



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 videos)

(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 $i = 0, 1, \dots, n$ and $x = 0, 1, \dots, nv_{\max}$ define $S_{i,x}$ = minimum total size needed to achieve value $\geq x$ while using only the first i items. (Or $+\infty$ if impossible)

Recurrence: ($i \geq 1$)

$$S_{i,x} = \min \begin{cases} S_{(i-1),x} & \text{Case 1, item } i \text{ not used in optimal solution} \\ w_i + S_{(i-1), (x-v_i)} & \text{Case 2, item } i \text{ used in optimal solution} \end{cases}$$

Interpret as 0 if $v_i > x$

The Algorithm

Let $A = 2$ -D array

[indexed by $i = 0, 1, \dots, n$ and $x = 0, 1, \dots, nv_{\max}$]

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

For $i = 1, 2, \dots, n \longrightarrow n^2 v_{\max}$ iterations

For $x = 0, 1, \dots, nv_{\max}$ Interpret as 0 if $v_i > x$

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

$O(1)$ work per iteration

Return the largest x such that $A[n, x] \leq W \leftarrow O(nv_{\max})$

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