Natural Language Processing Dependency parsing

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Credit to Yulia Tsvetkov for slides

Announcement(s)

- A3 is out!
- Quiz 7 goes out on Canvas on Wednesday 3/1 at the end of lecture
 - Available until Friday 3/3 at 2:20pm; you'll have 15 minutes to complete it once you start.
 - Remember that you can use your notes during the quiz
 - Will cover material from last Wednesday's lecture through the end of today's lecture (so, parsing)
- For **this Wednesday** and **this Friday's** lectures: you have a choice of whether to join live over zoom or come to our usual classroom!
 - Our 3/1 and 3/3 guest speakers will be joining virtually to give their talks, so I'll be sending the zoom link that they'll be using to all of you this afternoon via Ed
 - But if you'd rather come to the usual classroom, I'll be here and projecting their lectures on the big screen!

Dependency representation



Operations



Configuration:

• Stack, Buffer, Oracle, Set of dependency relations

Operations by a classifier at each step:

- Shift
 - remove w1 from the buffer, push it onto the stack as s1
- LeftArc or Reduce left
 - \circ assert a head-dependent relation between s1 and s2 (s1 \rightarrow s2)
 - pop s1 from the stack; pop s2 from the stack; then push (s2 \leftarrow s1) onto the stack
- RightArc or Reduce right
 - \circ assert a head-dependent relation between s2 and s1 (s2 \rightarrow s1)
 - pop s1 from the stack; pop s2 from the stack; then push $(s2 \rightarrow s1)$ onto the stack

Or as A3/Eisenstein do it...

Operations by a classifier at each step:

- Shift
 - remove w1 from the buffer, push it onto the stack as s1
- LeftArc or Reduce left
 - $\circ~$ assert a head-dependent relation between s1 and s2 (s1 \rightarrow s2)
 - pop s1 from the stack; pop s2 from the stack; then push (s2 ← s1) onto the stack
- RightArc or Reduce right
 - assert a head-dependent relation between s2 and s1 (s2 → s1)
 - pop s1 from the stack; pop s2 from the stack; then push $(s2 \rightarrow s1)$ onto the stack

Operations by a classifier at each step:

- Shift
 - remove w1 from the buffer, push it onto the stack as s1
- LeftArc or Reduce left
 - assert a head-dependent relation between b1 and s1 (b1 → s1)
 - pop s1 from the stack; pop b1 from the buffer; then push (s1 ← b1) onto the buffer
- RightArc or Reduce right
 - assert a head-dependent relation between s1 and b1 (s1 → b1)
 - pop s1 from the stack; pop b1 from the buffer; then push (s1 → b1) onto the buffer

Want to see an(other) example of transition-based parsing in action?

Slides 30-44 of <u>this slide deck</u> by Noah Smith do a really nice job of walking through the full transition-based assembly of a sentence's parse visually.

Shift-Reduce Parsing (Arc-standard)

$$C_{\text{initial}} = ([\text{ROOT}], \boldsymbol{w}, \varnothing)$$

Book me the morning flight

Step	Stack	Word List	Action	Relation Added



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



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



Step	Stack	Word List	Action	Relation Added
0	[root]	[book, me, the, morning, flight]	SHIFT	
1	[root, book]	[me, the, morning, flight]		



Step	Stack	Word List	Action	Relation Added
0	[root]	[book, me, the, morning, flight]	SHIFT	
1	[root, book]	[me, the, morning, flight]	SHIFT	



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]		



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	



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	$(book \rightarrow me)$



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]		



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	



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]		



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	



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]		



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	



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]	0		



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, norning]	[flight]	SHIFT	
6	[root, book, the, morning, flight]		LEFTARC	$(morning \leftarrow flight)$



Book me the morning flight

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]	0	LEFTARC	(morning \leftarrow flight)
7	[root, book, the, flight]	0	LEFTARC	(the \leftarrow flight)



Book me the morning flight

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]	0	LEFTARC	$(\text{the} \leftarrow \text{flight})$
8	[root, book, flight]	0	RIGHTARC	$(book \rightarrow flight)$



Book me the morning flight

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]	0	LEFTARC	(the \leftarrow flight)
8	[root, book sight]	[]	RIGHTARC	$(book \rightarrow flight)$
9	[root, book]		RIGHTARC	$(root \rightarrow book)$

