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# Transition-based dependency parsing

Syntactic analysis (5LN455)

2014-12-17

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Based on slides from Marco Kuhlmann





# Overview

- Arc-factored dependency parsing
  - Collins' algorithm
  - Eisner's algorithm
- **Transition-based dependency parsing**
  - The arc-standard algorithm**
- **Evaluation of dependency parsers**
- **Projectivity**



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# Transition-based dependency parsing



# Transition-based dependency parsing

- The parser starts in an **initial configuration**.
- At each step, it asks a **guide** to choose between one of several **transitions** (actions) into new configurations.
- Parsing stops if the parser reaches a **terminal configuration**.
- The parser returns the dependency tree associated with the terminal configuration.



# Transition-based dependency parsing

- Eisner's algorithm runs in time  $O(|w|^3)$ .  
This may be too much if a lot of data is involved.
- **Idea:** Design a dumber but really fast algorithm and let the machine learning do the rest.
- Eisner's algorithm searches over many different dependency trees at the same time.
- A transition-based dependency parser only builds *one* tree, in *one* left-to-right sweep over the input.



# Generic parsing algorithm

```
Configuration c = parser.getInitialConfiguration(sentence)

while c is not a terminal configuration do
    Transition t = guide.getNextTransition(c)
    c = c.makeTransition(t)

return c.getGraph()
```



# Variation

Transition-based dependency parsers differ with respect to the configurations and the transitions that they use.



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# The arc-standard algorithm





# The arc-standard algorithm

- The arc-standard algorithm is a simple algorithm for transition-based dependency parsing.
- It is very similar to shift–reduce parsing as it is known for context-free grammars.
- It is implemented in most practical transition-based dependency parsers, including MaltParser.



# Configurations

A **configuration** for a sentence  $w = w_1 \dots w_n$  consists of three components:

- a **buffer** containing words of  $w$
- a **stack** containing words of  $w$
- the **dependency graph** constructed so far



# Configurations

- **Initial configuration:**
  - All words are in the buffer.
  - The stack is empty.
  - The dependency graph is empty.
- **Terminal configuration:**
  - The buffer is empty.
  - The stack contains a single word.



# Possible transitions

- **shift (sh):** push  
the next word in the buffer onto the stack
- **left-arc (la):** add an arc  
from the topmost word on the stack,  $s_1$ ,  
to the second-topmost word,  $s_2$ , and pop  $s_2$
- **right-arc (ra):** add an arc  
from the second-topmost word on the stack,  $s_2$ ,  
to the topmost word,  $s_1$ , and pop  $s_1$



# Configurations and transitions

- **Initial configuration:**  $([], [0, \dots, n], [])$
- **Terminal configuration:**  $([0], [], A)$
- **shift (sh):**  
 $(\sigma, [i|\beta], A) \Rightarrow ([\sigma|i], \beta, A)$
- **left-arc (la):**  
 $([\sigma|i|j], B, A) \Rightarrow ([\sigma|j], B, A \cup \{j, l, i\})$       only if  $i \neq 0$
- **right-arc (ra):**  
 $([\sigma|i|j], B, A) \Rightarrow ([\sigma|i], B, A \cup \{i, l, j\})$

