# CS626: Speech, NLP and the Web

#### Softmax FFNN-BP and Neural Dependency Parsing Pushpak Bhattacharyya Computer Science and Engineering Department IIT Bombay

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#### Dependency parsing algorithms

Makin approaches

#### Rule based approach: CP to DP



Speech and Language Processing, Jurafksy & Martin, Ch-15, 2019

## Head directionality

- Head final languages
  - Head final structure
  - Head of a phrase follows its complements
  - Example: Hindi is strong head final language
- Head initial languages
  - Head initial structure
  - Head of a phrase precedes its complements
  - Example: English is strong head-initial language
- Head
  - Component of a phrase which determines the category of the phrase
  - Example: Head of a verb phrase is verb

#### Rule based approach: CP to DP

#### • Two main acts

- Identify all the head-dependent relations
- Identifying the correct dependency relations for above relations

#### Task involves

- Marking the head
- Make the head of each non-head depend on the head

Jurafsky, Dan. Speech & language processing. Pearson Education India, 2000.

Example

• Ram stayed at home in the evening

• raam shaam ko ghar par rahaa

#### English-Hindi: Head Initial-Head Final







#### English CP $\rightarrow$ DP cntd. S (stayed) stayed NP (Ram) VP (stayed) at IN Ram PP (in) NNP PP (at) VBD home evening stayed PNP (home) P NP (evening) the Ram Ν in DT Ν at home the evening root case pobj *d*ase nsubj pobj det evening in Ram home the stayed at

#### English Compact DP with labels



Relevant discussion: different types of subjects (nsubj, dsubj, gsubj)

- *nsubj*: Nominative subject
- Languages do have non-nominal subject
  - Dative subject is very common in Indian languages
  - Example
    - Hindi: मुझे आम चाहिए
    - English: I want Mango.
    - मुझे: Dative case (सम्प्रदान कारक) (dsubj: dative subject)
    - I: Nominative case (nsubj)

#### **Deeper representation**

- Deeper representation towards meaning
   *nsubj* is shallower than *agent*
  - nsubj will lead to agent
  - Semantic role labeled graph
    - Compact DP with labels
    - The relations captures the exact semantics of the situation
    - Case relationships: loc, time, agent
      - Indicate semantic relationships of the noun with the main verb

#### Hindi $CP \rightarrow DP$



#### Hindi Compact DP with labels



#### A Digression: Vauquois Triangle in Machine Translation



The deeper the NLP analysis, the closer the representations and the easier the analysis

# Grammatical head and Semantic Head

Grammatical Head: Word whose POS gives the Phrase its name

 Semantic Head: Word that carries the "semantic load"

#### Data Driven: two approaches

- Transition-based
  - State machine for mapping a sentence to its dependency graph
  - Inducing a model for predicting the next transition, given the current state and the transition history so far.
- Graph-based
  - Induce a model for assigning scores to the candidate dependency graphs for a sentence
  - Find the maximum-scoring dependency Tree
  - Maximum spanning tree (MST) parsing

#### **Basic Transition Based DP**



Examines top two elements of the stack and selects an action based on consulting an oracle that examines the current configuration.

Speech and Language Processing, Jurafksy & Martin, Ch-15, 2019

#### **Example: transition based**

Step	Stack	Word List	Action	Relation Added
0	[root]	[book, me, the, morning, flight]	SHIFT	
1	[root, book]	[me, the, morning, flight]	SHIFT	
2	[root, book, me]	[the, morning, flight]	RIGHTARC	$(book \rightarrow me)$
3	[root, book]	[the, morning, flight]	SHIFT	
4	[root, book, the]	[morning, flight]	SHIFT	
5	[root, book, the, morning]	[flight]	SHIFT	
6	[root, book, the, morning, flight]		LEFTARC	$(morning \leftarrow flight)$
7	[root, book, the, flight]		LEFTARC	$(\text{the} \leftarrow \text{flight})$
8	[root, book, flight]		RIGHTARC	$(book \rightarrow flight)$
9	[root, book]	[]	RIGHTARC	$(root \rightarrow book)$
10	[root]		Done	

Trace of a transition-based parse

Speech and Language Processing, Jurafksy & Martin, Ch-15, 2019

#### A neural transition based parser (chen and Manning 2014)



#### Notion of Projectivity

## Definition

- An arc from a head to a dependent is said to be projective if there is a path from the head to every word that lies between the head and the dependent in the sentence
- A dependency tree is then said to be projective if all the arcs that make it up are projective
- Intuition- the dependency graph can be drawn on a plane w/o crossing of arcs

#### **Example of projectivity**



#### Another example



#### Learning of transitions

Speech and NLP, J & M, Ch 15, 2019.

