

Dependency Parsing

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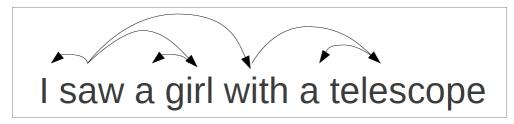
Lecture Outline

- Introduction and Formalisms
- Dependency Treebanks
- Transition-Based Dependency Parsing
- Graph-Based Dependency Parsing
- Evaluation
- Summary

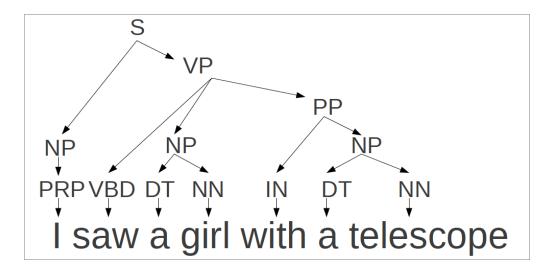


Two Types of Parsing

Dependency: focuses on relations between words



■ Phrase structure: focuses on identifying phrases and their recursive structure





Advantages of Dependency Parsing

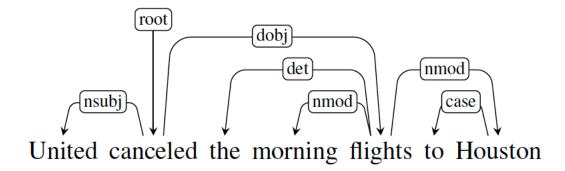
- Dependency grammar is more appropriate to languages that have a relatively free word order
 - ☐ Czech, Japanese
- Dependency relations are close to semantic relations between predicates/arguments
 - ☐ Useful for other NLP tasks: coreference resolution, question answering, information extraction



Dependency Relations

- Consists of head and dependent (head and modifier)
 - □ the head word is a central of a larger constituent
 - E.g., the primary noun in a noun phrase, or verb in a verb phrase

- Dependency relations can be categorized into grammatical functions
 - □ E.g., subject, direct object and indirect object



Clausal Argument Relations	Description
NSUBJ	Nominal subject
DOBJ	Direct object
IOBJ	Indirect object
CCOMP	Clausal complement
XCOMP	Open clausal complement
Nominal Modifier Relations	Description
NMOD	Nominal modifier
AMOD	Adjectival modifier
NUMMOD	Numeric modifier
APPOS	Appositional modifier
DET	Determiner
CASE	Prepositions, postpositions and other case markers
Other Notable Relations	Description
CONJ	Conjunct
CC	Coordinating conjunction

Selected dependency relations from the Universal Dependency set.

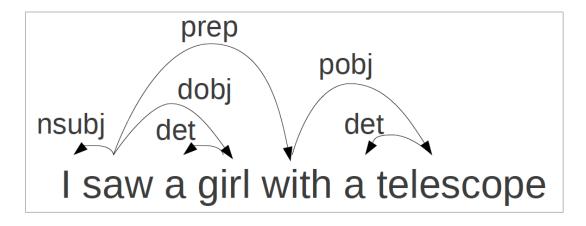
Relation	Examples with <i>head</i> and dependent
NSUBJ	United canceled the flight.
DOBJ	United diverted the flight to Reno.
	We booked her the first flight to Miami.
IOBJ	We booked her the flight to Miami.
NMOD	We took the morning <i>flight</i> .
AMOD	Book the cheapest <i>flight</i> .
NUMMOD	Before the storm JetBlue canceled 1000 flights.
APPOS	United, a unit of UAL, matched the fares.
DET	The flight was canceled.
	Which flight was delayed?
CONJ	We <i>flew</i> to Denver and drove to Steamboat.
CC	We flew to Denver and <i>drove</i> to Steamboat.
CASE	Book the flight through <i>Houston</i> .

Examples of core Universal Dependency relations.

