## 凹 WestlakeNLP

## Natural Language Processing

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## 凹 WestlakeNLP

## Chapter 11

## Transition-based Methods for Structured Prediction

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- 11.1 Transition-based Structured Prediction
- 11.1.1 Greedy Local Modelling
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- 11.2 Transition-based Constituent Parsing
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- 11.2.2 Feature Templates
- 11.3 Shift-reduce Dependency Parsing
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- 11.3.3 The Swap Action and Non-projective Trees
- 11.4 Joint Models
- 11.4.1 Joint POS-tagging and Dependency Parsing
- 11.4.2 Joint Word Segmentation, POS-tagging and Dependency Parsing


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## A Transition System

- 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 - $a_{i}$
- The operations that can be applied for state transition
- Construct output incrementally


## A Transition System

- Automata:
- State - $S_{i}$
- Action - $a_{i}$


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## Transition-based structured prediction § WestlakeNLP

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


## Example（Word Segmentation）

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


Next Action：SEP
$\sigma$ ：partial output $\quad w$ ：current partial word $\quad \beta$ ：list of next incoming characters

## Example（Word Segmentation）

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


Next Action：SEP
$\sigma$ ：partial output $\quad w$ ：current partial word $\quad \beta$ ：list of next incoming characters

## Example（Word Segmentation）

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


Next Action：APP
$\sigma$ ：partial output $\quad w$ ：current partial word $\quad \beta$ ：list of next incoming characters

## Example（Word Segmentation）

－Input：＂以（take）前（before）天（day）下（fall）雨（rain）为（be）
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## Example（Word Segmentation）

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## Example（Word Segmentation）

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


Next Action：SEP
$\sigma$ ：partial output $\quad w$ ：current partial word $\quad \beta$ ：list of next incoming characters

## Example（Word Segmentation）

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


Next Action：FIN
$\sigma$ ：partial output $\quad w$ ：current partial word $\quad \beta$ ：list of next incoming characters

## Example（Word Segmentation）

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

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

## Example (Dependency Parsing)



## Example (Dependency Parsing)


$\sigma$ : stack $\quad \beta$ : buffer

## Example (Dependency Parsing)

- NEXT ACTION: Shift



## Example (Dependency Parsing)

- NEXT ACTION: Left - Arc



## Example (Dependency Parsing)

- NEXT ACTION: Shift



## Example (Dependency Parsing)

- NEXT ACTION: Right - Arc



## Example (Dependency Parsing)

- NEXT ACTION: Reduce



## Example (Dependency Parsing)

- NEXT ACTION: Right - Arc



## Example (Dependency Parsing)

- NEXT ACTION: Reduce



## Example (Dependency Parsing)

- NEXT ACTION: Finish



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## Transition-based Modeling

- Given a state $s_{i-1}$, our goal is to disambiguate all possible actions $\mathrm{a}_{\mathrm{i}} \in \operatorname{POSSIBLEACTIONS}\left(s_{i-1}\right)$ by using a discriminative model to score transition actions:

$$
\operatorname{score}\left(a_{i} \mid s_{i-1}\right)=\vec{\theta} \cdot \vec{\phi}\left(s_{i-1}, a_{i}\right)
$$

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


## Greedy Local Method

- Training
- Training dataset $D=\left.\left\{\left(X_{i}, Y_{i}\right)\right\}\right|_{i=1} ^{N}$
$\checkmark$ break down
- A sequence of gold transitions: $\left(s_{j-1}^{(i)}, a_{j}^{(i)}\right)$
merge all the state-action pairs
- A training set for the discriminative model
- Testing
- Start from initial state $s_{0}(X)$
- Repeatedly find $\widehat{a_{i}}=\operatorname{argmax}_{\alpha} \vec{\theta} \cdot \vec{\phi}\left(s_{i-1}, \alpha\right)$


# Greedy Local Method 

- NEXT ACTION: Shift



## Greedy Local Method

- NEXT ACTION: Left - Arc

$\sigma$ : stack $\quad \beta$ : buffer


## Greedy Local Method

- NEXT ACTION: Shift



## Greedy Local Method

- NEXT ACTION: Right - Arc



## Greedy Local Method

- NEXT ACTION: Reduce



## Greedy Local Method

- NEXT ACTION: Right - Arc



## Greedy Local Method

- NEXT ACTION: Reduce



## Greedy Local Method

- NEXT ACTION: Finish



## Greedy Local Method

- An Example

He does it here

## Greedy Local Method

- An Example

He does it here $\longrightarrow S \quad \mathrm{He}$ does it here

## Greedy Local Method

- An Example



## Greedy Local Method

- An Example



## Greedy Local Method

- An Example



## Greedy Local Method

- An Example



## Greedy Local Method

- An Example



## Greedy Local Method

- An Example



## Problem of Greedy Local Modeling

- 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

- 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 $\operatorname{score}(A \mid X)$ directly, where $A_{1:|A|}=a_{1} a_{2} \ldots a_{|A|}$ is a sequence of transition actions $a_{i}$ for building an output structure for $X$.


