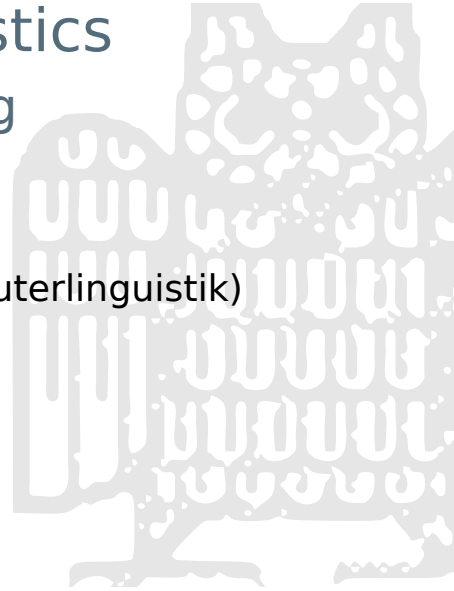


# Computational Linguistics

## Dependency-based Parsing

Dietrich Klakow & Stefan Thater  
FR 4.7 Allgemeine Linguistik (Computerlinguistik)  
Universität des Saarlandes

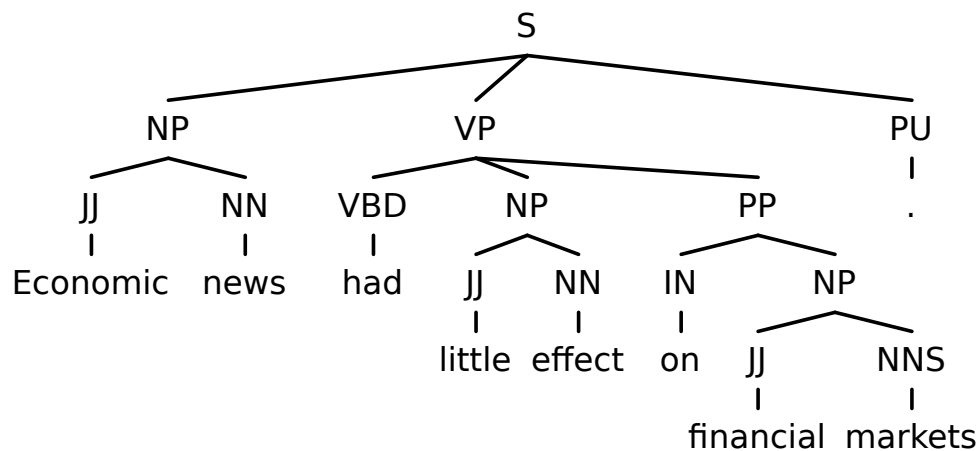
Summer 2012



## Acknowledgements

- These slides are heavily inspired by
  - an ESSLLI 2007 course by Joakim Nivre and Ryan McDonald
  - an ACL-COLING tutorial by Joakim Nivre and Sandra Kübler

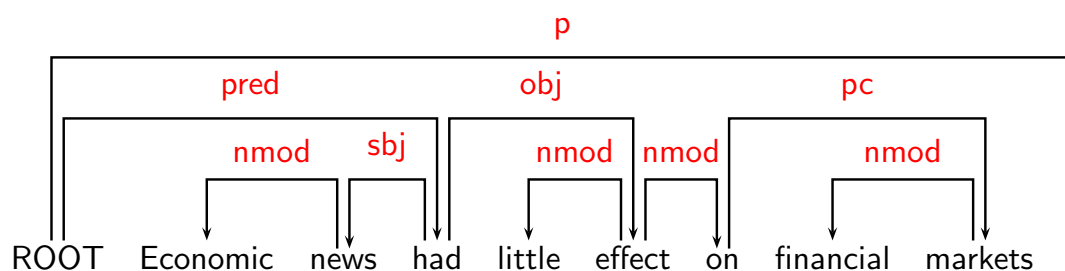
# Phrase-Structure Trees



3

# Dependency Trees

- **Basic idea:**
  - Syntactic structure = lexical items linked by relations
  - Syntactic structures are usually trees (... but not always)
- Relations  $H \rightarrow D$ 
  - H is the head (or governor)
  - D is the dependent