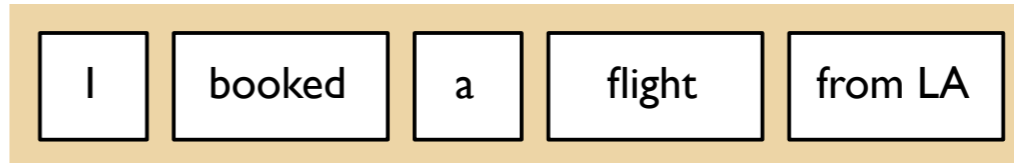


# Example run

Stack



Buffer



I

booked

a

flight

from LA

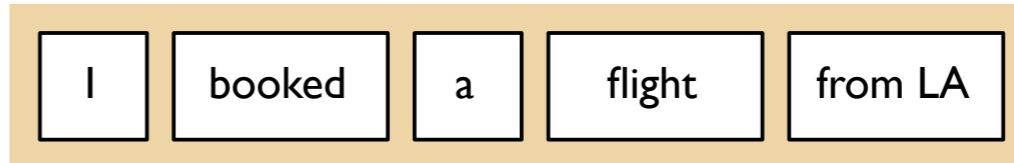


# Example run

Stack



Buffer



I

booked

a

flight

from LA

sh

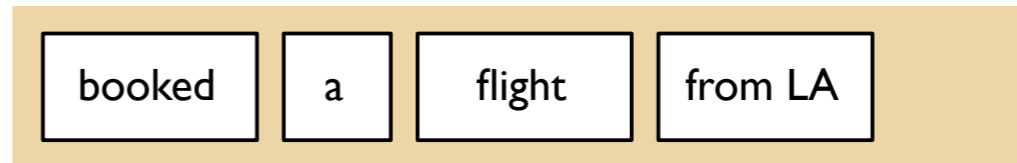


## Example run

Stack



Buffer



I

booked

a

flight

from LA



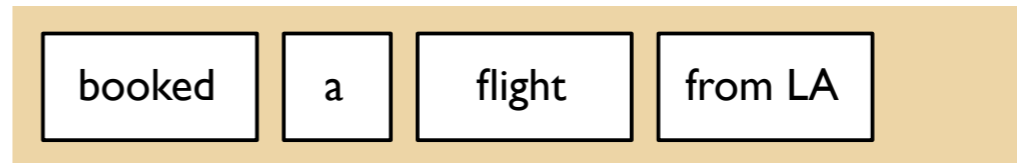


## Example run

Stack



Buffer



I

booked

a

flight

from LA

sh



## Example run

Stack



Buffer



I            booked

a            flight            from LA



## Example run

Stack



Buffer



I

booked

a

flight

from LA

la-subj

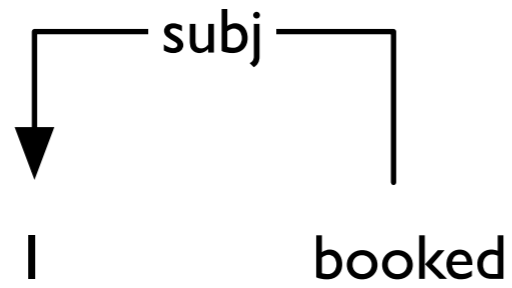


## Example run

Stack



Buffer



a                      flight                      from LA

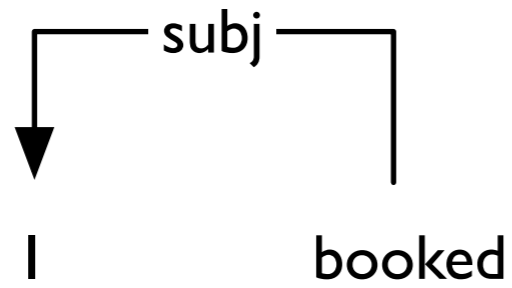


## Example run

Stack



Buffer



a                      flight                      from LA

sh

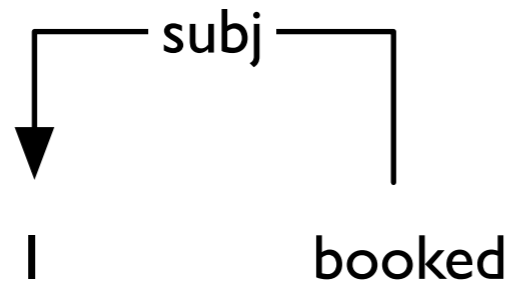


## Example run

Stack



Buffer



a flight from LA

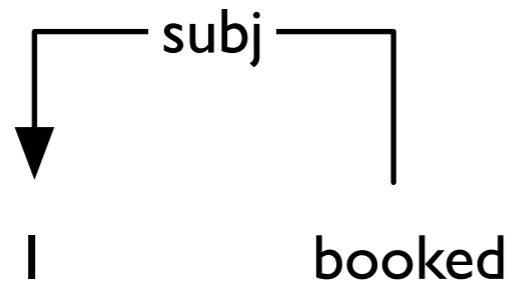


## Example run

Stack



Buffer



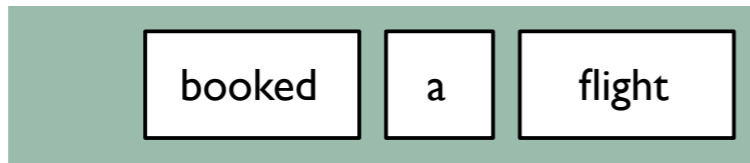
a flight from LA

sh

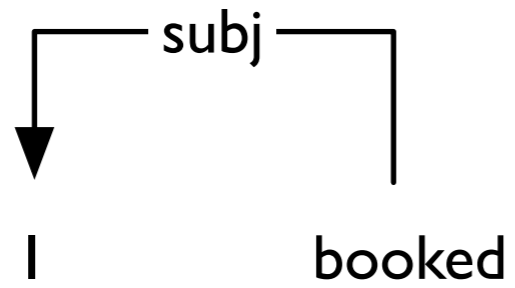


## Example run

Stack



Buffer



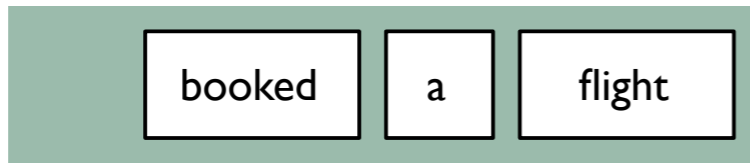
a flight from LA



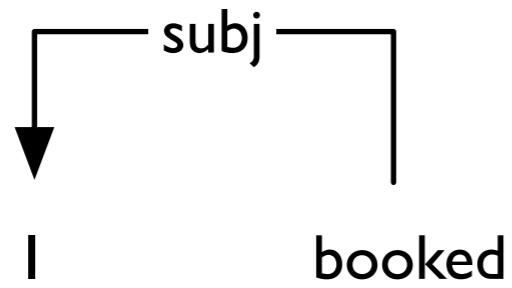


## Example run

Stack



Buffer



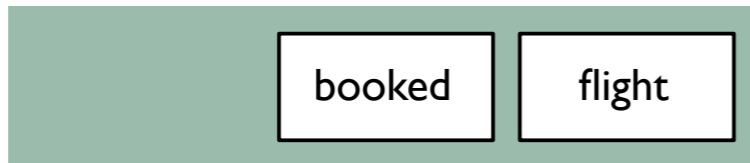
a                      flight                      from LA

la-det

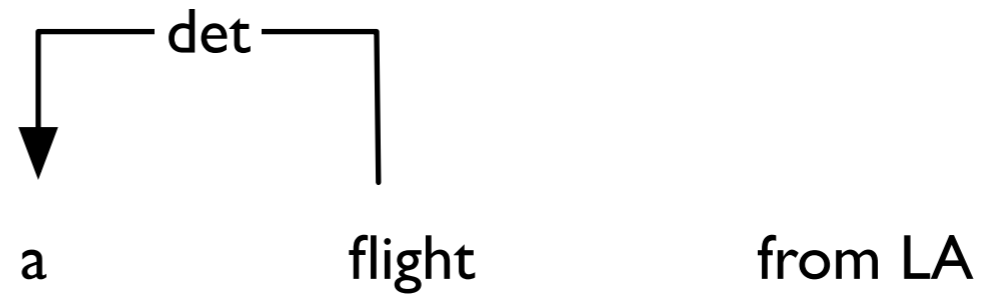
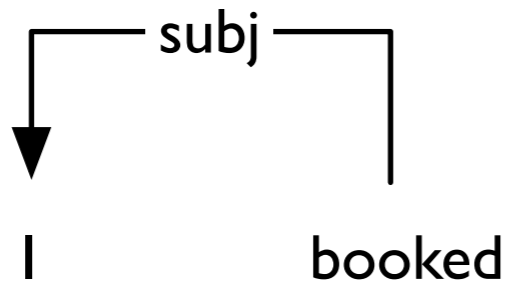


## Example run

Stack



Buffer

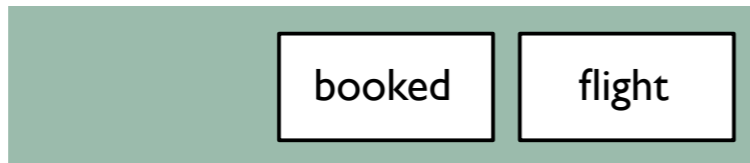


from LA

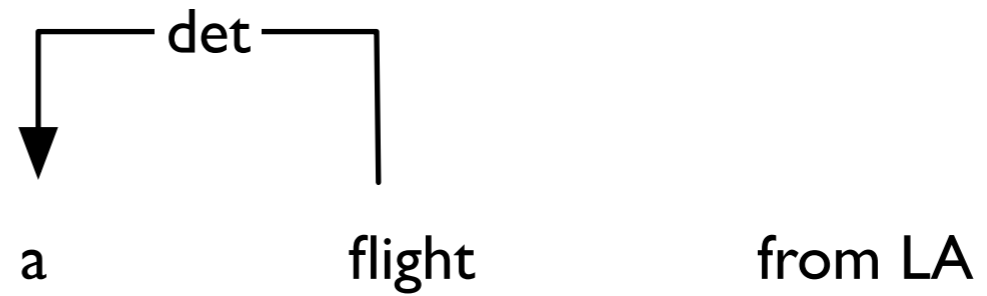
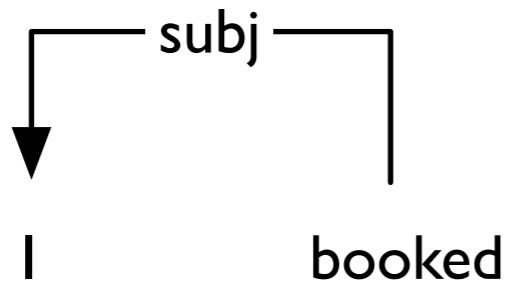


