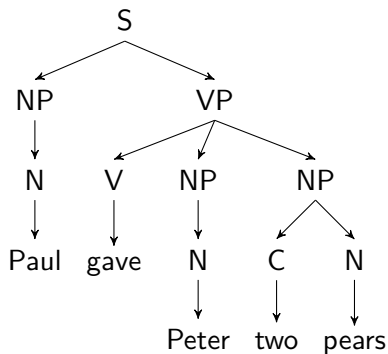
- Relations between sentence parts
- Sentence part = **token** (word, number, punctuation)
 - Advantages:
 - Token is easily recognizable
 - Unit of previous (morphological) level of processing
 - We do not restore **elided** (missing) constituents at this level
 - Drawbacks:
 - We must define relations for function words and punctuation
- Purpose:
 - Understand how meaning of words combines into meaning of sentence
 - Some theories also want to show how the sentence was **generated**

Syntactic Structure

- Different shapes in different theories
- Typically a hierarchical structure – tree
 - Phrasal (constituent) tree, parse tree
 - Dependency tree

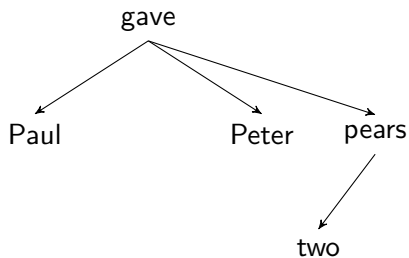
Constituent Tree

(S (NP (N Paul)) (VP (V gave) (NP (N Peter)) (NP (C two) (N pears))))



Dependency Tree

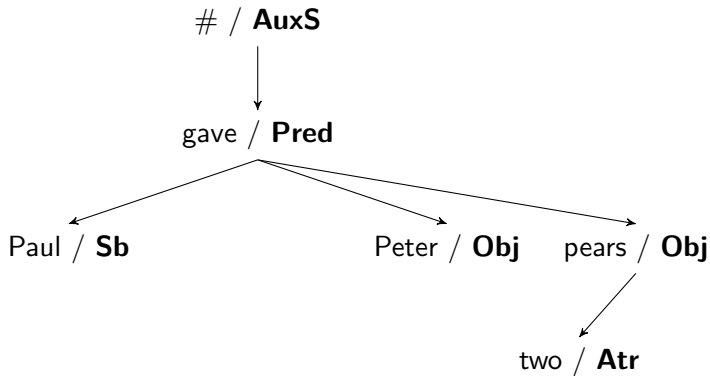
[gave,2] ([Paul,1], [Peter,3], [pears,5] ([two,4]))



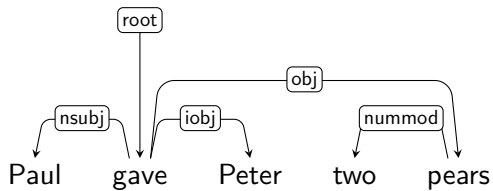
Paul gave Peter two pears.

Dependency Tree with Labels

[#,0] ([gave,2] ([Paul,1], [Peter,3], [pears,5] ([two,4])), [.,6])



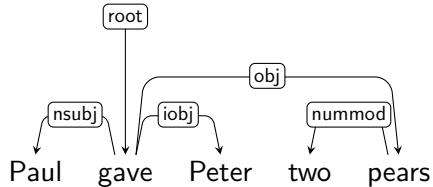
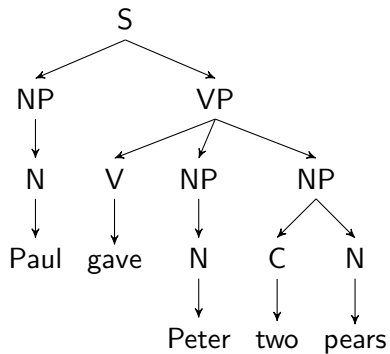
Dependency Tree with Labels



Constituents vs. Dependencies

- The two models are interconnected
- Sentence divided to **phrases** (constituents)
 - Recursive: phrases divided to smaller phrases
 - The smallest phrases are words
- There are **dependencies** (relations) between words (constituents)
 - **Head** of phrase = governing node, **parent** node
 - The other nodes are dependent nodes, **children** of the head