4

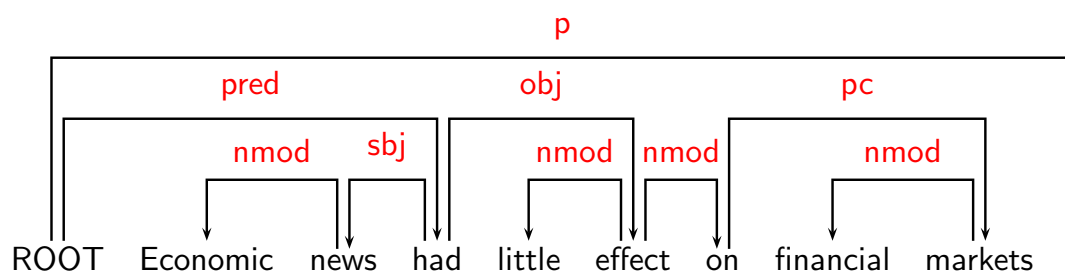
# Dependency Trees

## ■ Parsers

- are easy to implement and evaluate

## ■ Dependency-based representations

- are suitable for free word order languages
- are often close to the predicate argument structure



5

# Dependency Trees

## ■ Some criteria for dependency relations between a head H and a dependent D in a linguistic construction C:

- H determines the syntactic category of C; H can replace C.
- H determines the semantic category of C; D specifies H.
- H is obligatory; D may be optional.
- H selects D and determines whether D is obligatory.
- The form of D depends on H (agreement or government).
- The linear position of D is specified with reference to H.

6

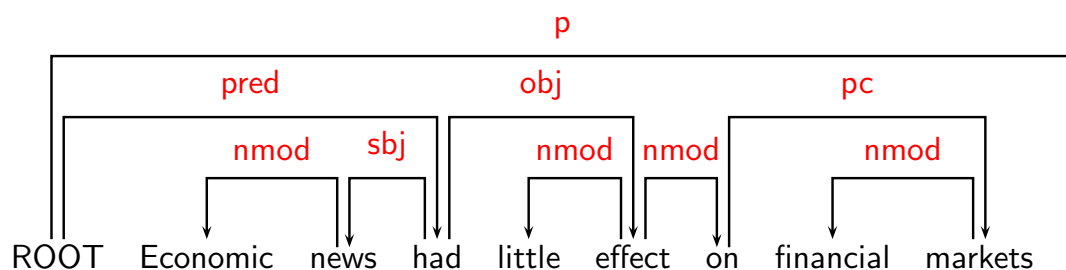
# Dependency Trees

- Clear cases:
  - Subject, Object, ...
- Less clear cases:
  - complex verb groups
  - subordinate clauses
  - coordination
  - ...

7

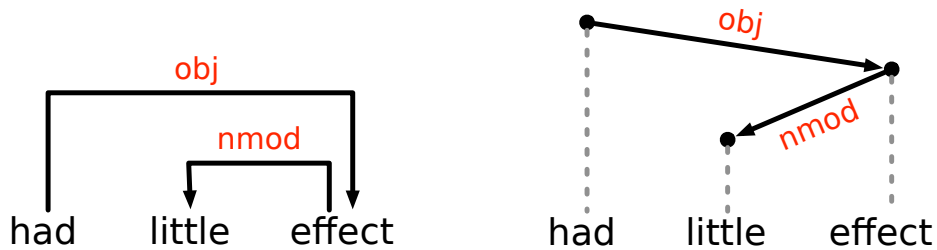
# Dependency Graphs

- Graph  $G = \langle V, A, L, < \rangle$ 
  - $V$  = a set of vertices (nodes)
  - $A$  = a set of arcs (directed edges)
  - $L$  = a set of edge labels
  - $<$  = a linear order on  $V$



8

# Dependency Trees – Notation



9

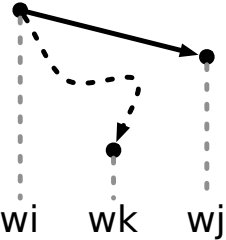
# Dependency Graphs / Trees

- Formal conditions on dependency graphs:
  - G is weakly connected
  - G is acyclic
  - Every node in G has at most one head
  - G is projective

