

Dynamic Programming

- Consider the instance: $i, W', (w_1, w_2, \dots, w_i)$;
- $opt[i, W']$: the optimum value of the instance

$$opt[i, W'] = \begin{cases} 0 & i = 0 \\ opt[i - 1, W'] & i > 0, w_i > W' \\ \max \left\{ \begin{array}{l} opt[i - 1, W'] \\ opt[i - 1, W' - w_i] + w_i \end{array} \right\} & i > 0, w_i \leq W' \end{cases}$$

Dynamic Programming

```
1: for  $W' \leftarrow 0$  to  $W$  do  
2:    $opt[0, W'] \leftarrow 0$   
3: for  $i \leftarrow 1$  to  $n$  do  
4:   for  $W' \leftarrow 0$  to  $W$  do  
5:      $opt[i, W'] \leftarrow opt[i - 1, W']$   
6:     if  $w_i \leq W'$  and  $opt[i - 1, W' - w_i] + w_i \geq opt[i, W']$  then  
7:        $opt[i, W'] \leftarrow opt[i - 1, W' - w_i] + w_i$   
8: return  $opt[n, W]$ 
```

Recover the Optimum Set

```
1: for  $W' \leftarrow 0$  to  $W$  do
2:    $opt[0, W'] \leftarrow 0$ 
3: for  $i \leftarrow 1$  to  $n$  do
4:   for  $W' \leftarrow 0$  to  $W$  do
5:      $opt[i, W'] \leftarrow opt[i - 1, W']$ 
6:      $b[i, W'] \leftarrow N$ 
7:     if  $w_i \leq W'$  and  $opt[i - 1, W' - w_i] + w_i \geq opt[i, W']$ 
   then
8:        $opt[i, W'] \leftarrow opt[i - 1, W' - w_i] + w_i$ 
9:        $b[i, W'] \leftarrow Y$ 
10: return  $opt[n, W]$ 
```

Recover the Optimum Set

```
1:  $i \leftarrow n, W' \leftarrow W, S \leftarrow \emptyset$   
2: while  $i > 0$  do  
3:   if  $b[i, W'] = Y$  then  
4:      $W' \leftarrow W' - w_i$   
5:      $S \leftarrow S \cup \{i\}$   
6:    $i \leftarrow i - 1$   
7: return  $S$ 
```

Running Time of Algorithm

```
1: for  $W' \leftarrow 0$  to  $W$  do  
2:    $opt[0, W'] \leftarrow 0$   
3: for  $i \leftarrow 1$  to  $n$  do  
4:   for  $W' \leftarrow 0$  to  $W$  do  
5:      $opt[i, W'] \leftarrow opt[i - 1, W']$   
6:     if  $w_i \leq W'$  and  $opt[i - 1, W' - w_i] + w_i \geq opt[i, W']$  then  
7:        $opt[i, W'] \leftarrow opt[i - 1, W' - w_i] + w_i$   
8: return  $opt[n, W]$ 
```

Running Time of Algorithm

```
1: for  $W' \leftarrow 0$  to  $W$  do  
2:    $opt[0, W'] \leftarrow 0$   
3: for  $i \leftarrow 1$  to  $n$  do  
4:   for  $W' \leftarrow 0$  to  $W$  do  
5:      $opt[i, W'] \leftarrow opt[i - 1, W']$   
6:     if  $w_i \leq W'$  and  $opt[i - 1, W' - w_i] + w_i \geq opt[i, W']$  then  
7:        $opt[i, W'] \leftarrow opt[i - 1, W' - w_i] + w_i$   
8: return  $opt[n, W]$ 
```

- Running time is $O(nW)$

Running Time of Algorithm

```
1: for  $W' \leftarrow 0$  to  $W$  do
2:    $opt[0, W'] \leftarrow 0$ 
3: for  $i \leftarrow 1$  to  $n$  do
4:   for  $W' \leftarrow 0$  to  $W$  do
5:      $opt[i, W'] \leftarrow opt[i - 1, W']$ 
6:     if  $w_i \leq W'$  and  $opt[i - 1, W' - w_i] + w_i \geq opt[i, W']$  then
7:        $opt[i, W'] \leftarrow opt[i - 1, W' - w_i] + w_i$ 
8: return  $opt[n, W]$ 
```

- Running time is $O(nW)$
- Running time is **pseudo-polynomial** because it depends on value of the input integers.

Running Time of Algorithm

```
1: for  $W' \leftarrow 0$  to  $W$  do
2:    $opt[0, W'] \leftarrow 0$ 
3: for  $i \leftarrow 1$  to  $n$  do
4:   for  $W' \leftarrow 0$  to  $W$  do
5:      $opt[i, W'] \leftarrow opt[i - 1, W']$ 
6:     if  $w_i \leq W'$  and  $opt[i - 1, W' - w_i] + w_i \geq opt[i, W']$  then
7:        $opt[i, W'] \leftarrow opt[i - 1, W' - w_i] + w_i$ 
8: return  $opt[n, W]$ 
```

- Running time is $O(nW)$
- Running time is **pseudo-polynomial** because it depends on value of the input integers.
- Game's running time:
<https://courses.csail.mit.edu/6.5440/fall23/>

Avoiding Unnecessary Computation and Memory Using Memoized Algorithm and Hash Map

compute-opt(i, W')

```
1: if  $opt[i, W'] \neq \perp$  then return  $opt[i, W']$ 
2: if  $i = 0$  then  $r \leftarrow 0$ 
3: else
4:    $r \leftarrow \text{compute-opt}(i - 1, W')$ 
5:   if  $w_i \leq W'$  then
6:      $r' \leftarrow \text{compute-opt}(i - 1, W' - w_i) + w_i$ 
7:     if  $r' > r$  then  $r \leftarrow r'$ 
8:  $opt[i, W'] \leftarrow r$ 
9: return  $r$ 
```

- Use hash map for opt

Outline

- 1 Weighted Interval Scheduling
- 2 Subset Sum Problem
- 3 Knapsack Problem**
- 4 Longest Common Subsequence
 - Longest Common Subsequence in Linear Space
- 5 Shortest Paths in Directed Acyclic Graphs
- 6 Matrix Chain Multiplication
- 7 Optimum Binary Search Tree
- 8 Summary

Knapsack Problem

Input: an integer bound $W > 0$

a set of n items, each with an integer weight $w_i > 0$

a value $v_i > 0$ for each item i

Output: a subset S of items that

$$\text{maximizes } \sum_{i \in S} v_i \quad \text{s.t. } \sum_{i \in S} w_i \leq W.$$

Knapsack Problem

Input: an integer bound $W > 0$

a set of n items, each with an integer weight $w_i > 0$

a value $v_i > 0$ for each item i

Output: a subset S of items that

$$\text{maximizes } \sum_{i \in S} v_i \quad \text{s.t. } \sum_{i \in S} w_i \leq W.$$

- Motivation: you have budget W , and want to buy a subset of items of maximum total value

DP for Knapsack Problem

- $opt[i, W']$: the optimum value when budget is W' and items are $\{1, 2, 3, \dots, i\}$.
- If $i = 0$, $opt[i, W'] = 0$ for every $W' = 0, 1, 2, \dots, W$.

$$opt[i, W'] = \begin{cases} & i = 0 \\ & i > 0, w_i > W' \\ & i > 0, w_i \leq W' \end{cases}$$

DP for Knapsack Problem

- $opt[i, W']$: the optimum value when budget is W' and items are $\{1, 2, 3, \dots, i\}$.
- If $i = 0$, $opt[i, W'] = 0$ for every $W' = 0, 1, 2, \dots, W$.

$$opt[i, W'] = \begin{cases} 0 & i = 0 \\ & i > 0, w_i > W' \\ & i > 0, w_i \leq W' \end{cases}$$

DP for Knapsack Problem

- $opt[i, W']$: the optimum value when budget is W' and items are $\{1, 2, 3, \dots, i\}$.
- If $i = 0$, $opt[i, W'] = 0$ for every $W' = 0, 1, 2, \dots, W$.

$$opt[i, W'] = \begin{cases} 0 & i = 0 \\ opt[i - 1, W'] & i > 0, w_i > W' \\ & i > 0, w_i \leq W' \end{cases}$$

DP for Knapsack Problem

- $opt[i, W']$: the optimum value when budget is W' and items are $\{1, 2, 3, \dots, i\}$.
- If $i = 0$, $opt[i, W'] = 0$ for every $W' = 0, 1, 2, \dots, W$.

$$opt[i, W'] = \begin{cases} 0 & i = 0 \\ opt[i - 1, W'] & i > 0, w_i > W' \\ \max \left\{ \begin{array}{l} opt[i - 1, W'] \\ opt[i - 1, W' - w_i] + v_i \end{array} \right\} & i > 0, w_i \leq W' \end{cases}$$

Exercise: Items with 3 Parameters

Input: integer bounds $W > 0$, $Z > 0$,
a set of n items, each with an integer weight $w_i > 0$
a size $z_i > 0$ for each item i
a value $v_i > 0$ for each item i

Output: a subset S of items that

$$\begin{aligned} & \text{maximizes } \sum_{i \in S} v_i && \text{s.t.} \\ & \sum_{i \in S} w_i \leq W \text{ and } \sum_{i \in S} z_i \leq Z \end{aligned}$$