Phrase vs. Dependency Trees



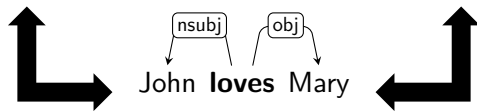
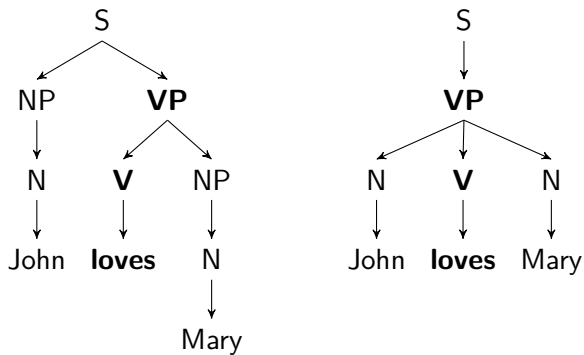
Phrase vs. Dependency Trees

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 - May not mark the **function** of the constituent in the superordinate constituent

Phrase vs. Dependency Trees

- Phrase trees
 - Usually do not mark the **head**
 - May not mark the **function** of the constituent in the superordinate constituent
- Dependency trees
 - Do not show **nonterminals** (phrase types)
 - Nor any other phrase-level features
 - Do not show “how the sentence is generated” (order, recursion, **proximity** of constituents)

Example



Phrases, Their Types and Their Heads

Phrase Replaceability

- A phrase can be replaced by another phrase of the same type
- The sentence stays grammatical
- Specifically, a phrase can be replaced by its head
 - This is related to the generation of the sentence

- Phrases x, y, z can be immediate constituents of a larger phrase f only if they are related to each other. This is however a matter of the particular phrase structure grammar.

- Example sentence: *This is the man that I talked about.*
 - The part *man that I* is not a whole noun phrase.
 - Cannot replace it by another noun phrase, e.g., *man*:
 - **This is the man talked about.*

- Sequence of immediate constituents (words or phrases)
- May be discontinuous in some languages
- Phrase types by their main word—**head**
 - Noun phrase: *the new **book** of my grandpa*
 - Adjectival phrase: *brand **new***
 - Adverbial phrase: *very **well***
 - Prepositional phrase: ***in** the classroom* (if preposition is considered the head—somewhat controversial (cannot replace whole phrase by preposition))
 - Verb phrase: *to **catch** a ball*

- A noun or a (substantive) pronoun is the head.
 - **water**
 - *the **book***
 - *new **ideas***
 - *two **millions** of inhabitants*
 - *one small **village***
 - *the greatest price **movement** in one year since the World War II*
 - *operating **system** that, regardless of all efforts by our admin, crashes just too often*
 - **he**
 - **whoever**

- An adjective or a determiner (attributive pronoun) is the head.
- Simple adjectives are very frequent, complex ADJPs are rare.
 - *old*
 - *very old*
 - *really very old*
 - *five times older than the oldest elephant in our zoo*
 - *sure that he will arrive first*

- (Substantive) pronouns: similar behavior as nouns
 - Personal pronouns (*I, you, they, oneself*)
 - Some demonstrative, interrogative, relative and negative pronouns (*who, what, somebody, something, everything, nothing*)
- Attributive pronouns (determiners): similar behavior as adjectives
 - Possessive pronouns in some languages (*my, your, his, whose*)
 - Articles (*the, a, an*)
 - Attributively used demonstrative, interrogative, relative and negative pronouns (*which, some, every, no*)

Numerals and Quantified Noun Phrases

- Slavic languages: not always clear what should be the head: the quantifier (number), or the counted noun?
 - Numeral inherits gender from counted noun.
 - Counted noun gets grammatical number from numeral (or in accord with it).
 - *jeden muž* “one man”, *jedna žena* “one woman”, *jedno dítě* “one child”
 - *dva muži* “two men”, *dvě ženy* “two women”, *dvě děti* “two children”
 - Numeral may govern the case of the counted noun.
 - *pět mužů* “five men” – noun in genitive, numeral in nominative, accusative, or vocative
 - Or both numeral and counted noun have case required by preposition or verb.
 - *pěti ženami* “five women” – instrumental case

- An adverb is the head.
- Simple adverbs are very frequent, complex ADVPs are rare.
 - ***quickly***
 - ***much more***
 - ***how***
 - ***louder than you can imagine***
 - ***yesterday***

Prepositional (Postpositional) Phrase

- Many theories: preposition is the head (it determines the case of the rest of the phrase)
- But we cannot replace the phrase by the preposition alone! (Nor can we replace it by the noun without the preposition.)
- PPs often have functions similar to adverbial phrases or noun phrases.
- Preposition in one language may correspond to case morphology in another language.
 - ***in** the city center*
 - ***in** January*
 - ***in** God*
 - ***around** five o'clock*
 - ***to** a better future*
 - *up **to** a situation where neither of them could back out*
 - *with respect **to** his nonage*

