

CS 7800: Advanced Algorithms

Class 6: Dynamic Programming III

- Knapsack
- Shortest Paths

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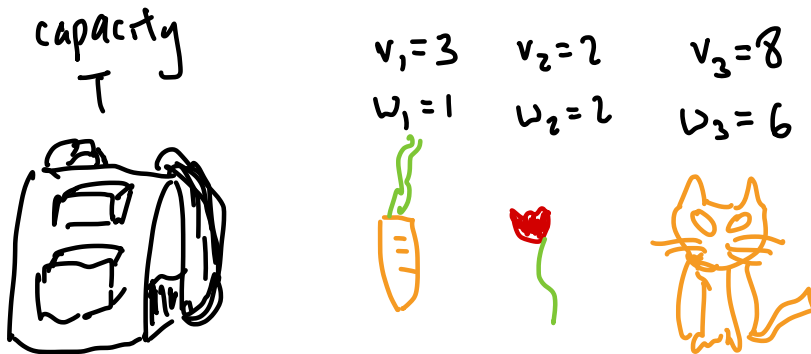
September 23, 2025

Housekeeping

- HW1 is returned
- HW2 due Friday (can use up to 2 late days)
- OH tomorrow → on Zoom 2-3pm
 ↪ will also add slots
- EXAM I TUESDAY OCT 7th
- HW3 out this Friday due next Friday 10/3

The Knapsack Problem

- **Input:** n items for your knapsack with value v_i and weight w_i and a capacity $T \in \mathbb{N}$ *a non-negative integer*
- **Output:** the most valuable subset of items that fits in the knapsack
 - **subset** $S \subseteq \{1, \dots, n\}$
 - **value** $V_S = \sum_{i \in S} v_i$ as large as possible
 - **weight** $W_S = \sum_{i \in S} w_i$ at most T

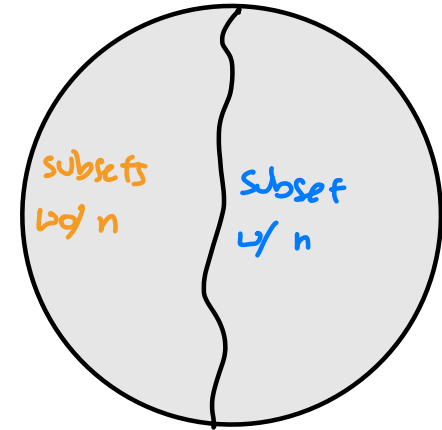


all subsets of $w \leq T$

Writing the Recurrence

$OPT(i) =$ value of the best subset
among items $\{1, \dots, i\}$ $i = 0, 1, \dots, n$

subproblems



$$OPT(n) = \max \left\{ \begin{array}{l} \text{value of best} \\ \text{solution w/o } n \end{array} , \begin{array}{l} \text{value of best} \\ \text{solution w/ } n \end{array} \right\}$$

$OPT(n-1)$

$v_n + OPT(n-1) ?$

capacity
 T



$v_1=3$
 $w_1=1$



$v_2=2$
 $w_2=2$



~~$v_3=8$
 $w_3=6$~~



capacity
 $T - w_n$



$v_1=3$
 $w_1=1$



$v_2=2$
 $w_2=2$



$v_3=8$
 $w_3=6$

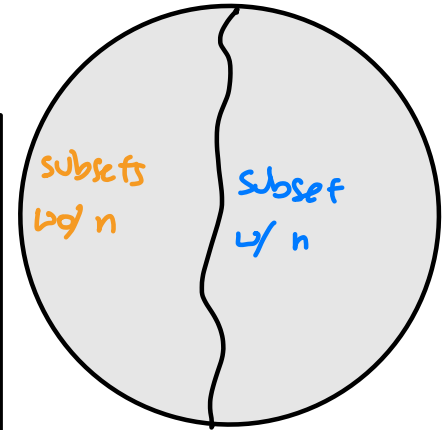
all subsets of wt ≤ T

Writing the Recurrence

$OPT(i, S) =$ maximum value of a subset of items $\{1, \dots, i\}$ with cap. $\leq S$ $i = 0, 1, \dots, n$
 $S = 0, 1, \dots, T$

subproblems

$\approx nT$ subproblems



$OPT(n, T) = \max \left\{ \begin{array}{l} \text{value of best} \\ \text{solution w/o } n \end{array} , \begin{array}{l} \text{value of best} \\ \text{solution w/ } n \end{array} \right\}$

