

CS5740: Natural Language Processing
Spring 2018

Dependency Parsing

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Overview

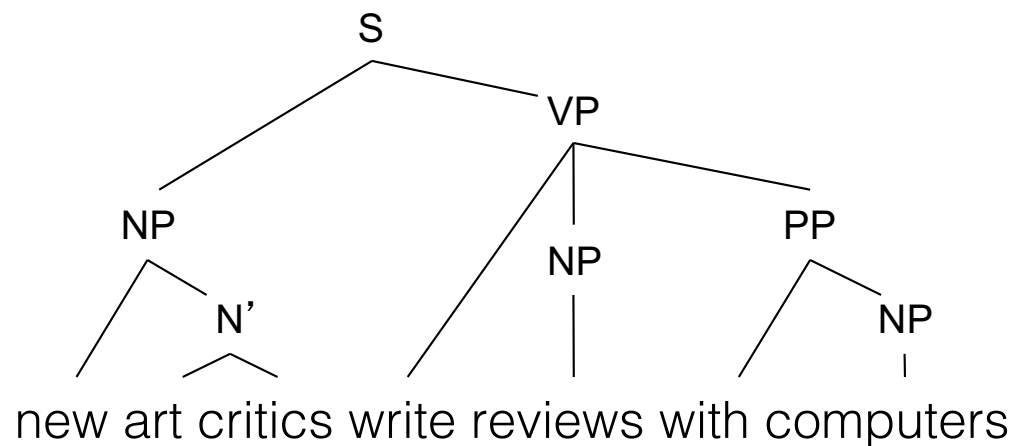
- The parsing problem
- Methods
 - Transition-based parsing
- Evaluation
- Projectivity

Parse Trees

- Part-of-speech Tagging:
 - Word classes
- Parsing:
 - From words to phrases to sentences
 - Relations between words
- Two views
 - Constituency
 - Dependency

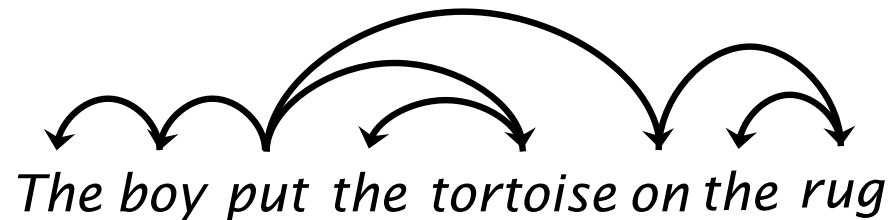
Constituency (Phrase Structure) Parsing

- Phrase structure organizes words into nested constituents
- Linguists can, and do, argue about details
- Lots of ambiguity



Dependency Parsing

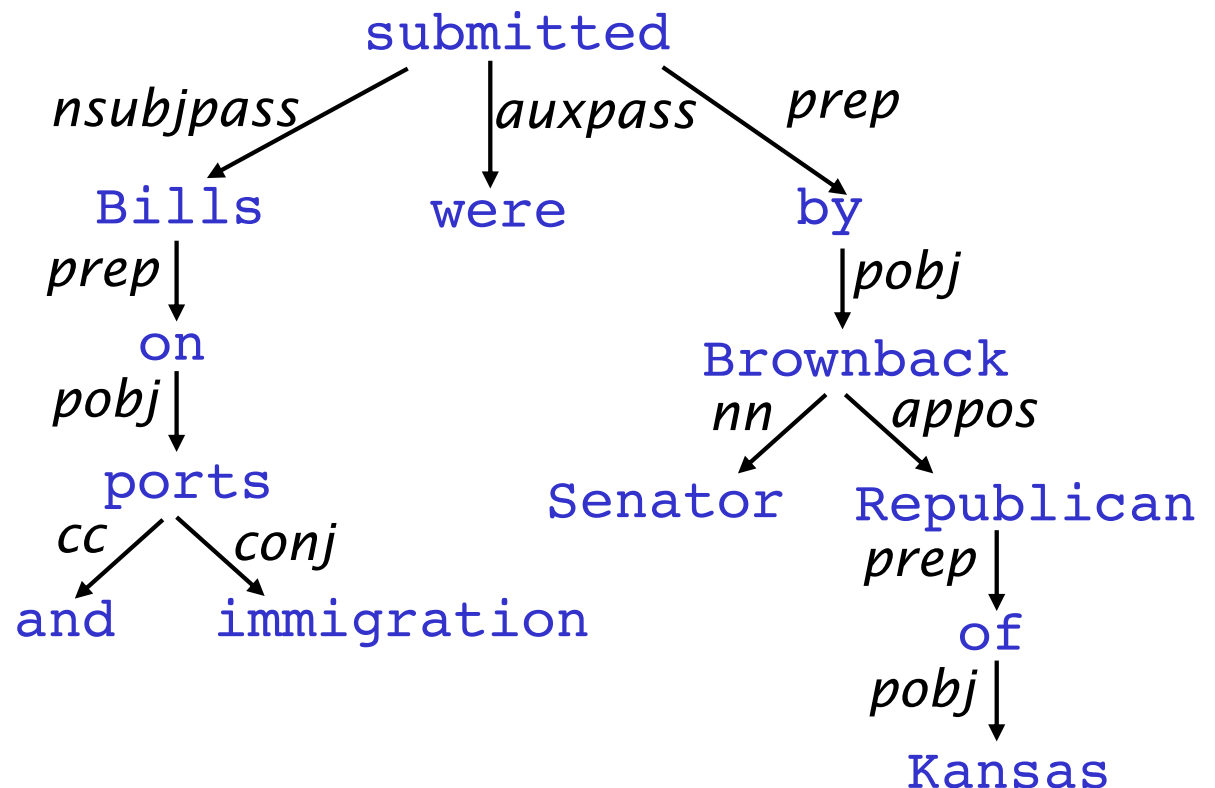
- Dependency structure shows which words depend on (modify or are arguments of) which other words.



