Lecture: Dynamic Programming III

Overview

- Subproblems for strings
- edit distance (& longest common subseq)
- knapsack

5 Easy Steps to Dynamic Programming:

- (1) define subproblems | count # subproblems
- (2) guess (part of solution) | count # choices
- (3) relate subproblems solutions | compute (time/subproblem)
- (4) recurse or memoize or build DP table bottom up | time = (time/subproblem) * #subproblems
- (5) **solve original problem** | a subproblem or a combination of subproblems

Fibonacci

Shortest Paths

Fk for
$$1 \le k \le n$$

$$\delta k(s, v)$$
 for $v \in V$, $0 \le k < |V|$

$$F_k = F_{k-1} + F_{k-2}$$

O(1) to compute

$$\delta k(s, v) = min\{\delta k-1(s, u)+w(u, v)\}$$

O(1 + indegree(v))

$$\delta_{|V|-1}$$
(s, v) for $v \in V$
O(V)

Subproblems for string and sequences:

- (1) suffixes $x[i:] \mid O(|x|)$
- (2) prefixes x[:j] O(|x|)
- (3) substrings x[i:j] $|O(|x|^2)$

Edit distance

Description:

Given two strings x & y, what is the cheapest possible sequence of character edits (insert c, delete c, replace $c \rightarrow c$) to transform x into y?

- cost of edit depends only on characters alphabet
- for example more common typo a -> s, then less cost
- cost of sequence = sum of costs of edits

Edit distance

Description:

Given two strings x & y, what is the cheapest possible sequence of character edits (insert c, delete c, replace $c \rightarrow c$) to transform x into y?

• If *insert* & *delete* cost 1, replace costs 0, minimum *edit distance* equivalent to finding *longest common subsequence*. Note that a subsequence is sequential but not necessarily contiguous.

• HIEROGLYPHOLOGY vs. MICHAELANGELO =⇒ HELLO

Edit distance

Subproblems for multiple strings/sequences:

- combine suffix/prefix/substring subproblems
- multiply state spaces

1) subproblems# subproblems

2) guess # choices

3) recurrence time / subproblem

Edit distance

```
c(i, j) = \text{edit-distance}(x[i: ], y[j: ]) for 0 \le i < |x|, 0 \le j < |y|
\Theta(|x| \cdot |y|) subproblems
```

guess whether, to turn x into y (3 choices):

- x[i] deleted
- y[j] inserted
- x[i] replaced by y[j]

c(i, j) = maximum of:

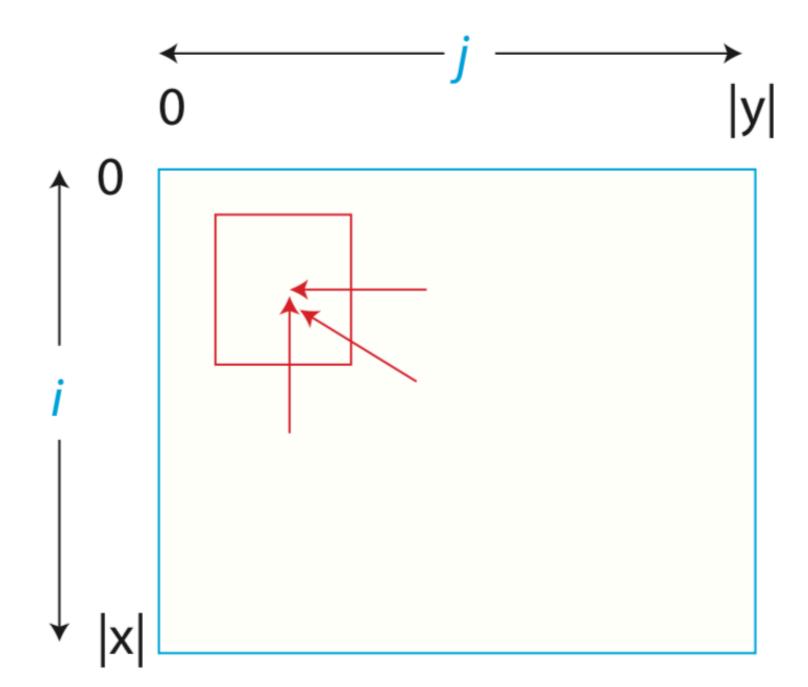
- cost(delete x[i]) + c(i + 1, j), if i < |x|,
- cost(insert y[j]) + c(i, j + 1), if j < |y|,
- $cost(replace x[i] \rightarrow y[j]) + c(i+1, j+1), if i < |x| & j < |y|$

4) topological order total time

Edit distance

DAG in 2D table:

- bottom-up OR right to left
- only need to keep last 2 rows/columns
- ⇒ linear space
- total time = $\Theta(|\mathbf{x}| \cdot |\mathbf{y}|)$



5) original subproblem

c(0, 0)

Knapsack

Description:

Knapsack of size S you want to pack

- item i has integer size si & real value vi
- goal: choose subset of items of maximum total value subject to total size ≤ S

- 1) subproblems# subproblems
- 2) guess # choices
- 3) recurrence time / subproblem

Knapsack

value for *suffix i:* | ⊖(|number of items|) subproblems

guessing = whether to include item i or not \Rightarrow # choices = 2

DP[i] = max(DP[i + 1],
$$vi$$
 + DP[i + 1]), if $si \le S$)

 not enough information. Don't know how much space is left!

Knapsack

1) subproblem# subproblems

value for *suffix i* given knapsack of size *X:* O(nS)

2) guess # choices

guessing = whether to include item i or not \Rightarrow # choices = 2

3) recurrence time / subproblem

DP[i, X] = max(DP[i + 1, X],
$$vi$$
 + DP[i + 1, X - si]), if $si \le X$)
DP[n, X] = 0

(4) topological order total time

for i in n,...,0: for X in 0,...S total time = O(nS) Examples: Knapsack

1) subproblems DP [0, S]

5 Easy Steps to Dynamic Programming:

- (1) define subproblems
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- (3) relate subproblems solutions
- (4) recurse or memoize or build DP table bottom up
- (5) solve original problem

Questions