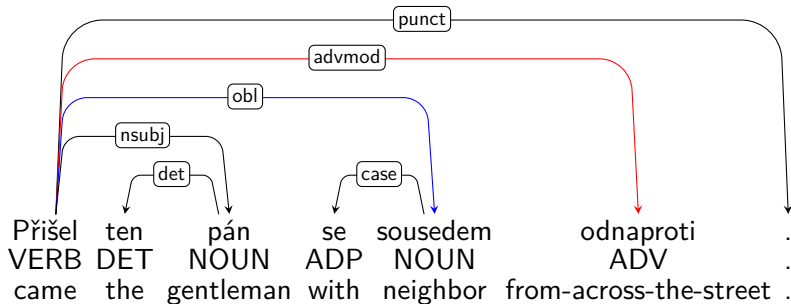
Ambiguity in Attachment of Prepositional Phrases

- Classic English example:
 - *I saw the man with a telescope.*
 - ① “Viděl jsem muže dalekohledem.”
 - ② “Viděl jsem muže s dalekohledem.”

Ambiguity in Attachment of Prepositional Phrases: Czech Example

Přišel ten pán se sousedem odnaproti. “The gentleman came with his neighbor from across the street.”

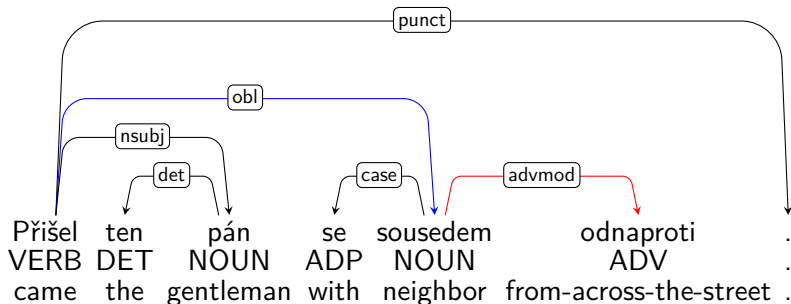
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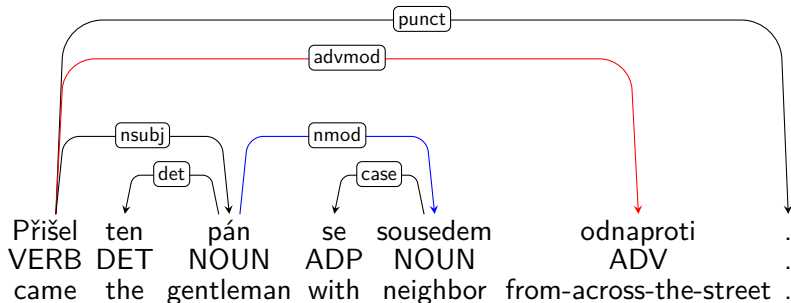
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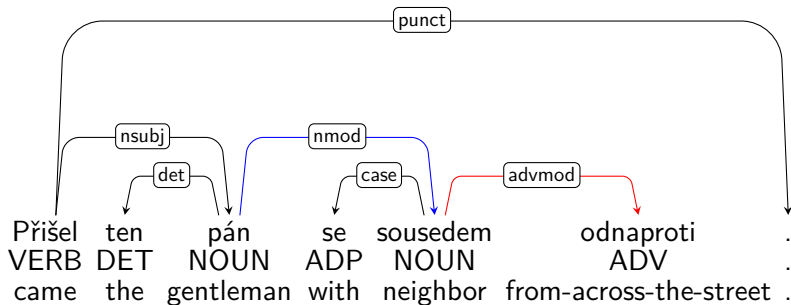
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Prepositional Phrase Attachment Ambiguity: Corpus Example

- *V letech 1991 – 1993 jsem absolvovala kurzy řízení a marketingu na Collège Bart v kanadském Québecu.*
- “In years 1991 – 1993 I attended classes of management and marketing at Collège Bart in Canadian Québec.”

(A Czech sentence from the Prague Dependency Treebank.)

