

TODAY: Dynamic Programming III (of 4)

- subproblems for strings
- parenthesization
- edit distance (& longest common subseq.)
- knapsack
- pseudopolynomial time

* 5 easy steps to dynamic programming:

- ① define subproblems
- ② guess (part of solution)
- ③ relate subprob. solutions
- ④ recurse + memoize

count # subprobs.

count # choices

compute time/subprob.

time = time/subprob.

• # subprobs.

OR build DP table bottom-up

- check subprobs. acyclic/topological order

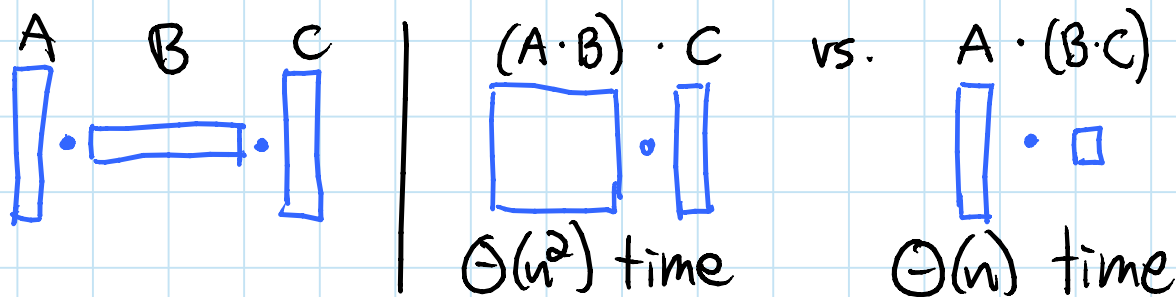
- ⑤ solve original problem: = a subproblem
OR by combining subprob. solutions (\Rightarrow extra time)

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

* useful subproblems for strings/sequences x :

- L20 \rightarrow
- suffixes $x[i:]$
 - prefixes $x[:i]$
 - substrings $x[i:j]$
- $\left. \begin{array}{l} \text{suffixes } x[i:] \\ \text{prefixes } x[:i] \end{array} \right\} \Theta(|x|) \leftarrow \text{cheaper} \Rightarrow \text{use if possible}$
- $\text{substrings } x[i:j] \left\} \Theta(|x|^2)\right.$

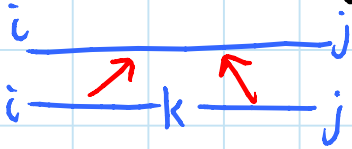
Parenthesization: optimal evaluation of an associative expression $A[0] \cdot A[1] \cdot \dots \cdot A[n-1]$
 - e.g. multiplying rectangular matrices



② guessing = outermost multiplication: $(\dots)(\dots)$
 \Rightarrow # choices = $O(n)$

① subproblems = ~~prefixes & suffixes?~~ **NO**
 = cost of substring $A[i:j]$
 \Rightarrow # subproblems = $\Theta(n^2)$

③ recurrence:

DAG: 

- $DP[i, j] = \min \left(DP[i, k] + DP[k, j] + \text{cost of } (A[i] \dots A[k-1]) \cdot (A[k] \dots A[j-1]) \right)$
 for k in range($i+1, j$)

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

④ topological order: increasing substring size
 - total time = $\Theta(n^3)$

⑤ original problem = $DP[\emptyset, n]$
 (& 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's the cheapest possible sequence of character edits to transform x into y ?

← insert c ↓ delete c → replace $c \rightarrow c'$

- cost of edit depends only on characters c, c'
- e.g. in DNA, $C \rightarrow G$ common mutation \Rightarrow low cost
- cost of sequence = sum of costs of edits

- if insert & delete cost 1, replace costs 0, min. edit distance equivalent to finding longest common subsequence
sequential but not necessarily contiguous

- e.g.: HIEROGLYPHOLOGY } HELLO
vs. MICHAELANGELO

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 $0 \leq i < |x|, 0 \leq j < |y|$

$\Rightarrow \Theta(|x| \cdot |y|)$ subproblems

② guess whether, to turn x into y ,

- $x[i]$ deleted

- $y[j]$ inserted

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

} 3 choices

③ recurrence: $c(i, j) = \max \{$
 $\text{cost}(\text{delete } x[i]) + c(i+1, j)$ 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: $c(|x|, |y|) = 0$

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

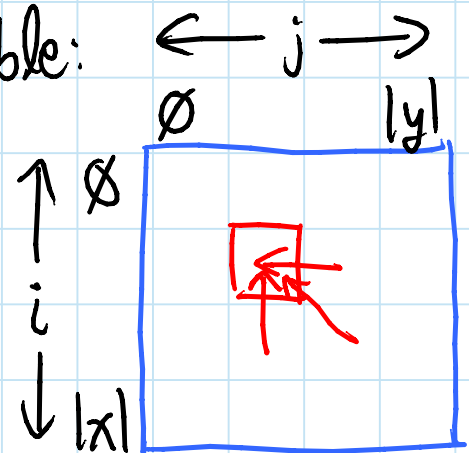
④ topological order: DAG in 2D table:

- bottom up OR right to left

- only need to keep last
2 rows/columns

\Rightarrow linear space

- total time = $\Theta(|x| \cdot |y|)$



⑤ original problem: $c(0, 0)$

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 max. total value subject to total size $\leq S$

First attempt:

- ① ~~subproblem = value for suffix i :~~ **WRONG**
 - ② guessing = whether to include item i
 \Rightarrow #choices = 2
 - ③ recurrence:
 - $DP[i] = \max(DP[i+1], v_i + DP[i+1])$ 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
 \Rightarrow #subproblems = $O(n \cdot S)$ (!)
- ③ recurrence:
 - $DP[i, X] = \max(DP[i+1, X], v_i + DP[i+1, X - s_i])$ if $s_i \leq X$
 - $DP[n, X] = \emptyset$
 - \Rightarrow time per subproblem = $O(1)$
- ④ topological order: for i in $n, \dots, 0$: for X in $0, \dots, S$
- total time = $O(n \cdot S)$
- 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 in input size
- here $\Theta(n)$ if number S fits in a word
- $\Theta(n \lg S)$ in general
- S is exponential in $\lg S$ (not polynomial)

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

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Remember: polynomial - GOOD
exponential - BAD
pseudopoly. - SO SO