## Beam search decoding example



## Beam search decoding example



## Beam search decoding example



## Beam search decoding example



## Beam search decoding example



## Beam search decoding example



## Beam search decoding example



## Beam search decoding example



## Solution

- Use global linear model to calculate $\operatorname{score}(A \mid X)$

$$
\begin{aligned}
& \operatorname{score}(A \mid X)=\vec{\theta} \cdot \vec{\phi}(A, X) \\
& \operatorname{decompose} \vec{\phi}(A \mid X) \text { for incremental decoding } \\
& \operatorname{score}(A \mid X)=\vec{\theta} \cdot \vec{\phi}(A, X)=\vec{\theta} \cdot\left(\sum_{i=1}^{|A|} \vec{\phi}\left(a_{i}, s_{i-1}\right)\right. \\
& =\sum_{i=1}^{|A|}(\vec{\theta}\} \cdot\left\{\vec{\phi}\left(a_{i}, s_{i-1}\right)\right) \\
&
\end{aligned}
$$

## Beam search decoding algorithm

```
Inputs: \(\vec{\theta}\)-discriminative linear model parameters;
\(X\) - task input;
\(K\) - beam size;
Initialization: agenda \(\leftarrow[(\operatorname{StartState}(X), 0)]\);
Algorithm:
while not AllTerminal(agenda) do
    to_expand \(\leftarrow\) agenda;
    agenda \(\leftarrow[]\);
    for \((\) state, score \() \in\) to_expand do
        for \(a \in \operatorname{PossibleActions(state)~do~}\)
            new_state \(\leftarrow \operatorname{EXPAND}(\) state,\(a)\);
            new_score \(\leftarrow\) score \(+\vec{\theta} \cdot \vec{\phi}(\) state,\(a)\);
            Append (agenda, (new_state, new_score));
        agenda \(\leftarrow\) Top-к \((\) agenda, \(K)\);
Output: Top-K(agenda, 1)[0];
```