$OPT(n, T) = \max \left\{ OPT(n-1, T), v_n + OPT(n-1, T - w_n) \right\}$
 $OPT(0, T) = 0 \quad OPT(i, 0) = 0$

capacity
T



$v_1 = 3$
 $w_1 = 1$



$v_2 = 2$
 $w_2 = 2$



~~$v_3 = 8$
 $w_3 = 6$~~



capacity
 $T - w_n$



$v_1 = 3$
 $w_1 = 1$



$v_2 = 2$
 $w_2 = 2$



$v_3 = 8$
 $w_3 = 6$

all subsets of $w \leq T$

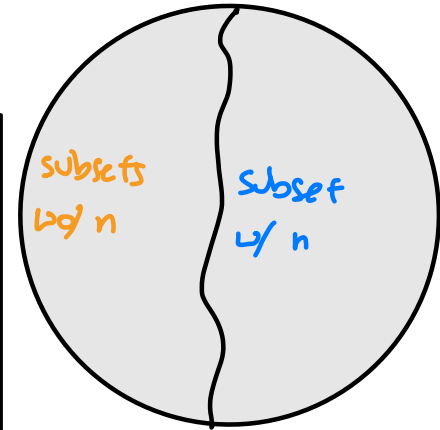
Writing the Recurrence

$OPT(i, S) =$ maximum value of a subset
of items $\{1, \dots, i\}$ with cap. $\leq S$

$i = 0, 1, \dots, n$
 $S = 0, 1, \dots, T$

subproblems

$\approx nT$ subproblems



$OPT(n, T) = \max \left\{ \begin{array}{l} \text{value of best} \\ \text{solution } w > n \end{array} , \begin{array}{l} \text{value of best} \\ \text{solution } w < n \end{array} \right\}$

$OPT(n, T) = \begin{cases} OPT(n-1, T) & \text{if } w_n > T \\ \max \{ OPT(n-1, T), v_n + OPT(n-1, T - w_n) \} & \text{if } w_n \leq T \end{cases}$

$OPT(0, T) = 0 \quad OPT(i, 0) = 0$

Solving the Recurrence

$OPT(i, S) =$ optimal value
items $\{1, \dots, i\}$
cap S

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

capacities

items

	0	1	2	...	T-1	T
n	0					v_n
n-1	0					v_{n-1}
...			
3	0					
2	0					
1	0					
0	0	0	0		0	0

Diagram illustrating the recurrence relation for the knapsack problem. The table shows the optimal value $OPT(i, S)$ for items i (rows) and capacity S (columns). The recurrence relation is $OPT(n, T) = \max \{ OPT(n-1, T), v_n + OPT(n-1, T-w_n) \}$. The diagram highlights the calculation of $OPT(n, T)$ by comparing the value of including item n (indicated by a red box around v_n) with the value of excluding item n (indicated by a blue box around $OPT(n-1, T-w_n)$). An arrow points from the blue box to the red box, indicating the comparison.

Knapsack Problem Recap

- Can solve the knapsack problem in time $O(nT)$
 - First example of **dynamic programming with multiple variables in the recurrence**
 - First example of a **pseudopolynomial-time algorithm**—compare to naïve $O(2^n)$ time algorithm
 - Later on we may see an **approximation algorithm** that solves knapsack in time $O(n)$ with small error



Input to a knapsack problem is $2n+1$ numbers

Single source

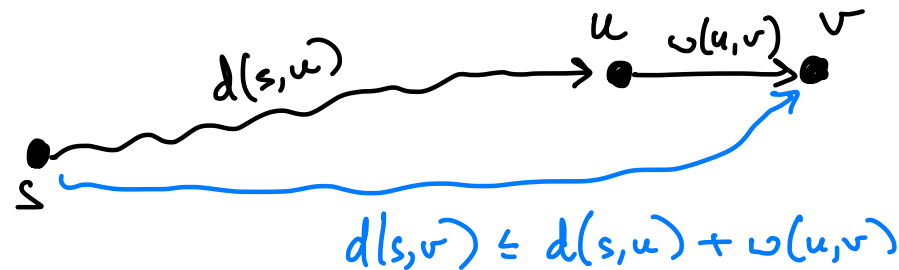
Shortest Paths

- **Input:** Directed, weighted graph $G = (V, E, \{w_e\})$, source node s
 - Possibly negative edge lengths $w_e \in \mathbb{R}$
 - No negative-length cycles!
- **Output:** Two arrays d, p
 - $d[u]$ is the length of the shortest $s \rightsquigarrow u$ path
 - $p[u]$ is the final hop on shortest $s \rightsquigarrow u$ path

