

# Dynamic Programming (DP)

- ① Optimization /  
max min longest
  - ② design recursion / (think recursively)
  - ③ run time /  
  
if impl using recursion naively  
Same computation
  - ④ - bottom up  
- top down w/ memoization

## Longest Common Subsequence (LCS)

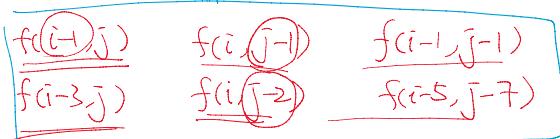
- input  $A = [\underline{3}, \underline{5}, \underline{1}, 2, \underline{4}]$   
 $B = [\underline{10}, 1, \underline{3}, \underline{5}, \underline{4}]$

sub seq      [3, 4]      sub seq of A

- optimization      length      (longest)
- size of      problem / instance

A = [      ]       $(i, j)$   
 B = [      ]      len of A      len of B

what we want given  $A[1:i]$  and  $B[i:j]$



"last" of something

common subseq



$$\begin{aligned} A &\in \{1, 3\} \\ B &\in \{1, 5\} \end{aligned}$$

case 1 :  $A[i] = B[j]$

$$f(i, j) = 1 + f(i-1, j-1)$$

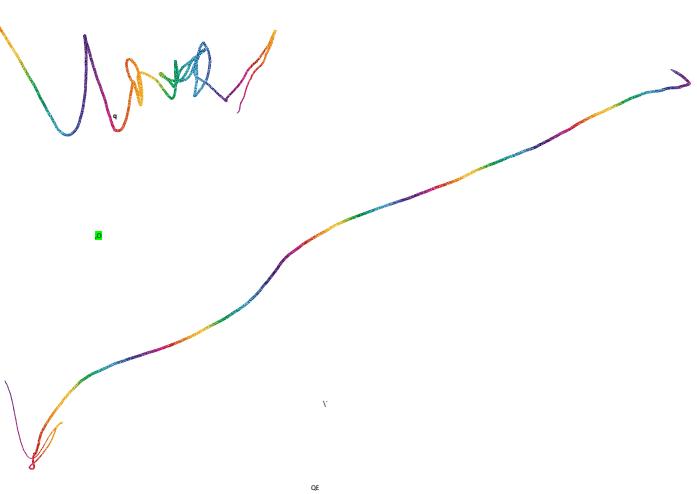
Case 2:  $A[i] \neq B[j]$

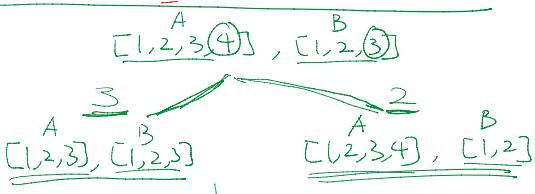
$$f(i, j) = \max\{f(\underline{i-1}, j), f(i, \underline{j-1})\}$$

Base Case:  $i=0$  or  $j=0$

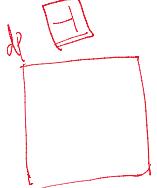
$$f(0, j) = f(i, 0) = 0$$







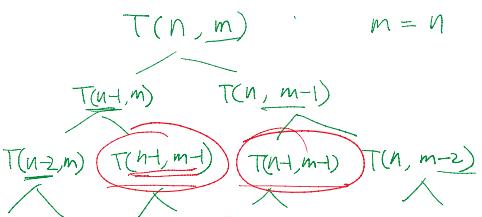
naive recursive impl.



```

LCS(A[1..i], B[1..j]) {
    if (i == 0 || j == 0) return 0;
    if (A[i] == B[j])
        return 1 + LCS(A[1..i-1], B[1..j-1]);
    else
        return max(LCS(A[1..i], B[1..j-1]),
                   LCS(A[1..i-1], B[1..j]));
}

