

The CKY algorithm part 2: Probabilistic parsing

Syntactic analysis/parsing

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Based on slides from Marco Kuhlmann





Recap: The CKY algorithm



The CKY algorithm

The CKY algorithm is an efficient bottom—up parsing algorithm for context-free grammars.

We use it to solve the following tasks:

- Recognition:Is there any parse tree at all?
- Probabilistic parsing:
 What is the most probable parse tree?

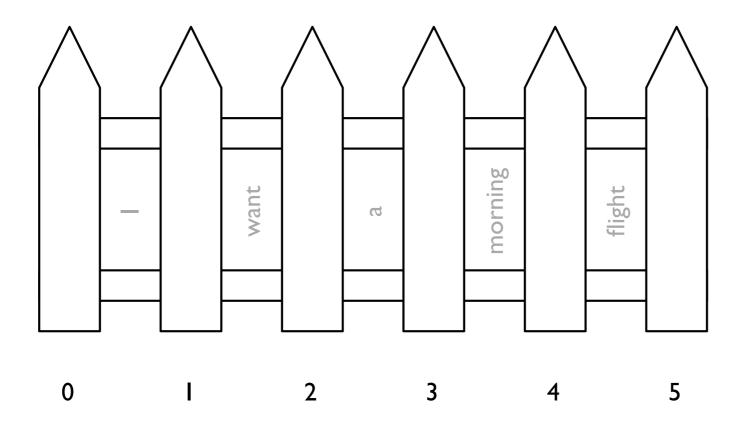
Restrictions

- The CKY algorithm as we present it here can only handle rules that are at most binary: $C \rightarrow w_i$, $C \rightarrow C_1$ C_2 , $(C \rightarrow C_1)$
- This restriction is not a problem theoretically, but requires preprocessing (binarization) and postprocessing (debinarization).
- A parsing algorithm that does away with this restriction is Earley's algorithm (J&M 13.4.2).



Fencepost positions

We view the sequence w as a fence with n holes, one hole for each token w_i , and we number the fenceposts from 0 till n.





Implementation

- The implementation uses a three-dimensional array chart.
- Whenever we have recognized a parse tree that spans all words between min and max and whose root node is labeled with C, we set the entry chart[min][max][C] to true.



Implementation

Implementation: preterminal rules

```
for each w<sub>i</sub> from left to right
  for each preterminal rule C -> w<sub>i</sub>
    chart[i - 1][i][C] = true
```



Implementation: Binary rules

```
for each max from 2 to n

for each min from max - 2 down to 0

for each syntactic category C

for each binary rule C -> C1 C2

for each mid from min + 1 to max - 1

if chart[min][mid][C1] and chart[mid][max][C2] then

chart[min][max][C] = true
```

Question

How do we need to extend the code in order to handle unary rules $C \rightarrow C_l$?

Unary rules

```
for each max from 1 to n

for each min from max - 1 down to 0

// First, try all binary rules as before.

...

// Then, try all unary rules.

for each syntactic category C

for each unary rule C -> C1

if chart[min][max][C1] then

chart[min][max][C] = true
```



Implementation

Question

This is not quite right.
Why, and how could we fix the problem?



Structure

- Is there any parse tree at all?
- What is the most probable parse tree?





What is the most probable parse tree?

- The number of possible parse trees grows rapidly with the length of the input.
- But not all parse trees are equally useful.

 Example: I booked a flight from Los Angeles.
- In many applications, we want the 'best' parse tree, or the first few best trees.
- Special case: 'best' = 'most probable'



Probabilistic context-free grammars

A probabilistic context-free grammar (PCFG) is a context-free grammar where

- each rule r has been assigned a probability p(r) between 0 and 1
- the probabilities of rules with the same left-hand side sum up to I

Example

	Made up probabilitie
Rule	Probability
S → NPVP	
NP → Pronoun	1/3
NP → Proper-Noun	1/3
NP → Det Nominal	1/3
Nominal → Nominal PP	1/3
Nominal → Noun	2/3
VP → Verb NP	8/9
VP → Verb NP PP	1/9
PP → Preposition NP	