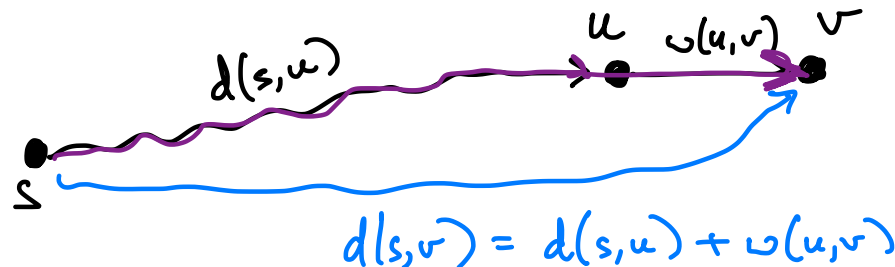
Structure of Shortest Paths

$d(u, v)$ = length of the shortest $u \rightsquigarrow v$ path

- If $(u, v) \in E$, then $d(s, v) \leq d(s, u) + w(u, v)$ for every node $s \in V$



- If $(u, v) \in E$, and $d(s, v) = d(s, u) + w(u, v)$ then there is a shortest $s \rightsquigarrow v$ -path ending with (u, v)



Writing the Recurrence

Subproblems

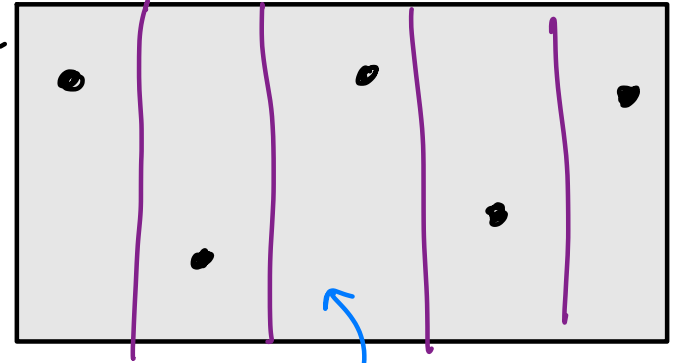
$\text{OPT}(v)$ = length of the shortest path from s to v

$v \in V$

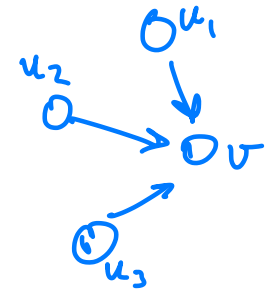
$$\text{OPT}(v) = \min_{u \text{ s.t. } (u,v) \in E} \{ \text{OPT}(u) + w(u,v) \}$$

$$\text{OPT}(s) = 0$$

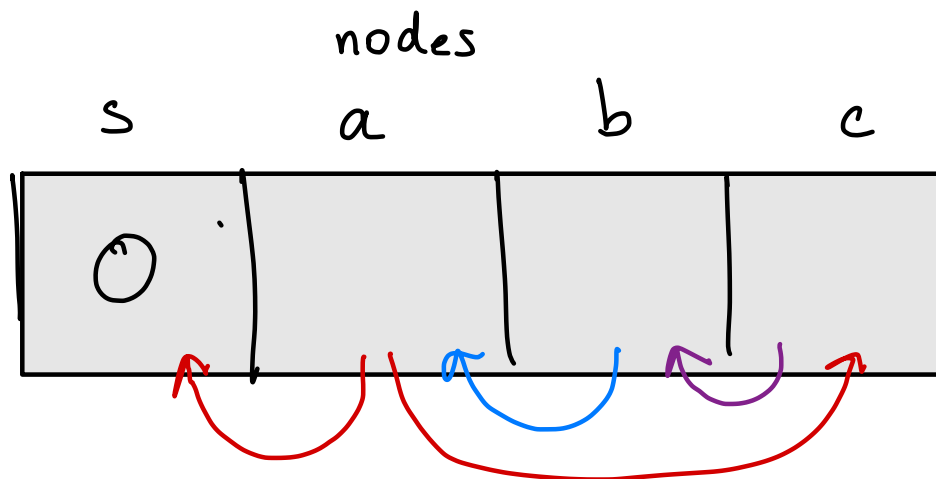
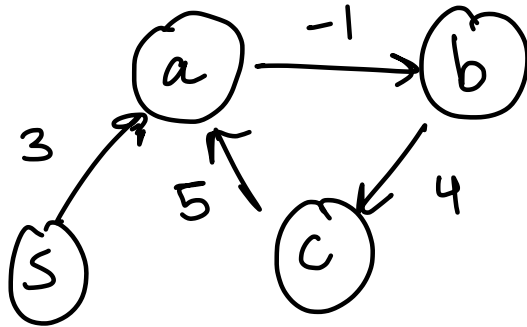
paths from s to v



paths from s to v
with (u,v) as final hop



Solving the Recurrence



Writing the Recurrence: Attempt 2

$\text{OPT}(v, i)$ = length of the shortest path from s to v
using at most i edges

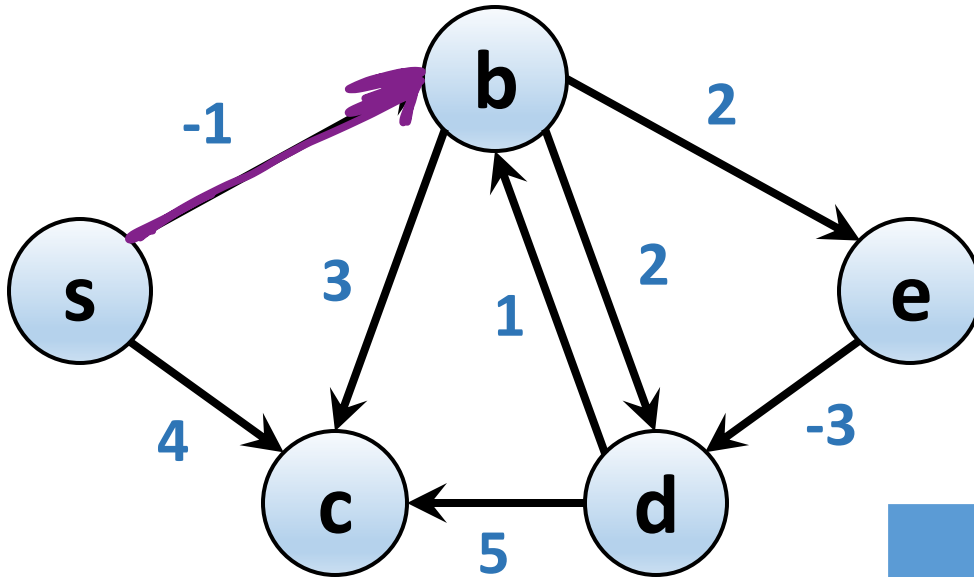
$v \in V$ $i = 0, 1, \dots, n-1$ n^2 subproblems

$$\text{OPT}(v, i) = \min_{u: (u, v) \in E} \left\{ w(u, v) + \text{OPT}(u, i-1) \right\}$$

$$\text{OPT}(s, i) = 0 \qquad \text{OPT}(v, 0) = \infty \quad \text{for } v \neq s$$

$\text{OPT}(v, n-1) = d(s, v)$ because no shortest path uses $> n-1$ hops

Solving the Recurrence: Attempt 2



$$\text{OPT}(b,1) = \min \{ \text{OPT}(s,0) - 1, \text{OPT}(d,0) + 1 \}$$

$$= \min \{ -1, \infty \} = -1$$

of hops

nodes

	0	1	2	3	4
s	0	0	0	0	0
b	∞	-1			
c	∞	4			
d	∞				
e	∞				

Shortest Paths Summary

- **Input:** Directed, weighted graph $G = (V, E, \{w_e\})$, and a source node s
- **Output:** Two arrays d, p
 - $d[u]$ is the length of the shortest $s \rightsquigarrow u$ path
 - $p[u]$ is the final hop on some shortest $s \rightsquigarrow u$ path
- **Negative lengths:** Bellman-Ford solves the single-source shortest paths problem in $O(nm)$ worst-case time, or finds a negative cycle
 - Often much faster in practice with suitable appropriate optimizations