## Beam search decoding example

- Dependency parsing



## Beam search decoding example

- Dependency parsing


[^0]
## Beam search decoding example

- Dependency parsing


AL-LEFT-ARC
AR-RIGHT-ARC

## Beam search decoding example

- Dependency parsing



## Beam search decoding example

- Dependency parsing



## Beam search decoding example

- Dependency parsing



## Beam search decoding example

- Dependency parsing



## Beam search decoding example

- Dependency parsing



## 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{0}\leftarrow0\mathrm{ ;
Algorithm:
for }t\in[1,\ldots,T]\mathrm{ do
    for (X,Y)\inD do
    G\leftarrowGoldActionSeq( }X,Y)
    agenda }\leftarrow[(StartState (X),0)]
    gold_state }\leftarrow\operatorname{StartSTATE}(X)
    i\leftarrow0;
    while not AllTerminal(agenda) do
        i\leftarrowi+1;
        to_expand }\leftarrow\mathrm{ agenda ;
        agenda }\leftarrow[]
        for (state, score) \in to_expand do
            for }a\in\operatorname{PossibleActions(state) do
                new_state \leftarrow EXPAND(state,a);
                new_score }\leftarrow\mathrm{ score + }\vec{0}\cdot\vec{\phi}(\mathrm{ state,a);
                APPEND(agenda,(new__state,new_score));
        agenda}\leftarrow\mathrm{ TOP-K(agenda,K);
        gold_state }\leftarrow\operatorname{ExPAND}(gold_state,G[i])
        if not Contain(agenda,gold__state) then
            pos }\leftarrow\mathrm{ gold__state;
            neg}\leftarrow\mathrm{ Top-k(agenda, 1)[0];
            \vec{0}}\leftarrow\vec{0}+\vec{\phi}(\mathrm{ pos ) - 房(neg);
            continue((W, G) \inD)
        if gold_state }\not=\mathrm{ TOP-K (agenda, 1)[0] then
        \vec{0}\leftarrow\vec{0}+\vec{\phi}(\mathrm{ gold_state ) - }\vec{\phi}(\mathrm{ TOP-K(agenda,1)[0]);}
```


## Contents

- 11.2 Transition-based Constituent Parsing


## Transition-based constituent parsing

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


## Transition-based constituent parsing

- State
$\sigma$ : stack of partially constituent outputs.
$\beta$ : buffer of the next incoming words.
- actions

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

## Transition-based constituent parsing

Shift-reduce constituent parsing

| Axiom: Goal: | $\underset{(\sigma,[])}{\left([], W_{1: n}\right)}$ | Shift: | $\frac{\left(\sigma, w_{0} \mid \beta\right)}{\left(\sigma \mid w_{0}, \beta\right)}$ |
| :---: | :---: | :---: | :---: |
|  | $\left(\sigma\left\|s_{1}\right\| s_{0}, \beta\right)$ |  | $\left(\sigma\left\|s_{1}\right\| s_{0}, \beta\right)$ |
| Reduce-l-x: |  | Reduce-R-x: |  |
|  | $\left(\sigma \mid s_{0}, \beta\right)$ |  | ( $\sigma,[$ ]) |
| Unary-x: | $\begin{gathered} \mathrm{X} \\ \left(\sigma \left\lvert\, \begin{array}{c} \mathrm{X} \\ \downarrow \\ s_{0} \end{array}\right., \beta\right) \end{gathered}$ | IdLe: | ( $\sigma,[])$ |

# Transition-based constituent parsing 

Shift-reduce constituent parsing

- Example
- Shift
stack

buffer



# Transition-based constituent parsing 

Shift-reduce constituent parsing

- Example
- Shift
stack

buffer



# Transition-based constituent parsing 

Shift-reduce constituent parsing

- Example
- Shift
stack

buffer



## Transition-based constituent parsing

Shift-reduce constituent parsing

- Example
- Reduce-R-NP



# Transition-based constituent parsing 

Shift-reduce constituent parsing

- Example
- Reduce-R-NP
stack

buffer



# Transition-based constituent parsing 

Shift-reduce constituent parsing

- Example
- Shift
stack

buffer



## Transition-based constituent parsing

Shift-reduce constituent parsing

- Example
- Shift
stack

buffer



# Transition-based constituent parsing 

Shift-reduce constituent parsing

- Example
- Shift
stack

buffer



## Transition-based constituent parsing

Shift-reduce constituent parsing

- Example
- Reduce-R-NP
stack

buffer
$\qquad$ -


## Transition-based constituent parsing

Shift-reduce constituent parsing

- Example
- Reduce-L-VP
stack

buffer
$\qquad$


# Transition-based constituent parsing 

Shift-reduce constituent parsing

- Example
- Shift
stack

buffer
$\qquad$

|
$\qquad$


## Transition-based constituent parsing

Shift-reduce constituent parsing

- Example
- Reduce-L-S
stack

buffer
$\qquad$
$\qquad$