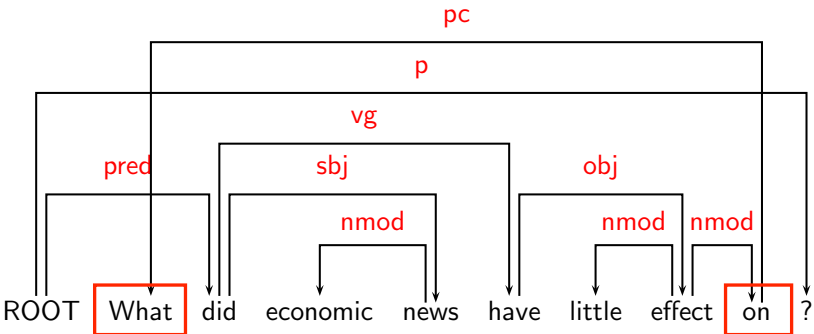
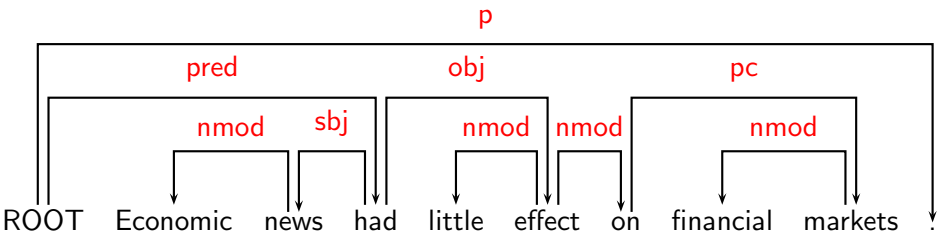
10

# Projectivity

- A dependency graph  $G$  is projective iff
  - if  $w_i \rightarrow w_j$ , then  $w_i \rightarrow^* w_k$  for all  $w_i < w_k < w_j$  or  $w_j < w_k < w_i$
  - if  $w_i$  is the head of  $w_j$ , then there must be a directed path from  $w_i$  to  $w_k$ , for all  $w_k$  between  $w_i$  and  $w_j$ .
- We need non-projectivity for
  - long distance dependencies
  - free word order



# Projectivity



# Projectivity

Language	Sentences	Dependencies
Arabic [Maamouri and Bies 2004]	11.2%	0.4%
Basque [Aduriz et al. 2003]	26.2%	2.9%
Czech [Hajic et al. 2001]	23.2%	1.9%
Danish [Kromann 2003]	15.6%	1.0%
Greek [Prokopidis et al. 2005]	20.3%	1.1%
Russian [Boguslavsky et al. 2000]	10.6%	0.9%
Slovenian [Dzeroski et al. 2006]	22.2%	1.9%
Turkish [Oflazer et al. 2003]	11.6%	1.5%

13

# Dependency-based Parsing

- Grammar-based
- Data-driven
  - Transition-based
  - Graph-based

14

# Transition-based Parsing

- Configurations  $\langle S, Q, A \rangle$ 
  - $S$  = a stack of partially processed tokens (nodes)
  - $Q$  = a queue of unprocessed input tokens
  - $A$  = a set of dependency arcs
- Initial configuration for input  $w_1 \dots w_n$ 
  - $\langle [w_0], [w_1, \dots, w_n], \{\} \rangle$ ,  $w_0 = \text{ROOT}$
- Terminal (accepting) configuration
  - $\langle \dots, [], \dots \rangle$

15

## Transitions („Arc-Standard“)

- Left-Arc( $r$ )
  - adds a dependency arc  $(w_j, r, w_i)$  to the arc set  $A$ , where  $w_i$  is the word on top of the stack and  $w_j$  is the first word in the buffer, and pops the stack.
- Right-Arc( $r$ )
  - adds a dependency arc  $(w_i, r, w_j)$  to the arc set  $A$ , where  $w_i$  is the word on top of the stack and  $w_j$  is the first word in the buffer, pops the stack and replaces  $w_j$  by  $w_i$  at the head of buffer.