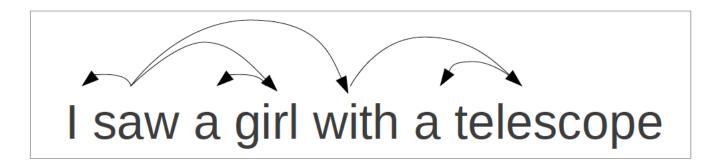


Dependencies

Typed: Label indicating relationship between words



Untyped: Only which words depend





Dependency Formalisms

Represent dependency structure as a directed graph G = (V, A)

 \square *V*: set of vertices

 \square A: set of ordered pairs

- A dependency tree is a directed graph
 - ☐ There is a single designated root node that has no incoming arcs.
 - Except for the root node, each vertex has exactly one incoming arc.
 - \Box There is a unique path from the root node to each vertex in V.



Dependency Parsing Methods

■ Shift-reduce

- □ Predict from left-to-right
- □ Fast (linear), but slightly less accurate?
- MaltParser

Spanning tree

- Calculate full tree at once
- ☐ Slightly more accurate, slower
- MSTParser, Eda (Japanese)

Cascaded chunking

- Chunk words into phrases, find heads, delete nonheads, repeat
- □ CaboCha (Japanese)



Shift-Reduce

Configuration

- □ Queue: of unprocessed words in input
- ☐ Stack: of partially processed words
- □ A set of relations

At each point choose

- ☐ SHIFT: move one word from queue to stack
- REDUCE LEFT: top word on stack is head of second word, remove the lower word
- □ REDUCE RIGHT: second word on stack is head of top word, remove the top word



Shift-Reduce Example

Initial configuration

Step	Stack	Queue	Action	Relation Added
0	[root]	[book, me, the, morning, flight]		

We rely on a trained oracle to choose the action

is a definition for an entity who knows the correct answer to all questions.

https://bit.ly/3I6kLni

Step	Stack	Queue	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]		

Step	Stack	Queue	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]	reduce right	book → me
3	[root, book]	[the, morning, flight]		

Ste	Stack	Queue	Action	Relation Added
p				
0	[root]	[book, me, the, morning, flight]	shift	
1	[root, book]	[me, the, morning, flight]	shift	
2	[root, book, me]	[the, morning, flight]	reduce right	(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]			

Ste p	Stack	Queue	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]	reduce right	(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]	D	reduce left	$(morning \leftarrow flight)$
7	[root, book, the, flight	[]		

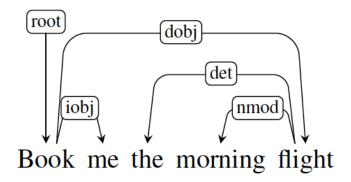
Ste p	Stack	Queue	Action	Relation Added
0	[root]	[book, me, the, morning, flight]	shift	
1	[root, book]	[me, the, morning, flight]	shift	
2	[root, book, me]	[the, morning, flight]	reduce right	(book \rightarrow me)
3	[root, book]	[the, morning, flight]	shift	
4	[root, book, the]	[morning, flight]	shift	
5	[root, book, the, morning]	[flight]	shift	
6	[root, book, the, morning, flight]		reduce left	$(morning \leftarrow flight)$
7	[root, book, the, flight]		reduce left	(the ← flight)
8	[root, book, flight]	[]		

Step	Stack	Queue	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]	reduce right	(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]		reduce left	$(morning \leftarrow flight)$
7	[root, book, the, flight]		reduce left	(the ← flight)
8	[root, book, flight]		reduce right	(book \rightarrow flight)
9	[root, book]		reduce right	$(root \rightarrow book)$
10	[root]	[]	Done	



Shift-Reduce Example

Resulting tree



- To produce labels for dependency relations: use actions with labels
 - ☐ REDUCE-LEFT[NSUBJ]
 - □ REDUCE-RIGHT[DOBJ]
 - □ ...



Shift-Reduce Analysis

- Predict from left-to-right
- Fast (linear complexity)
- Greedy (single choice at each point)



Creating an Oracle

Given a configuration

stack		queue
	[root, book, me]	[the, morning, flight]

Which action do we choose?

shift

reduce left

reduce right

We need a classifier!