Prepositional Phrase Attachment Ambiguity: Corpus Example

- *In years 1991 – 1993 I attended classes of management and marketing at Collège Bart in Canadian Québec.*
 - *attended at Collège Bart*
 - *classes at Collège Bart*
 - *management and marketing at Collège Bart*
 - *marketing at Collège Bart*
 - *Collège Bart in Québec*
 - *marketing in Québec*
 - *management and marketing in Québec*
 - *classes in Québec*
 - *attended in Québec*

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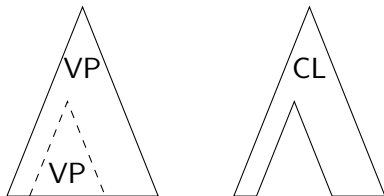
- *In years 1991 – 1993 I attended classes of management and marketing at Collège Bart in Canadian Québec.*
 - *attended (classes (of (management and marketing))) (at Collège Bart)*
 - *attended (classes (of (management and marketing)) (at Collège Bart))*
 - *attended (classes (of (management and marketing) (at Collège Bart)))*
 - *attended (classes (of (management and (marketing (at Collège Bart)))))*
 - *... ((at Collège Bart) (in Québec))*
 - Is Collège Bart in Québec or Québec in Collège Bart?

Prepositional Phrase Attachment Ambiguity: News Example

- *říjnové jednání OSN o klimatických změnách v Kodani* (Události ČT, 27.2.2009)
- “October UNO summit about climatic changes in Copenhagen” (Czech TV news, 2009-02-27)
- Question:
Were there climatic changes in Copenhagen?

- The repertory depends on the rules for analytical verb forms and varies greatly across languages.
 - *it rains*
 - *he could at all sight Mr. President*
 - *why we got wet so much*
 - *Go!*
 - *he has been transported to the hospital on Sunday*
 - *it began to rain*
 - *the law prohibits smoking in this room*
 - *give Mary the beads that we brought from the vacation in Morocco*
 - *the file could not be opened*

- **One predicate** together with its arguments and modifiers, e.g.:
 - *John loves Mary.*
 - *... that you are right.*
- Not recursive \Rightarrow not necessarily the same as a verb phrase (VP).
 - Nested phrases are part of the superordinate phrase.
 - Nested clauses are not parts of the main ('matrix') clause.



- There is **no real head** \Rightarrow difficult to capture in dependency trees.
- The coordinate phrases (conjuncts) are usually of the same type.
 - *chickens, hens, rabbits, cats **and** dogs*
 - *new **or** even newer*
 - *quickly **and** finely*
 - *he came to the conclusion that there is no point in hiding any more, **so** we might hear him here today*
 - *in the house **or** outside*
 - *to **and** from Prague*
 - ***either** now **or** later*
 - *not only on Monday **and** on Wednesday **but** also tomorrow **or** the day after tomorrow*
 - *veni, vidi, vici*

- A phrase omitted from the (surface form of the) sentence although it is present in the underlying meaning (deep structure).
- Often in dialogues: the elided information is known from context.
 - *Whom did you see there? — Peter.* (missing verb)
- Also often in coordination:
 - *Czech and German researchers discussed...* It probably means *Czech researchers and German researchers discussed*. Unlikely that each researcher was Czech and German at the same time.
 - *The Penguins are leading 4:0, while the Colorado Avalanches only 3:2.* (missing verb in the second part)
- Systemic elision of subject pronouns in pro-drop languages (it is marked on the verb).
 - *Sedím.* “(I) sit.”

Dependency Parsing

Dependency Parsing

- Automatically assign a dependency tree to a sentence.
- Machine-learning: manually annotated “gold standard” data needed!
- **Chart** (Eisner, CKY)
 - $O(n^3)$
 - Produces only projective parses
(if x directly depends on y , all words between x and y transitively depend on y)

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- **Transition-based** (shift-reduce)
 - $O(n)$ (fast!)
 - Can be extended to capture nonprojectivity
- **Graph-based** (MST)
 - $O(n^2)$
 - Can produce projective and nonprojective parses

Transition-Based Parsers: Malt

- Nivre et al., *Natural Language Engineering* 2007
- <http://maltparser.org/>
- Based on *transitions* from one configuration to another
- Configuration:
 - Input buffer (words of the sentence, left-to-right)
 - Stack
 - Output tree (words, edges, labels)
- Transitions:
 - **Shift**: move word from buffer to stack
 - **Larc**: connect two topmost stack words (higher is parent)
 - **Rarc**: connect two topmost stack words (lower is parent)

