L95: Natural Language Syntax and Parsing 6) N-best Parsing

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We have looked at the following algorithms:

- CKY
- Shift-Reduce
- A*

But so far we have discussed finding the best parse... what if we want to find the n-best parses?

• For the best parse we keep the most probable partial derivation for every non-terminal at each cell

1 2 3 0 $= \{S, NP, VP, VV, VM\}$ = $\{can, fish, they\}$ = SΣ S = { $S \rightarrow NP VP 1.0$ 1 $VP \rightarrow VM VV 0.9$ $VP \rightarrow VV NP 0.1$ $VV \rightarrow can 0.2 \mid fish 0.8$ $VM \rightarrow can 1.0$ $NP \rightarrow they 0.5 \mid fish 0.5 \}$ 2 thev can fish

$$1 \qquad 2 \qquad 3$$

$$NP_{(they)}^{0.5}$$

$$N = \{S, NP, VP, VV, VM\}$$

$$\Sigma = \{can, fish, they\}$$

$$S = S$$

$$P = \{S \rightarrow NP VP 1.0$$

$$VP \rightarrow VM VV 0.9$$

$$VP \rightarrow VV NP 0.1$$

$$VV \rightarrow can 0.2 | fish 0.8$$

$$VM \rightarrow can 1.0$$

$$NP \rightarrow they 0.5 | fish 0.5 \}$$

$$they \qquad can \qquad fish$$

$$1 2 3$$

$$NP_{(they)}^{0.5}$$

$$VV_{(can)}^{0.2}$$

$$VV_{(can)}^{1.0}$$

$$VM_{(can)}^{1.0}$$

$$VH_{(can)}^{0.2}$$

$$P = \begin{cases} S, NP, VP, VV, VM \\ \Sigma = \{can, fish, they\} \\ S = S \\ P = \{S \Rightarrow NP VP 1.0 \\ VP \Rightarrow VM VV 0.9 \\ VP \Rightarrow VM VV 0.9 \\ VP \Rightarrow VW NP 0.1 \\ VV \Rightarrow can 0.2 | fish 0.8 \\ VM \Rightarrow can 1.0 \\ NP \Rightarrow they 0.5 | fish 0.5 \}$$

$$they can fish$$

An example beam strategy:

- Discard partial derivations based on a score rather than their non-terminal type
- Discard all partial derivations whose score is less than α times the maximum score for that cell
- Typical value for α is 0.0001
- Strategy can cause some loss of accuracy



- Can apply beam dynamically at each cell
- To find n-best, select n most probable 5 parses from top right cell
- Alternatively, exploit fact that **2nd best parse will differ from best** parse by just 1 of its parsing decisions
- for nth-best parse all but one of its decisions will be involved in one of the 2nd through the (n -1)th-best parses.
- So first find the best parse, then find the second-best parse, then the third-best, and so on...
- Practically, at each cell keep an ordered list of n-best partial derivations, combine with n-best lists for adjacent partial derivations until you have exactly n to store in the new cell

Coarse-to-fine n-best strategies, Charniak

Charniak parser adopts a coarse-to-fine parsing strategy:

- 1 produce a parse forest using simple version of the grammar i.e. find possible parses using coarse-grained non-terminals, e.g. *VP*
- 2 refine most promising of coarse-grained parses using complex grammar i.e with feature-based, lexicalised non-terminals, e.g. *VP*[*buys*/*VBZ*]

Coarse-to-fine n-best strategies, Charniak

- Coarse-grained step can be efficiently parsed using e.g. CKY
- But the simple grammar **ignores contextual features** so best parse might not be accurate
- **Output a pruned packed parse** forest for the parses generated by the simple grammar (using a beam threshold)
- Evaluate remaining parses with complex grammar (i.e. each coarse-grained state is split into several fine-grained states)
- To create **n-best parses** fine-grained step keeps the n-best possibilities at each cell

Discriminative reranking is used to recover best parse

- Use parser to produce n-best list of parses
- Define an initial ranking of these parses based on original parse score
- Use **second model** (e.g. max-ent) to **improve the initial ranking** (using additional features)
- Collins re-ranking: http://www.aclweb.org/anthology/J05-1003
- Charniak re-ranking: https://dl.acm.org/citation.cfm?id=1219862
- Provides small improvements PARSEVAL metrics on Penn Treebank

Example of shift-reduce parse for the string *bacdfe*



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The shift-reduce parser is greedy

- Shift-reduce parser makes a single pass through the sentence making greedy decisions
- Makes the algorithm very efficient, O(n) for sentence length n
- Stuck with early decisions no matter how much later evidence contradicts them

Retrieve n-best shift-reduce parses using agenda

- To get the n-best parses we need to systematically explore and score alternative action sequences
- This gives rise to an exponential number of of potential sequences
- Solution is to score and filter possible sequences to within a **fixed beam size**
- Use an **agenda** to store possible buffer/stack configurations along with a score of the actions that let to that configuration
- Apply all actions to top item on the agenda and then score the resulting configurations
- Add new configurations to the agenda until the beam is full and then **replace lowest scoring items** with higher scoring ones
- Continue as long as non-terminating configurations exist on the agenda (guarantees best parse will be found)

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Score reflects action-sequences rather than actions

- In the greedy algorithm the classifier acted as an oracle actions are scored
- With the **beam search** we want to score action sequences action sequences are scored
- Notice that **beam** here is constrained by the size of the agenda

N-best dependency parse algorithm

function DEPENDENCYBEAMPARSE(words, width) returns dependency tree

```
state \leftarrow \{[root], [words], [], 0.0\}; initial configuration agenda \leftarrow \langle state \rangle; initial agenda
```

```
while agenda contains non-final states

newagenda \leftarrow \langle \rangle

for each state \in agenda do

for all {t | t \in VALIDOPERATORS(state)} do

child \leftarrow APPLY(t, state)

newagenda \leftarrow ADDTOBEAM(child, newagenda, width)

agenda \leftarrow newagenda

return BESTOF(agenda)
```

function ADDTOBEAM(state, agenda, width) returns updated agenda

```
if LENGTH(agenda) < width then
    agenda ← INSERT(state, agenda)
else if SCORE(state) > SCORE(WORSTOF(agenda))
    agenda ← REMOVE(WORSTOF(agenda))
    agenda ← INSERT(state, agenda)
return agenda
```

```
Psuedo code from Jurafsky and Martin version 3
```

n-best shift-reduce parser example in class



- Lexicalised PCFGs
- More on features and training...