	(root iobj C	f _{accept} = ($[ROOT], \varnothing, A)$
Sten	Stack	Word List	Action	Relation Added
	[root]	[book me the morning flight]	SHIET	Relation Added
1	[root book]	[book, me, me, morning, mgnt]	SHIFT	
2	[root book me]	[the morning flight]	RIGHTARC	$(book \rightarrow me)$
3	[root, book]	[the, morning, flight]	SHIFT	(book / me)
4	[root, book, the]	[morning, flight]	SHIFT	
5	[root, book, the, morning]	[flight]	SHIFT	
6	[root, book, the, morning, flight]	Π	LEFTARC	(morning \leftarrow flight)
7	[root, book, the, flight]	ŭ	LEFTARC	(the \leftarrow flight)
8	[root, book, flight]	0	RIGHTARC	$(book \rightarrow flight)$
9	[root, book]	0	RIGHTARC	$(root \rightarrow book)$
10	[root]		Done	

Or as A3/Eisenstein do it...



Configuration:

• Stack, Buffer, Oracle, Set of dependency relations

Operations by a classifier at each step:

- Shift
 - remove w1 from the buffer, push it onto the stack as s1
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 - pop s1 from the stack; pop s2 from the stack; then push (s2 \leftarrow s1) onto the stack
- RightArc or Reduce right
 - assert a head-dependent relation between s2 and s1 (s2 \rightarrow s1)
 - pop s1 from the stack; pop s2 from the stack; then push $(s2 \rightarrow s1)$ onto the stack

Complexity?

Oracle decisions can correspond to unlabeled or labeled arcs

- Oracle is a supervised classifier that learns a function from the configuration to the next operation
- How to extract the training set?

- Given a gold tree, how to extract the gold set of steps for training?
 - $\circ \quad \text{if LeftArc} \rightarrow \text{LeftArc}$
 - $\circ \quad \text{if RightArc} \quad$
 - if s1 dependents have been processed →
 RightArc
 - $\circ \quad \mathsf{else} \to \mathsf{Shift}$



Book the flight through Houston

- Given a gold tree, how to extract the gold set of steps for training?
 - $\circ \quad \text{if LeftArc} \rightarrow \text{LeftArc}$
 - \circ if RightArc
 - if s1 dependents have been processed \rightarrow

RightArc

 $\circ \quad \mathsf{else} \to \mathsf{Shift}$

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

Figure 13.8 Generating training items consisting of configuration/predicted action pairs by simulating a parse with a given reference parse.



Book the flight through Houston

- Oracle is a supervised classifier that learns a function from the configuration to the next operation
- Given a gold tree, how to extract the gold set of steps for training?
 - $\circ \quad \text{if LeftArc} \rightarrow \text{LeftArc}$
 - \circ if RightArc
 - if s1 dependents have been processed \rightarrow RightArc
 - $\circ \quad \mathsf{else} \to \mathsf{Shift}$
- What features to use?

Features

- POS, word-forms, lemmas on the stack/buffer
- morphological features for some languages
- previous relations
- conjunction features (e.g. Zhang&Clark'08; Huang&Sagae'10; Zhang&Nivre'11)

Source	Feature templates	5	
One word	<i>s</i> ₁ . <i>w</i>	<i>s</i> ₁ . <i>t</i>	s ₁ .wt
	<i>s</i> ₂ . <i>w</i>	<i>s</i> ₂ . <i>t</i>	s ₂ .wt
	$b_1.w$	$b_1.w$	$b_0.wt$
Two word	$s_1.w \circ s_2.w$	$s_1.t \circ s_2.t$	$s_1.t \circ b_1.w$
	$s_1.t \circ s_2.wt$	$s_1.w \circ s_2.w \circ s_2.t$	$s_1.w \circ s_1.t \circ s_2.t$
	$s_1.w \circ s_1.t \circ s_2.t$	$s_1.w \circ s_1.t$	

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

In other words, features are usually some combination of...

- Information about the word(s) left in the buffer (often just the first)
- Information about the top elements in the stack (often just the top two)

Or as A3/Eisenstein do it...

- Information about the word(s) left in the buffer (often just the first)
- Information about the top elements in the stack (often just the top two)

- Information about the word(s) left in the buffer (often just the first two)
- Information about the top elements in the stack (often just the top one)

Learning

- Before 2014: Support Vector Machines (SVMs),
- After 2014: Neural Nets



Slides by Danqi Chen & Chris Manning

Softmax probabilities



- Features
 - o s1, s2, s3, b1, b2, b3
 - leftmost/rightmost children of s1 and s2
 - leftmost/rightmost
 grandchildren of
 s1 and s2
 - \circ POS tags for the above
 - arc labels for children/grandchildren