#### Recall: transition based DP

Step	Stack	Word List	Action	Relation Added
0	[root]	[book, me, the, morning, flight]	SHIFT	
1	[root, book]	[me, the, morning, flight]	SHIFT	
2	[root, book, me]	[the, morning, flight]	RIGHTARC	$(book \rightarrow me)$
3	[root, book]	[the, morning, flight]	SHIFT	
4	[root, book, the]	[morning, flight]	SHIFT	
5	[root, book, the, morning]	[flight]	SHIFT	
6	[root, book, the, morning, flight]		LEFTARC	$(morning \leftarrow flight)$
7	[root, book, the, flight]		LEFTARC	$(\text{the} \leftarrow \text{flight})$
8	[root, book, flight]		RIGHTARC	$(book \rightarrow flight)$
9	[root, book]	[]	RIGHTARC	$(root \rightarrow book)$
10	[root]		Done	

Trace of a transition-based parse

Speech and Language Processing, Jurafksy & Martin, Ch-15, 2019

#### **Basic Transition Based DP**



Examines top two elements of the stack and selects an action based on consulting an oracle that examines the current configuration.

Speech and Language Processing, Jurafksy & Martin, Ch-15, 2019

#### Operators: shift, leftarc, rightarc

function DEPENDENCYPARSE(words) returns dependency tree

```
state \leftarrow {[root], [words], [] }; initial configuration
while state not final
```

 $t \leftarrow ORACLE(state)$ ; choose a transition operator to apply state  $\leftarrow APPLY(t, state)$ ; apply it, creating a new state **return** state

#### **Generation of Training Data**

Step	Stack	Word List	Predicted Action
0	[root]	[book, the, flight, through, houston]	SHIFT
1	[root, book]	[the, flight, through, houston]	SHIFT
2	[root, book, the]	[flight, through, houston]	SHIFT
3	[root, book, the, flight]	[through, houston]	LEFTARC
4	[root, book, flight]	[through, houston]	SHIFT
5	[root, book, flight, through]	[houston]	SHIFT
6	[root, book, flight, through, houston]	[]	LEFTARC
7	[root, book, flight, houston ]		RIGHTARC
8	[root, book, flight]	Π	RIGHTARC
9	[root, book]		RIGHTARC
10	[root]		Done

Training data

#### How are operators generated

- LEFTARC(r): if  $(S_1 r S_2) \in R_p$ RIGHTARC(r): if  $(S_2 r S_1) \in R_p$  and  $\forall r', w s.t.(S_1 r' w) \in R_p$  then  $(S_1 r' w) \in R_c$
- SHIFT: otherwise

#### **Generation of Training Data**

Step	Stack	Word List	Predicted Action
0	[root]	[book, the, flight, through, houston]	SHIFT
1	[root, book]	[the, flight, through, houston]	SHIFT
2	[root, book, the]	[flight, through, houston]	SHIFT
3	[root, book, the, flight]	[through, houston]	LEFTARC
4	[root, book, flight]	[through, houston]	SHIFT
5	[root, book, flight, through]	[houston]	SHIFT
6	[root, book, flight, through, houston]		LEFTARC
7	[root, book, flight, houston ]		RIGHTARC
8	[root, book, flight]		RIGHTARC
9	[root, book]		RIGHTARC
10	[root]		Done

Training data

#### A neural transition based parser (chen and Manning 2014)



#### Features: example sentence "cancelled flights to Houston"

 $\langle s_1.w = flights, op = shift \rangle$  $\langle s_2.w = canceled, op = shift \rangle$  $\langle s_1.t = NNS, op = shift \rangle$  $\langle s_2.t = VBD, op = shift \rangle$  $\langle b_1.w = to, op = shift \rangle$  $\langle b_1.t = TO, op = shift \rangle$  $\langle s_1.wt = flightsNNS, op = shift \rangle$ 

#### **DP** across languages

• "people in front of the house told me"

 "gharaasamorchyaanii malaa saaMgitle"

 "ghar ke saamnewaloM ne mujhe batayaa"

## **Multilingual DP**

• *"people in front of the house told me"* 



Examines top two elements of the stack and selects an action based on consulting an oracle that examines the current configuration.

Speech and Language Processing, Jurafksy & Martin, Ch-15, 2019

#### Essence of DP

 Cannot pop a *head* out of the stack if any of its dependents remains on the stack

• The above works if the sentence is projective