

### Natural Language Processing

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Chapter 11

### **Transition-based Methods for Structured Prediction**

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- 11.1 Transition-based Structured Prediction
  - 11.1.1 Greedy Local Modelling
  - 11.1.2 Structured Modelling
- 11.2 Transition-based Constituent Parsing
  - 11.2.1 Shift-reduce Constituent Parsing
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- 11.3 Shift-reduce Dependency Parsing
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  - 11.3.2 Arc-eager Projective Parsing
  - 11.3.3 The Swap Action and Non-projective Trees
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  - 11.4.1 Joint POS-tagging and Dependency Parsing
  - 11.4.2 Joint Word Segmentation, POS-tagging and Dependency Parsing

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### **VestlakeNLP**

#### • 11.1 Transition-based Structured Prediction

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- A different view from graph-based models for structured prediction
  - Encodes rich unbounded features without complexity constraints
  - A general framework that is easy to adapt to different tasks
- Maps output building process into a state transducer
  - State  $S_i$ 
    - Corresponds to partial results during decoding
  - Action *a<sub>i</sub>* 
    - The operations that can be applied for state transition
    - Construct output incrementally



- Automata:
  - State  $S_i$
  - Action  $a_i$





- Automata:
  - State  $S_i$
  - Action  $a_i$





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  - Action  $a_i$



#### Transition-based structured prediction **V**estlakeNLP

Transition-based structured prediction is a state transition process.

#### Automata:

- State
  - Start state an empty structure
  - End state —— the output structure
  - Intermediate state partially constructed structures
- Transition actions
  - Incremental steps that build output structures, change one state to another

#### **WestlakeNLP**

Input: "以(take) 前(before) 天(day) 下(fall) 雨(rain) 为(be)
例(example)"



Next Action: SEP

 $\sigma$ : partial output w: current partial word  $\beta$ : list of next incoming characters

Input: "以(take) 前(before) 天(day) 下(fall) 雨(rain) 为(be)
例(example)"



Next Action: SEP

 $\sigma$ : partial output w: current partial word  $\beta$ : list of next incoming characters

#### • Input: "以(take) 前(before) 天(day) 下(fall) 雨(rain) 为(be) 例(example)"



Next Action: APP

 $\sigma$ : partial output w: current partial word  $\beta$ : list of next incoming characters

#### **WestlakeNLP**

Input: "以(take) 前(before) 天(day) 下(fall) 雨(rain) 为(be)
例(example)"



Next Action: SEP

 $\sigma$ : partial output w: current partial word  $\beta$ : list of next incoming characters

• Input: "以(take) 前(before) 天(day) 下(fall) 雨(rain) 为(be) 例(example)"



Next Action: APP

 $\sigma$ : partial output w: current partial word  $\beta$ : list of next incoming characters

Input: "以(take) 前(before) 天(day) 下(fall) 雨(rain) 为(be)
例(example)"



Next Action: SEP

 $\sigma$ : partial output w: current partial word  $\beta$ : list of next incoming characters

## Input: "以(take) 前(before) 天(day) 下(fall) 雨(rain) 为(be) 例(example)"



Next Action: SEP

 $\sigma$ : partial output w: current partial word  $\beta$ : list of next incoming characters

# Input: "以(take) 前(before) 天(day) 下(fall) 雨(rain) 为(be) 例(example)"



Next Action: FIN

 $\sigma$ : partial output w: current partial word  $\beta$ : list of next incoming characters

## Input: "以(take) 前(before) 天(day) 下(fall) 雨(rain) 为(be) 例(example)"



 $\sigma$ : partial output w: current partial word  $\beta$ : list of next incoming characters









• NEXT ACTION: *Shift* 



#### **VestlakeNLP**

• NEXT ACTION: Left - Arc



#### **WestlakeNLP**

• NEXT ACTION: *Shift* 



#### **VestlakeNLP**

• NEXT ACTION: *Right – Arc* 



#### **VestlakeNLP**

• NEXT ACTION: *Reduce* 



#### **WestlakeNLP**

• NEXT ACTION: *Right – Arc* 



#### **VestlakeNLP**

• NEXT ACTION: Reduce



#### **WestlakeNLP**

• NEXT ACTION: Finish



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#### **Transition-based Modeling**



