

CS7800: Advanced Algorithms

Dynamic Programming I:

- Weighted Interval Scheduling
- Log cutting

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What is Dynamic Programming?

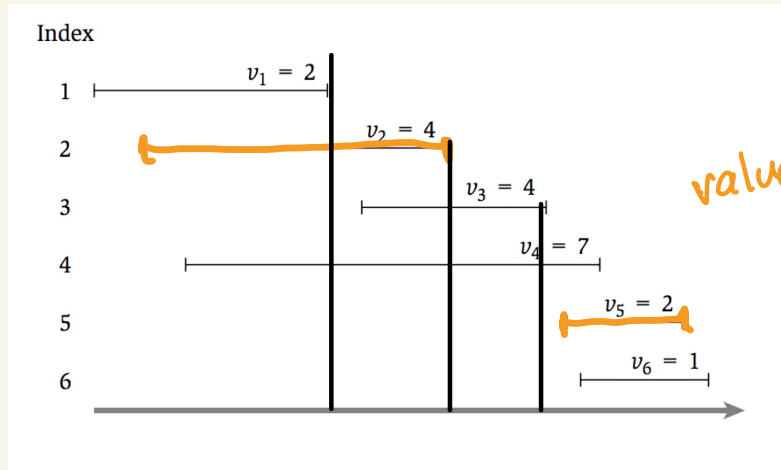
- Don't think too hard about the name
 - *I thought dynamic programming was a good name. It was something not even a congressman could object to. So I used it as an umbrella for my activities. -Bellman*
- Dynamic programming is careful recursion
 - Break the problem up into small pieces
 - Recursively solve the smaller pieces
 - **Key Challenge:** identifying the pieces

Weighted Interval Scheduling

Input: n intervals $[s_i, f_i]$ with values v_i

Output: a set of nonoverlapping intervals

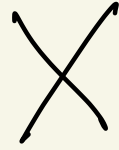
$S \subseteq \{1, 2, \dots, n\}$ maximizing $\sum_{i \in S} v_i$



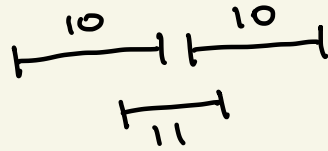
value = 4 + 2 = 6

A Greedy Algorithm?

- Sort by finish time?



- Sort by value?



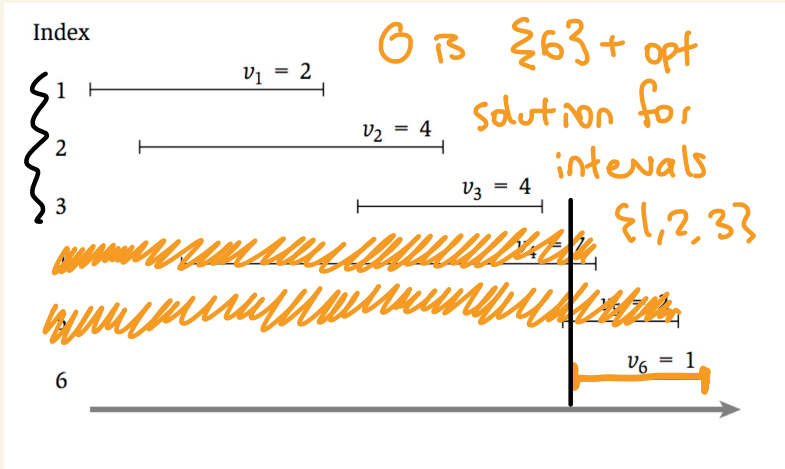
Dynamic Programming — Thinking Recursively

