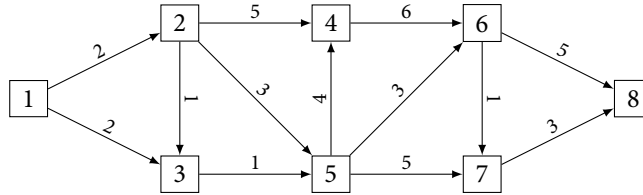


# The principle of optimality and formulating DP recursions

## 0 Warm up

**Example 1.** Consider the following directed graph. The labels on the edges are edge lengths.



In this order:

- a. Find the shortest path from node 1 to node 8.

- b. Find the shortest path from node 3 to node 8.

- c. Find the shortest path from node 5 to node 8.

## 1 The principle of optimality

- In Example 1, we found that the shortest path from node 1 to node 8 is  $1 \rightarrow 3 \rightarrow 5 \rightarrow 6 \rightarrow 7 \rightarrow 8$  with length 10
- Now, consider paths from node 3 to node 8
  - For example,  $3 \rightarrow 5 \rightarrow 6 \rightarrow 7 \rightarrow 8$  is such a path with length 8
- Could there be a shorter path from node 3 to node 8?
  - Suppose we had a path from 3 to 8 with length  $< 8$
  - Consider edge  $(1, 3)$  + this path

$\Rightarrow$

### The principle of optimality (for shortest path models)

In a directed graph with no negative cycles, optimal paths must have optimal subpaths.

- Consider a directed graph  $(N, E)$  with target node  $t \in N$  and edge lengths  $c_{ij}$  for  $(i, j) \in E$
- By the principle of optimality, the shortest path from node  $i$  to node  $t$  must be:

edge  $(i, j)$  + shortest path from  $j$  to  $t$      for some  $j \in N$  such that  $(i, j) \in E$

- How can we exploit this?

- Let  $f(i) =$

- Then we can write the following **boundary conditions** and **recursion**

- For example, in Example 1,  $f(5)$  is

## 2 Formulating DP recursions

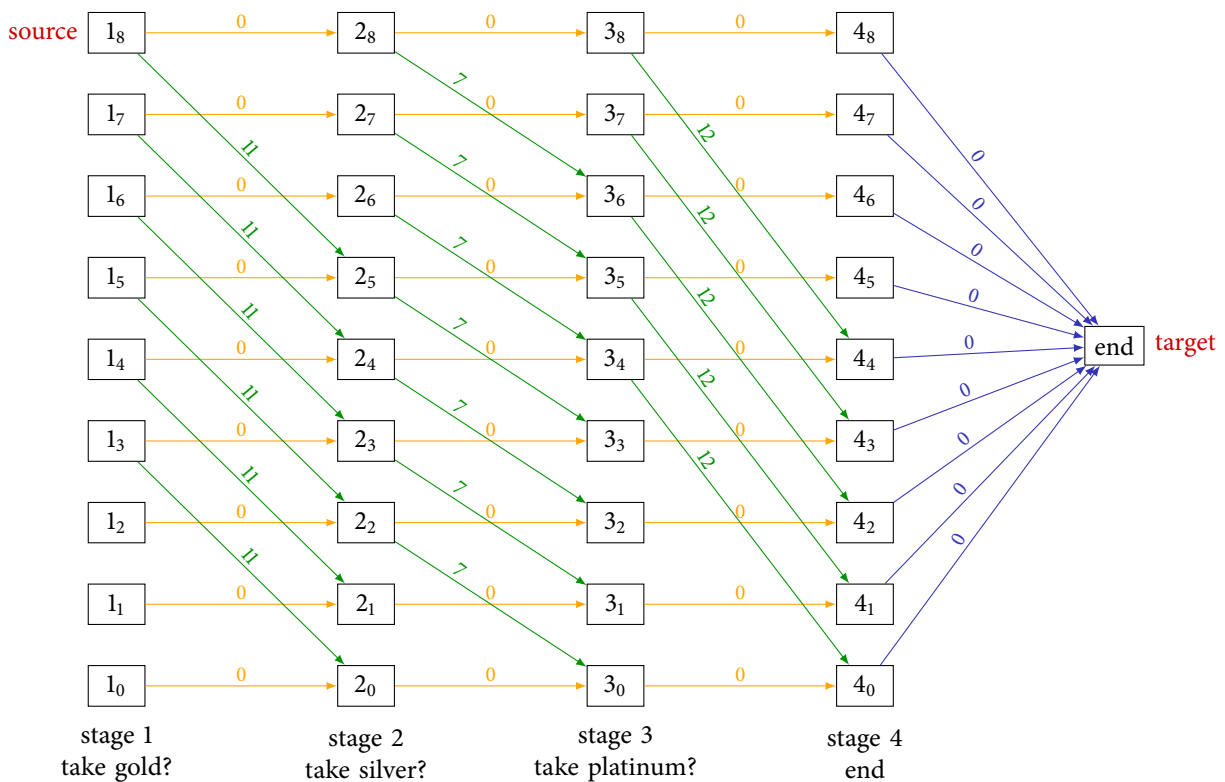
- Dynamic programs are not usually given as shortest/longest path problems as we have done over the past few lessons
- Instead, DPs are usually given as **recursions**
- Let's revisit the following knapsack problem that we studied back in Lesson 7