## Example run

Stack



Buffer



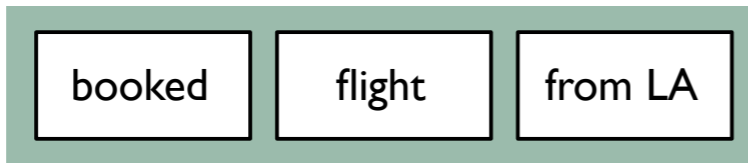
from LA

sh

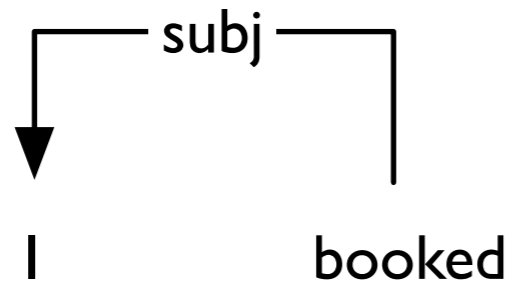


## Example run

Stack



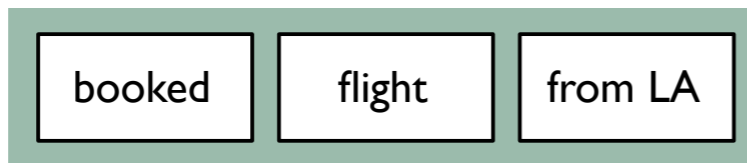
Buffer



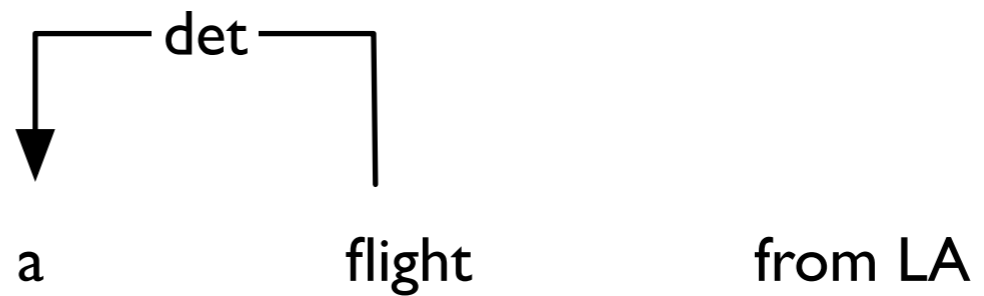
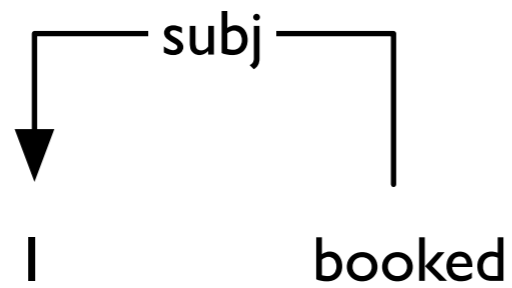


## Example run

Stack



Buffer

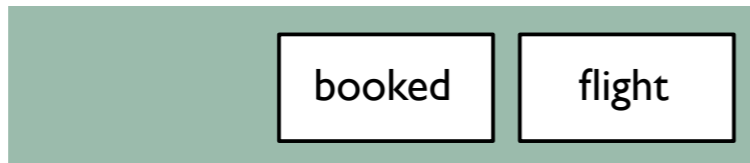


ra-pmod

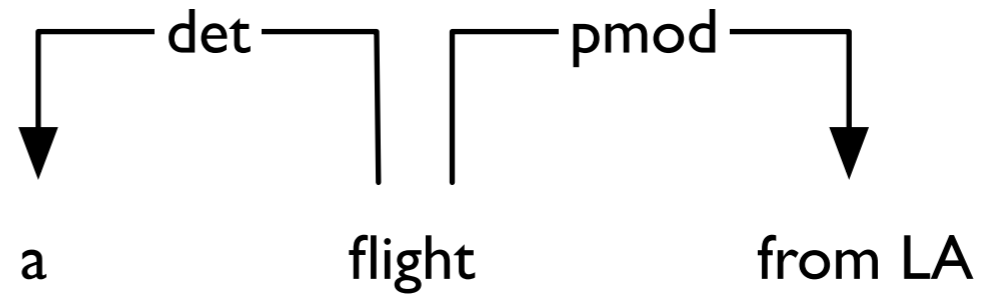
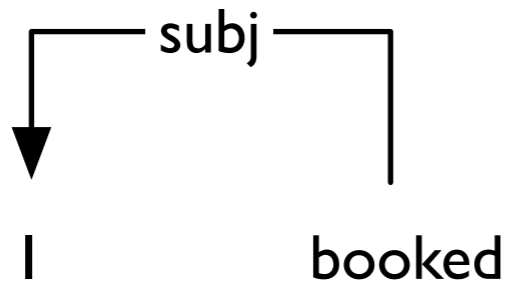


## Example run

Stack



Buffer



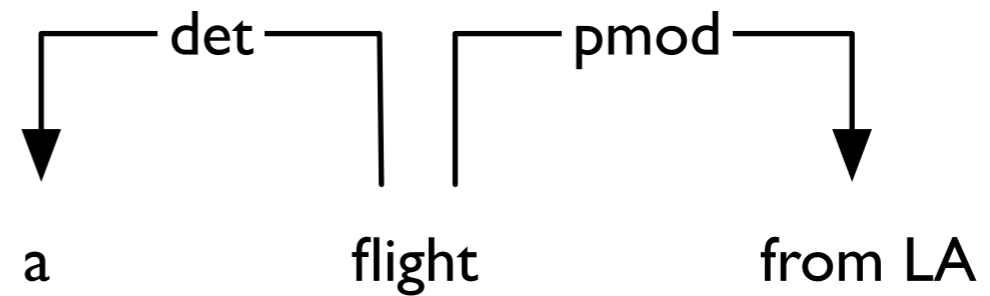
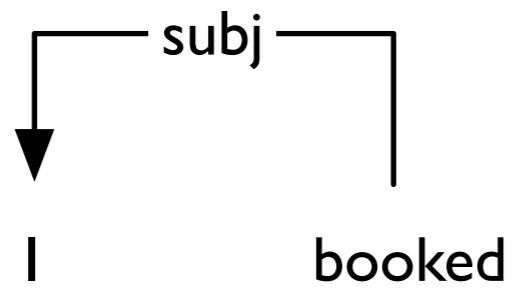