- Let Θ be an optimal solution

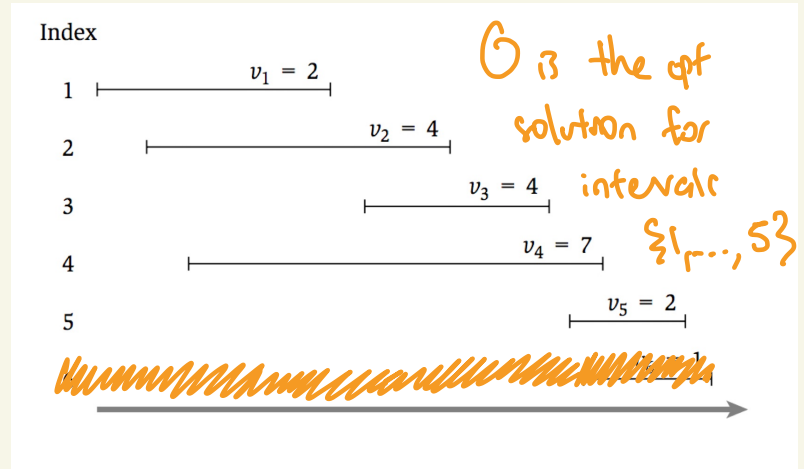
1 + value of opt for $\{1, 2, 3\} = 7$

0 + value of opt for $\{1, 3, 5\} = 8$

Case 1: $6 \in \Theta$



Case 2: $6 \notin \Theta$



Dynamic Programming - Thinking Recursively

Let $OPT(i)$ be the optimal value of a schedule using only intervals $\{1, 2, \dots, i\}$ (for $i = 0, 1, 2, \dots, n$)

$OPT(n)$ is the thing I want

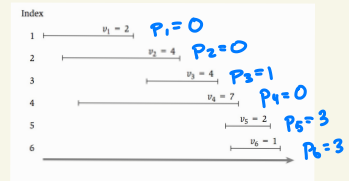
Case 1: i is in the optimal solution for $\{1, 2, \dots, i\}$

$$OPT(i) = v_i + OPT(p_i) \quad (f_1 \leq f_2 \leq \dots \leq f_n)$$

where $p_i =$ largest j such that $f_j < s_i$

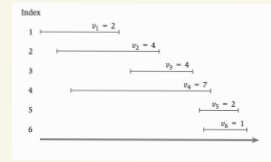
Case 2: i is not in the optimal solution

$$OPT(i) = OPT(i-1)$$



$$OPT(i) = \max \{ OPT(i-1), v_i + OPT(p_i) \} \quad OPT(0) = 0 \quad OPT(1) = v_1$$

Evaluating the Recurrence : Take I



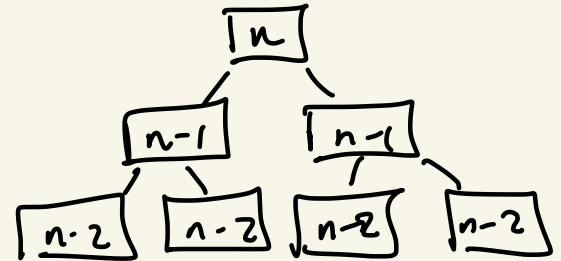
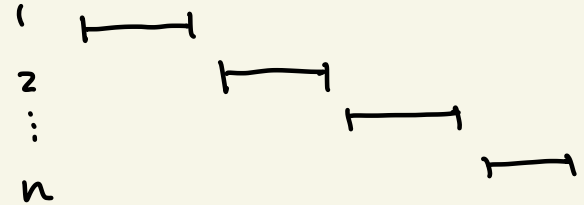
$$\text{OPT}(i) = \max \{ \text{OPT}(i-1), v_i + \text{OPT}(p_i) \} \quad \text{OPT}(0) = 0 \quad \text{OPT}(i) = v_i$$

main :

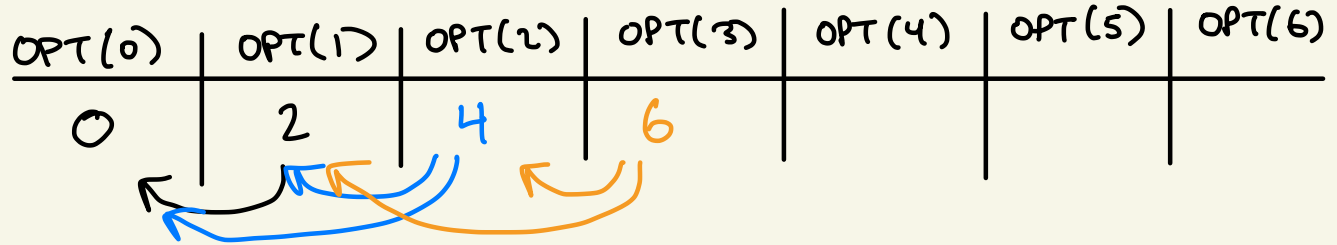
return $f(n)$

$f(i)$:
if $(i=0)$ return 0
return $\max \{ f(i-1), v_i + f(p_i) \}$

Key Observation: Don't recompute the same subproblem.