Given a state s<sub>i-1</sub>, our goal is to disambiguate all possible
actions a<sub>i</sub> ∈POSSIBLEACTIONS(s<sub>i-1</sub>) by using a discriminative
model to score transition actions:

$$score(a_i|s_{i-1}) = \vec{\theta} \cdot \vec{\phi}(s_{i-1}, a_i)$$

- $\vec{\theta}$ : model parameter vector
- $\vec{\phi}(s_{i-1}, a_i)$ : feature vector on the input-output pair  $(s_{i-1}, a_i)$

### **Greedy Local Method**

### **VestlakeNLP**

- Training
  - Training dataset  $D = \{(X_i, Y_i)\}|_{i=1}^N$

break down

• A sequence of gold transitions:  $(s_{j-1}^{(i)}, a_j^{(i)})$ 

merge all the state-action pairs

- A training set for the discriminative model
- Testing
  - Start from initial state  $s_0(X)$
  - Repeatedly find  $\hat{a}_i = argmax_{\alpha}\vec{\theta} \cdot \vec{\phi}(s_{i-1}, \alpha)$

#### **Greedy Local Method**



• NEXT ACTION: *Shift* 




• NEXT ACTION: Left - Arc





• NEXT ACTION: *Shift* 





• NEXT ACTION: *Right – Arc* 





• NEXT ACTION: *Reduce* 





• NEXT ACTION: *Right – Arc* 





• NEXT ACTION: Reduce





• NEXT ACTION: Finish





• An Example

\_\_\_\_ He does it here

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• An Example

#### $He does it here \longrightarrow He does it here$

























#### **Problem of Greedy Local Modeling**

**WestlakeNLP** 

- In a globally optimal action sequence, each action may not necessarily be the optimal choice locally.
- Result in error propagation.
- Model does not see incorrect states during training.

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# **Problem of Greedy Local Modeling**

**WestlakeNLP** 

- In a **globally** optimal action sequence, each action may not necessarily be the optimal choice **locally**.
- Result in error propagation.

#### Solution:

- Beam search use a globally trained model:
  - Given an input *X*, a global transition-based model calculates score(A|X) directly, where  $A_{1:|A|} = a_1a_2 \dots a_{|A|}$  is a sequence

of transition actions  $a_i$  for building an output structure for *X*.

















#### Solution

# **VestlakeNLP**

• Use global linear model to calculate *score*(*A*|*X*)

 $score(A|X) = \vec{\theta} \cdot \vec{\phi}(A, X)$   $\stackrel{\bullet}{\checkmark} decompose \vec{\phi}(A|X) \text{ for incremental decoding}$   $\stackrel{\bullet}{\Rightarrow} \vec{\phi}(A|X) = \sum_{i=1}^{|A|} \vec{\phi}(a_i, s_{i-1})$   $\stackrel{\bullet}{\Rightarrow} score(A|X) = \vec{\theta} \cdot \vec{\phi}(A, X) = \vec{\theta} \cdot \left(\sum_{i=1}^{|A|} \vec{\phi}(a_i, s_{i-1})\right)$   $= \sum_{i=1}^{|A|} \left(\vec{\theta}\right) \cdot \left\{\vec{\phi}(a_i, s_{i-1})\right\}$ 

# **Beam search decoding algorithm**

# **VestlakeNLP**

```
Inputs: \vec{\theta} —discriminative linear model parameters;
X - \text{task input};
K — beam size;
Initialization: agenda \leftarrow [(STARTSTATE(X), 0)];
Algorithm:
while not ALLTERMINAL(agenda) do
    to\_expand \leftarrow agenda;
   agenda \leftarrow [];
   for (state, score) \in to\_expand do
        for a \in \text{POSSIBLEACTIONS}(state) do
            new\_state \leftarrow EXPAND(state, a);
           new\_score \leftarrow score + \vec{\theta} \cdot \vec{\phi}(state, a);
            APPEND(agenda, (new_state, new_score));
    agenda \leftarrow \text{TOP-K}(agenda, K);
Output: TOP-K(agenda, 1)[0];
```











• Dependency parsing



AL–LEFT-ARC AR–RIGHT-ARC













# **VestlakeNLP**






## **Beam search decoding example**



Gold Sequence of Action:

## Beam search training algorithm

```
Inputs: D — gold standard training set;
K — beam size;
T — number of training iterations;
Initialisation: \vec{\theta} \leftarrow 0;
Algorithm:
for t \in [1, ..., T] do
    for (X, Y) \in D do
         G \leftarrow \text{GOLDACTIONSEQ}(X, Y);
         agenda \leftarrow [(STARTSTATE(X), 0)];
         gold\_state \leftarrow \text{STARTSTATE}(X);
         i \leftarrow 0;
         while not ALLTERMINAL(agenda) do
             i \leftarrow i + 1;
             to expand \leftarrow agenda ;
             agenda \leftarrow [];
             for (state, score) \in to expand do
                  for a \in \text{POSSIBLEACTIONS}(state) do
                      new\_state \leftarrow EXPAND(state, a);
                      new score \leftarrow score + \vec{\theta} \cdot \vec{\phi}(state, a);
                      APPEND(agenda, (new_state, new_score));
             agenda \leftarrow \text{TOP-K}(agenda, K);
             gold\_state \leftarrow EXPAND(gold\_state, G[i]);
             if not CONTAIN(agenda, gold_state) then
                  pos \leftarrow gold\_state;
                  neg \leftarrow \text{TOP-K}(agenda, 1)[0];
                  \vec{\theta} \leftarrow \vec{\theta} + \vec{\phi}(pos) - \vec{\phi}(neg);
                 continue((W, G) \in D)
         if gold\_state \neq TOP-K(agenda, 1)[0] then
             \vec{\theta} \leftarrow \vec{\theta} + \vec{\phi}(gold\_state) - \vec{\phi}(\text{TOP-K}(agenda, 1)[0]);
Output: \theta;
```

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- Three steps in transition-based modeling
  - Find the state transition process
  - Define global feature vector
  - Apply the standard learning and search framework

• State

 $\sigma$ : stack of partially constituent outputs.

 $\beta$ : buffer of the next incoming words.

actions

SHIFT, REDUCE-L/R-X, UNARY-X, IDLE



#### Shift-reduce constituent parsing

Axiom: Goal:	$([], W_{1:n}) \\ (\sigma, [])$	SHIFT:	$\frac{(\sigma, w_0 \beta)}{(\sigma w_0, \beta)}$
REDUCE-L-X:	$(\sigma s_1 s_0, \beta)$		$(\sigma s_1 s_0, \ \beta)$
	Х	REDUCE-R-X:	Х
	$(\sigma   \checkmark \land \beta)$		$(\sigma   \land , \beta)$
	$s_1$ $s_0$		$s_1$ $s_0$
UNARY-X:	$(\sigma s_0,\ eta)$		$(\sigma, [])$
	Х	Idle:	
	$(\sigma   \downarrow, eta)$		$(\sigma, [])$
	$s_0$		

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# Transition-based constituent parsing Use VestlakeNLP