16



# Transitionen („Arc-Standard“)

- Left-Arc(r)

$$\frac{\langle [\dots, w_i], [w_j, \dots], A \rangle}{\langle [\dots], [w_j, \dots], A \cup \{(w_j, r, w_i)\} \rangle} \quad i \neq 0, \neg \exists k \exists l' (wk, l', w_i) \in A$$

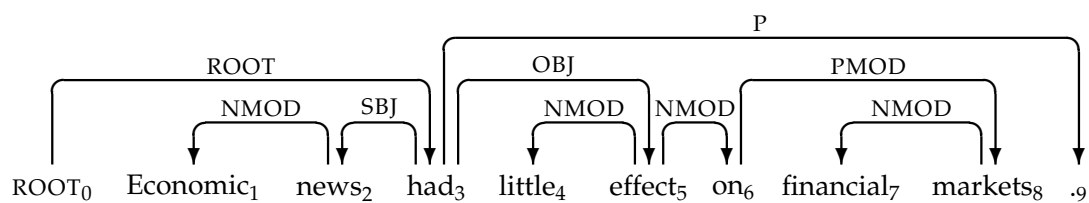
- Right-Arc(r)

$$\frac{\langle [\dots, w_i], [w_j, \dots], A \rangle}{\langle [\dots], [w_i, \dots], A \cup \{(w_i, r, w_j)\} \rangle} \quad \neg \exists k \exists l' (wk, l', w_j) \in A$$

- Shift

$$\frac{\langle [\dots], [w_i, \dots], A \rangle}{\langle [\dots, w_i], [\dots], A \rangle}$$

# An Example



**Figure 2**  
Dependency graph for an English sentence from the Penn Treebank.

# An Example

	(	[0,	[1, ..., 9],	$\emptyset$	)
SHIFT $\Rightarrow$	(	[0, 1],	[2, ..., 9],	$\emptyset$	)
LEFT-ARC <sub>NMOD</sub> $\Rightarrow$	(	[0],	[2, ..., 9],	$A_1 = \{(2, \text{NMOD}, 1)\}$	)
SHIFT $\Rightarrow$	(	[0, 2],	[3, ..., 9],	$A_1$	)
LEFT-ARC <sub>SBJ</sub> $\Rightarrow$	(	[0],	[3, ..., 9],	$A_2 = A_1 \cup \{(3, \text{SBJ}, 2)\}$	)
SHIFT $\Rightarrow$	(	[0, 3],	[4, ..., 9],	$A_2$	)
SHIFT $\Rightarrow$	(	[0, 3, 4],	[5, ..., 9],	$A_2$	)
LEFT-ARC <sub>NMOD</sub> $\Rightarrow$	(	[0, 3],	[5, ..., 9],	$A_3 = A_2 \cup \{(5, \text{NMOD}, 4)\}$	)
SHIFT $\Rightarrow$	(	[0, 3, 5],	[6, ..., 9],	$A_3$	)
SHIFT $\Rightarrow$	(	[0, ..., 6],	[7, 8, 9],	$A_3$	)
SHIFT $\Rightarrow$	(	[0, ..., 7],	[8, 9],	$A_3$	)
LEFT-ARC <sub>NMOD</sub> $\Rightarrow$	(	[0, ..., 6],	[8, 9],	$A_4 = A_3 \cup \{(8, \text{NMOD}, 7)\}$	)
RIGHT-ARC <sub>PMOD</sub> <sup>s</sup> $\Rightarrow$	(	[0, 3, 5],	[6, 9],	$A_5 = A_4 \cup \{(6, \text{PMOD}, 8)\}$	)
RIGHT-ARC <sub>NMOD</sub> <sup>s</sup> $\Rightarrow$	(	[0, 3],	[5, 9],	$A_6 = A_5 \cup \{(5, \text{NMOD}, 6)\}$	)
RIGHT-ARC <sub>OBJ</sub> <sup>s</sup> $\Rightarrow$	(	[0],	[3, 9],	$A_7 = A_6 \cup \{(3, \text{OBJ}, 5)\}$	)
SHIFT $\Rightarrow$	(	[0, 3],	[9],	$A_7$	)
RIGHT-ARC <sub>P</sub> <sup>s</sup> $\Rightarrow$	(	[0],	[3],	$A_8 = A_7 \cup \{(3, \text{P}, 9)\}$	)
RIGHT-ARC <sub>ROOT</sub> <sup>s</sup> $\Rightarrow$	(	[],	[0],	$A_9 = A_8 \cup \{(0, \text{ROOT}, 3)\}$	)
SHIFT $\Rightarrow$	(	[0],	[],	$A_9$	)

19

# Deterministic Parsing

- oracle(c):
  - predicts the next transition
- parse( $w_1 \dots w_n$ ):
  - $c := \langle [w_0 = \text{ROOT}], [w_1, \dots, w_n], \{\} \rangle$
  - while c is not terminal
    - $t := \text{oracle}(c)$
    - $c := t(c)$
  - return  $G = \langle \{w_0, \dots, w_n\}, A_c \rangle$

20

# Deterministic Parsing

- Linear time complexity: the algorithm terminates after  $2n$  steps for input sentences with  $n$  words.
- The algorithm is complete and correct for the class of projective dependency trees:
  - For every projective dependency tree  $T$  there is a sequence of transitions that generates  $T$
  - Every sequence of transition steps generates a projective dependency tree
- Whether the resulting dependency tree is correct or not depends of course on the oracle.