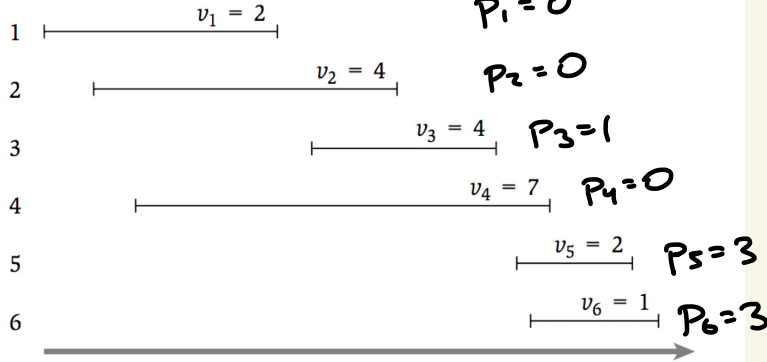


Evaluating the Recurrence: Take II



$$OPT(i) = \max \{ OPT(i-1), v_i + OPT(p_i) \} \quad OPT(0) = 0 \quad OPT(1) = v_1$$

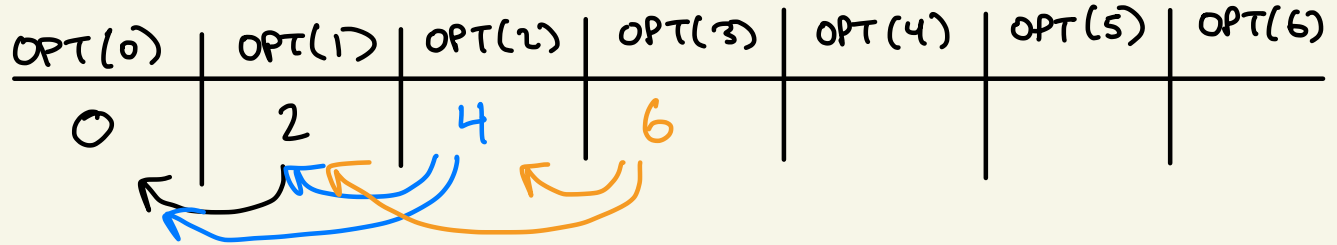
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"Bottom up":

fill in the table of subproblems one by one

Evaluating the Recurrence: Take II



main:

$opt(0) = 0$

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

$opt(i) = \max \{ opt(i-1), v_i + opt(p_i) \}$

Return $opt(n)$

running time
is just $O(n)$

(once you sort
the input and
compute P_i .)

Finding the Optimal Schedule

$$\text{OPT}(i) = \max \{ \text{OPT}(i-1), v_i + \text{OPT}(p_i) \}$$

\mathcal{O}_i = the better solution of $\{i\} + \mathcal{O}_{p_i}$
and \mathcal{O}_{i-1}

\mathcal{O}_0	\mathcal{O}_1	\mathcal{O}_2	\mathcal{O}_3	\mathcal{O}_4	\mathcal{O}_5	\mathcal{O}_6
\emptyset	$\{1\}$	$\{2\}$	$\{1,3\}$	$\{4\}$	$\{1,2,5\}$	$\{1,3,5\}$
	yes	yes	yes	yes	yes	no

$\mathcal{O}_1 = \{1\} + \emptyset$ $\mathcal{O}_3 = \{3\} + \mathcal{O}_1$ $\mathcal{O}_5 = \{5\} + \mathcal{O}_3$ $\mathcal{O}_6 = \mathcal{O}_5$

Which branch in
the recurrence gives this bit.