Training Data for Shift-Reduce

- We need to train the classifier for shift-reduce
- Training data contains configurations (stack, queue)
 paired with action label

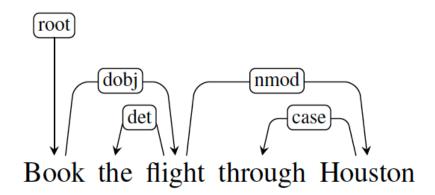
stack	queue	label
[root]	[book, me, the, morning, flight]	shift
[root, book]	[me, the, morning, flight]	shift
[root, book, me]	[the, morning, flight]	reduce right



Generating Training Data for Shift-Reduce

- Dependency Treebank includes sentences with annotaed dependency parses
 - □ Not directly provides our required training data
- We simulate shift-reduce parsing for sentences given their reference parses
- Given a reference parse and a configuration
 - Choose reduce left if it produces a correct headdependent relation
 - Otherwise, choose reduce right if
 - (1) it produces a correct head-dependent and
 - (2) all the dependents of the word at the top of the stack have already been assigned
 - ☐ Otherwise, choose shift

Step	Stack	Queue	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]	reduce left
4	[root, book, flight]	[through, houston]	shift
5	[root, book, flight, through]	[houston]	shift
6	[root, book, flight, through, houston]		reduce left
7	[root, book, flight, houston]		reduce right
8	[root, book, flight]		reduce right
9	[root, book]		reduce right
10	[root]		Done





Features for Shift-Reduce

Features should generally cover at least the last stack entries and first queue entry

Stack	Queue
[root, canceled, flights]	[to Houston]

Use Word, POS information

	stack[1]	stack[2]	queue[1]
Word	flights	canceled	to
POS	NNS	VBD	TO



Features for Shift-Reduce

Stack	Queue
[root, canceled, flights]	[to Houston]

	stack[1]	stack[2]	queue[1]
Word	flights	canceled	to
POS	NNS	VBD	TO

Some features

- □ s1.w=flights
- □ s2.w=canceled
- □ s1.pos=NNS
- □ s2.pos=VBD
- □ q1.w=to
- □ q1.pos=TO
- ☐ s1.wpos=flights-NNS
- □ ...



Training

- Multinomial logistic regression
- Support Vector Machines
- Neural Networks (Chen and Manning, 2014)



CoNLL File Format

- Standard format for dependencies
- Tab-separated columns, sentences separated by space

```
# sent_id = train-s1
# text = manh đất của đan bom không còn người nghèo.
        mảnh
                 mảnh
                         NOUN
                                  Nc
                                                            compound
        đất
                 đất
                         NOUN
                                                            nsubi
                                  Ν
        của
                 của
                         ADP
                                                            case
        đan
                 đan
                         NOUN
                                  Ν
                                                            nmod
        bom
                 bom
                         NOUN
                                                            compound
6
        không
                 không
                         Χ
                                  R
                                           Polarity=Neg
                                                                     advmod
        còn
                 còn
                         VERB
                                                            root
        người
                 người
                         NOUN
                                  Ν
                                                            obj
        nghèo
                 nghèo
                         ADJ
                                                                              SpaceAfter=No
                                                    8
                                                            amod
                                  Α
10
                          PUNCT
                                                            punct
```

Reference: https://github.com/UniversalDependencies/UD_Vietnamese-VTB/blob/master/vi_vtb-ud-train.conllu



Graph-Based Dependency Parsing

Search for the tree that maximizes some score

$$\widehat{T}(S) = \underset{t \in g_S}{\operatorname{argmax}} score(t, S)$$

Edge-factored: score for a tree is based on scores of its edges

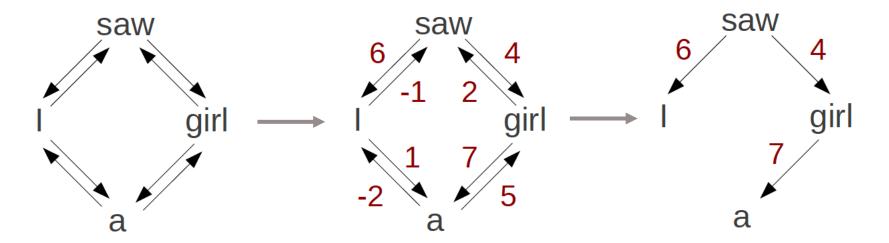
$$score(t,S) = \sum_{t \in T} score(e)$$