#### Shift-reduce constituent parsing

- Example ٠
  - Shift



bu	ffer
υu	1101

DT	ADJ	NN	VV	ADJ	NNS	
The	little	boy	likes	red t	omatoes	

#### Shift-reduce constituent parsing

- Example
  - Shift



#### Shift-reduce constituent parsing

- Example
  - Shift



#### Shift-reduce constituent parsing

- Example
  - Reduce-R-NP



#### **WestlakeNLP**

#### Shift-reduce constituent parsing

- Example
  - Reduce-R-NP







#### Shift-reduce constituent parsing

- Example
  - Shift







#### **WestlakeNLP**

#### Shift-reduce constituent parsing

- Example ٠
  - Shift



buffer

#### Shift-reduce constituent parsing

- Example ٠
  - Shift



buffer

### **WestlakeNLP**

#### Shift-reduce constituent parsing

- Example
  - Reduce-R-NP



#### Shift-reduce constituent parsing

- Example
  - Reduce-L-VP





#### Shift-reduce constituent parsing

- Example
  - Shift



buffer

#### **WestlakeNLP**

#### Shift-reduce constituent parsing

- Example
  - Reduce-L-S



buffer

#### Shift-reduce constituent parsing

- Example
  - Reduce-R-S





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## **WestlakeNLP**

#### Shift-reduce constituent parsing

- Example
  - Terminate stack



buffer

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#### 

- Feature Templates for Shift-reduce Constituent Parser
- Example state



- **WestlakeNLP**
- Feature Templates for Shift-reduce Constituent Parser

Feature type	Feature Template
unigrams	$s_0ps_0c, s_0ws_0c, s_1ps_1c, s_1ws_1c, s_2ps_2c, s_2ws_2c, s_3ps_3c,$
	$s_3ws_3c$ , $b_0wb_0p$ , $b_1wb_1p$ , $b_2wb_2p$ , $b_3wb_3p$ , $s_{0.l}ws_{0.l}c$ ,
	$s_{0.r}ws_{0.r}c, s_{0.u}ws_{0.u}c, s_{1.l}ws_{1.l}c, s_{1.r}ws_{1.r}c, s_{1.u}ws_{1.u}c$
bigrams	$s_0ws_1w, s_0ws_1c, s_0cs_1w, s_0cs_1c, s_0wb_0w, s_0wb_0p, s_0cb_0w,$
	$s_0cb_0p, b_0wb_1w, b_0wb_1p, b_0pb_1w, b_0pb_1p, s_1wb_0w, s_1wb_0p,$
	$s_1 c b_0 w,  s_1 c b_0 p$
trigrams	$s_0cs_1cs_2c, \ s_0ws_1cs_2c, \ s_0cs_1wb_0p, \ s_0cs_1cs_2w, \ s_0cs_1cb_0p,$
	$s_0ws_1cb_0p,  s_0cs_1wb_0p,  s_0cs_1cb_0w$

*s<sub>i</sub>*: top node of the stack; *b<sub>i</sub>*: front word on the buffer; *xw*: the word form of *x*; *xp*: the POS tag; *xc*: the constituent label of a non-terminal node *x*; *xl*, *xr*, *xu*: the left child, the right child and the unary child of *x*, respectively.



- Feature Templates for Shift-reduce Constituent Parser
- Example

In the 10-th step of the example sentence "*The little boy likes red tomatoes*", the  $s_0 ps_0 c$  feature is VBZ | VP and the  $s_1 ps_1 c$  feature is NN | NP.



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## **WestlakeNLP**

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• Projective dependency tree



• Non-projective dependency tree



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- Arc-standard Parsing
  - State

 $\sigma$ : stack of partially dependency outputs.

 $\beta$ : buffer of the next incoming words.

• actions

SHIFT, LEFT-ARC-X, RIGHT-ARC-X

## **VestlakeNLP**

#### Actions



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#### Arc-standard Dependency Parsing



State:  $(\sigma, \beta, A)$ ;  $\sigma$ : stack;  $\beta$ : buffer; A: the set of dependency arcs that have been constructed



Example

Sentence "He gave her a tomato"

Next action: SHIFT

He gave her a tomato



Example

Sentence "He gave her a tomato"

Next action: SHIFT

He

gave her a tomato



Example

Sentence "He gave her a tomato"

Next action: LEFT-ARC-SUBJ

He gave

her a tomato



Example

Sentence "He gave her a tomato"

Next action: SHIFT



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Example

Sentence "He gave her a tomato"

Next action: RIGHT-ARC-IOBJ





Example

Sentence "He gave her a tomato"

Next action: SHIFT



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Example

Sentence "He gave her a tomato"

Next action: SHIFT




Example

Sentence "He gave her a tomato" Next action: LEFT-ARC-DET



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Example

Sentence "He gave her a tomato" Next action: RIGHT-ARC-DOBJ





Example

Sentence "He gave her a tomato"

Next action: END





Arc-standard Dependency Parsing

• Example state



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#### **WestlakeNLP**

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#### **Feature Templates for arc-standard dependency parsing**