## Transition-based constituent parsing

Shift-reduce constituent parsing

- Example
- Reduce-R-S

$\qquad$


## Transition-based constituent parsing

Shift-reduce constituent parsing

- Example
- Terminate
stack

buffer
$\qquad$


## Transition-based constituent parsing

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

likes redtomatoes


## Transition-based constituent parsing

- Feature Templates for Shift-reduce Constituent Parser

| Feature type | Feature Template |
| :---: | :--- |
| unigrams | $s_{0} p s_{0} c, s_{0} w s_{0} c, s_{1} p s_{1} c, s_{1} w s_{1} c, s_{2} p s_{2} c, s_{2} w s_{2} c, s_{3} p s_{3} c$, |
|  | $s_{3} w s_{3} c, \quad b_{0} w b_{0} p, \quad b_{1} w b_{1} p, \quad b_{2} w b_{2} p, \quad b_{3} w b_{3} p, \quad s_{0 . l} w s_{0 . l} c$, |
|  | $s_{0 . r} w s_{0 . r} c, s_{0 . u} w s_{0 . u} c, s_{1 . l} w s_{1 . l} c, s_{1 . r} w s_{1 . r} c, s_{1 . u} w s_{1 . u} c$ |
| bigrams | $s_{0} w s_{1} w, s_{0} w s_{1} c, s_{0} c s_{1} w, s_{0} c s_{1} c, s_{0} w b_{0} w, s_{0} w b_{0} p, s_{0} c b_{0} w$, |
|  | $s_{0} c b_{0} p, b_{0} w b_{1} w, b_{0} w b_{1} p, b_{0} p b_{1} w, b_{0} p b_{1} p, s_{1} w b_{0} w, s_{1} w b_{0} p$, |
|  | $s_{1} c b_{0} w, s_{1} c b_{0} p$ |
| trigrams | $s_{0} c s_{1} c s_{2} c, s_{0} w s_{1} c s_{2} c, s_{0} c s_{1} w b_{0} p, s_{0} c s_{1} c s_{2} w, s_{0} c s_{1} c b_{0} p$, |
|  | $s_{0} w s_{1} c b_{0} p, s_{0} c s_{1} w b_{0} p, s_{0} c s_{1} c b_{0} w$ |

$s_{i}$ : top node of the stack; $\quad b_{i}$ : front word on the buffer; $x w$ : the word form of $x$; $x p$ : the POS tag; $x c$ : the constituent label of a non-terminal node $x$;
$x l, x r, x u$ : the left child, the right child and the unary child of $x$, respectively.

## Transition-based constituent parsing

- Feature Templates for Shift-reduce Constituent Parser
- Example

In the 10-th step of the example sentence "The little boy likes
red tomatoes", the $s_{0} p s_{0} c$ feature is VBZ|VP and the $s_{1} p s_{1} c$ feature is NN INP.


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- 11.3 Shift-reduce Dependency Parsing
$\qquad$
11.4.2 Joint Word Segmentation, POS-tagging and Dependency Parsing


## Shift-reduce dependency parsing

- Projective dependency tree

- Non-projective dependency tree



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## Shift-reduce dependency parsing

- 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

# Shift-reduce dependency parsing 

Actions

$b_{0} b_{1} \ldots$

SHIFT


$$
b_{1} \ldots
$$

LEFT-ARC

$b_{0} b_{1} \ldots$

RIGHT-ARC

$b_{0} b_{1} \ldots$

## Shift-reduce dependency parsing

Arc-standard Dependency Parsing

|  | $\left([], W_{1: n}, \phi\right)$ | Left-ARC-X | $\left(\sigma\left\|s_{1}\right\| s_{0}, \beta, A\right)$ |
| :---: | :---: | :---: | :---: |
| Goal: | $\left(\left[s_{0}\right],[], A\right)$ |  | $\left(\sigma \mid s_{0}, \beta, A \cup\left\{s_{1} \stackrel{\stackrel{x}{\curvearrowleft}}{\left.s_{0}\right\}}\right. \text { ) }\right.$ |
| SHIFT: | $\left(\sigma, b_{0} \mid \beta, A\right)$ | Right-ARC-X: | $\left(\sigma\left\|s_{1}\right\| s_{0}, \beta, A\right)$ |
|  | $\left(\sigma \mid b_{0}, \beta, A\right)$ |  | $\left(\sigma \mid s_{1}, \beta, A \cup\left\{s_{1} \stackrel{\mathrm{X}}{\curvearrowright} s_{0}\right\}\right)$ |
| State: $(\sigma, \beta, A)$; |  |  |  |
| $\sigma$ : stack; | $\beta$ : buffer; $\quad$ : | the set of dependenc | arcs that have been constructed |

# Arc-standard dependency parsing 

Example
Sentence "He gave her a tomato"
Next action: SHIFT

He gave her a tomato

# Arc-standard dependency parsing 

Example
Sentence "He gave her a tomato"
Next action: SHIFT


# Arc-standard dependency parsing 

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

# Arc-standard dependency parsing 

Example
Sentence "He gave her a tomato"
Next action: SHIFT

her a tomato

# Arc-standard dependency parsing 

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

a tomato

# Arc-standard dependency parsing 

Example
Sentence "He gave her a tomato"
Next action: SHIFT

a tomato

# Arc-standard dependency parsing 

Example
Sentence "He gave her a tomato"
Next action: SHIFT

tomato

# Arc-standard dependency parsing 

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


