



Algorithms: Design  
and Analysis, Part II

## Approximation Algorithms for NP-Complete Problems

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