Feature Type	Feature Template	Feature Type	Feature Template
from single words	$s_0wp; s_0w; s_0p; b_0wp;$	valency	$s_0wv_r; s_0pv_r; s_0wv_l;$
	$b_0w; b_0p; b_1wp; b_1w;$		$s_0 p v_l; b_0 w v_l; b_0 p v_l;$
	$b_1p; b_2wp; b_2w; b_2p;$		
from word pairs	$s_0wpb_0wp; s_0wpb_0w;$	unigrams	$s_{0.h}w; s_{0.h}p; s_{0.l};$
	$s_0wpb_0p; s_0wb_0wp;$		$s_{0.l}w; s_{0.l}p; s_{0.l}l;$
	$s_0pb_0wp; s_0wb_0w;$		$s_{0.r}w; s_{0.r}p; s_{0.r}l;$
	$s_0 p b_0 p; b_0 p b_1 p;$		$b_{0.l}w; b_{0.l}p; b_{0.l}l;$
from three words	$b_0pb_1pb_2p; \ s_0pb_0pb_1p;$	third-order	$s_{0.h_2}w; s_{0.h_2}p; s_{0.h}l;$
	$s_{0.h}ps_0pb_0p;$		$s_{0,l_2}p; s_{0,l_2}l; s_{0,r_2}w;$
	$s_0ps_{0.l}pb_0p;$		$s_{0.r_2}p; s_{0.r_2}l; b_{0.l_2}w;$
	$s_0 p s_{0.r} p b_0 p;$		$b_{0.l_2}p; b_{0.l_2}l;$
	$s_0pb_0pb_{0.l}p;$		$s_0 p s_{0.l} p s_{0.l_2} p;$
			$s_0 p s_{0.r} p s_{0.r_2} p;$
			$s_0 p s_{0.h} p s_{0.h_2} p;$
			$b_0 p b_{0.l} p b_{0.l_2} p;$
distance	$s_0wd; s_0pd; b_0wd;$	label set	$s_0ws_r; s_0ps_r; s_0ws_l;$
	$b_0pd; s_0wb_0wd;$		$s_0 p s_l; n_0 w s_l; n_0 p s_l$
	$s_0 p b_0 p d;$		

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- Arc-eager Parsing
  - State

 $\sigma$ : stack of partially constituent outputs.

 $\beta$ : buffer of the next incoming words.

• actions

SHIFT, LEFT-ARC-X, RIGHT-ARC-X, REDUCE

# **VestlakeNLP**

Arc-eager dependency parsing

• Example



# **WestlakeNLP**



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#### Arc-eager dependency parsing



Main differences comparing with arc-standard:

- Left Arc X/Right Arc X: construct a dependency arc from the front word on the buffer to the top word on the stack, but not as that (from the top two words on the stack) in arc-standard.
- *Reduce*: pop the top word off the stack.

**WestlakeNLP** 



• NEXT ACTION: *Shift* 



# **VestlakeNLP**

• NEXT ACTION: Left - Arc - SUBJ



# **WestlakeNLP**

• NEXT ACTION: *Shift* 



# **VestlakeNLP**

• NEXT ACTION: *Right – Arc – OBJ* 



# **VestlakeNLP**

• NEXT ACTION: *Reduce* 



# **VestlakeNLP**

• NEXT ACTION: *Right – Arc – MOD* 



# **VestlakeNLP**

• NEXT ACTION: Reduce



# **WestlakeNLP**

• NEXT ACTION: Finish



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- The Swap Action and Non-projective Trees
- To allow constructing non-projective trees, the arc-standard system can be extended by adding a new action:
  - *Swap*: remove the second top word from the stack, pushing

it onto the buffer front.



#### Arc-standard Dependency Parsing with Swap Action



*State*:  $(\sigma, \beta, A)$ ;  $\sigma$ : stack;  $\beta$ : buffer; A: the set of dependency arcs that have been constructed; *IDX*(*w*): return the index of *w* in the sentence  $W_{1:n}$ .

