

# Lecture 21: Dynamic Programming III

## Lecture Overview

- Subproblems for strings
- Parenthesization
- Edit distance (& longest common subseq.)
- Knapsack
- Pseudopolynomial Time

## Review:

- \* 5 easy steps to dynamic programming

- |   |                                 |
|---|---------------------------------|
| (a) define subproblems                      | count # subproblems             |
| (b) guess (part of solution)                | count # choices                 |
| (c) relate subproblem solutions             | compute time/subproblem         |
| (d) recurse + memoize                       | time = time/subproblem · # sub- |
| problems                                    |                                 |
| OR build DP table bottom-up                 |                                 |
| check subproblems acyclic/topological order |                                 |
| (e) solve original problem: = a subproblem  |                                 |
| OR by combining subproblem solutions        | ⇒ extra time                    |

- \* problems from L20 (text justification, Blackjack) are on sequences (words, cards)

- \* useful problems for strings/sequences  $x$ :

suffixes $x[i:]$	}	$\Theta( x )$	← cheaper ⇒ use if possible
prefixes $x[:i]$			
substrings $x[i:j]$	}	$\Theta(x^2)$	

## Parenthesization:

Optimal evaluation of associative expression  $A[0] \cdot A[1] \cdots A[n-1]$  — e.g., multiplying rectangular matrices

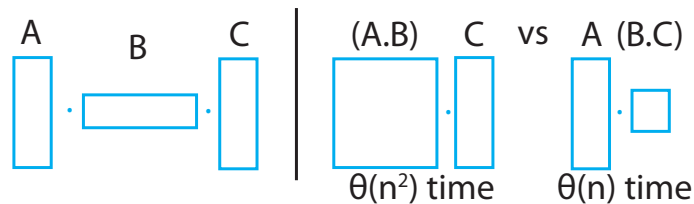


Figure 1:

2. guessing = outermost multiplication  $\underbrace{(\dots)}_{\uparrow_{k-1}} \underbrace{(\dots)}_{\uparrow_k}$

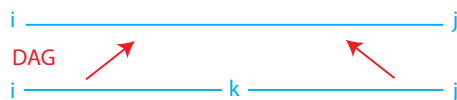
$\Rightarrow$  # choices =  $O(n)$

1. subproblems = ~~prefixes & suffixes?~~ **NO**  
= cost of substring  $A[i : j]$

$\Rightarrow$  # subproblems =  $\Theta(n^2)$

3. recurrence:

- $DP[i, j] = \min(DP[i, k] + DP[k, j] + \text{cost of multiplying } (A[i] \cdots A[k-1]) \text{ by } (A[k] \cdots A[j-1]))$  for  $k$  in range( $i+1, j$ )



- $DP[i, i+1] = 0$   
 $\Rightarrow$  cost per subproblem =  $O(j-i) = O(n)$

4. topological order: increasing substring size. Total time =  $O(n^3)$

5. original problem =  $DP[0, n]$   
(& use parent pointers to recover parens.)

NOTE: Above DP is not shortest paths in the subproblem DAG! Two dependencies  $\Rightarrow$  not path!

## Edit Distance

Used for DNA comparison, diff, CVS/SVN/..., spellchecking (typos), plagiarism detection, etc.

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  $c, c'$
- for example in DNA,  $C \rightarrow G$  common mutation  $\implies$  low cost
- cost of sequence = sum of costs of edits
- 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.
- for example **H I E R O G L Y P H O L O G Y** vs. **M I C H A E L A N G E L O**  
 $\implies$  **HELLO**

### Subproblems for multiple strings/sequences

- combine suffix/prefix/substring subproblems
- multiply state spaces
- still polynomial for  $O(1)$  strings

### Edit Distance DP

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

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

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

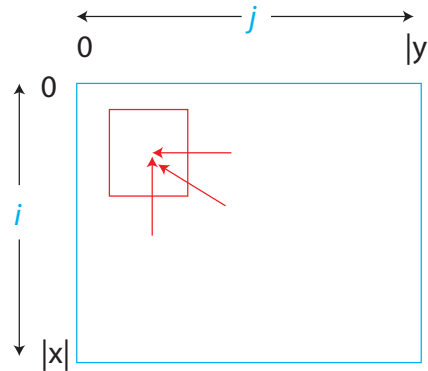
(3) recurrence:  $c(i, j) = \text{maximum of}$ :

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

base case:  $c(|x|, |y|) = 0$

$\implies \Theta(1)$  time per subproblem

(4) topological order: DAG in 2D table:



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

(5) original problem:  $c(0, 0)$

## Knapsack:

Knapsack of size  $S$  you want to pack

- item  $i$  has integer size  $s_i$  & real value  $v_i$
- goal: choose subset of items of maximum total value subject to total size  $\leq S$

### First Attempt:

1. ~~subproblem = value for suffix  $i$~~ : **WRONG**
2. guessing = whether to include item  $i \implies \# \text{ choices} = 2$
3. recurrence:
  - $DP[i] = \max(DP[i+1], v_i + DP[i+1] \text{ if } s_i \leq S?)$
  - **not enough information to know whether item  $i$  fits — how much space is left?**  
**GUESS!**

### Correct:

1. subproblem = value for suffix  $i$ :  
     given knapsack of size  $X$   
 $\implies \# \text{ subproblems} = O(nS)$       **!**

3. recurrence:

- $DP[i, X] = \max(DP[i + 1, X], v_i + DP[i + 1, X - s_i] \text{ if } s_i \leq X)$
- $DP[n, X] = 0$   
 $\implies$  time per subproblem  $= O(1)$

4. topological order: for  $i$  in  $n, \dots, 0$ : for  $X$  in  $0, \dots, S$

total time  $= O(nS)$

5. original problem  $= DP[0, S]$

(& use parent pointers to recover subset)

AMAZING: effectively trying all possible subsets! ...but is this actually fast?

## Polynomial time

Polynomial time = polynomial in input size

- here  $\Theta(n)$  if number  $S$  fits in a word
- $O(n \lg S)$  in general
- $S$  is exponential in  $\lg S$  (not polynomial)

## Pseudopolynomial Time

Pseudopolynomial time = polynomial in the problem size AND the numbers (here:  $S$ ,  $s_i$ 's,  $v_i$ 's) in input.  $\Theta(nS)$  is pseudopolynomial.

Remember:  
 polynomial — GOOD  
 exponential — BAD  
 pseudopoly — SO SO

MIT OpenCourseWare  
<http://ocw.mit.edu>

6.006 Introduction to Algorithms  
Fall 2011

For information about citing these materials or our Terms of Use, visit: <http://ocw.mit.edu/terms>.