We assign scores for edges with machine learning



Parsing by Maximum Spanning Tree (MST)

- A spanning tree of a graph G
 - \square A subgraph of G
 - \square includes all the vertices of G.



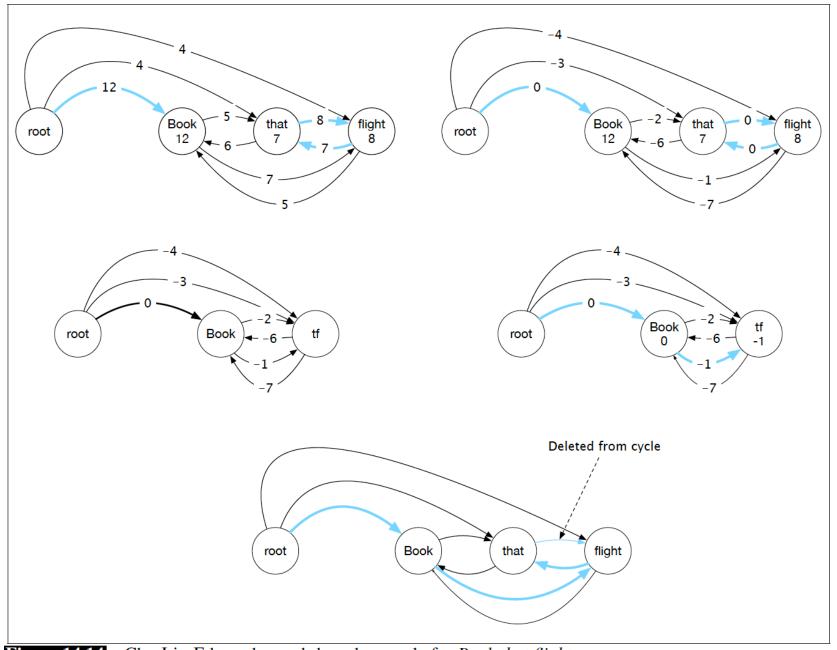


Figure 14.14 Chu-Liu-Edmonds graph-based example for *Book that flight*



Features and Training

- We need a model to compute edge score
- Use weighted sum of features $score(S, e) = w \cdot f$
- Features are similar to features used to train the shift-reduce classifier
- State-of-the-art algorithms in multilingual parsing are based on recurrent neural networks (RNNs)
 (Zeman et al. 2017, Dozat et al. 2017).

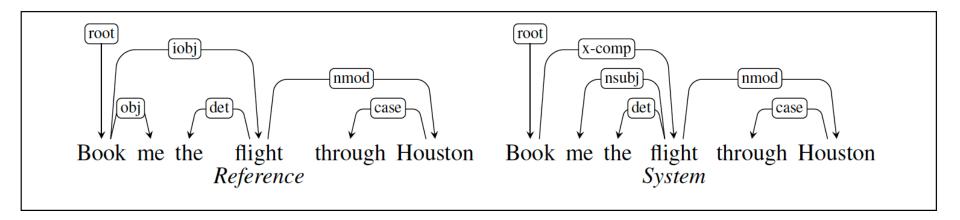


Evaluation Metrics

- Percentage of words in an input that are assigned the correct head with the correct relation.
- Two types
 - □ Labeled attachment score (LAS)
 - Consider labels of dependency relations
 - □ Unlabeled attachment score (UAS)
 - Ignore labels of dependency relations



Example of LAS and UAS



Reference and system parses for *Book me the flight through Houston*, resulting in an LAS of 2/3 and an UAS of 5/6.