**Example 2.** You are a thief deciding which precious metals to steal from a vault:

	Metal	Weight (kg)	Value
1	Gold	3	11
2	Silver	2	7
3	Platinum	4	12

You have a knapsack that can hold at most 8kg. If you decide to take a particular metal, you must take all of it. Which items should you take to maximize the value of your theft?

- We formulated the following dynamic program for this problem by giving the following longest path representation:



- Let's formulate this as a dynamic program, but now by giving its recursion representation

- Let

$w_t =$  weight of metal  $t$

$v_t =$  value of metal  $t$

for  $t = 1, 2, 3$

- Stages:

- States:

- Allowable decisions  $x_t$  at stage  $t$  and state  $n$ :

- Reward of decision  $x_t$  at stage  $t$  and state  $n$ :

- Reward-to go function  $f_t(n)$  at stage  $t$  and state  $n$ :

- Boundary conditions:

- Recursion:

- Desired reward-to-go function value:

- In general, to formulate a DP by giving its recursive representation:

#### Dynamic program – recursive representation

- **Stages**  $t = 1, 2, \dots, T$  and **states**  $n = 0, 1, 2, \dots, N$
- Allowable **decisions**  $x_t$  at stage  $t$  and state  $n$   $(t = 1, \dots, T - 1; n = 0, 1, \dots, N)$
- **Cost** of decision  $x_t$  at stage  $t$  and state  $n$   $(t = 1, \dots, T; n = 0, 1, \dots, N)$
- **Cost-to-go** function  $f_t(n)$  at stage  $t$  and state  $n$   $(t = 1, \dots, T; n = 0, 1, \dots, N)$
- **Boundary conditions** on  $f_T(n)$  at state  $n$   $(n = 0, 1, \dots, N)$
- **Recursion** on  $f_t(n)$  at stage  $t$  and state  $n$   $(t = 1, \dots, T - 1; n = 0, 1, \dots, N)$

$$f_t(n) = \min \{ (\text{cost of decision at stage } t) + f_{t+1}(\text{new state at stage } t + 1) \}$$

- **Desired cost-to-go function value**

### 3 Solving DP recursions

- To improve our understanding of how this recursive representation works, let's solve the DP we just wrote for the knapsack problem
- We solve the DP backwards:
  - start with the boundary conditions in stage  $T$
  - compute values of the cost-to-go function  $f_t(n)$  in stages  $T - 1, T - 2, \dots, 3, 2$
  - ... until we reach the desired cost-to-go function value

- Stage 4 computations – boundary conditions:

- Stage 3 computations:

$f_3(8) =$

$f_3(7) =$

$f_3(6) =$

$f_3(5) =$

$f_3(4) =$

$f_3(3) =$

$f_3(2) =$

$f_3(1) =$

$f_3(0) =$

- Stage 2 computations:

$f_2(8) =$

$f_2(7) =$

$f_2(6) =$

$f_2(5) =$

$f_2(4) =$

$f_2(3) =$

$f_2(2) =$

$f_2(1) =$

$f_2(0) =$

- Stage 1 computations – desired cost-to-go function: