

# Lecture #11

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CS 170

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Spring 2021

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# Dynamic Programming (DP)

General Approach to many problems:

Solve a big problem by breaking it into smaller subproblems, solve subproblems in order from "small" to "large"

Isn't this recursion?

```
func Fib(n)
  if  $n \leq 1$  return n
  else Fib(n-1)
    + Fib(n-2)
```

cost =  $O(\text{Fib}(n)) \sim 1.6^n$

fix-memoization

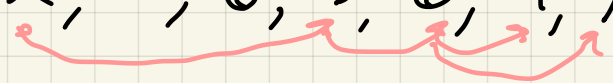
```
Fib(0) = 0, Fib(1) = 1
for i = 2 to n
  Fib(i) =
    Fib(i-1) + Fib(i-2)
```

cost =  $O(n)$

# Dynamic Programming - Examples

- Shortest Path in DAG
- Longest Increasing Subsequence

5, 2, 8, 6, 3, 6, 9, 7



LIS length = 4 vertices

- Edit Distance = fewest # edits to change one string to another (delete, insert, substitute)
  - Spell checking
  - How similar is my DNA to your DNA
  - Cheating detection
- Knapsack: given  $n$  items with weights  $w_1, \dots, w_n$  values  $v_1, \dots, v_n$ , total weight limit =  $W$   
how to choose subset  $S$  of items with  $\max \sum_{i \in S} v_i$   
with  $\sum_{i \in S} w_i \leq W$
- All-pairs-Shortest-Paths (better than  $n \times$  Bellman-Ford)
- TSP - Traveling Salesperson Problem ...

# Shortest Paths in DAGs - DP point of view

- Given  $G(V, E)$ ,  $w(e) \in \mathbb{Z}$ , find shortest path from  $s \in V$  to  $v \in V$  (negative  $w(e)$  OK, no cycles)

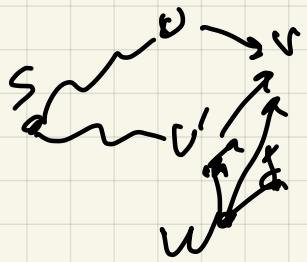
- Approach

1) Define subproblems: find shortest path from  $s$  to  $v'$  for  $v'$  "closer to"  $s$  than  $v$

2) Show how to solve a problem given solutions to subproblems:  $\text{dist}(v) = \min_{u: (u,v) \in E} \text{dist}(u) + w(u,v)$

3) Base case:  $\text{dist}(s) = 0$ ,  $\text{dist}(w) = \infty$  if  $w$  is a source-

4) Choose order to solve subproblems  
- topologically sort vertices starting at  $s$

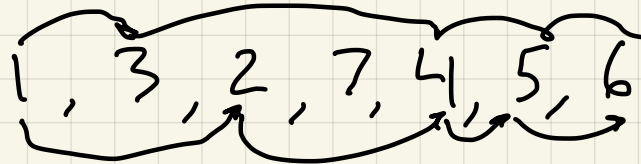


What if we wanted longest path?

$w \rightarrow -w$ ,  $\min \rightarrow \max$

# Longest Increasing Subsequence (LIS)

Given  $n$  unsorted numbers  $x_1, \dots, x_n$  find LIS



~~1) Subproblems:  $f(i) = \text{LIS in } x_1, \dots, x_i$~~

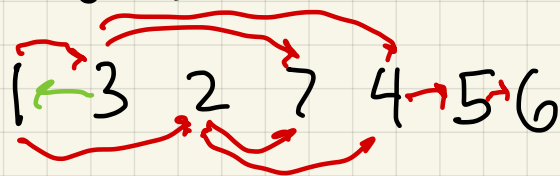
~~2) Solve using subproblems~~      5 6 7 1 2 3 4      oops

1) Subproblems:  $L(i) = \text{length of LIS in } x_1, \dots, x_i \text{ ending } x_i$

2) Solve using subprobs:  $L(i) = \max(1, \max_{\substack{j < i \\ x_j < x_i}} (L(j) + 1))$

3) Base:  $L(1) = 1$

4) Order to solve: increasing  $i$



$$\text{Cost} = \sum_{i=1}^n O(i) = O(n^2)$$

Red edges form DAG  $x_i \rightarrow x_j$  if  $x_i < x_j$ : Longest Path in DAG

# Edit Distance

- How similar are 2 strings  $[x_1, \dots, x_n], [y_1, \dots, y_m]$ ?
- How many "edits" needed to change  $x$  to  $y$ ?

Edit means insert, delete or substitute a char.

snowy    \_ s n o w \_ y  
sunny    s u n \_ \_ n y  
          +1 +1    +1 +1 +1  
          #edits = 5

s \_ n o w y    s \_ n o w y  
s u n n \_ y    s u n \_ n y  
          +1    +1 +1    +1 +1  
          #edits = 3    #edits = 3

Motivation: spell-checking - suggest fixes  
DNA matching  
cheat detection  
spam filtering

Edit Distance (ED) between  $x = [x_1, \dots, x_n]$  and  $y = [y_1, \dots, y_m]$

1) Subproblems: for all  $1 \leq i \leq n$ ,  $1 \leq j \leq m$   
 $f(i, j) = \text{ED}([x_1, \dots, x_i], [y_1, \dots, y_j])$

2) Look at last char in optimal alignment, could be

$$\begin{array}{ccc} \overline{x_i} & \overline{y_j} & \overline{x_i} \\ \text{remove } x_i & \text{insert } y_j & x_i \end{array}$$

Cost =  $f(i-1, j) + 1$        $f(i, j-1) + 1$        $f(i-1, j-1) + \begin{cases} 0 & \text{if } x_i = y_j \\ 1 & \text{if } x_i \neq y_j \end{cases}$

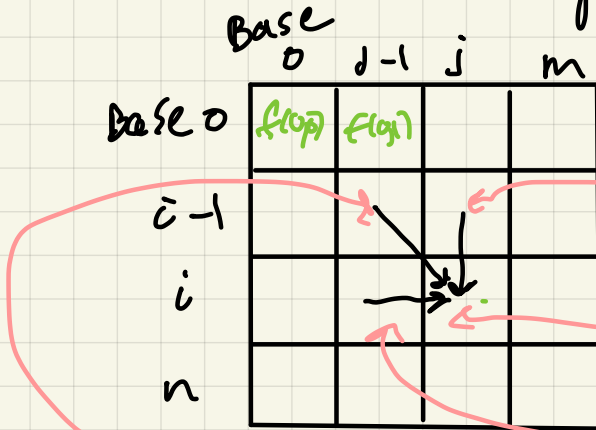
$d(i, j) = \begin{cases} 0 & x_i = y_j \\ 1 & \text{otherwise} \end{cases}$

$$f(i, j) = \min(f(i-1, j) + 1, f(i, j-1) + 1, f(i-1, j-1) + \delta(i, j))$$

3) Base case:  $f(i, 0) = i$  (deletes),  $f(0, j) = j$  (insert)

4) Order: for  $f(i, j)$  need  $f(i-1, j)$ ,  $f(i, j-1)$ ,  
 $f(i-1, j-1)$

# Order of Subproblems for ED



DAG for computing  $f(i,j)$

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

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

$$f(i-1,j-1) + d_{ij})$$

y = S h o w y  
0 1 2 3 4 5  $\rightarrow$  insert y  
j

rowwise, columnwise

$$\text{cost} = O(n \cdot m)$$

$$\text{Memory} = O(\min \begin{pmatrix} m & \text{rowwise} \\ n & \text{columnwise} \end{pmatrix})$$

Implicitly shortest path in DAG with edge weight  $\rightarrow \downarrow \searrow d_{ij}$

delete  $x_i$

x	0	0	1	2	3	4	5
h	1	1	0	1	2	3	4
o	2	2	1	1	2	3	4
w	3	3	2	1	2	3	4
y	4	4	3	2	2	3	4
y	5	5	4	3	3	3	3
i							