```



Bottom Up

	0	1	2	m
0	0	0	0	-
1	0	0	0	-
2	0	0	0	-
i	0	0	0	-
n	0	0	0	-

$f(n, m)$

$\text{LCS}(A[1..n], B[1..m])$

```

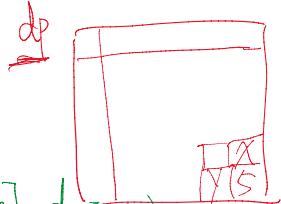
for (i=0 ... n)
    dp[i, 0] = 0;
for (j=0 ... m)
    dp[0, j] = 0;
for (i = 1 ... n) {
    for (j=1 ... m) {
        if (AC[i] == BC[j])
            dp[i, j] = 1 + dp[i-1, j-1];
        else
            dp[i, j] = max(dp[i-1, j], dp[i, j-1]);
    }
}
return dp[n, m];

```

## LCS of A & B

length of LCS

② A 1 ③ ← ⌂ DF1 ⌂ J



LCS(A[0..i], B[0..j], dp[i][j])

if((i==0 || j==0)) return 0;

If ( $dp[i][j] \neq -1$ ) return  $dp[i][j]$ ;

if ( $A[i] == B[j]$ )

$$dp[i][j] = 1 + \underline{LCS(A[1..i-1], B[1..j-1], dp)}$$

$$dp[i][j] = \max \left( lcs(A[1..i-1], B[1..j], dp[i-1]) \right. \\ \left. lcs(A[1..i], B[1..j-1], dp[i]) \right)$$

return dp[i][j];

A diagram illustrating a dependency or flow between two concepts. A curved red arrow originates from the word "Sub seq" and points towards the phrase "length of subseq". Both terms are enclosed in rectangular boxes.

$$A = \{1, 3, 5, 7, 9\}$$

$$B = \underline{\underline{[3, 7, 1, 2]}}$$

$$\text{E)} \quad \sqrt{A} = 4$$

dp	0	1	2	3
0	0	0	0	0
1	0	0	0	1
2	0			
3	0			
4	0			
5	0			



length of LCS

find-lcs(A[1..n], B[1..m], d[i][j])  
 $i=n, j=m;$   
 $d = [ ];$

while ( $i > 0 \text{ and } j > 0$ )  
 $i = i - 1;$   
 $j = j - 1;$

if ( $A[i] == B[j]$ )

$d[i].append(A[i]);$

$i = i - 1;$

$j = j - 1;$

else if ( $d[i-1][j] < d[i-1][j-1]$ )

$j = j - 1;$

else {

$i = i - 1;$

}

return  $d.reverse();$

}

$O(n+m)$

$A = [1, 2, 3]$

$B = [1, 3]$

	0	1	2
0	0	0	0
1	0	1	1
2	0	1	1
3	0	1	2

$i=0, j=0$

$l=[3, 1]$

$[1, 3]$

0-1 knapsack

-  $n$  items

- each item  $i$  value[i] weight[i]

- capacity of knapsack is  $C$

- what is  $\max \text{ #}$

naive enumerate all possible subsets



naive enumerate all possible subsets

$$2^n$$

$f(i, j) := \max \$$  given first  $i$  items  
and knapsack capacity is  $j$

smaller problems

item  $\frac{1}{\text{item}}$   $\frac{i}{\text{item}}$

less items

smaller knapsack

take  $i$ th item

$$\underline{f(i-1, j - \text{weight}[i]) + \text{value}[i]}$$

leave  $i$ th item alone

$$\underline{f(i-1, j)}$$

$$\underline{f(i, j) = \max (\underline{f(i-1, j - \text{weight}[i]) + \text{value}[i]}, \underline{f(i-1, j)})}$$