## Example run

Stack



Buffer



ra-dobj

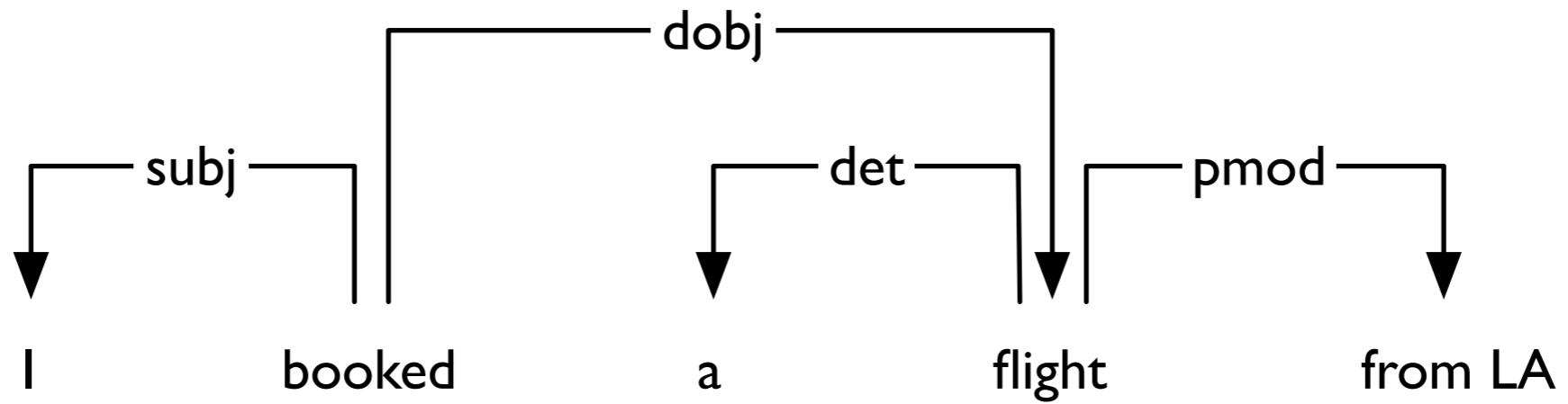


# Example run

Stack



Buffer





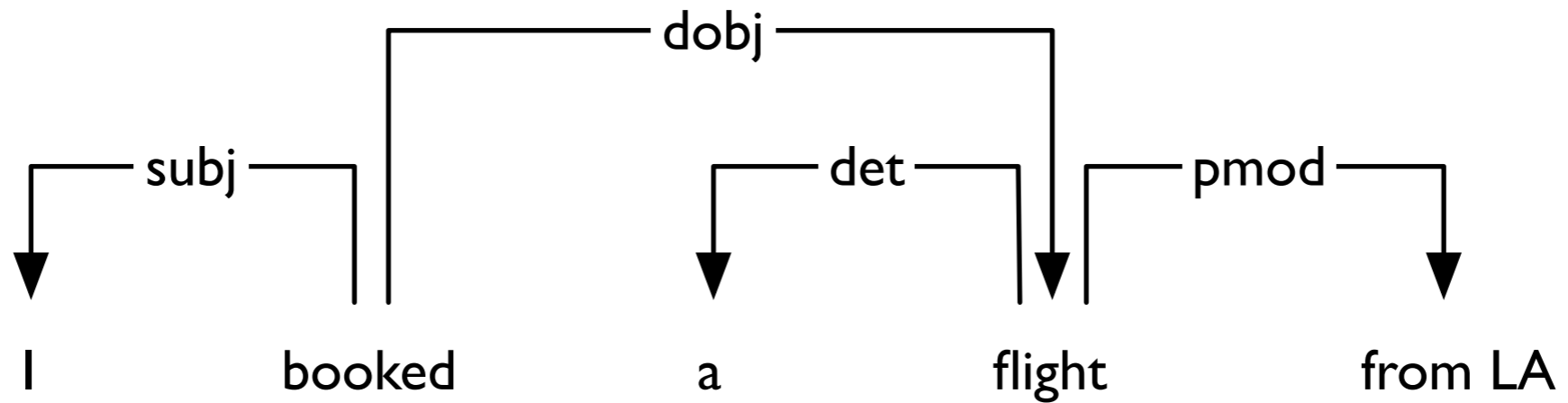


## Example run

Stack



Buffer



done!



# Complexity and optimality

- Time complexity is linear,  $O(n)$ , since we only have to treat each word once
- This can be achieved since the algorithm is greedy, and only builds one tree, in contrast to Eisner's algorithm, where all trees are explored
- There is no guarantee that we will even find the best tree given the model, with the arc-standard model
- There is a risk of error propagation
- An advantage is that we can use very informative features, for the ML algorithm



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# Training a guide



# Guides

- We need a guide that tells us what the next transition should be.
- The task of the guide can be understood as **classification**: Predict the next transition (class), given the current configuration.



# Training a guide

- We let the parser run on gold-standard trees.
- Every time there is a choice to make, we simply look into the tree and do ‘the right thing’<sup>TM</sup>.
- We collect all (configuration, transition) pairs and train a classifier on them.
- When parsing unseen sentences, we use the trained classifier as a guide.



# Training a guide

- The number of (configuration, transition) pairs is far too large.
- We define a set of **features** of configurations that we consider to be relevant for the task of predicting the next transition.