Dependency Structure

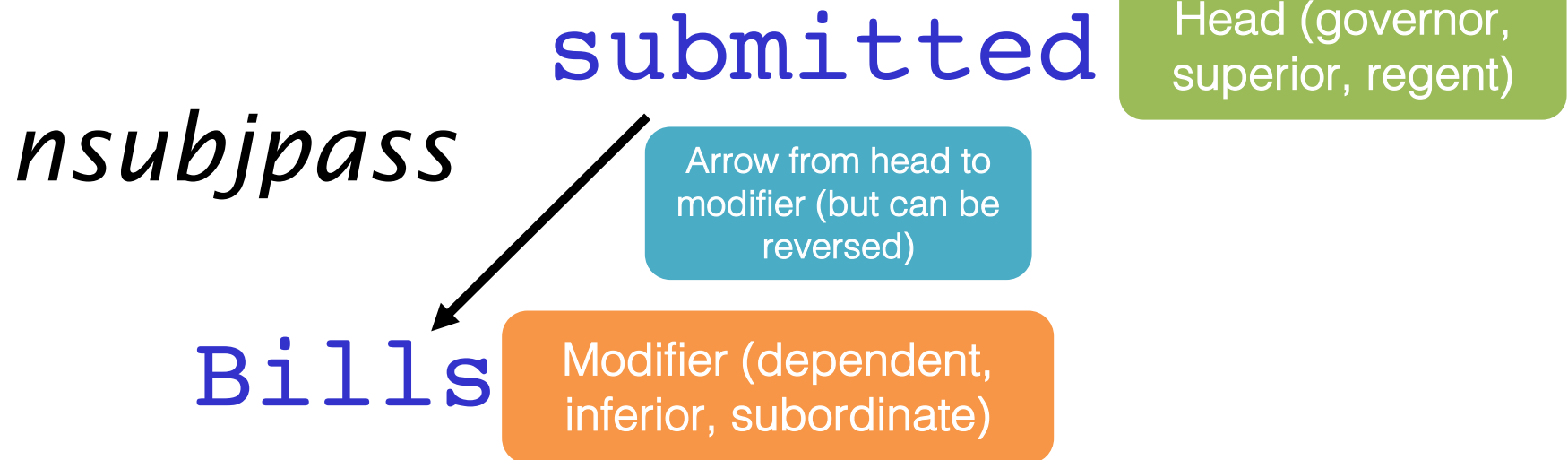
- Syntactic structure consists of:
 - Lexical items
 - Binary asymmetric relations → dependencies