Finding the Optimal Schedule

$$\text{OPT}(i) = \max \{ \text{OPT}(i-1), v_i + \text{OPT}(p_i) \}$$

↑
if this is the max
then $O_i = O_{i-1}$

↖
if this is the
max then $O_i = O_{p_i} + \xi_i$

Calc OPT:

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

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

$\left\{ \begin{array}{l} \text{OPT}(i) = [\text{recurrence}] \\ \text{Branch}(i) = \begin{cases} \text{yes if } v_i + \text{OPT}(p_i) \geq \text{OPT}(i-1) \\ \text{no o.w.} \end{cases} \end{array} \right.$

Find Schedule(i)

$\left\{ \begin{array}{l} \text{if } i=0 \text{ return } \emptyset \\ \text{else if } \text{Branch}(i) = \text{yes} \\ \quad \text{[return } \xi_i + \text{FS}(p_i) \\ \text{else} \\ \quad \text{[return } \text{FS}(i-1) \end{array} \right.$

What were the steps?

requires
creativity

①

We asked a good question about the solution

②

We found a recurrence relation for the value of opt

③

We implemented that recurrence efficiently

- there is a small set of subproblems

- avoided recomputing the same subproblem

④

Add a step to find the optimal solution

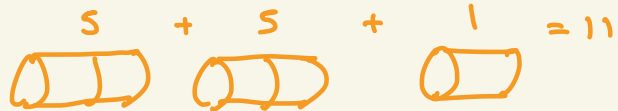
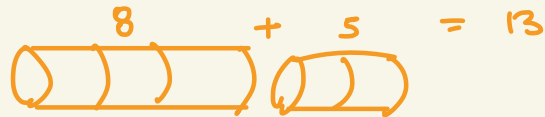
- remember which branch you took and interpret its meaning

The Log Cutting Problem

Input: The length n of a log,
prices $\{p_l\}_{l=1 \dots n}$ for selling an l -foot log

Output: Integer lengths l_1, \dots, l_k such that
 $\sum_{i=1}^k l_i = n$ and $\sum_{i=1}^k p_{l_i}$ is maximized

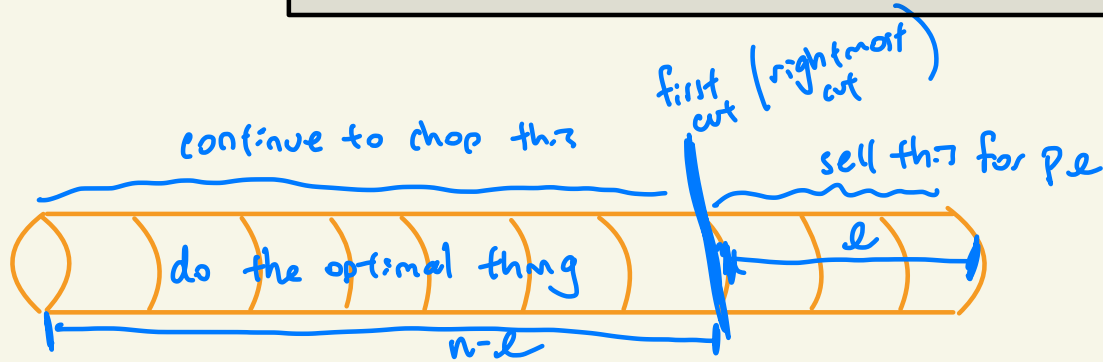
length l	1	2	3	4	5
price p_l	1	5	8	9	10



Dynamic Programming

Let $OPT(i)$ be the most money I can get with a log of length i (for $i = 0, 1, \dots, n$)

$$OPT(i) = \max_{1 \leq l \leq i} \{ p_l + OPT(i-l) \} \quad OPT(0) = 0$$



length l	1	2	3	4	5	6	7	8	9	10
price p_l	1	5	8	9	10	17	17	20	24	30

Calc OPT :

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

for $i = 1, 2, \dots, n$:

[let $\text{OPT}(i) = \max_{1 \leq e \leq i} \{ p_i + \text{OPT}(i-e) \}$

return $\text{OPT}(n)$

Running time is $O(n^2)$