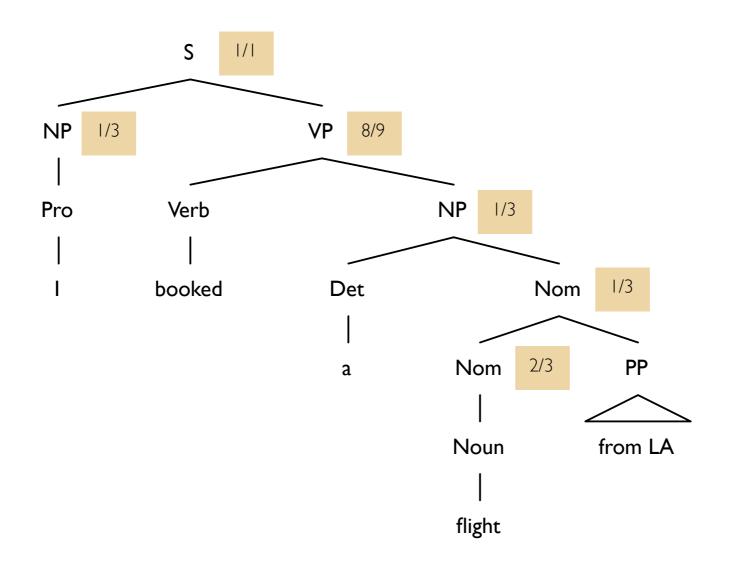


The probability of a parse tree

The probability of a parse tree is defined as the product of the probabilities of the rules that have been used to build the parse tree.



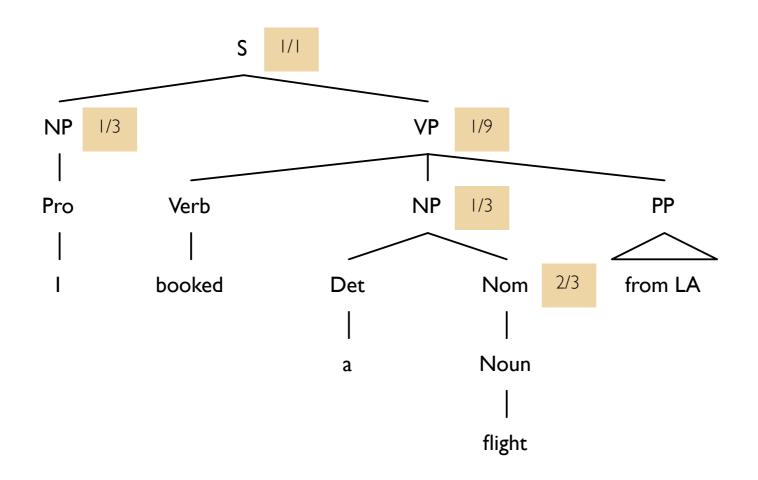
Example



Probability: 16/729



Example



Probability: 6/729



Small trees





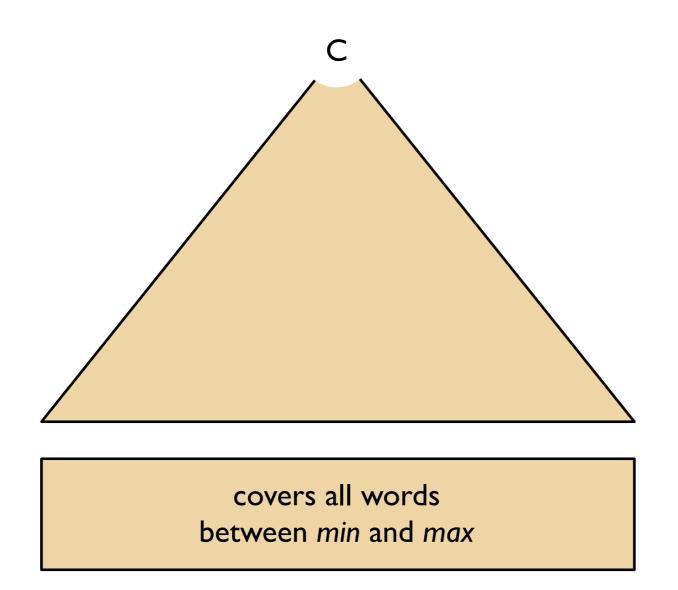
Small trees



Choose the most probable rule!