Dependencies are
typed with name of
grammatical relation



Dependency Structure

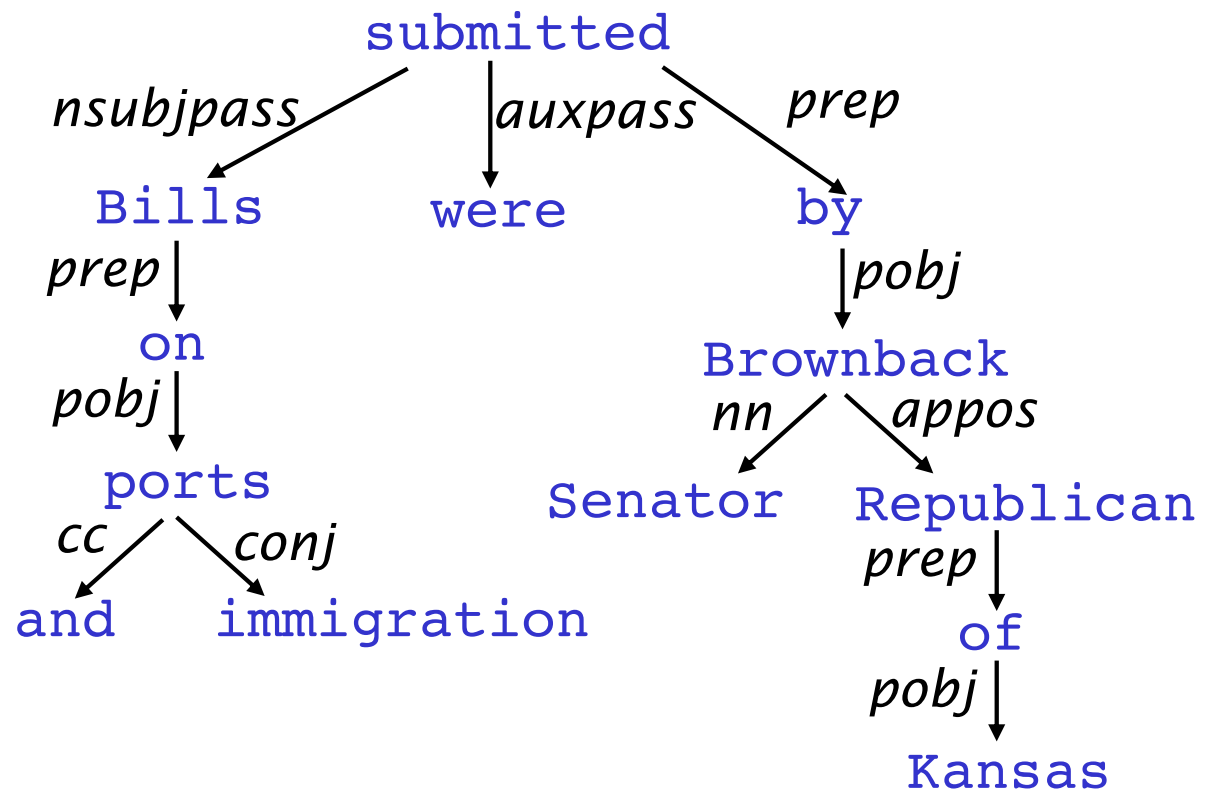
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Dependency Structure