base case  $f(0, j) = f(i, 0) = \underline{0}$

$$T(n, C)$$

$$T(n-1, C-1) \quad T(n-1, C)$$

$$T(n-2, C-2)$$

$$T(n-2, C-1)$$

$$T(n-2, C-1)$$

$$T(n-2, C)$$

$n$

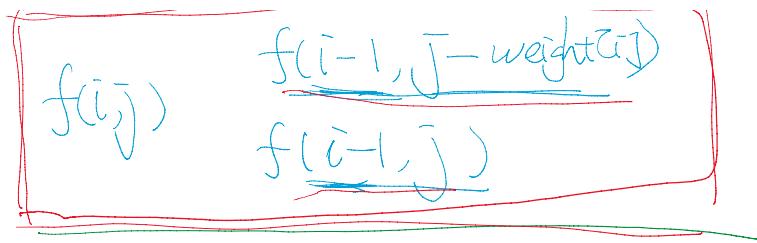
$n$

$f(i-1)$

$$\underline{f(i-1, j - \text{weight}[i])}$$

eight(i)≤j

$i$	0	1	2	3	4	$\dots$	$n$	$C$
0	0	0	0	0	0	$\dots$	0	0
1	0	0	0	0	0	$\dots$	0	0
2	0	0	0	0	0	$\dots$	0	0
3	0	0	0	0	0	$\dots$	0	0
4	0	0	0	0	0	$\dots$	0	0
$\vdots$								
$n$	0	0	0	0	0	$\dots$	0	0



for ( $i = 1 \dots n$ ) {

    for ( $j = 1 \dots C$ ) {

        if ( $weight[i] \leq j$ )

$dp[i,j] = \max(dp[i-1,j - weight[i]], dp[i-1,j])$

        else

$dp[i,j] = dp[i-1,j]$

} } return  $dp[n,C]$

knapsack(  $\dots$ ,  $weight[1 \dots i]$ ,  $value[1 \dots i]$ ,  $j$ ,  $dp[i]$  ) {

    if ( $i == 0 \text{ || } j == 0$ ) return 0

    if ( $dp[i,j] \neq -1$ ) return  $dp[i,j]$

$dp[i,j] = knapsack( \dots, weight[1 \dots i], value[1 \dots i], j - weight[i], dp[i-1] )$

$\alpha[i] + \text{value}[i]$ ,  
 $\beta[i]$

$i - \underline{i}$ ,

$\beta$ :

$[i, j]$ ;

$\text{right}[\underline{i}] - \underline{i}$

$\text{map}$   
 $i$

$(i, j) \Rightarrow \$$



$\boxed{\underline{dp[i][j] = knapsack(\dots, weight[i-1], value[i-1], \dots)}}$   
 if ( $weight[i] \leq j$ ) {  
 $\boxed{dp[i][j] = \max(dp[i-1][j], knapsack(\dots, weight[i-1], value[i-1], \dots))}$   
 }  
 return  $\underline{dp[i][j]}$