21

# The oracle

- Approximate the oracle by a classifier
- Represent configurations by feature vectors; for instance
  - lexical properties (word form, lemma)
  - category (part of speech)
  - labels of partial dependency trees
  - ...

22

# An Example

$f(c_0)$	=	(ROOT	Economic	news	NULL	NULL	NULL	NULL)
$f(c_1)$	=	(Economic	news	had	NULL	NULL	NULL	NULL)
$f(c_2)$	=	(ROOT	news	had	NULL	NULL	ATT	NULL)
$f(c_3)$	=	(news	had	little	ATT	NULL	NULL	NULL)
$f(c_4)$	=	(ROOT	had	little	NULL	NULL	SBJ	NULL)
$f(c_5)$	=	(had	little	effect	SBJ	NULL	NULL	NULL)
$f(c_6)$	=	(little	effect	on	NULL	NULL	NULL	NULL)
$f(c_7)$	=	(had	effect	on	SBJ	NULL	ATT	NULL)
$f(c_8)$	=	(effect	on	financial	ATT	NULL	NULL	NULL)
$f(c_9)$	=	(on	financial	markets	NULL	NULL	NULL	NULL)
$f(c_{10})$	=	(financial	markets	.	NULL	NULL	NULL	NULL)
$f(c_{11})$	=	(on	markets	.	NULL	NULL	ATT	NULL)
$f(c_{12})$	=	(effect	on	.	ATT	NULL	NULL	ATT)
$f(c_{13})$	=	(had	effect	.	SBJ	NULL	ATT	ATT)
$f(c_{14})$	=	(ROOT	had	.	NULL	NULL	SBJ	OBJ)
$f(c_{15})$	=	(had	.	NULL	SBJ	OBJ	NULL	NULL)
$f(c_{16})$	=	(ROOT	had	NULL	NULL	NULL	SBJ	PU)
$f(c_{17})$	=	(NULL	ROOT	NULL	NULL	NULL	NULL	PRED)
$f(c_{18})$	=	(ROOT	NULL	NULL	NULL	PRED	NULL	NULL)

23

# Non-projective Parsing

- Configurations  $\langle L_1, L_2, Q, A \rangle$ 
  - $L_1, L_2$  are stacks of partially processed nodes
  - $Q$  = a queue of unprocessed input tokens
  - $A$  = a set of dependency arcs
- Initial configuration for input  $w_1 \dots w_n$ 
  - $\langle [w_0], [], [w_1, \dots, w_n], \{\} \rangle$ ,  $w_0 = \text{ROOT}$
- Terminal configuration:
  - $\langle [w_0, w_1, \dots, w_n], [], [], A \rangle$

24

# Transitions

- Left-Arc(l)

$$\frac{\langle [\dots, w_i], [\dots], [w_j, \dots], A \rangle}{\langle [\dots], [w_i, \dots], [w_j, \dots], A \cup \{(w_j, l, w_i)\} \rangle}$$

$$\begin{aligned} & i \neq 0 \\ & \neg \exists k \exists l' (w_k, l', w_i) \in A \\ & \neg w_i \rightarrow^* w_j \end{aligned}$$

- Right-Arc(l)

$$\frac{\langle [\dots, w_i], [\dots], [w_j, \dots], A \rangle}{\langle [\dots], [w_i, \dots], [w_j, \dots], A \cup \{(w_i, l, w_j)\} \rangle}$$

$$\begin{aligned} & \neg \exists k \exists l' (w_k, l', w_j) \in A \\ & \neg w_i \rightarrow^* w_j \end{aligned}$$

25

# Transitions

- No-Arc

$$\frac{\langle [\dots, w_i], [\dots], [\dots], A \rangle}{\langle [\dots], [w_i, \dots], [\dots], A \rangle}$$