# Arc-standard dependency parsing 

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


# Arc-standard dependency parsing 

Example
Sentence "He gave her a tomato"
Next action: END


# Shift-reduce dependency parsing 

Arc-standard Dependency Parsing

- Example state



## Arc-standard dependency parsing

Feature Templates for arc-standard dependency parsing

| Feature Type | Feature Template | Feature Type | Feature Template |
| :---: | :---: | :---: | :---: |
| from single words | $s_{0} w p ; s_{0} w ; s_{0} p ; b_{0} w p ;$ $b_{0} w ; b_{0} p ; b_{1} w p ; b_{1} w ;$ $b_{1} p ; b_{2} w p ; b_{2} w ; b_{2} p ;$ | valency | $\begin{aligned} & s_{0} w v_{r} ; s_{0} p v_{r} ; s_{0} w v_{l} ; \\ & s_{0} p v_{l} ; b_{0} w v_{l} ; b_{0} p v_{l} ; \end{aligned}$ |
| from word pairs | $s_{0} w p b_{0} w p ; s_{0} w p b_{0} w$; <br> $s_{0} w p b_{0} p ; s_{0} w b_{0} w p ;$ <br> $s_{0} p b_{0} w p ; s_{0} w b_{0} w ;$ <br> $s_{0} p b_{0} p ; b_{0} p b_{1} p ;$ | unigrams | $s_{0 . h} w ; s_{0 . h} p ; s_{0 . l}$; <br> $s_{0 . l} w ; s_{0 . l} p ; s_{0 . l} l$; <br> $s_{0 . r} w ; s_{0 . r} p ; s_{0 . r} l ;$ <br> $b_{0 . l} w ; b_{0 . l} p ; b_{0 . l} l$; |
| from three words | $\begin{aligned} & b_{0} p b_{1} p b_{2} p ; s_{0} p b_{0} p b_{1} p ; \\ & s_{0 . h} p s_{0} p b_{0} p ; \\ & s_{0} p s_{0 . l} p b_{0} p ; \\ & s_{0} p s_{0 . r} p b_{0} p ; \\ & s_{0} p b_{0} p b_{0 . l} p ; \end{aligned}$ | third-order | $\begin{aligned} & s_{0 . h_{2}} w ; s_{0 . h_{2}} p ; s_{0 . h} l ; \\ & s_{0 . l_{2}} p ; s_{0 . l_{2}} l ; s_{0 . r_{2}} w ; \\ & s_{0 . r_{2}} p ; s_{0 . r_{2}} l ; b_{0 . l_{2}} w ; \\ & b_{0 . l_{2}} p ; b_{0 . l_{2}} l ; \\ & s_{0} p s_{0 . l} p s_{0 . l_{2}} p ; \\ & s_{0} p s_{0 . r} p s_{0 . r_{2}} p ; \\ & s_{0} p s_{0 . h} p s_{0 . h_{2}} p ; \\ & b_{0} p b_{0 . l} p b_{0 . l_{2}} p ; \end{aligned}$ |
| distance | $s_{0} w d ; s_{0} p d ; b_{0} w d ;$ $b_{0} p d ; s_{0} w b_{0} w d ;$ $s_{0} p b_{0} p d ;$ | label set | $\begin{aligned} & s_{0} w s_{r} ; s_{0} p s_{r} ; s_{0} w s_{l} ; \\ & s_{0} p s_{l} ; n_{0} w s_{l} ; n_{0} p s_{l} \end{aligned}$ |

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## Shift-reduce dependency parsing

- 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

## Shift-reduce dependency parsing

Arc-eager dependency parsing

- Example



## Shift-reduce dependency parsing



## Shift-reduce dependency parsing

## Arc-eager dependency parsing

| Axiom: ([ ], $\left.W_{1: n}, \phi\right)$ | LEFT-ARC-X: | $\left(\sigma\left\|s_{0}, b_{0}\right\| \beta, A\right)$, such that $\neg\left(\exists(k, \mathrm{~L}) w_{k} \stackrel{\mathrm{~L}}{\curvearrowright} s_{0} \in A\right)$ |
| :---: | :---: | :---: |
| Goal: $\quad\left(\left[s_{0}\right],[], A\right)$ |  | $\left(\sigma, b_{0} \mid \beta, A \cup\left\{s_{0} \stackrel{\text { x }}{n} b_{0}\right\}\right)$ |
| SHIFT: $\frac{\left(\sigma, b_{0} \mid \beta, \phi\right)}{\left(\sigma \mid b_{0}, \beta, \phi\right)}$ | Right-ARC-X: | $\left(\sigma\left\|s_{0}, b_{0}\right\| \beta, A\right)$ |
|  |  | $\left(\sigma\left\|s_{0}\right\| b_{0}, \beta, A \cup\left\{s_{0} \stackrel{\mathrm{x}}{\curvearrowright} b_{0}\right\}\right)$ |
|  | Reduce: | $\left(\sigma \mid s_{0}, \beta, A\right)$, such that $\left(\exists(k, \mathrm{~L}) w_{k} \stackrel{\mathrm{~L}}{\curvearrowright} s_{0} \in A\right)$ |
|  |  | $(\sigma, \beta, A)$ |

Main differences comparing with arc-standard:

- Left - Arc $-X /$ Right $-\operatorname{Arc}-X$ : 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.


## Example (Dependency Parsing)

- NEXT ACTION: Shift



## Example (Dependency Parsing)

- NEXT ACTION: Left - Arc - SUBJ



## Example (Dependency Parsing)

- NEXT ACTION: Shift



## Example (Dependency Parsing)

- NEXT ACTION: Right - Arc - OBJ



## Example (Dependency Parsing)

- NEXT ACTION: Reduce



## Example (Dependency Parsing)

- NEXT ACTION: Right - Arc - MOD



## Example (Dependency Parsing)

- NEXT ACTION: Reduce



## Example (Dependency Parsing)

- NEXT ACTION: Finish



## Contents

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11.4.1 Joint POS-tagging and Dependency Parsing

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## Shift-reduce dependency parsing

- 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.



## Shift-reduce dependency parsing

Arc-standard Dependency Parsing with Swap Action

| $\begin{array}{ll} \text { Axiom: } & \left([], W_{1: n}, \phi\right) \\ \text { Goal: } & \left(\left[s_{0}\right],[], A\right) \\ \hline \end{array}$ | Left-ARC-X: | $\left(\left[\sigma\left\|s_{1}\right\| s_{0}\right], \beta, A\right)$ |
| :---: | :---: | :---: |
|  |  | $\left(\left[\sigma \mid s_{0}\right], \beta, A \cup\left\{s_{1}{ }^{\curvearrowleft} s_{0}\right\}\right)$ |
| Shift: $\frac{\left(\sigma,\left[b_{0} \mid \beta\right], \phi\right)}{\left(\left[\sigma \mid b_{0}\right], \beta, \phi\right)}$ | Right-ARC-X: | $\left(\left[\sigma\left\|s_{1}\right\| s_{0}\right], \beta, A\right)$ |
|  |  | $\left(\left[\sigma \mid s_{1}\right], \beta, A \cup\left\{s_{1} \stackrel{\stackrel{\wedge}{\wedge}}{ } s_{0}\right\}\right)$ |
|  |  | $\left(\left[\sigma\left\|s_{1}\right\| s_{0}\right], \beta, A\right)$, such that $\operatorname{Idx}\left(s_{1}\right)<\operatorname{IDX}\left(s_{0}\right)$ |
|  | Sw | $\left(\left[\sigma \mid s_{0}\right],\left[s_{1} \mid \beta\right], A\right)$ |

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

# Shift-reduce dependency parsing 

Arc-standard Dependency Parsing with Swap Action

- Example

Next action: SHIFT

A hearing was scheduled on this today

## Shift-reduce dependency parsing

Arc-standard Dependency Parsing with Swap Action

- Example

Next action: SHIFT

A
hearing was scheduled on this today

# Shift-reduce dependency parsing 

Arc-standard Dependency Parsing with Swap Action

- Example

Next action: LEFT-ARC-DET


# Shift-reduce dependency parsing 

Arc-standard Dependency Parsing with Swap Action

- Example

Next action: SHIFT

was scheduled on this today

## Shift-reduce dependency parsing

Arc-standard Dependency Parsing with Swap Action

- Example

Next action: SHIFT

scheduled on this today

# Shift-reduce dependency parsing 

Arc-standard Dependency Parsing with Swap Action

- Example

Next action: SHIFT

on this today

# Shift-reduce dependency parsing 

Arc-standard Dependency Parsing with Swap Action

- Example

Next action: SWAP

this today

## Shift-reduce dependency parsing

Arc-standard Dependency Parsing with Swap Action

- Example

Next action: SWAP

scheduled this today

# Shift-reduce dependency parsing 

Arc-standard Dependency Parsing with Swap Action

- Example

Next action: SHIFT

was scheduled this today

# Shift-reduce dependency parsing 

Arc-standard Dependency Parsing with Swap Action

- Example

Next action: SHIFT

scheduled this today

# Shift-reduce dependency parsing 

Arc-standard Dependency Parsing with Swap Action

- Example

Next action: SHIFT
this today

# Shift-reduce dependency parsing 

Arc-standard Dependency Parsing with Swap Action

- Example

Next action: SWAP


# Shift-reduce dependency parsing 

Arc-standard Dependency Parsing with Swap Action

- Example

Next action: SWAP

scheduled today

## Shift-reduce dependency parsing

Arc-standard Dependency Parsing with Swap Action

- Example

Next action: RIGHT-ARC-POBJ


## Shift-reduce dependency parsing

Arc-standard Dependency Parsing with Swap Action

- Example

Next action: RIGHT-ARC-NMOD

was scheduled today

# Shift-reduce dependency parsing 

Arc-standard Dependency Parsing with Swap Action

- Example

Next action: SHIFT


## Shift-reduce dependency parsing

Arc-standard Dependency Parsing with Swap Action

- Example

Next action: LEFT-ARC-SUBJ
 scheduled today

# Shift-reduce dependency parsing 

Arc-standard Dependency Parsing with Swap Action

- Example

Next action: SHIFT

scheduled today

# Shift-reduce dependency parsing 

Arc-standard Dependency Parsing with Swap Action

- Example

Next action: SHIFT


# Shift-reduce dependency parsing 

Arc-standard Dependency Parsing with Swap Action

- Example

Next action: RIGHT-ARC-TMP


# Shift-reduce dependency parsing 

Arc-standard Dependency Parsing with Swap Action

- Example

Next action: LEFT-ARC-VG


# Shift-reduce dependency parsing 

Arc-standard Dependency Parsing with Swap Action

- Example

Next action: END


## Contents

11.1 Transition-based Structured Prediction


- 11.4 Joint Models
$\qquad$


## Joint Models

- 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.