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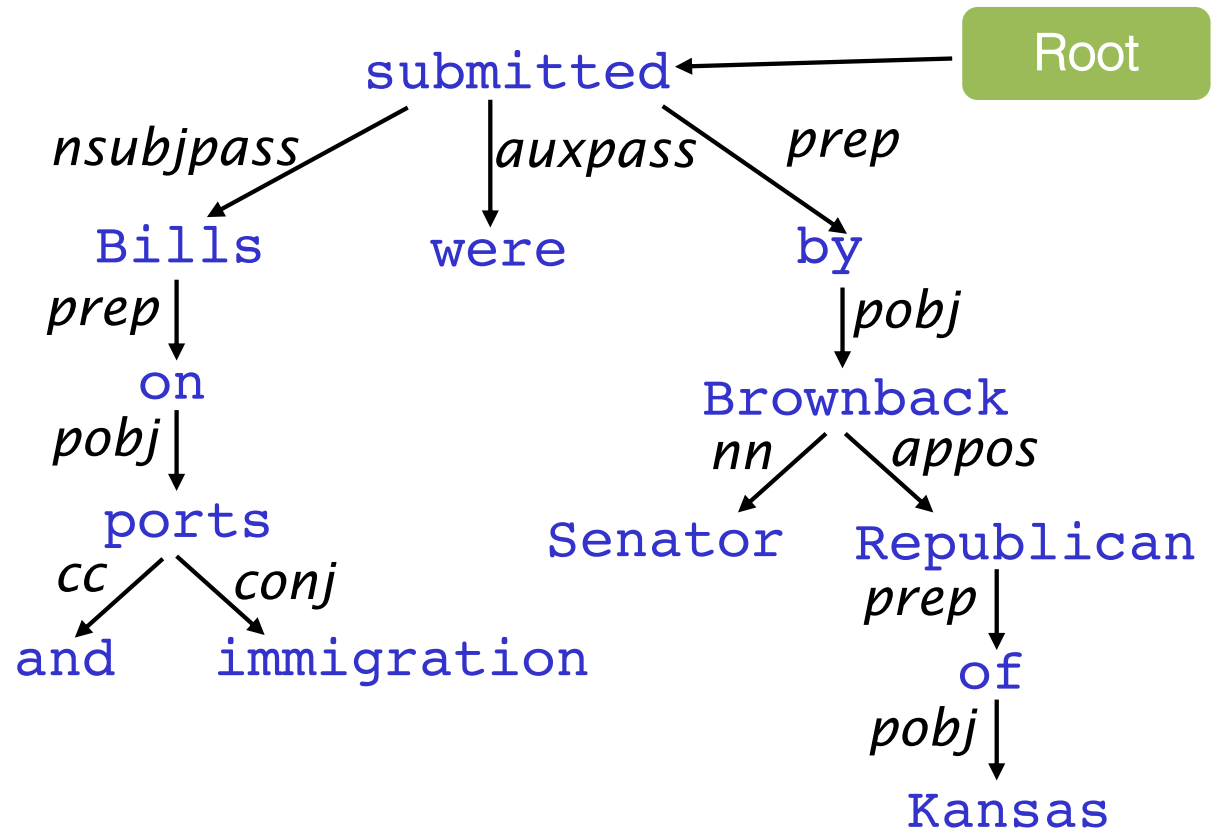
Dependencies
form a tree



Dependency Structure

- Syntactic structure consists of:
 - Lexical items
 - Binary asymmetric relations → dependencies

Dependencies
form a tree





Let's Parse

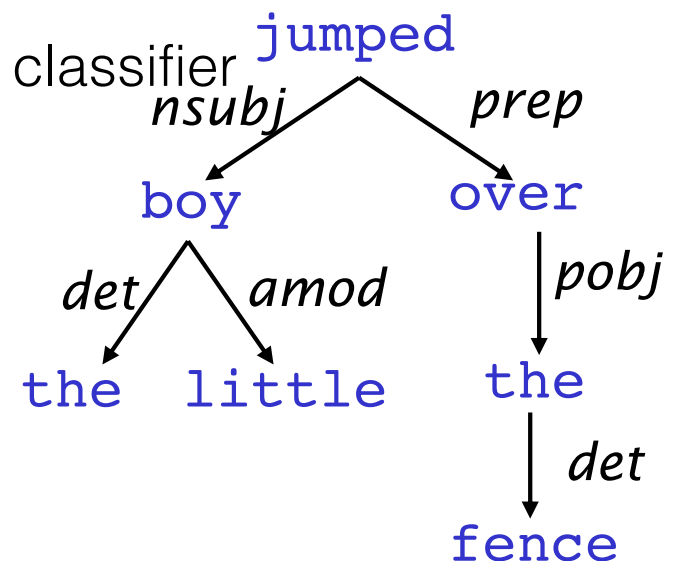
Start with main verb, and draw dependencies. Don't worry about labels. Just try the modifiers right.

John saw Mary

He said that the boy who was wearing the blue shirt with the white pockets has left the building

Methods for Dependency Parsing

- Dynamic programming
 - Eisner (1996): $O(n^3)$
- Graph algorithms
 - McDonald et al. (2005): score edges independently using classifier and use maximum spanning tree
- Constraint satisfaction
 - Start with all edges, eliminate based on hard constraints
- “Deterministic parsing”
 - Left-to-right, each choice is done with a classifier



Making Decisions

What are the sources of information for dependency parsing?

1. Bilexical affinities
 - [issues → the] is plausible
2. Dependency distance
 - mostly with nearby words
3. Intervening material
 - Dependencies rarely span intervening verbs or punctuation
4. Valency of heads
 - How many dependents on which side are usual for a head?