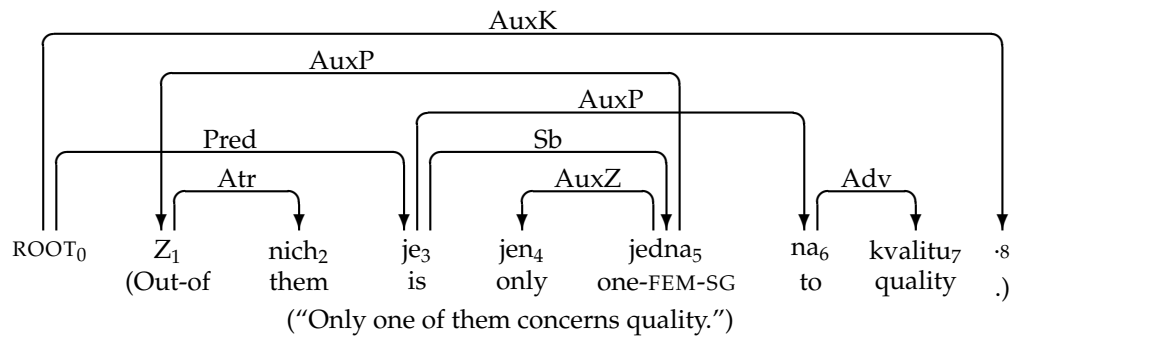
- Shift

$$\frac{\langle [\dots]_{L_1}, [\dots]_{L_2}, [w_i, \dots], A \rangle}{\langle [\dots]_{L_1} \cdot [\dots, w_i]_{L_2}, [], [\dots], A \rangle}$$

- $L_1 \cdot L_2 =$  the concatenation of  $L_1$  and  $L_2$

26

# An Example



**Figure 1**  
 Dependency graph for a Czech sentence from the Prague Dependency Treebank.

# An Example

	( [0,	[ ],	[1, ..., 8],	∅	)
SHIFT <sup>λ</sup> ⇒	( [0, 1],	[ ],	[2, ..., 8],	∅	)
RIGHT-ARC <sup>n</sup> <sub>Atr</sub> ⇒	( [0,	[1,	[2, ..., 8],	A <sub>1</sub> = {(1, Atr, 2)}	)
SHIFT <sup>λ</sup> ⇒	( [0, 1, 2],	[ ],	[3, ..., 8],	A <sub>1</sub>	)
NO-ARC <sup>n</sup> ⇒	( [0, 1],	[2,	[3, ..., 8],	A <sub>1</sub>	)
NO-ARC <sup>n</sup> ⇒	( [0,	[1, 2],	[3, ..., 8],	A <sub>1</sub>	)
RIGHT-ARC <sup>n</sup> <sub>Pred</sub> ⇒	( [ ],	[0, 1, 2],	[3, ..., 8],	A <sub>2</sub> = A <sub>1</sub> ∪ {(0, Pred, 3)}	)
SHIFT <sup>λ</sup> ⇒	( [0, ..., 3],	[ ],	[4, ..., 8],	A <sub>2</sub>	)
SHIFT <sup>λ</sup> ⇒	( [0, ..., 4],	[ ],	[5, ..., 8],	A <sub>2</sub>	)
LEFT-ARC <sup>n</sup> <sub>AuxZ</sub> ⇒	( [0, ..., 3],	[4,	[5, ..., 8],	A <sub>3</sub> = A <sub>2</sub> ∪ {(5, AuxZ, 4)}	)
RIGHT-ARC <sup>n</sup> <sub>Sb</sub> ⇒	( [0, 1, 2],	[3, 4],	[5, ..., 8],	A <sub>4</sub> = A <sub>3</sub> ∪ {(3, Sb, 5)}	)
NO-ARC <sup>n</sup> ⇒	( [0, 1],	[2, 3, 4],	[5, ..., 8],	A <sub>4</sub>	)
LEFT-ARC <sup>n</sup> <sub>AuxP</sub> ⇒	( [0,	[1, ..., 4],	[5, ..., 8],	A <sub>5</sub> = A <sub>4</sub> ∪ {(5, AuxP, 1)}	)
SHIFT <sup>λ</sup> ⇒	( [0, ..., 5],	[ ],	[6, 7, 8],	A <sub>5</sub>	)
NO-ARC <sup>n</sup> ⇒	( [0, ..., 4],	[5],	[6, 7, 8],	A <sub>5</sub>	)