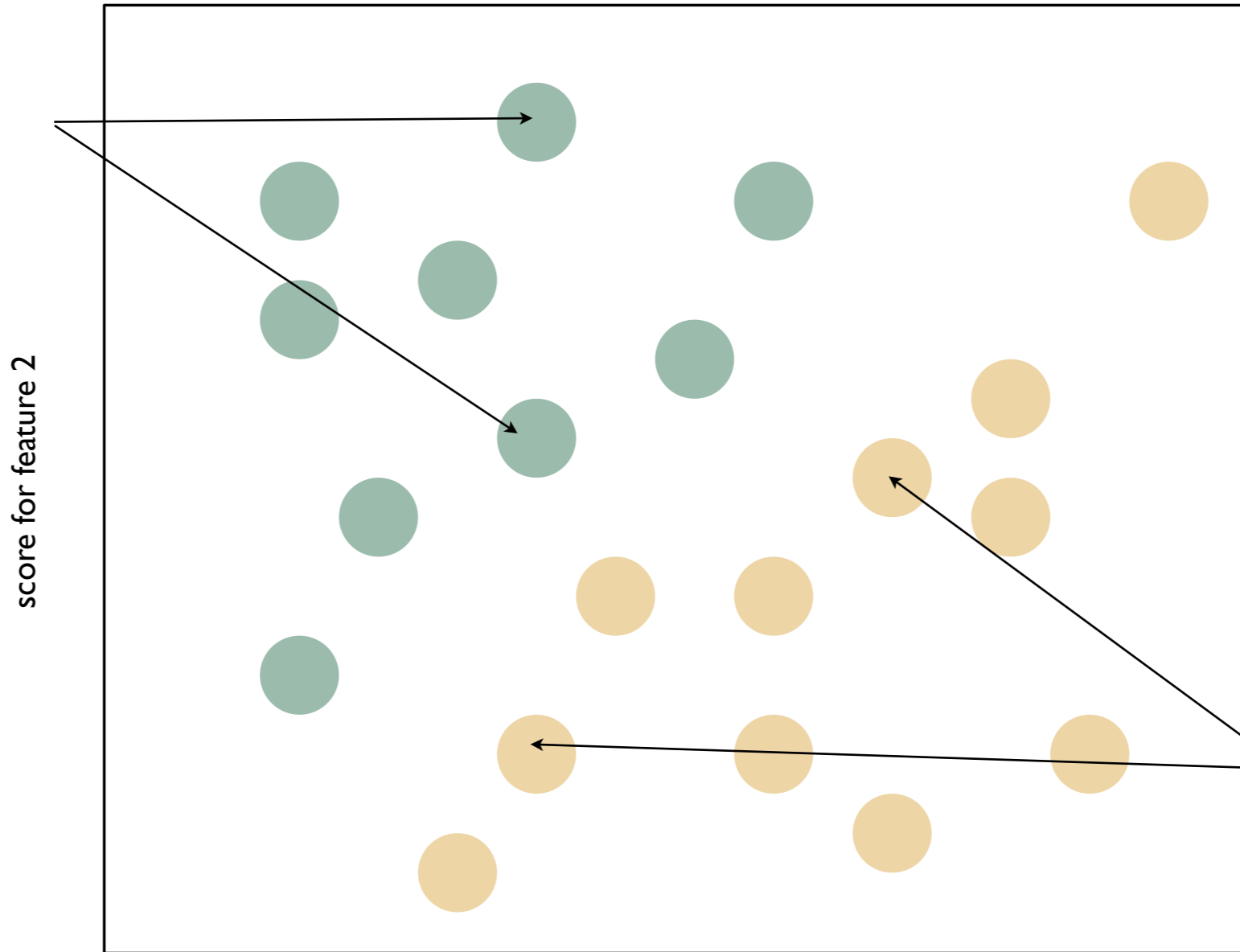
*Example:* word forms of the topmost two words on the stack and the next two words in the buffer

- We can then describe every configuration in terms of a **feature vector**.