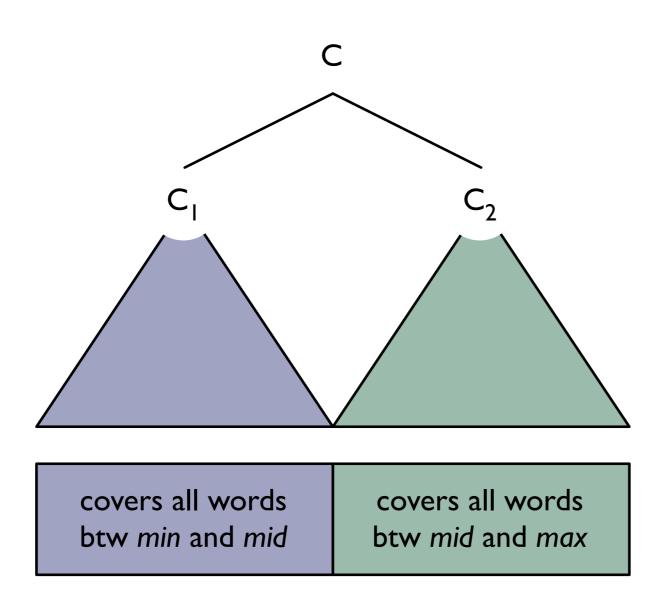


Small trees





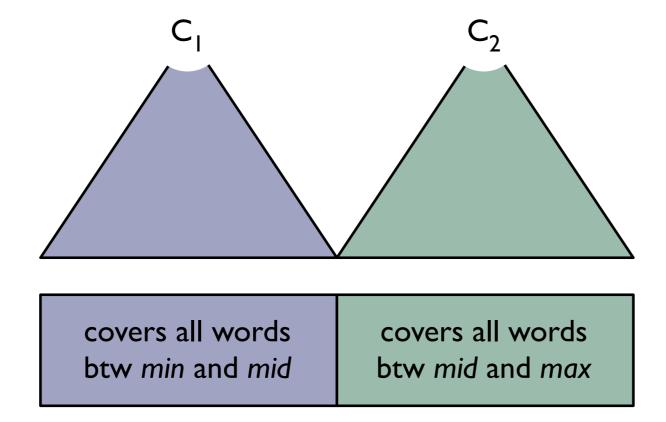
Big trees



Big trees



Choose the most probable rule!





Idea

- For trees built using preterminal rules:
 Find a most probable rule. (apply all rules!)
- For trees built using binary rules: Find a binary rule r and a split point mid such that $p(r) \times p(t_1) \times p(t_2)$ is maximal, where t_1 is a most probable left subtree and t_2 is a most probable right subtree.



Implementation

- Instead of an array with Boolean values, we now have an array with probabilities, i.e., *floats*.
- When all is done, we want to have
 chart[min][max][C] = p
 if and only if a most probable parse tree with
 signature [min, max, C] has probability p.

Preterminal rules

```
for each w_i from left to right

for each preterminal rule C \rightarrow w_i

chart[i - 1][i][C] = p(C \rightarrow w_i)
```



Binary rules

```
for each max from 2 to n
  for each min from max - 2 down to 0
    for each syntactic category C
       double best = undefined
       for each binary rule C -> C<sub>1</sub> C<sub>2</sub>
         for each mid from min + 1 to max - 1
           double t_1 = chart[min][mid][C_1]
           double t<sub>2</sub> = chart[mid][max][C<sub>2</sub>]
           double candidate = t_1 * t_2 * p(C \rightarrow C_1 C_2)
            if candidate > best then
              best = candidate
       chart[min][max][C] = best
```



Question

How should we treat unary rules?





One more question

The only thing that we have done so far is to compute the *probability* of the most probable parse tree. But how does that parse tree look like?



Backpointers

- When we find a new best parse tree,
 we want to remember how we built it.
- For each element t = chart[min][max][C],
 we also store backpointers to those elements
 from which t was built.
- Besides the ordinary chart of floats, we also have a backpointer chart

Backpointers

```
double best = undefined
Backpointer backpointer = undefined
if candidate > best then
 best = candidate
  // We found a better tree; update the backpointer!
 backpointer = (C \rightarrow C_1 C_2, min, mid, max)
chart[min][max][C] = best
backpointerChart[min][max][C] = backpointer
```

Implementation

Implementation ideas

```
# defaultdict is a suitable datastructure for charts!

pi = defaultdict(float)

bp = defaultdict(tuple)

# Recognize all parse trees built with with preterminal rules.

# Recognize all parse trees built with binary rules.

# "S" is not always the top category, the below is a simplification return backtrace(bp[0, n, "S"], bp);
```



Assignment 1/2: CKY parsing

• Tips:

- Use print statements to make sure your code does what you think it should
- Use a small test set, and possibly a small grammar during dvelopment. The parser is slow
- Start on the assignment now! Do not leave it until the last week!



Advanced models

- The CKY model is used in many competitive parsers
- To improve performance the grammar is often modified, e.g. by
 - Parent annotation
 - Lexicalized rules



Summary

- The CKY algorithm is an efficient parsing algorithm for context-free grammars.
- Today, we have used it for probabilistic parsing:
 The task of computing the most probable parse tree for a given sentence.

Coming up

- Thursday:
 - Lecture on treebanks and PCFGs
- Next week:
 - Tuesday: Earley's algorithm
 - Thursday:
 - Advanced PCFG parsing (~ 45 minutes)
 - Time for supervision (~ 45 minutes)

Own work

- Reading:
 - CKY: J&M 14.1, 14.2
 - Treebanks: J&M12.4, 14.3, 14.7
- Work on assignments
 - Start working on assignment 1/2: CKY
 - Contact me if you need help
 - You can also ask questions in the lectures

Deadlines

- Ordinary deadlines
 - Assignment I+2 (b) I (m): I7-I2-04
 - Assignment 2 (m): 17-12-18
 - Assignment 3+4: 18-01-12
- Resubmission deadline
 - All assignments: 18-02-09
- If you fail to meet these deadlines you will have to wait for the next time the course is given (Jan-Mar 2018)
- Assignments will only be graded in connection with each deadline
- In case of special circumstances, contact me BEFORE the deadline it concerns!