

Dynamic Programming

What is it?

- similar to Divide & Conquer
 - ↳ solves problems by combining solⁿs to subproblems
- when subproblems overlap
- solve subproblem only once
 - ↳ store in memory (table)
- optimization problems
 - ↳ many solutions → want the optimal one

4 Steps in Dynamic Programming

- ① Characterize the structure of an optimal solⁿ.
- ② Recursively define the value of an optimal solⁿ.
- ③ Compute the value of an optimal solⁿ (typically Bottom up)
- ④ Compute optimal solⁿ from computed information

Longest Common sequence

- given 2 sequences

1. $X = \{x_1, \dots, x_m\}$

2. $Y = \{y_1, \dots, y_n\}$

Goal: Find a subsequence common to both whose length is longest.

A subsequence doesn't have to be consecutive, but it has to be in order

example: horseback
 | | |
 snowflake

Brute Force Approach:

→ For every subsequence of X ,
check whether it's a subsequence

of Y .

ex: $X = \{a, b, c, d\}$

subsequences: a, b, c, d

ab, ac, ad, bc, bd, cd

abc, acd, abd, bcd

$abcd$

$$\text{Total} = 16 = 2^m$$

→ each subsequence takes $\Theta(n)$ time to check:

• check y_1, y_2, \dots, y

$$\text{Total} = \Theta(n \cdot 2^m)$$

Dynamic Programming Approach

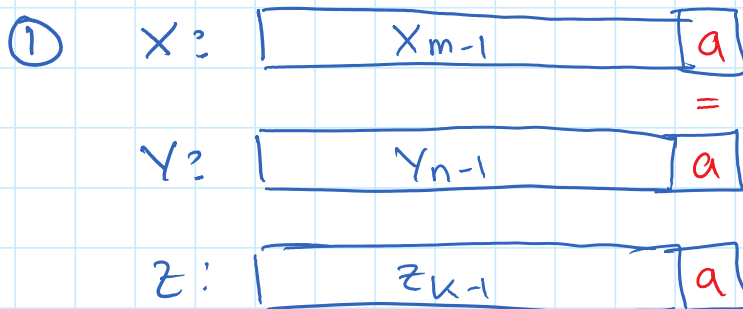
$X_i = \text{prefix}(x_1, \dots, x_i)$

$Y_i = \text{prefix}(y_1, \dots, y_i)$

Thm: Let $Z = \langle z_1, \dots, z_k \rangle$ be any LCS of X & Y

- ① If $x_m = y_n$: $z_k = x_m = y_n$ and $z_{k-1} = \text{LCS}(X_{m-1}, Y_{n-1})$
- ② If $x_m \neq y_n$: $z_k \neq x_m$ and $z_{k-1} = \text{LCS}(X_{m-1}, Y)$
- ③ If $x_m \neq y_n$: $z_k \neq y_n$ and $z_{k-1} = \text{LCS}(X, Y_{n-1})$

Visual Interpretation



$$z_{k-1} = \text{LCS}(X_{m-1}, Y_{n-1})$$

$$c_{k-1} = \text{LCS}(x_{m-1}, y_{n-1})$$

②

X: x_{m-1} a

\neq

Y: y_{n-1} b

Z: z_{k-1} b

$$z_{k-1} = \text{LCS}(x_{m-1}, y)$$

③

X: x_{m-1} a

\neq

Y: y_{n-1} b

Z: z_{k-1} a

$$z_{k-1} = \text{LCS}(x, y_{n-1})$$

Recursive Formulation

Define (memory) $c[i, j]$ = length of $\text{LCS}(x_i, y_j)$
 want: $c[m, n]$

$$c[i, j] = \begin{cases} 0 & \text{if } i=0 \text{ or } j=0 \\ c[i-1, j-1] + 1 & \text{if } i, j > 0 \text{ and } x_i = y_i \\ \max(c[i-1, j], c[i, j-1]) & \text{if } i, j > 0 \text{ and } x_i \neq y_i \end{cases}$$

Longest common sequence algorithm:

LCS-LENGTH(X, Y, m, n)

for $i \leftarrow 1$ to m

do $c[i, 0] \leftarrow 0$

for $j \leftarrow 0$ to n

do $c[0, j] \leftarrow 0$

for $i \leftarrow 1$ to m

do for $j \leftarrow 1$ to n

do if $x_i = y_j$

then $c[i, j] \leftarrow c[i - 1, j - 1] + 1$

$b[i, j] \leftarrow \nwarrow$

else if $c[i - 1, j] \geq c[i, j - 1]$

then $c[i, j] \leftarrow c[i - 1, j]$

$b[i, j] \leftarrow \uparrow$

else $c[i, j] \leftarrow c[i, j - 1]$

$b[i, j] \leftarrow \leftarrow$

return c and b

$O(m \cdot n)$

$O(n)$

$X = \{A, B, C, B, D, A, B\}$ $m = 7$

$Y = \{B, D, C, A, B, A\}$ $n = 6$

$Z = \{B, C, B, A\}$ $k = 4$

		0	1	2	3	4	5	6
	y_j	B	D	C	A	B	A	
x_i	0	0	0	0	0	0	0	0
→ 1	A	0	0	↑	0	↑	1	↑
→ 2	B	0	↑	1	↑	1	0	↑
→ 3	C	0	↑	↑	2	↑	2	2
→ 4	B	0	↑	↑	2	↑	3	↑
→ 5	D	0	↑	2	↑	2	↑	3
→ 6	A	0	↑	2	↑	3	↑	4
→ 7	B	0	↑	2	↑	3	↑	4

for $i = 1$ to 7

for $j = 1$ to 6

$(1, 1) : 0$

$(1, 2) : 1$

$(1, 3) :$

Sol: ①: B C B A

②: B C A B

- Naive approach: exponential ($O(2^m \cdot n)$)
- Dynamic programming: polynomial ($O(m \cdot n)$)

* Read Chapter 15: Dynamic Programming.