Announcements:

- Midterm: Tuesday, Oct 27, in class.
 - one handwritten cheat sheet.
 - through dynamic programming.
 - midterm reviews:
 - Taggart, Sunday,
 - Hartline, Monday, 6-8pm,

Reading: 6.4, 6.8

Last time:

- Dynamic Programming
- \bullet Weighted Interval scheduling

Today:

- D.P. (cont.)
- Integer Knapsack
- Interval Pricing.

Suggested Approach

I. identify subproblem in english

OPT(i) = "optimal schedule of $\{i, ..., n\}$ (sorted by start time)"

II. specify subproblem recurrence

 $\begin{array}{lll}
\text{OPT}(i) &= & \max(\text{OPT}(i + 1), v_i + \\
\text{OPT}(\text{next}(i)))
\end{array}$

III. identify base case

OPT(n+1) = 0

IV. write iterative DP.

(see last thurs)

Interval Pricing

input: • n customers $S = \{1, \ldots, n\}$

- \bullet 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 $p[t_i] \leq v_i$.
- revenue = $\sum_{i \text{ that buys}} p[t_i]$.
- goal: maximize revenue.

Dynamic Programming: Finding Subproblems

"find a first decision you can make which breaks problem into pieces that

- (a) do not interact (across subproblems)
- (b) can be describe succinctly."

Example: Integer Knapsack

input:

- $n \text{ objects } S = \{1, ..., n\}$
- $s_i = \text{size of object } i \text{ (integer)}.$
- v_i = value of object i.
- \bullet capacity C of knapsack (integer)

output:

- subset $K \subseteq S$ of objects that
 - (a) fit in knapsack together (i.e., $\sum_{i \in K} s_i \leq C$)
 - (b) maximize total value (i.e., $\sum_{i \in K} v_i$)

Greedy fails, e.g.,

• largest value/size:

$$\mathbf{v} = (C/2 + 2, C/2, C/2).$$

$$\mathbf{s} = (C/2 + 1, C/2, C/2).$$

• smallest value/size:

$$\mathbf{v} = (1, C/2, C/2).$$

$$\mathbf{s} = (2, C/2, C/2).$$

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

Answer: Is item 1 in the knapsack or not?

• if 1 in knapsack:

value of knapsack is v_i + optimal knapsack value on $S \setminus \{1\}$ with capacity $C - s_1$.

• if 1 not in knapsack:

value of knapsack is optimal knapsack on $S \setminus \{1\}$ with capacity C.

Succinct description:

- remaining objects $\{j, \ldots, n\}$ represented by "j"
- remaining capacity represented by $D \in \{0, \ldots, C\}$.

Step I: identify subproblem in Correctness English

induction

= "value of optimal size
$$D$$
 knapsack on **Runtime** $\{j,\ldots,n\}$ "

$$T(n, C) = O(\# \text{ of subprobs} \times \text{cost per subprob})$$

= $O(nC)$.

Note: not polynomial time.

Step II: write recurrence

$$OPT(j, D) = \max(\underbrace{v_j + OPT(j+1, D - s_j)}_{\text{if } s_j \le D}, OPT(j+1, D))$$

Step III: base case

$$OPT(n+1, D) = 0$$
 (for all D)

Step IV: iterative DP

Algorithm: knapsack

- 1. $\forall D, \text{ memo}[n+1, D] = 0.$
- 2. for i = n down to 1,

for D = C down to 0,

(a) if i fits (i.e., $s_i \leq D$)

memo[j, D] = max[memo[j + 1, D],

$$v_j + \text{OPT}(j+1, D-s_j)$$

(b) else,

$$memo[j, D] = memo[j + 1, D]$$

3. return memo[1, C]

Example: Interval Pricing

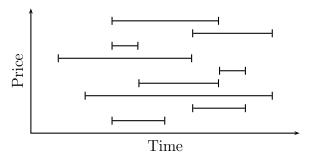
input: • n customers $S = \{1, \dots, n\}$

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

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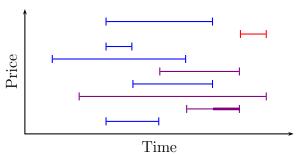
Example:



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 intervals strictly between s and f with minimum price at least p"

Step II: write recurrence

$$= \max_{s < t < f, q \ge p} \operatorname{Rev}(t, p)$$

$$+ \operatorname{OPT}(s, t, q)$$

$$+ \operatorname{OPT}(t, f, q).$$

Step III: base case

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

Step IV: iterative DP

(exercise)

Correctness

induction

Runtime

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

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