**WestlakeNLP** 



Arc-standard Dependency Parsing with Swap Action

• Example

Next action: SHIFT

A hearing was scheduled on this today

#### **WestlakeNLP**

#### Arc-standard Dependency Parsing with Swap Action

• Example

Next action: SHIFT

hearing was scheduled on this today

А



#### Arc-standard Dependency Parsing with Swap Action

• Example

Next action: LEFT-ARC-DET

A hearing

was scheduled on this today

#### **WestlakeNLP**

#### Arc-standard Dependency Parsing with Swap Action

• Example

Next action: SHIFT



was scheduled on this today

### **WestlakeNLP**

#### Arc-standard Dependency Parsing with Swap Action

• Example

Next action: SHIFT



scheduled on this today

#### **WestlakeNLP**

#### Arc-standard Dependency Parsing with Swap Action

• Example

Next action: SHIFT



on this today

#### **WestlakeNLP**

#### Arc-standard Dependency Parsing with Swap Action

• Example

Next action: SWAP



this today

#### **WestlakeNLP**

#### Arc-standard Dependency Parsing with Swap Action

• Example

Next action: SWAP



scheduled this today

#### **WestlakeNLP**

#### Arc-standard Dependency Parsing with Swap Action

• Example

Next action: SHIFT



was scheduled this today

### **WestlakeNLP**

#### Arc-standard Dependency Parsing with Swap Action

• Example

Next action: SHIFT



scheduled this today

#### **WestlakeNLP**

#### Arc-standard Dependency Parsing with Swap Action

• Example

Next action: SHIFT



this today

#### **WestlakeNLP**

#### Arc-standard Dependency Parsing with Swap Action

• Example

Next action: SWAP



#### **WestlakeNLP**

#### Arc-standard Dependency Parsing with Swap Action

• Example

Next action: SWAP



scheduled today

- Arc-standard Dependency Parsing with Swap Action
- Example

Next action: RIGHT-ARC-POBJ



WestlakeNLP

# **WestlakeNLP**

#### Arc-standard Dependency Parsing with Swap Action

• Example

Next action: RIGHT-ARC-NMOD



was scheduled today
### **WestlakeNLP**

#### Arc-standard Dependency Parsing with Swap Action

• Example

Next action: SHIFT



was scheduled today

### **WestlakeNLP**

#### Arc-standard Dependency Parsing with Swap Action

• Example

Next action: LEFT-ARC-SUBJ



# **WestlakeNLP**

#### Arc-standard Dependency Parsing with Swap Action

• Example

Next action: SHIFT



# **WestlakeNLP**

Arc-standard Dependency Parsing with Swap Action

• Example

Next action: SHIFT





Arc-standard Dependency Parsing with Swap Action

• Example

Next action: RIGHT-ARC-TMP



### **WestlakeNLP**

Arc-standard Dependency Parsing with Swap Action

• Example

Next action: LEFT-ARC-VG



# **WestlakeNLP**

Arc-standard Dependency Parsing with Swap Action

Next action: END



### Contents

# **WestlakeNLP**

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# **WestlakeNLP**

- Motivations
  - Cross-task information sharing
  - Reduction of error propagation
- Example
  - For joint POS-tagging and syntactic parsing, the *SHIFT* action of arc-standard algorithm can be replaced with *SHIFT X* action, where *X* refers to the POS label.

# **VestlakeNLP**

### Joint POS-tagging and Dependency Parsing

- Input:  $W_{1:n} = w_1, ..., w_n$
- Output:  $(T_{1:n}, A)$ , where  $t_i$  is POS for  $w_i$ .
- The arc-standard algorithm can be extended by replacing *SHIFT* action with *SHIFT X* action
  - SHIFT X, which removes the front word from the buffer, assigning the POS label X to the word, and pushing it onto the stack.



#### Joint POS-tagging and Dependency Parsing

• Example

Next action: SHIFT—PRP

Stack [S]

Buffer [B]

 $He_1 won_2 the_3 game_4$ 



#### Joint POS-tagging and Dependency Parsing

• Example

Next action: SHIFT—VBD





#### Joint POS-tagging and Dependency Parsing

• Example

Next action: LEFT—ARC—NSUBJ

Stack [S]

Buffer [B]

 $He_{1/PRP}$  won<sub>2</sub>

 $the_3$  game<sub>4</sub>

# **VestlakeNLP**

#### Joint POS-tagging and Dependency Parsing

• Example

Next action: SHIFT——DT





#### Joint POS-tagging and Dependency Parsing

• Example

Next action: SHIFT——NN





#### Joint POS-tagging and Dependency Parsing

• Example

Next action: LEFT—ARC—DET



# **VestlakeNLP**

#### Joint POS-tagging and Dependency Parsing

• Example

Next action: RIGHT—ARC—DOBJ



# **VestlakeNLP**

#### Joint POS-tagging and Dependency Parsing

• Example

Next action: END





#### Joint Word Segmentation, POS-tagging and Dependency Parsing

- Input: $C_{1:n} = C_1 \dots C_n$
- Output:  $(W_{1:m}, T_{1:m}, A)$  where  $w_1 \dots w_m$  are words,  $t_1 \dots t_n$  are POS, and A is the set of dependency arcs.
- State:
  - $\sigma$  -- partially built outputs
  - $\delta$  -- words
  - $\beta$  -- incoming characters
  - $A_c$  -- character dependencies
  - $A_w$ -- word dependencies

