

CS4800: Algorithms & Data

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Lecture 9:

- Dynamic Programming:
 - Tug of War / Subset Sum / Knapsacks
 - Edit Distance / Alignments

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Tug of War
Subset-Sum
Knapsack

Tug-of-War

Goal: Lighter team should have wt as close to $\frac{T}{2}$ as possible using

$$T = 148$$

integer weight

- We have n students with weights $w_1, \dots, w_n \in \mathbb{N}$, need to split as evenly as possible into two teams

- e.g. $\{21, 42, 33, 52\}$

$$A_1 = \{21, 42\} \quad 63$$

$$A_2 = \{33, 52\} \quad 85$$

$$\Delta = 22$$

$$\left. \begin{array}{l} A_1 = \{42, 33\} \quad 75 \\ A_2 = \{21, 52\} \quad 73 \\ \Delta = 2 \end{array} \right\}$$

$$T = \sum_{i=1}^n w_i$$

the lighter team has wt at most $\frac{T}{2}$.



Tug-of-War

- **Input:** weights $w_1, \dots, w_n \in \mathbb{N}$ for n students
 - Define $T = \sum_i w_i$
- **Output:** a subset of students S with weight as large as possible but not more than $T/2$
 - $S \subseteq \{1, \dots, n\}, W_S = \sum_{i \in S} w_i$



Subset Sum (generalization of ToW)

$$M \text{ ToW } T = \frac{1}{2} \sum_{i=1}^n w_i$$

- **Input:** weights $w_1, \dots, w_n \in \mathbb{N}$ for n items, and a maximum weight $T > 0$
- **Output:** a subset of students S with weight as large as possible but not more than T
 - $S \subseteq \{1, \dots, n\}, W_S = \sum_{i \in S} w_i$

Tug-of-War

$w_1, \dots, w_n \in \mathbb{N}$ $T > 0$ is the max w.t.
(not necessarily sorted)

- Let O be the **optimal** subset (wt as close to T as possible w/o going over)
- O either contains n or it doesn't.
- Case I ($n \notin O$): O is the optimal set among (w_1, \dots, w_{n-1}, T)
- Case II ($n \in O$): O is the [optimal set among $(w_1, \dots, w_{n-1}, T - w_n)$] + w_n
- $\text{OPT}(i, t)$ the weight of the optimal sol'n w/ items $1, \dots, i$, max wt t .

$$\text{OPT}(n, T) = \max \{ \text{OPT}(n-1, T), w_n + \text{OPT}(n-1, T - w_n) \}$$

Ask the Audience

$OPT(i, t)$: value of OPT for v_1, \dots, v_i and wt t .

$$OPT(n, T) = \max \left\{ OPT(n-1, T), v_n + OPT(n-1, T - v_n) \right\}$$

- Input: $T = 7, w_1 = 1, w_2 = 3, w_3 = 5$
- Fill the dynamic programming table

$OPT(0, t) = 0$
 $OPT(i, 0) = 0$
 $OPT(i, -w) = -\infty$



items

3	0	1	1	3	4	5	6	6
2	0	1	1	3	4	4	4	4
1	0	1	1	1	1	1	1	1
0	0	0	0	0	0	0	0	0
-	0	1	2	3	4	5	6	7

weights



OPT sdn
 $\rightarrow \{1, 3\}$
 w/ wt ~~4~~ 6.

Tug-of-War (Bottom-Up)

ToW(w_1, \dots, w_n, T):

Let $M[0:n, 0:T]$ be an array to store the solutions

$M[0, t] \leftarrow 0$ for all $0 \leq t \leq T$ $M[i, 0] = 0 \quad 1 \leq i \leq n$

For $i = 1, \dots, n$:

For $t = 1, \dots, T$:

if $w_i \leq t$ $M[i, t] \leftarrow \max\{M[i-1, t], w_i + M[i-1, t - w_i]\}$

if $w_i > t$ $M[i, t] \leftarrow M[i-1, t]$

Return $M[n, T]$

After running ToW, need another alg to trace through M finding the optimal set.

Knapsack ^{ToW is} $[w_i = v_i]$

- **Input: weights and values** $(w_1, v_1), \dots, (w_n, v_n)$ for n items, and a maximum weight $T > 0$
- **Output:** a subset of students S with **value** as large as possible but **weight** at most T

- $S \subseteq \{1, \dots, n\}$, $W_S = \sum_{i \in S} w_i$, $V_S = \sum_{i \in S} v_i$

$$\leq T$$

thing you're trying to optimize

$OPT(n, T)$ is the value of the opt soln

$$OPT(n, T) = \max \left\{ OPT(n-1, T), v_n + OPT(n-1, T-w_n) \right\}$$

Ask the Audience

- Let $OPT(i, t)$ be the optimal solution for items $1, \dots, i$ with weight at most t .
- $OPT(i, t) = ???$

Summary

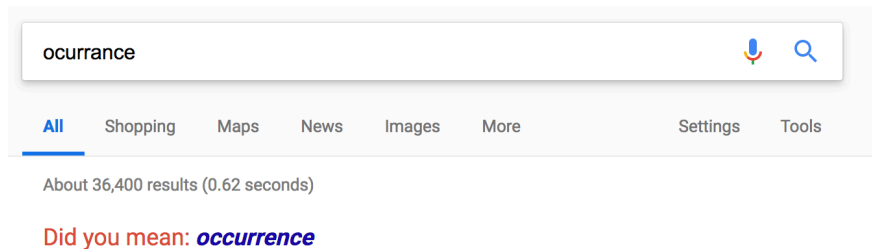
→ table has size $(n+1)(T+1)$

- Can also solve subset sum and knapsack problems in $\Theta(nT)$ time where T is the ~~maximum~~ maximum weight
 - Note, dependence on T is rather undesirable, what if $M = 2^{168}$?
- Tradeoff btw precision of sol'n and running time

Edit Distance Alignments

Distance Between Strings

- Autocorrect is (usually) pretty good



- **ocurrance** and **occurrence** seem similar, but only if we define similarity carefully

ocurrance
occurrence

7 apart

oc **urrance**
occurrence

2 apart

Edit Distance / Alignments

minimum

for today $\Sigma = \{a, \dots, z\}$

- Given two strings $x \in \Sigma^n$, $y \in \Sigma^m$, the **edit distance** is the number of **insertions**, **deletions**, and **swaps** required to turn x into y .
replace a w/ b.

- Given an **alignment**, the cost is the number of positions where the two strings don't agree

x	o	c		u	r	r	a	n	c	e
y	o	c	c	u	r	r	e	n	c	e

edit distance is 2.

edit distance $d \iff$ alignment of cost d .

Ask the Audience

- What is the minimum cost alignment of the strings **moon** and **money**

m	o	o	n		
m		o	n	e	y

cost 3

m o o n

m o n e y

cost 3

Edit Distance / Alignments

$O(1)$ time to check if

$$x_i = y_j$$

- **Input:** Two strings $x \in \Sigma^n, y \in \Sigma^m$
- **Output:** The minimum cost alignment of x and y
edit distance

Dynamic Programming $x_1 \dots x_n$

$y_1 \dots y_m$

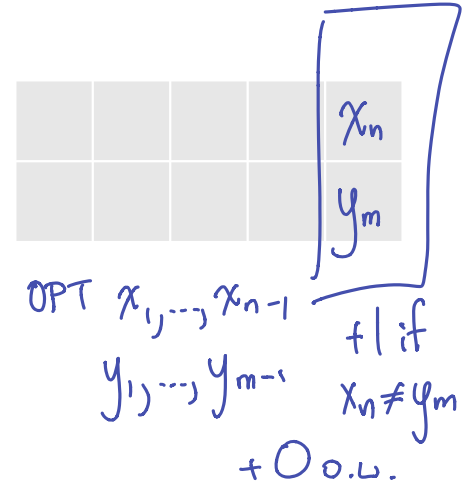
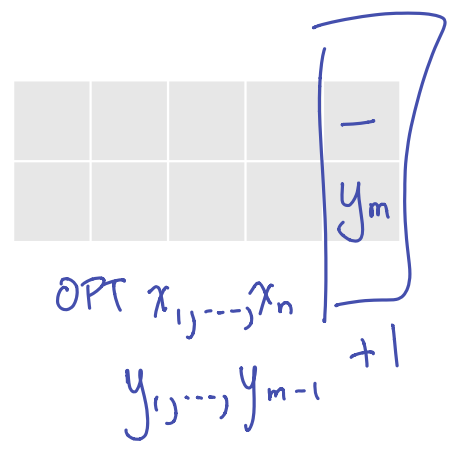
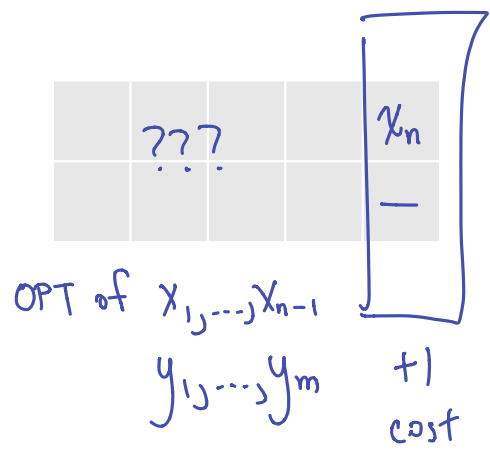
- As always, consider the **optimal** alignment...
- Three choices for the final column
 - Case I: only use x ($x_n, -$)
 - Case II: only use y ($-, y_m$)
 - Case III: use one symbol from each (x_n, y_m)

Do not consider $(-, -)$

case I

case II

case III



Recurrence:

Let $OPT(i, j)$ be the edit dist / cost of opt. alignment
of the prefixes x_1, \dots, x_i
 y_1, \dots, y_j

$$OPT(i, j) = \begin{cases} \min \{ OPT(i-1, j) + 1, OPT(i, j-1) + 1, OPT(i-1, j-1) \} \\ \text{if } x_i = y_j \\ \\ 1 + \min \{ OPT(i-1, j), OPT(i, j-1), OPT(i-1, j-1) \} \\ \text{if } x_i \neq y_j \end{cases}$$

$$OPT(i, 0) = i$$

$$OPT(0, j) = j$$

$$OPT(0, 0) = 0$$

Base cases where ^{at least} one string is empty

Dynamic Programming

- As always, consider the **optimal** alignment...
- Three choices for the final column
 - Case I: only use $x (x_n, -)$
 - Case II: only use $y (-, y_m)$
 - Case III: use one symbol from each (x_n, y_m)

Example

x = **peas**

y = **east**

P | e | a | s | t

	-	e	a	s	t
-	0	1	2	3	4
p	1	1	2	3	4
e	2	1	2	3	4
a	3	2	1	2	3
s	4	3	2	1	2

Ask the Audience

- Suppose adding a space costs $\delta > 0$ and aligning a, b costs $c_{a,b} > 0$. Write a recurrence for the minimum cost alignment.

Summary

- Can compute the **edit distance**, or **minimum cost alignment** between two strings in time $O(nm)$
 - Works for any alphabet, assuming we can decide if two symbols are equal in $O(1)$ time
- There was nothing special about our notion of alignment cost
 - **In general:** **cost δ for using a space (cases I/II)** and **costs $c_{a,b}$ for aligning $a, b \in \Sigma$ (case III)**
 - Can still solve in $O(nm)$ time
- Uses $O(nm)$ **space**, more prohibitive in practice