MaltParse (Nivre et al. 2008)

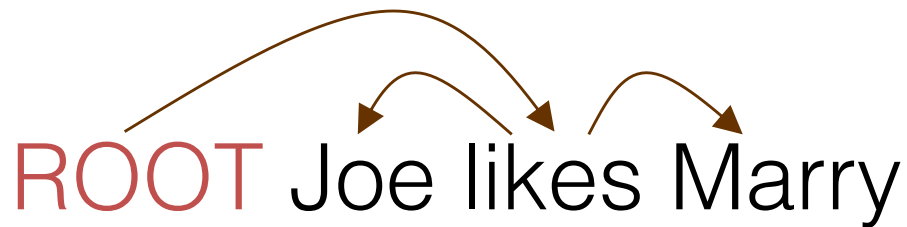
- Greedy transition-based parser
- Each decision: how to attach each word as we encounter it
 - If you are familiar: like shift-reduce parser
- Select each action with a classifier
- The parser has:
 - a **stack σ** , written with the top to the right
 - which starts with the ROOT symbol
 - a **buffer β** , written with the top to the left
 - which starts with the input sentence
 - a **set of dependency arcs A**
 - which starts off empty
 - a **set of actions**

Arc-standard Dependency Parsing

Start: $\sigma = [\text{ROOT}]$, $\beta = w_1, \dots, w_n$, $A = \emptyset$



- Shift $\sigma, w_i | \beta, A \rightarrow \sigma | w_i, \beta, A$
- Left-Arc_r $\sigma | w_i, w_j | \beta, A \rightarrow \sigma, w_j | \beta, A \cup \{r(w_j, w_i)\}$
- Right-Arc_r $\sigma | w_i, w_j | \beta, A \rightarrow \sigma, w_i | \beta, A \cup \{r(w_i, w_j)\}$

Finish: $\beta = \emptyset$

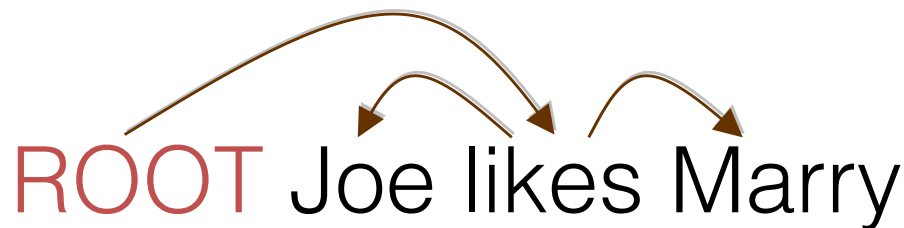


Arc-standard Dependency Parsing

Start: $\sigma = [\text{ROOT}]$, $\beta = w_1, \dots, w_n$, $A = \emptyset$

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Finish: $\beta = \emptyset$



	[ROOT]	[Joe, likes, marry]	\emptyset
Shift	[ROOT, Joe]	[likes, marry]	\emptyset
Left-Arc	[ROOT]	[likes, marry]	$\{(likes, Joe)\} = A_1$
Shift	[ROOT, likes]	[marry]	A_1
Right-Arc	[ROOT]	[likes]	$A_1 \cup \{(likes, Marry)\} = A_2$
Right-Arc	[]	[ROOT]	$A_2 \cup \{(ROOT, likes)\} = A_3$
Shift	[ROOT]	[]	A_3

Arc-standard Dependency Parsing

Start: $\sigma = [\text{ROOT}]$, $\beta = w_1, \dots, w_n$, $A = \emptyset$

- Shift $\sigma, w_i | \beta, A \rightarrow \sigma | w_i, \beta, A$
- Left-Arc_r $\sigma | w_i, w_j | \beta, A \rightarrow \sigma, w_j | \beta, A \cup \{r(w_j, w_i)\}$ 
- Right-Arc_r $\sigma | w_i, w_j | \beta, A \rightarrow \sigma, w_i | \beta, A \cup \{r(w_i, w_j)\}$ 

Finish: $\beta = \emptyset$



Arc-eager Dependency Parsing

Start:

$\sigma = [\text{ROOT}], \beta = w_1, \dots, w_n, A = \emptyset$

- Left-Arc_r $\sigma|w_i, w_j|\beta, A \rightarrow \sigma, w_j|\beta, A \cup \{r(w_j, w_i)\}$
 - Precondition: $r'(w_k, w_i) \notin A, w_i \neq \text{ROOT}$
- Right-Arc_r $\sigma|w_i, w_j|\beta, A \rightarrow \sigma|w_i|w_j, \beta, A \cup \{r(w_i, w_j)\}$
- Reduce $\sigma|w_i, \beta, A \rightarrow \sigma, \beta, A$
 - Precondition: $r'(w_k, w_i) \in A$
- Shift $\sigma, w_i|\beta, A \rightarrow \sigma|w_i, \beta, A$