## Joint Models

Joint POS-tagging and Dependency Parsing

- Input: $W_{1: n}=w_{1}, \ldots, w_{n}$
- Output: $\left(T_{1: n}, A\right)$, 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 Models

Joint POS-tagging and Dependency Parsing

- Example

Next action: SHIFT——PRP

Stack [S]


Buffer [B]
$\mathrm{He}_{1}$ won $_{2}$ the $_{3}$ game $_{4}$

## Joint Models

Joint POS-tagging and Dependency Parsing

- Example

Next action: SHIFT——_VBD

Stack [S]


| Buffer [B] |
| :---: |
| won $_{2}$ the $_{3}$ game $_{4}$ |

## Joint Models

Joint POS-tagging and Dependency Parsing

- Example

Next action: LEFT——ARC——NSUBJ

Stack [S]
$\begin{array}{r} \\ \mathrm{He}_{1 / \mathrm{RRP}} \text { Won }_{2} \\ \hline\end{array}$
Buffer [B]
the $_{3}$ game $_{4}$

Joint POS-tagging and Dependency Parsing

- Example

Next action: SHIFT——DT


## Joint Models

Joint POS-tagging and Dependency Parsing

- Example

Next action: SHIFT——NN


## Joint Models

Joint POS-tagging and Dependency Parsing

- Example

Next action: LEFT——ARC——DET


## Joint Models

Joint POS-tagging and Dependency Parsing

- Example

Next action: RIGHT——ARC——DOBJ


Buffer [B]
$\square$

## Joint Models

Joint POS-tagging and Dependency Parsing

- Example

Next action: END


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

- Input: $C_{1: n}=C_{1} \ldots C_{n}$
- Output: $\left(W_{1: m}, T_{1: m}, A\right)$ where $w_{1} \ldots w_{m}$ are words, $t_{1} \ldots t_{n}$ 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


## Joint Models

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

| Axiom: | ([], [ ], $\left.C_{1: n}, \phi, \phi\right)$ |
| :---: | :---: |
| Goal: | ([S $\left.\left.S_{0}\right],[],[], A_{c}, A_{w}\right)$ |
| LEFT-ARC-C: | $\left(\sigma, \delta\left\|d_{0}, b_{0}\right\| \beta, A_{c}, A_{w}\right)$ such that $\neg\left(\exists d \in \delta, d^{\curvearrowright} d_{0} \in A_{c}\right)$ |
|  | $\left(\sigma, \delta, b_{0} \mid \beta, A_{c} \cup\left\{d_{0} \curvearrowleft b_{0}\right\}, A_{w}\right)$ |
| LEFT-ARC-X: | $\left(\sigma \mid s_{0},\left[d_{0}\right], \beta, A_{c}, A_{w}\right)$ such that $\neg\left(\exists s \in \sigma, s^{\stackrel{l}{\curvearrowright}} s_{0} \in A_{w}\right)$ |
|  | $\left(\sigma,\left[d_{0}\right], \beta, A_{c}, A_{w} \cup\left\{s_{0} \curvearrowleft d_{0}\right\}\right)$ |
| SHIFT: | $\left(\sigma,\left[d_{0}\right], \beta, A_{c}, A_{w}\right)$ |
|  | $\left(\sigma \mid d_{0},[], \beta, A_{c}, A_{w}\right)$ |
| SHIFT-C: | $\left(\sigma, \delta, b_{0} \mid \beta, A_{c}, A_{w}\right)$ |
|  | $\left(\sigma, \delta \mid b_{0}, \beta, A_{c}, A_{w}\right)$ |
| Right-ARC-C: | $\left(\sigma, \delta\left\|d_{0}, b_{0}\right\| \beta, A_{c}, A_{w}\right)$ |
|  | $\left(\sigma, \delta\left\|d_{0}\right\| b_{0}, \beta, A_{c} \cup\left\{d_{0} \curvearrowright b_{0}\right\}, A_{w}\right)$ |
| Right-ARC-X: | $\left(\sigma \mid s_{0},\left[d_{0}\right], \beta, A_{c}, A_{w} \cup\left\{s_{0} \curvearrowright d_{0}\right\}\right)$ |
|  | $\left(\sigma\left\|s_{0}\right\| d_{0},[], \beta, A_{c}, A_{w}\right)$ |
| Pop-X: | $\left(\sigma,\left[d_{0}\right], \beta, A_{c}, A_{w}\right)$ |
|  | $\left(\sigma,\left[\operatorname{SuBTREE}\left(d_{0}, A_{c}\right) / X\right], \beta, A_{c}, A_{w}\right)$ |
| REDUCE-C: | $\left(\sigma, \delta \mid d_{0}, \beta, A_{c}, A_{w}\right)$ such that $\exists d \in \delta, d^{\curvearrowright} d_{0} \in A_{c}$ |
|  | $\left(\sigma, \delta, \beta, A_{c}, A_{w}\right)$ |
| Reduce: | $\left(\sigma \mid s_{0}, \delta, \beta, A_{c}, A_{w}\right)$ such that $\exists s \in \sigma, s^{\curvearrowright} s_{0} \in A_{w}$ |
|  | $\left(\sigma, \delta, \beta, A_{c}, A_{w}\right)$ |

State: $\left(\sigma, \delta, \beta, A_{-} c, A_{-} w\right) \quad \sigma$ : stack; $\quad \beta$ :buffer; $\quad \delta$ : partial-word buffer;
$A_{c}$ : the set of character dependencies; $\quad A_{w}$ : the set of word dependencies ${ }^{164}$

## Joint Models

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

－Example

| Step $\sigma$ $\delta$ $\beta$ $A_{c}$ $A_{w}$ Action <br> 0   我，来，到，会，客，室   SHIFT－C |
| :--- |
|  |

## Joint Models

Joint Word Segmentation, POS-tagging and Dependency Parsing

- Example



## Joint Models

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