# An Example

NO-ARC <sup>n</sup>	⇒	( [0, ..., 4], [5], [6, 7, 8], A <sub>5</sub> )
NO-ARC <sup>n</sup>	⇒	( [0, ..., 3], [4, 5], [6, 7, 8], A <sub>5</sub> )
RIGHT-ARC <sup>n</sup> <sub>AuxP</sub>	⇒	( [0, 1, 2], [3, 4, 5], [6, 7, 8], A <sub>6</sub> = A <sub>5</sub> ∪ {(3, AuxP, 6)} )
SHIFT <sup>λ</sup>	⇒	( [0, ..., 6], [], [7, 8], A <sub>6</sub> )
RIGHT-ARC <sup>n</sup> <sub>Adv</sub>	⇒	( [0, ..., 5], [6], [7, 8], A <sub>7</sub> = A <sub>6</sub> ∪ {(6, Adv, 7)} )
SHIFT <sup>λ</sup>	⇒	( [0, ..., 7], [], [8], A <sub>7</sub> )
NO-ARC <sup>n</sup>	⇒	( [0, ..., 6], [7], [8], A <sub>7</sub> )
NO-ARC <sup>n</sup>	⇒	( [0, ..., 5], [6, 7], [8], A <sub>7</sub> )
NO-ARC <sup>n</sup>	⇒	( [0, ..., 4], [5, 6, 7], [8], A <sub>7</sub> )
NO-ARC <sup>n</sup>	⇒	( [0, ..., 3], [4, ..., 7], [8], A <sub>7</sub> )
NO-ARC <sup>n</sup>	⇒	( [0, 1, 2], [3, ..., 7], [8], A <sub>7</sub> )
NO-ARC <sup>n</sup>	⇒	( [0, 1], [2, ..., 7], [8], A <sub>7</sub> )
NO-ARC <sup>n</sup>	⇒	( [0], [1, ..., 7], [8], A <sub>7</sub> )
RIGHT-ARC <sup>n</sup> <sub>AuxK</sub>	⇒	( [], [0, ..., 7], [8], A <sub>8</sub> = A <sub>7</sub> ∪ {(0, AuxK, 8)} )
SHIFT <sup>λ</sup>	⇒	( [0, ..., 8], [], [] )

29

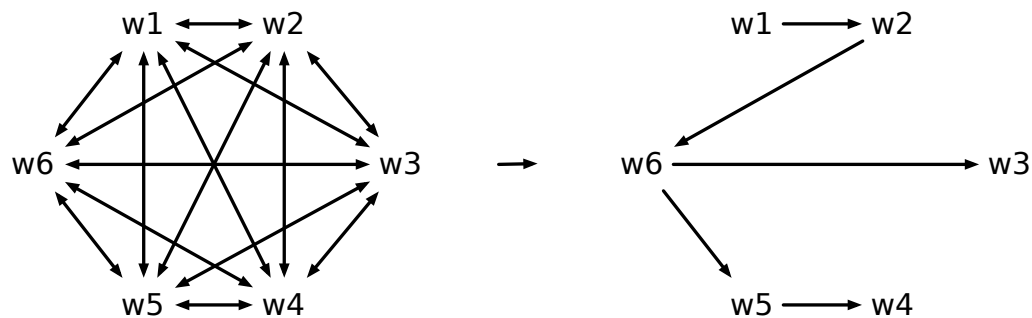
# Non-projective Parsing

- The algorithm is sound and complete for the class of dependency forests
- Time complexity is O(n<sup>2</sup>)
  - at most n Shift-transitions
  - between the i-th and (i+1)-th Shift-transition there are at most i transitions (left-arc, right-arc, no-arc)

30

# Graph-based Parsing

- Basic idea:
  - consider the complete graph where the nodes are the words from the input and edges are annotated with scores
  - Parsing = compute the maximum spanning tree



31

## Literature

- Sandra Kübler, Ryan McDonald and Joakim Nivre (2009). Dependency Parsing.
- Joakim Nivre (2008). Algorithms for Deterministic Incremental Dependency Parsing. Computational Linguistics 34(4), 513-553.

32