Finish:

$\beta = \emptyset$

This is the common “arc-eager” variant: a head can immediately take a right dependent, before *its* dependents are found



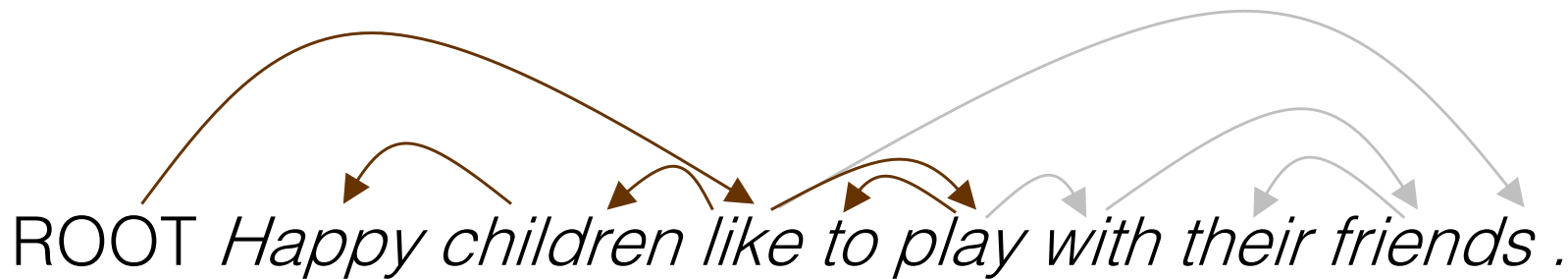
Arc-eager

1. Left-Arc_r $\sigma | w_i w_j | \beta, A \rightarrow \sigma, w_j | \beta, A \cup \{r(w_i w_j)\}$
Precondition: $r(w_i w_j) \notin A, w_i \neq \text{ROOT}$
2. Right-Arc_r $\sigma | w_i w_j | \beta, A \rightarrow \sigma | w_i | w_j \beta, A \cup \{r(w_i w_j)\}$
3. Reduce $\sigma | w_i \beta, A \rightarrow \sigma, \beta, A$
Precondition: $r(w_i) \in A$
4. Shift $\sigma, w_i | \beta, A \rightarrow \sigma | w_i \beta, A$



Arc-eager

1. Left-Arc_r $\sigma | w_i | w_j | \beta, A \rightarrow \sigma, w_j | \beta, A \cup \{r(w_i, w_j)\}$
Precondition: $r(w_i, w_j) \notin A, w_i \neq \text{ROOT}$
2. Right-Arc_r $\sigma | w_i | w_j | \beta, A \rightarrow \sigma | w_i | w_j | \beta, A \cup \{r(w_i, w_j)\}$
3. Reduce $\sigma | w_i | \beta, A \rightarrow \sigma, \beta, A$
Precondition: $r(w_i, w_j) \in A$
4. Shift $\sigma, w_i | \beta, A \rightarrow \sigma | w_i | \beta, A$



	[ROOT]	[Happy, children, ...]	\emptyset
Shift	[ROOT, Happy]	[children, like, ...]	\emptyset
LA _{amod}	[ROOT]	[children, like, ...]	$\{\text{amod}(\text{children}, \text{happy})\} = A_1$
Shift	[ROOT, children]	[like, to, ...]	A_1
LA _{nsubj}	[ROOT]	[like, to, ...]	$A_1 \cup \{\text{nsubj}(\text{like}, \text{children})\} = A_2$
RA _{root}	[ROOT, like]	[to, play, ...]	$A_2 \cup \{\text{root}(\text{ROOT}, \text{like})\} = A_3$
Shift	[ROOT, like, to]	[play, with, ...]	A_3
LA _{aux}	[ROOT, like]	[play, with, ...]	$A_3 \cup \{\text{aux}(\text{play}, \text{to})\} = A_4$
RA _{xcomp}	[ROOT, like, play]	[with their, ...]	$A_4 \cup \{\text{xcomp}(\text{like}, \text{play})\} = A_5$

Arc-eager

1. Left-Arc_r $\sigma | w_i w_j | \beta, A \rightarrow \sigma, w_j | \beta, A \cup \{r(w_i w_j)\}$
Precondition: $r(w_i w_j) \notin A, w_i \neq \text{ROOT}$
2. Right-Arc_r $\sigma | w_i w_j | \beta, A \rightarrow \sigma | w_i | w_j \beta, A \cup \{r(w_i w_j)\}$
3. Reduce $\sigma | w_i \beta, A \rightarrow \sigma, \beta, A$
Precondition: $r(w_i) \in A$
4. Shift $\sigma, w_i | \beta, A \rightarrow \sigma | w_i \beta, A$