# **VestlakeNLP**

#### Joint Word Segmentation, POS-tagging and Dependency Parsing

	Axiom:		$([], [], C_{1:n}, \phi, \phi)$	$\phi$ )
	Goal:	(	$[S_0], [], [], A_c, A$	$(1_w)$
	LEET-ABC-C'	$(\sigma, \ \delta   d_0, \ b_0   \beta, \ A_a)$	$(A_w)$ such that $\neg$	$(\exists d \in \delta, d^{\frown} d_0 \in A_c)$
	LEF I-AIO-O.	$(\sigma, \delta,$	$b_0 \beta, A_c \cup \{d_0 \frown b$	$_{0}\}, A_{w})$
	Left-arc-X:	$(\sigma s_0, [d_0], \beta, A_c)$	$(A_w)$ such that $\neg$	$(\exists s \in \sigma, s \stackrel{l}{\frown} s_0 \in A_w)$
		$(\sigma, \lfloor d$	$[0], \ \beta, \ A_c, \ A_w \cup \{s\}$	$(a_0^{\psi} d_0))$
	Shift: -	/	$(\sigma, [d_0], \beta, A_c, A$	
			$\frac{\sigma d_0, [], \beta, A_c, A}{\sigma d_0, [], \beta, A_c, A}$	$\mathbf{I}_w$ )
	SHIFT-C:		$\frac{(\sigma, \ \delta, \ b_0 \beta, \ A_c, \ A}{(\sigma, \ \delta, \ b_0 \beta, \ A_c, \ A_c)}$	w)
			$(\sigma, \delta   b_0, \beta, A_c, A$	$\frac{w}{d}$
	RIGHT-ARC-C: -		$\sigma, \ \delta   d_0, \ b_0   \beta, \ A_c, \ A$	$A_w$
		$(\sigma, \delta   d_0)$	$b_0   b_0, \ \beta, \ A_c \cup \{d_0'\}$	$\{b_0\}, A_w$
	RIGHT-ARC-X:	$(\sigma s_0, [$	$[d_0], \beta, A_c, A_w \cup \cdot$	$\frac{\{s_0 \land \forall d_0\}}{\langle A \rangle}$
		$(\sigma$	$\frac{ s_0 d_0, [], \beta, A_c,}{ \beta   \beta   \beta   \beta   \beta   \beta }$	$\frac{A_w}{2}$
	Pop-X:		$(\sigma, [d_0], \beta, A_c, A$	(w)
		$(\sigma, [SUB])$	$\frac{\operatorname{FREE}(a_0, A_c)/X}{4}$	$\frac{5}{1-5}, \frac{A_c}{A_c}, \frac{A_w}{A_w}$
	REDUCE-C:	$(\sigma, \ \delta   d_0, \ \beta, \ A_c$	$\frac{A_w}{(\Sigma_{w})}$ such that $\exists a$	$\underbrace{d \in \delta, \ d' \ \ }_{O} d \in A_c}_{O}$
			$(\sigma, 0, \beta, A_c, A_w$	)
	REDUCE:	$(\sigma s_0, \ \delta, \ \beta, \ A_c$	$(\sigma, \delta, \beta, A_c, A_w)$ such that $\exists (\sigma, \delta, \beta, A_c, A_w)$	$\frac{s \in \sigma, s' * s_0 \in A_w}{)}$
State: (σ,	$\delta, \beta, A_c, A_w$	) $\sigma$ : stack;	β:buffer;	$\delta$ : partial-word buffer;

 $A_c$ : the set of character dependencies;