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- Training: decompose the training tree to a sequence of transitions
 - Sometimes more than one possibility
 - Various learning strategies: e.g. create dependencies eagerly, as soon as possible
- The oracle learns based on the features of the configuration
 - E.g. word, lemma, POS, case, number...
 - n^{th} word from the top of the stack
 - k^{th} word remaining in the buffer
 - particular node in output tree part created so far

- Machine learning responsible for training, here the **Support Vector Machines (SVM)**
 - Classifier. Input vectors: values of all features of the current configuration
 - In addition, during training there is the output value, i.e. action identifier (`shift` / `larc` / `rarc`)
 - The trained oracle (SVM) tells the output value during parsing
- Training on the whole PDT may take weeks!
 - Complexity $O(n^2)$ where n is number of training examples
 - Over 3 million training examples can be extracted from PDT
- Parsing comparatively faster (~ 1 sentence / second) and can be parallelized

Example of Malt Parsing

STACK		BUFFER
ROOT		Pavel dal Petrovi dvě hrušky .

Pavel dal Petrovi dvě hrušky .
Pavel gave Petr two pears .

Example of Malt Parsing

STACK		BUFFER
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SHIFT

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LARC


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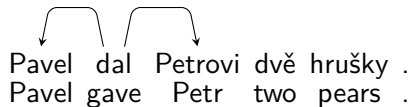
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STACK		BUFFER
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RARC



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SHIFT

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The diagram illustrates dependency arcs between the Czech sentence and its English translation. Two arcs are shown: one connecting 'Pavel' to 'Petr' and another connecting 'dal' to 'gave'.

Example of Malt Parsing

STACK	BUFFER
ROOT dal dvě	hrušky .

SHIFT

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Example of Malt Parsing

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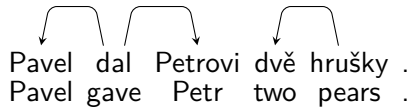
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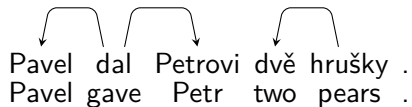
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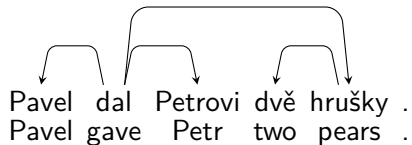
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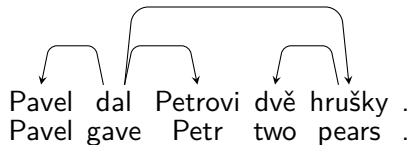
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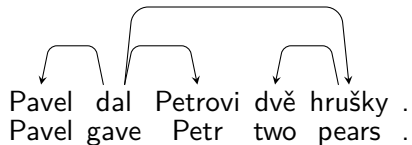
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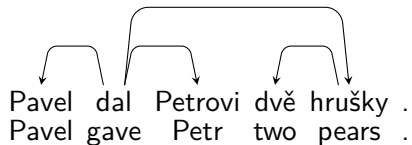
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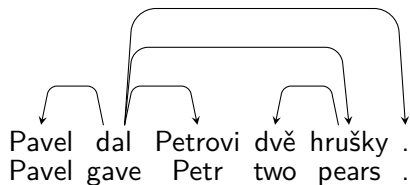
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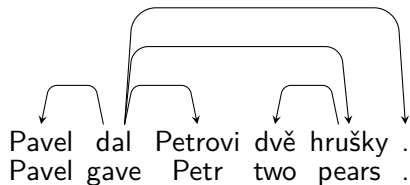
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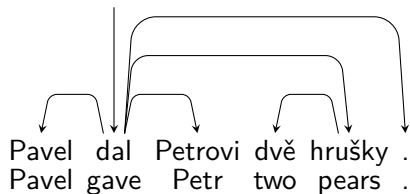
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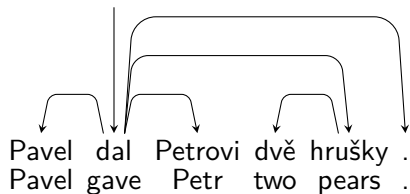
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So Far Only Projective Trees

- It can be proven that the above transition system is
 - **correct**
 - resulting graph is always a tree (connected, cycle-free)

- There are extensions that can produce non-projective trees
- Non-projective constructions occur in natural languages but they are rare

So Far Only Projective Trees

- It can be proven that the above transition system is
 - **correct**
 - resulting graph is always a tree (connected, cycle-free)
 - **complete** for the set of **projective trees**
 - every projective tree can be expressed as a sequence of transitions
- There are extensions that can produce non-projective trees
- Non-projective constructions occur in natural languages but they are rare