$\uparrow$   
ED

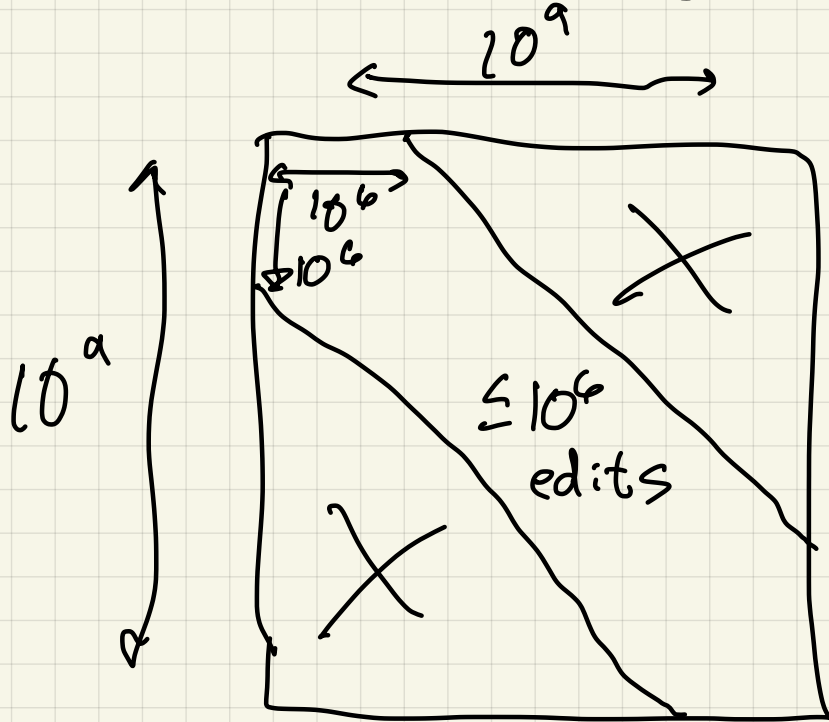


What about  $O(m \cdot n)$  cost if  $m, n = O(10^9)$ ?

Why? sequencing DNA

any 2 people have 99.9% same DNA

$$\Rightarrow ED = O(.001 \cdot 10^9) = O(10^6)$$



only compute  $f(\bar{i}, j)$

for  $|\bar{i} - j| = O(10^6)$

$$\Rightarrow \text{cost}(10^6 \cdot 10^9) = O(10^{15})$$

Smith-Waterman

Needleman-Wunsch

Metaphlan

# Knapsack Problem

- Suppose you are robbing a jewelry store

deciding how to invest

You have

Knapsack can carry  
 $W$  lbs

$\$W$  to invest

You have to choose among  
 $n$  jewels, values  $v_1, \dots, v_n$   
weights  $w_1, \dots, w_n$

$n$  investments  
likely payoffs  $v_1, \dots, v_n$   
costs  $w_1, \dots, w_n$

Which should you pick?

$S \in \{1, 2, \dots, n\}$

to maximize  $\sum_{i \in S} v_i$  subject to  $\sum_{i \in S} w_i \leq W$

Does a greedy algorithm work?

$$W = 20$$

$$w_1 = 11$$

$$v_1 = 15$$

$$w_2 = 10$$

$$v_2 = 8$$

$$w_3 = 10$$

$$v_3 = 8$$

$$\left. \begin{array}{l} w_1 = 11 \\ w_2 = 10 \\ w_3 = 10 \end{array} \right\} \begin{array}{l} v_1 = 15 \\ v_2 = 8 \\ v_3 = 8 \end{array} \quad \begin{array}{l} w_2 + w_3 = 20 \\ v_2 + v_3 = 16 \end{array}$$

Greedy: choose  $i$  to maximize  $v_i \rightarrow v_1 = 15$   
" " " "  $\frac{v_i}{w_i} \rightarrow v_1 = 15$

# Knapsack by Dynamic Programming

- 1) Subproblems:  $f(i, v) = \text{max value packing}$   
subset of  $1, \dots, i$  max weight  $v \leq W$
- 2) Solve  $f(i, v) =$  if  $w_i > v \dots w_i \text{ too heavy}$   
 $f(i-1, v)$   
else ... could pack  $w_i$   
 $\max(f(i-1, v), v_i + f(i-1, v - w_i))$   
Don't pack  $w_i$       pack  $w_i$
- 3) Base Case:  $f(0, v) = 0$ ,  $f(i, 0) = 0$
- 4) Order: for  $i = 1$  to  $n$   
for  $v = 1$  to  $W$   
 $f(i, v) = \dots$  step 2

# Cost of Knapsack

for  $i=0$  to  $n$ ,  $f(i,0)=0$ ; for  $v=0$  to  $W$ ,  $f(0,v)=0$

for  $i=1$  to  $n$   
for  $v=1$  to  $W$

6 (i) { if  $w_i > v$   
           $f(i,v) = f(i-1,v)$   
      else  
           $f(i,v) = \max(f(i-1,v), f(i-1,v-w_i) + v_i)$   
      end if

Cost =  $O(n \cdot W)$ : Polynomial Time?

In "size of input"  $v_1, \dots, v_n, w_1, \dots, w_n, W$

Size of input  $O(n(\log_2 \max_i v_i + \log_2 W))$

$n \cdot W$  can be exponentially larger

Knapsack is NP-complete Chap 8