# Training a guide

configurations in which  
we want to do la

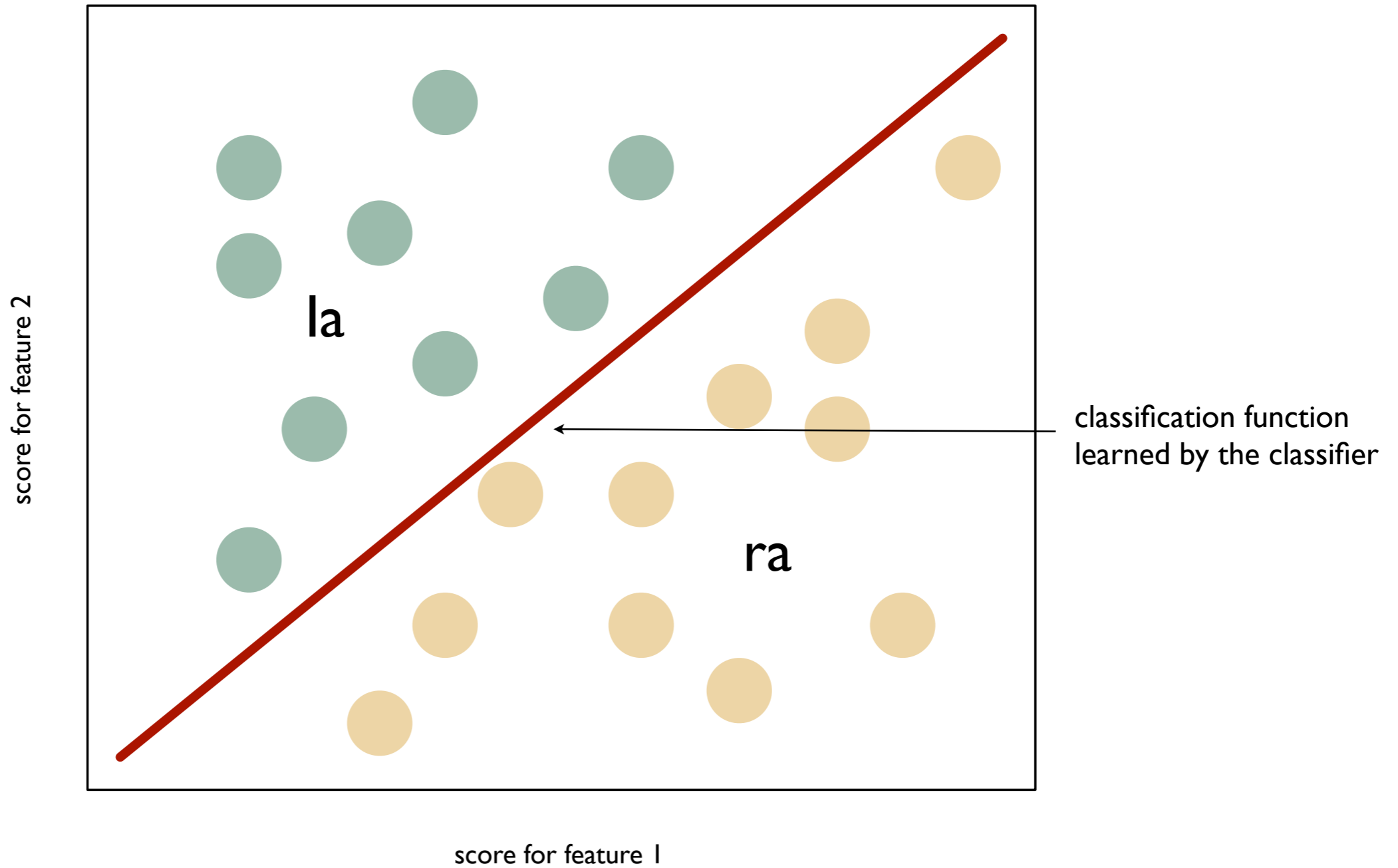


configurations in which  
we want to do ra

score for feature 1



# Training a guide







# Training a guide

- In practical systems, we have thousands of features and hundreds of transitions.
- There are several machine-learning paradigms that can be used to train a guide for such a task.

*Examples:* perceptron, decision trees,  
support-vector machines, memory-based learning



# Example features

	Attributes				
Adress	FORM	LEMMA	POS	FEATS	DEPREL
Stack[0]	X	X	X	X	
Stack[1]			X		
Ldep(Stack[0])					X
Rdep(Stack[0])					X
Buffer[0]	X	X	X	X	
Buffer[1]			X		
Ldep(Buffer[0])					X
Ldep(Buffer[0])					X
...					

- Combinations of addresses and attributes (e.g. those marked in the table)
- Other features, such as distances, number of children, ...



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# Alternative transition models



# Alternatives

- The arc-standard model as I presented it is just one example of a transition model
  - In the book you can see another version of the arc-standard model, where arcs are added between the topmost word on the stack and the topmost word on the buffer
- There are many other alternatives



# Arc-eager model

- Contains four transitions:
  - Shift
  - Reduce
  - Left-arc
  - Right-arc
- Advantage: not strictly bottom-up, can create arcs earlier than in the arc-standard model



# Non-projective transition model

- Allows non-projective parsing by adding a swap transition
- Contains four transitions:
  - Shift
  - Swap
  - Left-arc
  - Right-arc
- Runtime is  $O(n^2)$  in the worst case (but usually less in practice)



# Transition models in Maltparser

- Nivre family
  - Arcs created between stack and buffer
    - arc-eager model
    - arc-standard (variant from course book)
- Stack family
  - Arcs between two topmost words on stack
    - arc-standard model (from slides)
    - models with swap transition
- Other families available as well



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# Evaluation of dependency parsing





# Evaluation of dependency parsers

- **labelled attachment score (LAS):**  
percentage of correct arcs,  
relative to the gold standard
- **labelled exact match (LEM):**  
percentage of correct dependency trees,  
relative to the gold standard
- **unlabelled attachment score/exact match (UAS/  
UEM):**  
the same, but ignoring arc labels



# Word- vs sentence-level AS

- **Example: 2 sentence corpus**  
sentence 1: 9/10 arcs correct  
sentence 2: 15/45 arcs correct
- **Word-level (*micro-average*):**  
 $(9+15)/(10+45) = 24/55 = 0.436$
- **Sentence-level (*macro-average*):**  
 $(9/10+15/45)/2 = (0.9+0.33)/2 = 0.617$
- Word-level evaluation is normally used



