

## Lecture 11

1. Generative grammars, PCFG
  - a. CYK algorithm
2. Limitations of PCFG
  - a. Lexicalized PCFG
  - b. Other grammars
3. Dependency parsing: Graph-based parsing
  - a. Scoring
  - b. Inference
  - c. Learning

### Two inference problems for PCFG

1.  $T^* = \operatorname{argmax}_{T \in T(S)} p(S, T)$ ; parsing
2.  $p(S) = \sum_{T \in T(S)} p(S, T)$ ; language modeling

Recall that any PCFG can be written in Chomsky Normal Form:

$$X \rightarrow Y_1 Y_2, X, Y_1, Y_2 \in N$$

$$X \rightarrow Y, X \in N, Y \in \Sigma$$

Advantage of PCFG: Efficient calculation via dynamic programming

Limitation: Can only capture short term dependencies

### CYK Algorithm

$\pi[i, j, k]$  = max probability of a tree that starts at position  $i$  and ends at position  $j$  from non-terminal  $k$

Assume non-terminals are numbered:  $N = \{N_1, \dots, N_K\}$ ,  $N_1 = S$

Want to find  $\pi[1, n, S]$ .

Base Case:

$$\pi[i, i, k] = p(N_k \rightarrow w_i | N_k)$$

Recursive Case:

$$\pi[i, j, k] = \max_{l, m, s} \pi[i, s, l] \pi[s, j, m] p(N_l, N_m | N_k)$$

Complexity:  $O(n^3 k^3)$ , can be made  $O(n^3 |G|)$

# of subproblems:  $n^2 K$

Time per subproblem:  $n K^2$

How can we modify the algorithm for language modeling? Take sum instead of max in recursion.

### **Limitations**

She ate pasta with a fork.

She ate pasta with butter.

The grammars do not care about the actual words in the sentence – it only says which tree is more likely. So in the ambiguity above, it will always parse one of the sentences above incorrectly.

### **Lexicalization**

Recursively propagate head information

How to compute? Compute as normal, but do some linear interpolation as well.