RA _{xcomp}	[ROOT, like, play]	[with their, ...]	A ₄ ∪ {xcomp(like, play) = A ₅ }
RA _{prep}	[ROOT, like, play, with]	[their, friends, ...]	A ₅ ∪ {prep(play, with) = A ₆ }
Shift	[ROOT, like, play, with, their]	[friends, .]	A ₆
LA _{poss}	[ROOT, like, play, with]	[friends, .]	A ₆ ∪ {poss(friends, their) = A ₇ }
RA _{pobj}	[ROOT, like, play, with, friends]	[.]	A ₇ ∪ {pobj(with, friends) = A ₈ }
Reduce	[ROOT, like, play, with]	[.]	A ₈
Reduce	[ROOT, like, play]	[.]	A ₈
Reduce	[ROOT, like]	[.]	A ₈
RA _{punc}	[ROOT, like, .]	[]	A ₈ ∪ {punc(like, .) = A ₉ }

You terminate as soon as the buffer is empty. Dependencies = A₉

MaltParser (Nivre et al. 2008)

- Selecting the next action:
 - Discriminative classifier (SVM, MaxEnt, etc.)
 - Untyped choices: 4
 - Typed choices: $|R| * 2 + 2$
- Features: POS tags, word in stack, word in buffer, etc.
- Greedy → no search
 - But can easily do beam search
- Close to state of the art
- Linear time parser → **very fast!**

Parsing with Neural Networks

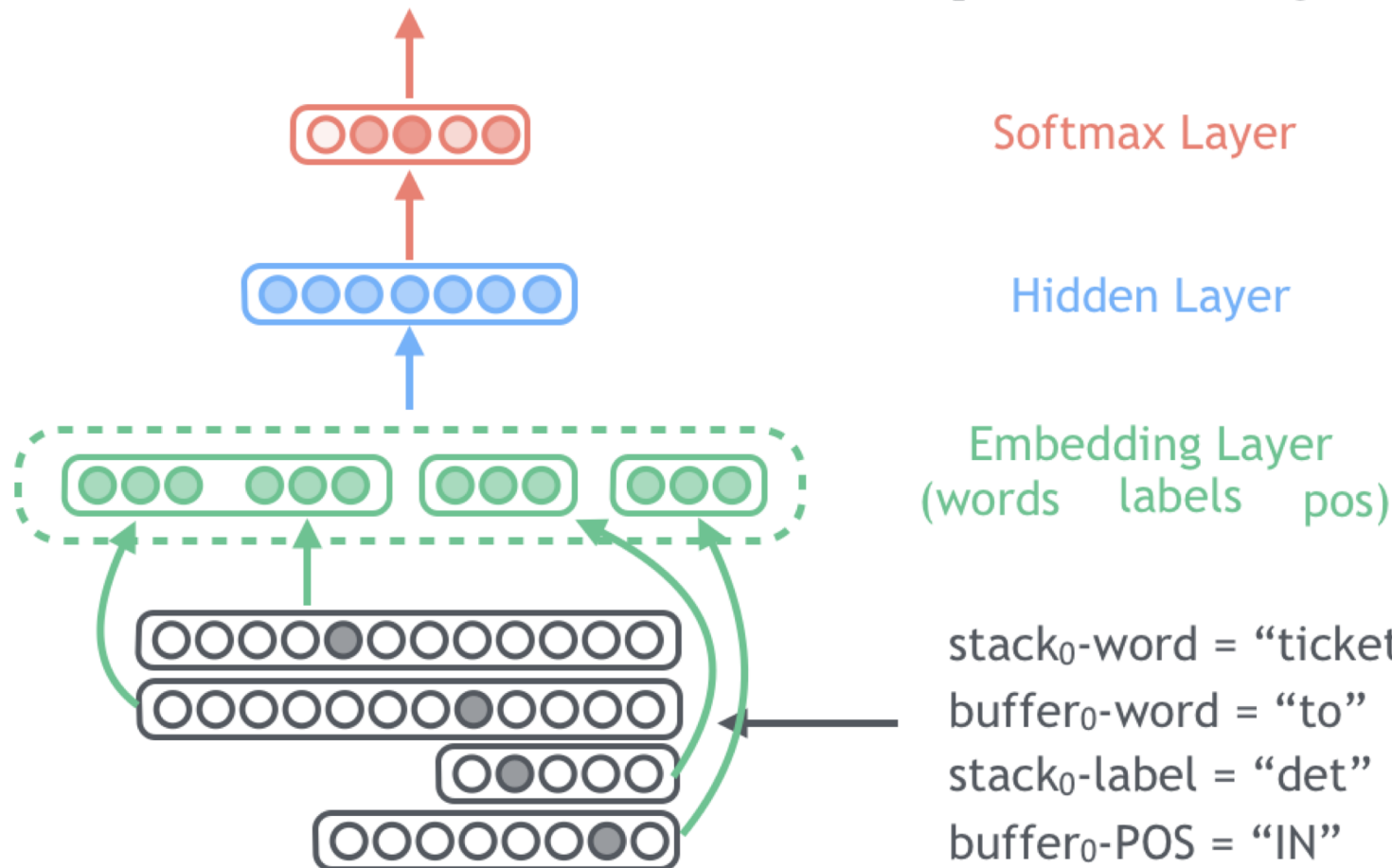
Chen and Manning (2014)