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# Projectivity

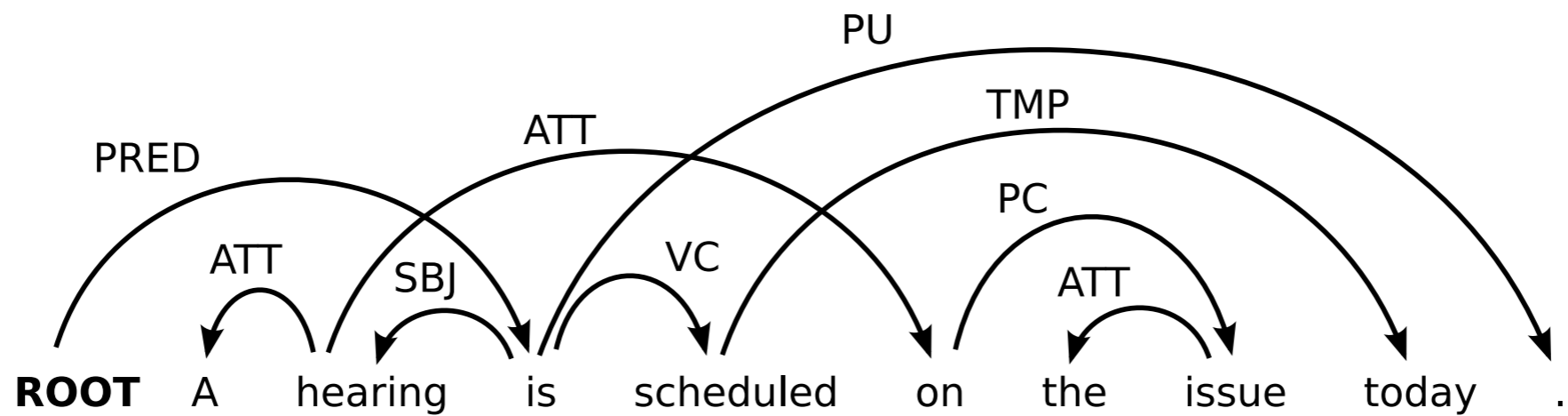
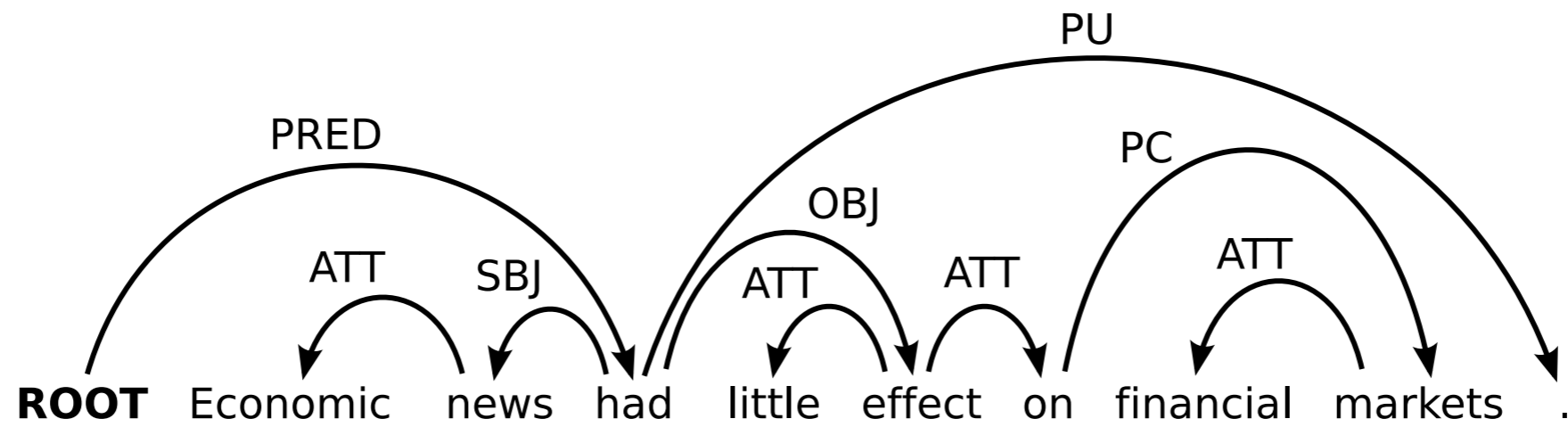


# Projectivity

- A dependency tree is projective if:
  - For every arc in the tree, there is a directed path from the head of the arc to all words occurring between the head and the dependent (that is, the arc  $(i,l,j)$  implies that  $i \rightarrow^* k$  for every  $k$  such that  $\min(i, j) < k < \max(i, j)$ )



# Projective and non-projective trees





# Projectivity and dependency parsing

- Many dependency parsing algorithms can only handle projective trees
  - Including all algorithms we have discussed in detail
- Non-projective trees do occur in natural language
  - How often depends on language (and treebank)



# Non-projective dependency parsing

- Variants of transition-based parsing
  - Using a swap-transition
  - Using more than one stack
  - Pseudo-projective parsing (seminar 2)
- Graph-based parsing
  - Minimum spanning tree algorithms
- ...



# Summary

- In transition-based dependency parsing one does not score graphs but computations, sequences of (configuration, transition) pairs.
- In its simplest form, transition-based dependency parsing uses classification.
- One specific instance of transition-based dependency parsing is the arc-standard algorithm.





# The end of the course

- Seminar 2, Pseudo-projective parsing
  - Easier and shorter article than last seminar, some more general questions
- Assignment 3: Disambiguation in arc-factored and transition-based parsing
- Assignment 4: Try and evaluate MaltParser
- Course evaluation in the student portal