

EECS 336: Lecture 6: Introduction to Algorithms Example: Interval Pricing

Dynamic Programming (cont) interval pricing

Reading: 6.5

Announcements:

- peer grading

Last Time:

- Approach: isolating previous decisions
- Shortest-paths (Bellman-Ford Alg)

Today:

- interval pricing
- summary of dynamic programming
- comparison to divide and conquer

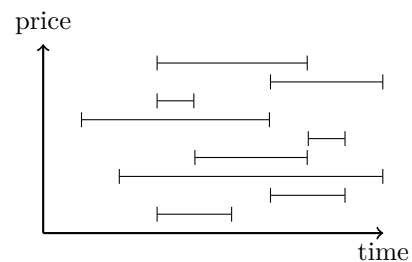
input:

- n customers $S = \{1, \dots, n\}$
- T days.
- i 's ok days: $I_i = \{s_i, \dots, f_i\}$
- i 's value: $v_i \in \{1, \dots, V\}$

output:

- prices $p[t]$ for day t .
- consumer i buys on day $t_i = \operatorname{argmin}_{t \in I_i} p[t]$ if $p[t] \leq v_i$.
- revenue = $\sum_{i \text{ that buys } p[t_i]}$.
- goal: maximize revenue.

Example:

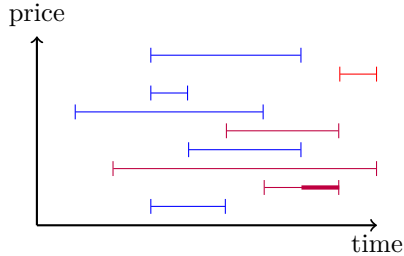


let's use dynamic programming. subproblem?

Question: What is "first decision we can make" to separate into subproblems?

Answer: day and price of smallest price.

Example:



Step I: identify subproblem in English

$OPT(s, f, p)$ = "optimal revenue from customers i with intervals $\{s_i, \dots, f_i\}$ contained within interval $\{s + 1, \dots, f - 1\}$ with minimum price at least p ."

Step II: write recurrence

$$OPT(s, f, p) = \max_{t \in \{s+1, \dots, f-1\}; q \in \{p, \dots, V\}} Rev(s, t, f, p) + OPT(s, t, q) + OPT(t, f, q).$$

$Rev(s, t, f, p)$ = "the revenue from customers i with intervals $\{s_i, \dots, f_i\}$ contained within interval $\{s + 1, \dots, f - 1\}$ with minimum price at least p ."

with

Step III: value of optimal solution

- optimal interval pricing = $OPT(1, T, 0)$

Step IV: base case

- $OPT(s, s + 1, p) = 0$.
- $OPT(s, t, V + 1) = 0$.

Step V: iterative DP

(exercise)

Correctness

induction

Step VI: Runtime

- precompute $Rev(s, t, f, p)$ in $O(T^3 V n)$ time.
- size of table: $O(T^2 V)$
- cost of combine: $O(TV)$
- total: $O(T^3 V(V + n))$

Note: without loss of generality T, V are $O(n)$ so runtime is $O(n^5)$.

Note: can be improved to $O(n^4)$ with slightly better program.

Step VII: implementation

(exercise)

Summary of Dynamic Programming

“divide the problem into small number of subproblems and memoize solution to avoid redundant computation.”

Finding Subproblems

- identify a first decision, subproblems for each outcome of decision.
- partition problem, summarize information from one part needed to solve other part.

Subproblem Properties

1. succinct (only a polynomial number of them)
2. efficiently combinable.
3. depend on “smaller” subproblems (avoid infinite loops), e.g.,
 - process elements “once and for all” [today]
 - “measure of progress/size” [coming soon]

Runtime Analysis

runtime = initialization + size of table \times cost to combine

Finding Solution

- write DP to identify value of optimal solution.
- traverse memoization table to determine actual solution.