



Algorithms: Design
and Analysis, Part II

Approximation Algorithms for NP-Complete Problems

Dynamic Programming for Knapsack, Revisited

Two Dynamic Programming Algorithms

Dynamic programming algorithm #1 (see earlier video)

- ① Assume sizes w_i and capacity W are integers
- ② Running time = $O(nW)$

Dynamic programming algorithm #2 (this video)

- ① Assume values v_i are integers
- ② Running time = $O(n^2 V_{\max})$, where $V_{\max} = \max_i v_i$

The Subproblems and Recurrence

Subproblems: For $i = 0, 1, 2, \dots, n$ and
 $x = 0, 1, 2, 3, \dots, n \cdot v_{\max}$

define $S_{i,x}$ = minimum total size needed to achieve
value $\geq x$ while using only the first i items. (or ∞ if impossible)

Recurrence: ($i \geq 1$)

$$S_{i,x} = \min \left\{ \begin{array}{l} S_{i-1,x} \\ w_i + S_{i-1,x-v_i} \end{array} \right.$$

Case 1, item i not used in optimal solution

Case 2, item i used in optimal solution

interpret as 0 if $v_i \geq x$

The Algorithm

Let $A = 2\text{-D array}$ [indexed by $i = 0, 1, 2, \dots, n$
and $x = 0, 1, 2, \dots, n \cdot v_{\max}$]

Base case: $A[0, x] = \begin{cases} 0 & \text{if } x = 0 \\ \infty & \text{otherwise} \end{cases}$

For $i = 1, 2, 3, \dots, n$:

For $x = 0, 1, 2, \dots, n \cdot v_{\max}$:

$$A[i, x] = \min \{ A[i-1, x], w_i + A[i-1, x - v_i] \}$$

} $n \cdot v_{\max}$ iterations
interpret as 0 if $v_i > x$
← $O(1)$ work per iteration

Return the largest x such that $A[n, x] \leq W$. ← $O(n \cdot v_{\max})$

Running time: $O(n^2 v_{\max})$.