 $A_w$ : the set of word dependencies <sup>164</sup>



#### Joint Word Segmentation, POS-tagging and Dependency Parsing

• Example

Step	$\sigma$	δ	β	$A_c$	$A_w$	Action
0			我,来,到,会,客,室			SHIFT-C

我, 来, 到, 会, 客, 室



#### Joint Word Segmentation, POS-tagging and Dependency Parsing

Step	$\sigma$	δ	eta	$A_c$	$A_w$	Action
1		我	来,到,会,客,室			POP-PN

我	来, 到, 会, 客, 室



#### Joint Word Segmentation, POS-tagging and Dependency Parsing

Step	σ	δ	$\beta$	$A_c$	$A_w$	Action
2		我/PN	来,到,会,客,室			SHIFT-W

我/PN	来, 到, 会, 客, 室



#### Joint Word Segmentation, POS-tagging and Dependency Parsing

Step	σ	δ	β	$A_c$	$A_w$	Action
3	我/PN		来,到,会,客,室			SHIFT-C





#### Joint Word Segmentation, POS-tagging and Dependency Parsing

Step	σ	δ	β	$A_c$	$A_w$	Action
4	我/PN	来	到, 会, 客, 室			LEFTARC-C

我/PN	来	到, 会, 客, 室



#### Joint Word Segmentation, POS-tagging and Dependency Parsing

Step	σ	δ	β	$A_c$	$A_w$	Action
5	我/PN		到, 会, 客, 室	来← 到		SHIFT





#### Joint Word Segmentation, POS-tagging and Dependency Parsing

Step	$\sigma$	δ	eta	$A_c$	$A_w$	Action
6	我/PN	到	到, 会, 客, 室	来← 到		POP-VV



#### Joint Word Segmentation, POS-tagging and Dependency Parsing

Step	σ	δ	β	$A_c$	$A_w$	Action
7	我/PN	来到/VV	会, 客, 室	来← 到		LEFTARC- $SUBJ$





#### Joint Word Segmentation, POS-tagging and Dependency Parsing

Step	σ	δ	β	$A_c$	$A_w$	Action
8		来到/VV	会, 客, 室	来← 到	我/PN← <i>SUBJ</i> 来到/VV	SHIFT





#### Joint Word Segmentation, POS-tagging and Dependency Parsing

Step	σ	δ	β	$A_c$	$A_w$	Action
9	来到/VV		会, 客, 室	来← 到	我/PN← <i>SUBJ</i> 来到/VV	SHIFT-C





#### Joint Word Segmentation, POS-tagging and Dependency Parsing

Step	σ	δ	β	$A_c$	$A_w$	Action
10	来到/VV	会	客,室	来← 到	我/PN← <i>SUBJ</i> 来到/VV	RIGHTARC-C





#### Joint Word Segmentation, POS-tagging and Dependency Parsing

Step	σ	δ	β	$A_c$	$A_w$	Action
11	来到/VV	会,客	室	来← 到, 会→客	我/PN← <i>subj</i> 来到/VV	REDUCE-C





#### Joint Word Segmentation, POS-tagging and Dependency Parsing

Step	σ	δ	β	$A_c$	$A_w$	Action
12	来到/VV	会	室	来← 到, 会→客	我/PN← SUBJ 来到/VV	LEFTARC-C





#### Joint Word Segmentation, POS-tagging and Dependency Parsing

Step	σ	δ	β	$A_c$	$A_w$	Action
13	来到/VV		室	来← 到, 会→客, 会←室	我/PN← <i>SUBJ</i> 来到/VV	SHIFT-C





#### Joint Word Segmentation, POS-tagging and Dependency Parsing

• Example

Step	σ	δ	β	$A_c$	$A_w$	Action
14	来到/VV	室		来← 到, 会→客, 会←室	我/PN← SUBJ 来到/VV	POP-NN



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#### Joint Word Segmentation, POS-tagging and Dependency Parsing

Step	σ	δ	β	$A_c$	$A_w$	Action
15	来 到/VV	会客 室/NN		来← 到, 会→客, 会 ←室	我/PN <i>← SUBJ</i> 来 到/VV	RIGHTARC-DOBJ


# **Joint Models**



### Joint Word Segmentation, POS-tagging and Dependency Parsing

• Example

Step	σ	δ	β	$A_c$	$A_w$	Action
16	来到/VV,会客 室/NN			来← 到, 会→ 客, 会←室	我/PN← <i>SUBJ</i> 来到/VV,来到/VV <i>DOBJ</i> →会客室/NN	REDUCE



# **Joint Models**



### Joint Word Segmentation, POS-tagging and Dependency Parsing

• Example

Step	σ	δ	β	$A_c$	$A_w$	Action
17	来 到/VV			来← 到, 会→客, 会 ←室	我/PN← <i>SUBJ</i> 来到/VV,来到/VV <i>DOBJ</i> →会客 室/NN	END



### Summary



- What is transition-based method?
- Apply transition-based methods on different tasks.
- Joint modeling with transition-based methods.