Evaluation of Dependency Parsers

- LAS labeled attachment score
- UAS unlabeled attachment score

 $\frac{\#correct\ dependencies}{}$

 $\# of \ dependencies$



				-			
Go	old			Pa	rse	d	
1	2	She	nsubj	1	2	She	nsubj
2	0	saw	root	2	0	saw	root
3	5	the	det	3	4	the	det
4	5	video	nn	4	5	video	nsubj
5	2	lecture	obj	5	2	lecture	ccomp
				1			

Parser	UAS	LAS	sent. / s
MaltParser	89.8	87.2	469
MSTParser	91.4	88.1	10
TurboParser	92.3*	89.6*	8
C & M 2014	92.0	89.7	654

Follow-up

Method	UAS	LAS (PTB WSJ SD 3.3
Chen & Manning 2014	92.0	89.7
Weiss et al. 2015	93.99	92.05
Andor et al. 2016	94.61	92.79

Stack LSTMs (Dyer et al. 2015)

Transition-Based Dependency Parsing with Stack Long Short-Term Memory



Arc-Eager version (in A3/Eisenstein speak)

- LEFTARC: Assert a head-dependent relation from b1 to s1; pop the stack.
- RIGHTARC: Assert a head-dependent relation from s1 to b1; shift b1 to be s1 instead of dropping it from the stack/buffer.
- SHIFT: Remove b1 and push it to be the new s1.
- REDUCE: Pop the stack.



Arc-Eager

root dobj det det case							
Step Stack Word List Action Relation Added							
$\frac{\partial \alpha p}{\partial}$	[root]	[book, the, flight, through, houston]	RIGHTARC	$(root \rightarrow book)$			
1	[root, book]	[the, flight, through, houston]	SHIFT				
2	[root, book, the]	[flight, through, houston]	LEFTARC	(the \leftarrow flight)			
3	[root, book]	[flight, through, houston]	RIGHTARC	$(book \rightarrow flight)$			
4	[root, book, flight]	[through, houston]	SHIFT				
5	[root, book, flight, through]	Nouston]	LEFTARC	$(through \leftarrow houston)$			
6	[root, book, flight]	[houston]	RIGHTARC	(flight \rightarrow houston)			
7	[root, book, flight, houston]		REDUCE				
8	[root, book, flight]		REDUCE				
9	[root, book]	0	REDUCE				
10	[root]		Done				

Parsing algorithms

- Transition based
 - o greedy choice of local transitions guided by a good classifier
 - \circ deterministic
 - MaltParser (Nivre et al. 2008), Stack LSTM (Dyer et al. 2015)
- Graph based
 - Minimum Spanning Tree for a sentence
 - non-projective
 - \circ globally optimized
 - McDonald et al.'s (2005) MSTParser
 - Martins et al.'s (2009) Turbo Parser

Summary

- Transition-based
 - + Fast
 - + Rich features of context
 - - Greedy decoding
- Graph-based
 - + Exact or close to exact decoding
 - - Weaker features

Well-engineered versions of the approaches achieve comparable accuracy (on English), but make different errors

 \rightarrow combining the strategies results in a substantial boost in performance

Previewing the last few lectures of the quarter

Remaining lecture topics

- Wednesday 3/1: <u>Saadia Gabriel</u> on commonsense reasoning, factuality, also touches on toxic language detection [virtual; your choice whether to join Zoom meeting or come to classroom]
- Friday 3/3: <u>Alane Suhr</u> on multimodal NLP and grounding for NLP [virtual; your choice whether to join Zoom meeting or come to classroom]
- Monday 3/6: <u>Sewon Min</u> on prompting and in-context learning using large language models [back to in-person]
- Wednesday 3/8: <u>Akari Asai</u> on multilingual NLP [in-person]
- Friday 3/10: I'll be giving the wrap-up lecture and leaving time for Q&A/chatting about NLP more broadly. Have questions about NLP? Bring 'em!! :)

Preview of remaining announcements for the quarter

- Remember: this Wednesday's quiz is on parsing!
- A2 grades should be out sometime late tomorrow or on Wednesday; regrade requests can be submitted (via private Ed post) up to a week after A2 grades are posted on Canvas
- We'll be holding extra office hours next week (3/6-3/10) leading up to the A3 deadline (which is 11:59pm on Friday 3/10, the last day of class)
 - Mind your late days! The number of late days you have available to use on A3 is min(3, 5 - your total late days used so far this quarter). Send us a message to ask what your total late day usage so far has been if you're not sure!
- The last quiz of the quarter, Wednesday of next week (3/8), will be on a combination of topics from Saadia, Alane, and Sewon's lectures