－Example

| Step | $\sigma$ | $\delta$ | $\beta$ | $A_{c}$ | $A_{w}$ | Action |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 2 |  | 我／PN | 来，到，会，客，室 |  |  | SHIFT－W |



## Joint Models

Joint Word Segmentation, POS-tagging and Dependency Parsing

- Example



## Joint Models

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

| Step | $\sigma$ | $\delta$ | $\beta$ | $A_{c}$ | $A_{w}$ | Action |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 4 | 我／PN | 来 | 到，会，客，室 |  |  | LEFTARC－C |

$\square$
$\qquad$到，会，客，室

## Joint Models

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

| Step | $\sigma$ | $\delta$ | $\beta$ | $A_{c}$ | $A_{w}$ | Action |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 5 | 我／PN |  | 到，会，客，室 | 来 $\leftarrow$ 到 |  | SHIFT |
|  |  |  |  |  |  |  |
| 我／PN |  |  |  |  | 来 到, | ，客，室 |

## Joint Models

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

| Step | $\sigma$ | $\delta$ | $\beta$ | $A_{c}$ | $A_{w}$ | Action |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 6 | 我／PN | 到 | 到，会，客，室 | 来 $\leftarrow$ 到 |  | POP－VV |
|  |  |  |  |  |  |  |
| 我／PN |  |  | 来 到 |  |  | ，客，室 |

## Joint Models

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

| Step | $\sigma$ | $\delta$ | $\beta$ | $A_{c}$ | $A_{w}$ | Action |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 7 | 我／PN | 来到／VV | 会，客，室 | 来 $\leftarrow$ 到 |  | LEFTARC－SUBJ |

$\square$

会, 客, 室

## Joint Models

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

| Step | $\sigma$ | $\delta$ | $\beta$ | $A_{c}$ | $A_{w}$ | Action |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 8 |  | 来到／VV | 会，客，室 | 来 $\leftarrow$ 到 | 我／PN $\leftarrow$ SUBJ 来到／VV | SHIFT |




## Joint Models

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

| Step | $\sigma$ | $\delta$ | $\beta$ | $A_{c}$ | $A_{w}$ | Action |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 9 | 来到／VV |  | 会，客，室 | 来 $\leftarrow$ 到 | 我／PN $\leftarrow$ SUBJ 来到／VV | SHIFT－C |



## Joint Models

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

－Example

| Step | $\sigma$ | $\delta$ | $\beta$ | $A_{c}$ | $A_{w}$ | Action |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 10 | 来到／VV | 会 | 客，室 | 来 $\leftarrow$ 到 | 我／PN $\leftarrow$ SUBJ 来到／VV | RIGHTARC－C |

## Joint Models

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

| Step | $\sigma$ | $\delta$ | $\beta$ | $A_{c}$ | $A_{w}$ | Action |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 11 | 来到／VV | 会，客 | 室 | 来 $\leftarrow$ 到，会 $\rightarrow$ 客 | 我／PN $\leftarrow S U B J$ 来到／VV | $R E D U C E-C$ |

## Joint Models

Joint Word Segmentation, POS-tagging and Dependency Parsing

- Example



## Joint Models

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

| Step | $\sigma$ | $\delta$ | $\beta$ | $A_{c}$ | $A_{w}$ | Action |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 13 | 来到／VV |  | 室 | 来 $\leftarrow$ 到，会 $\rightarrow$ 客，会 $\leftarrow$ 室 | 我／PN $\leftarrow$ SUBJ 来到／VV | SHIFT－C |



## Joint Models

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

| Step | $\sigma$ | $\delta$ | $\beta$ | $A_{c}$ | $A_{w}$ | Action |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 14 | 来到 $/ \mathrm{VV}$ | 室 |  | 来 $\leftarrow$ 到，会 $\rightarrow$ 客，会 $\leftarrow$ 室 | 我／PN $\leftarrow S U B J$ 来到 $/ \mathrm{VV}$ | $P O P-N N$ |



## Joint Models

Joint Word Segmentation, POS-tagging and Dependency Parsing

- Example



## Joint Models

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

| Step | $\sigma$ | $\delta$ | $\beta$ | $A_{c}$ | $A_{w}$ | Action |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 16 | 来到／VV，会客 <br> 室／NN |  |  | 来 $\leftarrow$ 到，会 $\rightarrow$ <br> 客，会室 | 我／PN $\leftarrow$ SUBJ 来到／VV，来到／VV <br> DOBJ $\rightarrow$ 会客室／NN | REDUCE |



## Joint Models

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

－Example

| Step | $\sigma$ | $\delta$ | $\beta$ | $A_{c}$ | $A_{w}$ | Action |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 17 | 来 <br> 到 $/ \mathrm{VV}$ |  |  | 来 $\leftarrow$ 到，会 $\rightarrow$ 客，会 <br> $\leftarrow$ 室 | 我／PN $\leftarrow S U B J$ 来到 $/ V V$, 来到／VV DOBJ $\rightarrow$ 会客 <br> 室／NN | END |



## Summary

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


[^0]:    S-SHIFT