- Arc-standard Transitions
 - Shift
 - Left-Arc_r
 - Right-Arc_r
- Selecting the next actions:
 - Untyped choices: 3
 - Typed choices: $|R| * 2 + 1$
 - Neural network classifier
- With a few model improvements and very careful hyper-parameter tuning gives SOTA results

Parsing with Neural Networks

Chen and Manning (2014)

[Chen & Manning, 2014]





Hyperparameters?

- Regularization
- Loss function





Hyperparameters?

- Regularization
- Loss function
- Dimensions
- Activation function
- Initialization
- Adagrad
- Dropout





Hyperparameters?

- Regularization
- Loss function
- Dimensions
- Activation function
- Initialization
- Adagrad
- Dropout
- Mini-batch size
- Initial learning rate
- Learning rate schedule
- Momentum



- Stopping time
- Parameter averaging



Hyperparameters?



Slide from David Weiss



Hyperparameters?

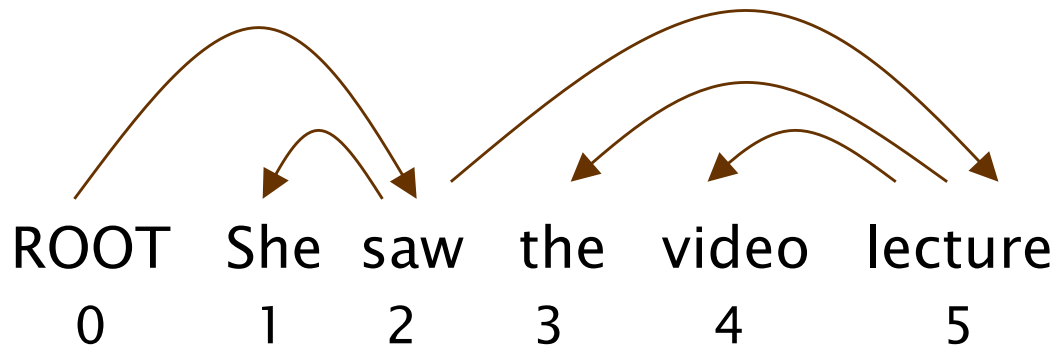
Use random restarts, grid search
Pick best using holdout data

Tune: WSJ S24 (grid search)

Dev: WSJ S22 (development)

Test: WSJ S23 (final results)

Evaluation



$$\text{Acc} = \frac{\# \text{ correct deps}}{\# \text{ of deps}}$$

$$\text{UAS} = 4 / 5 = 80\%$$

$$\text{LAS} = 2 / 5 = 40\%$$

Gold

1	2	She	nsubj
2	0	saw	root
3	5	the	det
4	5	video	nn
5	2	lecture	dobj

Parsed

1	2	She	nsubj
2	0	saw	root
3	4	the	det
4	5	video	nsubj
5	2	lecture	ccomp

Projectivity

- Dependencies from CFG trees with head rules must be projective
 - Crossing arcs are not allowed
- But: theory allows to account for displaced constituents → non-projective structures



Projectivity

- Arc-eager transition system:
 - Can't handle non-projectivity
- Possible directions:
 - Give up!
 - Post-processing
 - Add new transition types
 - Switch to a different algorithm
 - Graph-based parsers (e.g., MSTParser)