bottom up VS

run time

easy to see

space complexity

$O(nC)$

$f(i, j)$

$\frac{f(i-1, j-w_i)}{f(i-1, j)}$

~~weight[1..i-1]~~  
~~j, dp[i])~~

~~DP~~

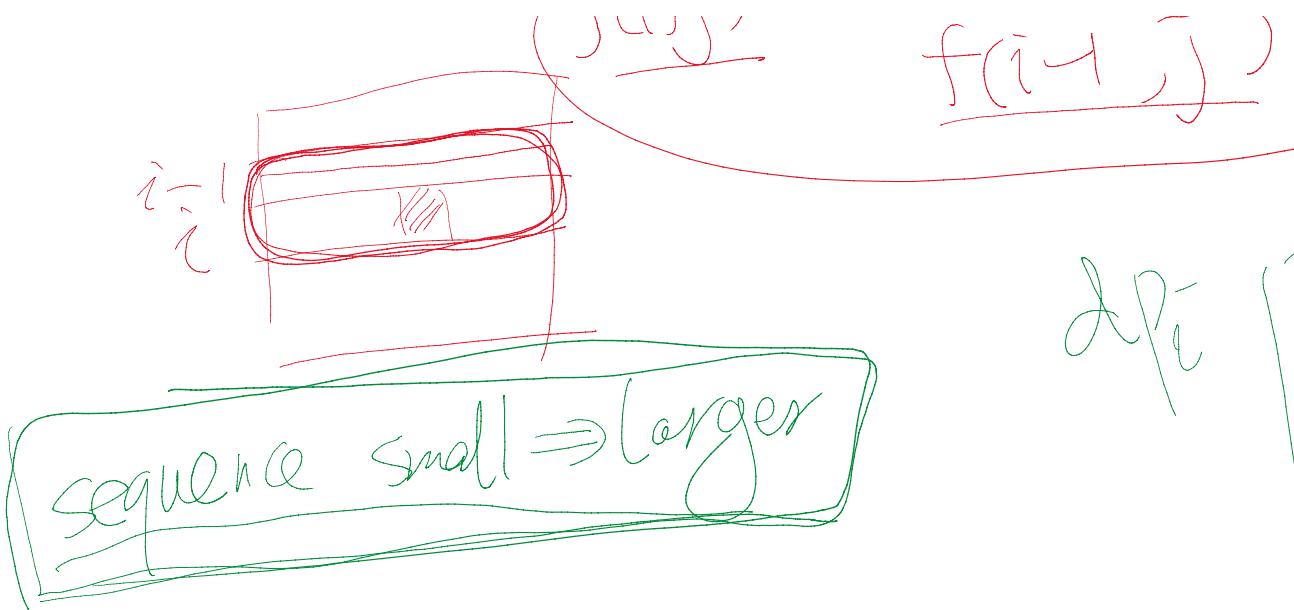
~~... , weight[1..i-1],  
net[1..i+1], j-weight[i], dp[i]) + value[i])~~

top down

harder to see

$O(nC)$

$i, j$



personal preferenc.

n questions in the ex

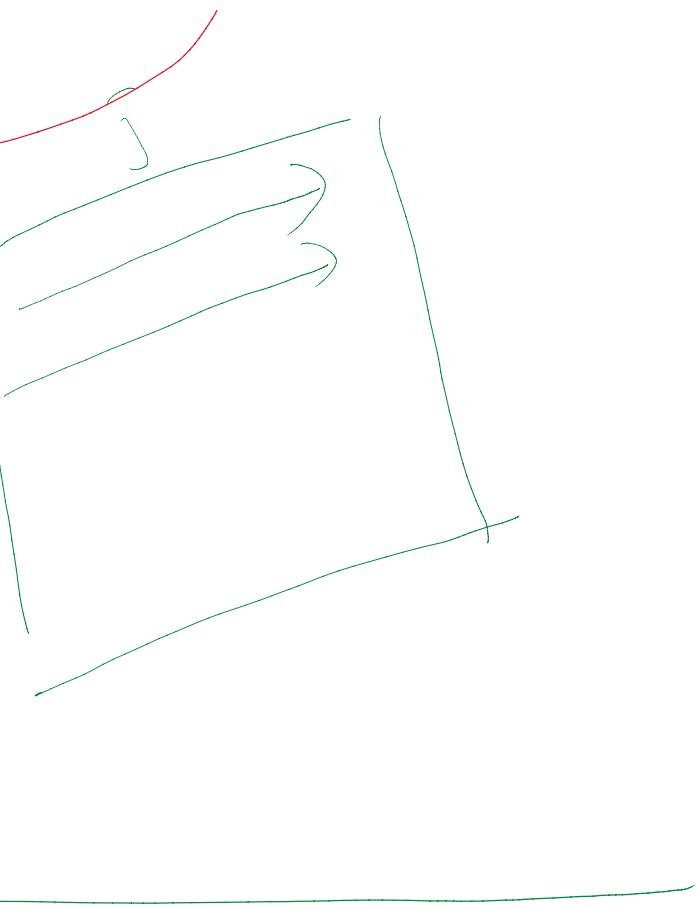
soreti]

timeTi]

I don't give partial

T min

max



in.

credit

~~Sub Seq~~

$A = [1, 2, 3]$

$[1, 3]$

$[3, 17]$

---

3, 4]

✓

✗