

LECTURE #12

CS 170

Spring 2021



Last time:

- introduction to dynamic programming ↗ \approx recursion + memoization (with explicit solving order)
- examples
 - shortest paths in a DAG
 - longest increasing subsequence
 - edit distance
 - Knapsack (w/o repetition)

DP Recipe:

1. define problems
2. set boundaries
3. give recurrence
4. specify order (and give time)

Today:

More examples of DP solutions:

- Knapsack with repetition
- chain matrix multiplication
- all-pairs shortest paths
- traveling salesperson problem

Knapsack With Repetition

Given a maximum weight W and value-weight pairs $(v_1, w_1), \dots, (v_n, w_n)$, output a multi-set of weight $\leq W$ of highest value.

(ok to pick the same item more than once)

- Problems: $K(w) := \text{max value achievable with capacity } \leq w$
- Boundaries: $K(0) = 0$
- Recurrence: $K(w) := \max_{i=1, \dots, n} \{ K(w - w_i) + v_i \mid w_i \leq w \}$
- Efficiency:
 - solve in increasing weight w : $K(0), K(w_1), \dots, K(W)$
 - each computation takes $O(n)$ time
 - total time is $O(n \cdot W)$

The algorithm is only weakly polynomial time.

Polynomial time would have been $\text{poly}(n, \log W)$,

but we don't expect this because the problem is NP-complete.

Chain Matrix Multiplication

Multiplying x -by- y matrix A and y -by- z matrix B takes xyz ops.

(We have also seen how to do better but this is not important for now.)

Today: find strategy to multiply matrices A_1, A_2, \dots, A_n as cheaply as possible.

Example: 50-by-1 A , 1-by-50 B , 50-by-1 C . The result is 50-by-1 R .

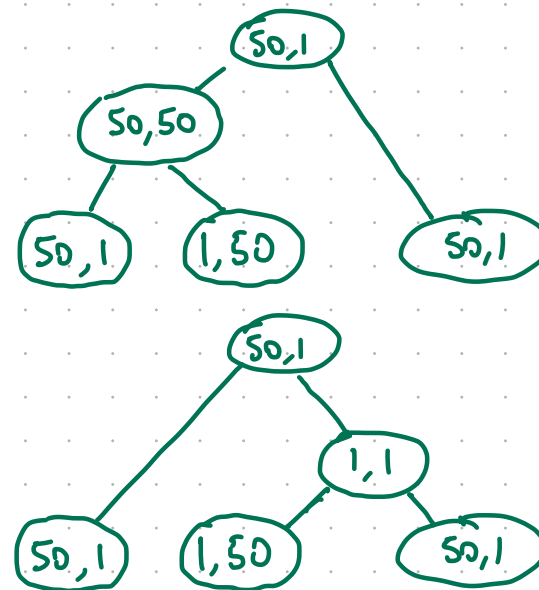
• option 1: $(A \cdot B) \cdot C = D \cdot C$

$50 \cdot 1 \cdot 50 + 50 \cdot 50 \cdot 1 = 5000$

• option 2: $A \cdot (B \cdot C) = A \cdot E$

$1 \cdot 50 \cdot 1 + 50 \cdot 1 \cdot 1 = 100$

100x better!



The association order matters!

In general, the goal is to find the **optimal association order** for

$$A_1, A_2, \dots, A_n$$

$$m_0 \text{--} b\text{y--} m_1, \quad m_1 \text{--} b\text{y--} m_2, \quad \dots, \quad m_{n-1} \text{--} b\text{y--} m_n$$

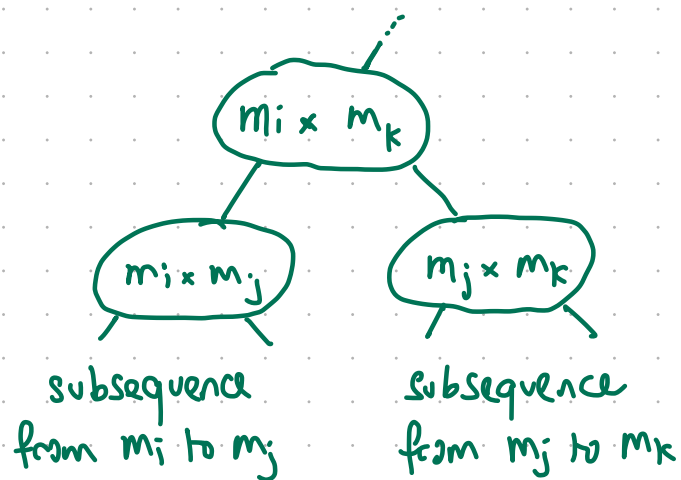
in the example the input is
 $(m_0, m_1, m_2, m_3) = (50, 1, 50, 1)$

input: list of positive integers m_0, m_1, \dots, m_n representing matrix dimensions

(the matrices themselves don't matter)

output: parenthetization of the list (a tree on it)

We solve the problem via **dynamic programming**:



- subproblems:

$C(i, k) :=$ optimal cost for subsequence m_i, \dots, m_k
 [multiplying $A_{i+1} \dots A_k$]

- boundaries:

$$C(0, 1) = C(1, 2) = \dots = C(n-1, n) = 0$$

- recurrence:

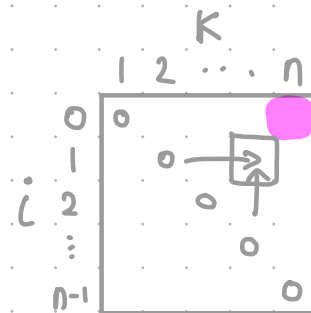
$A_{i+1} \dots A_j$ $A_j \dots A_k$ combine

$$C(i, k) := \min_{i < j < k} \{ C(i, j) + C(j, k) + m_i m_j m_k \}$$

The cost so far is:

$$m_i m_j m_k + \text{cost_left} + \text{cost_right}$$

- cost: $O(n^2)$ subproblems, computing each is $O(n)$
 $\Rightarrow O(n^3)$ cost to find $C(0, n)$.



All-Pairs Shortest Paths

Given graph $G=(V,E)$ and lengths $l:E \rightarrow \mathbb{Z}$, find $\{ \text{dist}(u,v) \}_{u,v \in V}$.
distance from u to v

Idea: Run Bellman-Ford for every possible source: $|V| \cdot O(|V| \cdot |E|) = O(|V|^3 \cdot |E|)$.

Better: $O(|V|^3)$ via dynamic programming ($|E| \geq |V|-1$ if G is connected)

Floyd-Warshall algorithm

- The subproblems are:

$d(u,v,i)$ = shortest path from u to v using intermediate nodes in $\{1,2,\dots,i\}$

- Boundaries are:
$$d(u,v,0) = \begin{cases} \text{if } (u,v) \in E: l(u,v) \\ \text{if } (u,v) \notin E: \infty \end{cases}$$

- The recurrence is:

$$d(u,v,i) = \min \{ d(u,v,i-1), d(u,i,i-1) + d(i,v,i-1) \}$$

- Efficiency:

$\left. \begin{array}{l} \cdot \text{init boundaries: } O(|V|^2) \\ \cdot \text{table has } |V|^3 \text{ entries and filling each is } O(1) \end{array} \right\} \text{total cost is } O(|V|^3)$

Traveling Salesperson Problem

cycle visiting each vertex once
↓

Given graph $G=(V,E)$ and lengths $\ell:E \rightarrow \mathbb{Z}$, output a shortest tour of G .

Straightforward approach: try all tours.

wlog start at vertex 1. There are $\leq (n-1)!$ tours. Evaluating a tour costs $O(n)$.

$$\Rightarrow O(n!) \approx O\left(\frac{n^n}{e^n}\right).$$

Better: $O(2^n \cdot n^2)$ via dynamic programming (cannot expect poly-time also because TSP is NP-complete)

• The subproblems are:

$$\text{answer is } \min_{j=2,\dots,n} \{ C(\{2,\dots,n\}, j) + \ell(j, 1) \}$$

$C(S, j)$ = shortest path from 1 to $j \in S$ visiting each vertex in $S \subseteq \{2, \dots, n\}$ once

• Boundaries: $C(\{j\}, j) = \ell(1, j)$

• Recurrence: $C(S, j) = \min_{i \in S \setminus \{j\}} \{ C(S - \{i\}, i) + \ell(i, j) \}.$

• Efficiency: compute in order of increasing $|S|$

table has $O(2^n \cdot n)$ entries and each takes $O(n)$ to compute $\Rightarrow O(2^n \cdot n^2)$.