

Data Structures and Algorithms

Dynamic Programming

Dr. Chung-Wen Albert Tsao

Overview

- Introductory Example
- Shortest Path Problem
- Dynamic Programming Concept
- Longest Common Subsequence
- General Alignment
- Knapsack
- Scheduling

A motivating example: Fibonacci numbers

- 1, 1, 2, 3, 5, 8, 13, 21, 34, 55, 89, ...

$$\begin{cases} F(0) = 1; \\ F(1) = 1; \\ F(n) = F(n-1) + F(n-2) \end{cases}$$

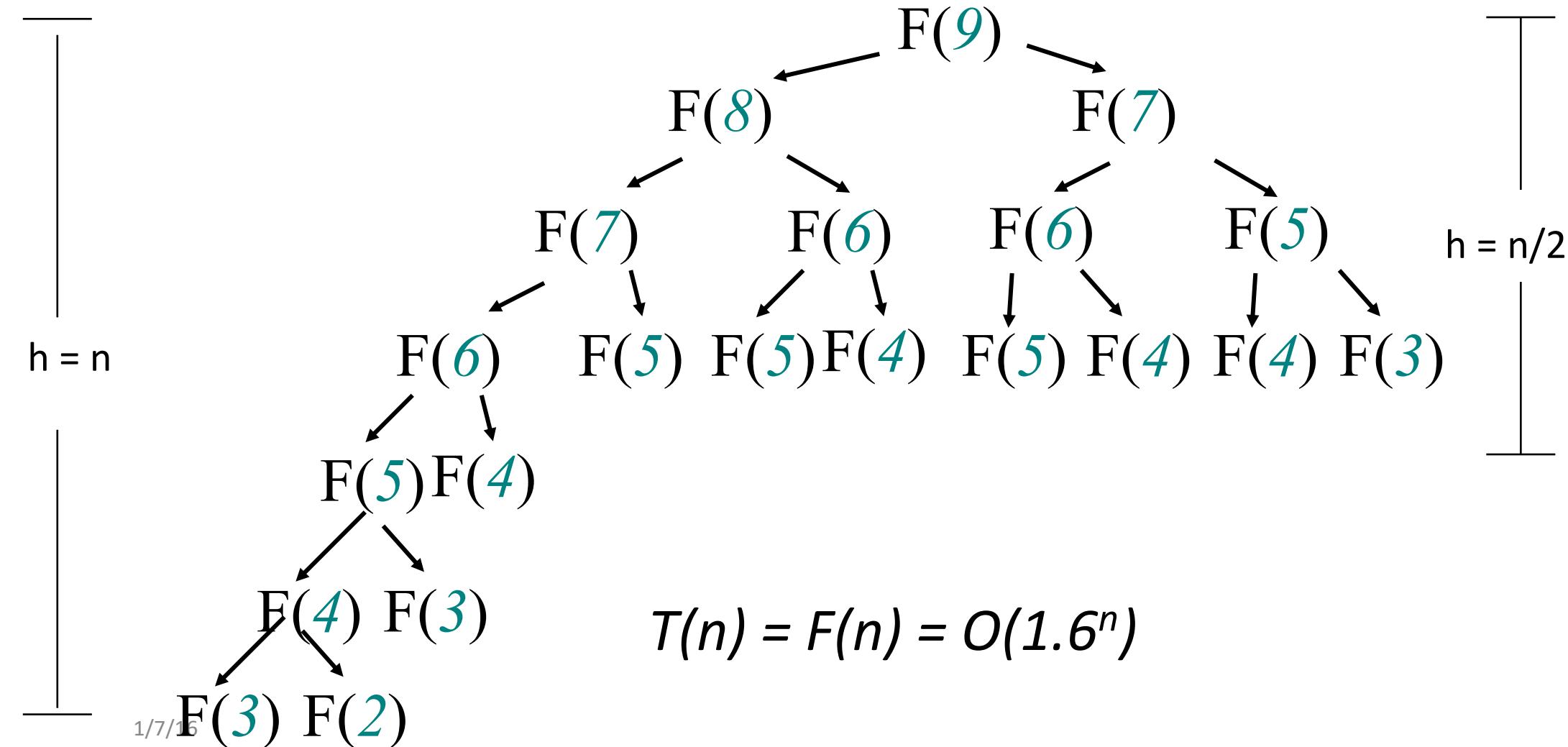
- How to compute $F(n)$?

A Top-Down Recursive Algorithm

```
function fib(n)
    if (n == 0 or n == 1) return 1;
    else return fib(n-1) + fib(n-2);
```

What's the time complexity?

- Time complexity between $2^{n/2} = (1.414)^n$ and 2^n



An Bottom-Up **iterative** algorithm

```
function fib(n)
    F[0] = 1;      F[1] = 1;
    for i = 2 to n
        F[i] = F[i-1] + F[i-2];
    Return F[n];
```

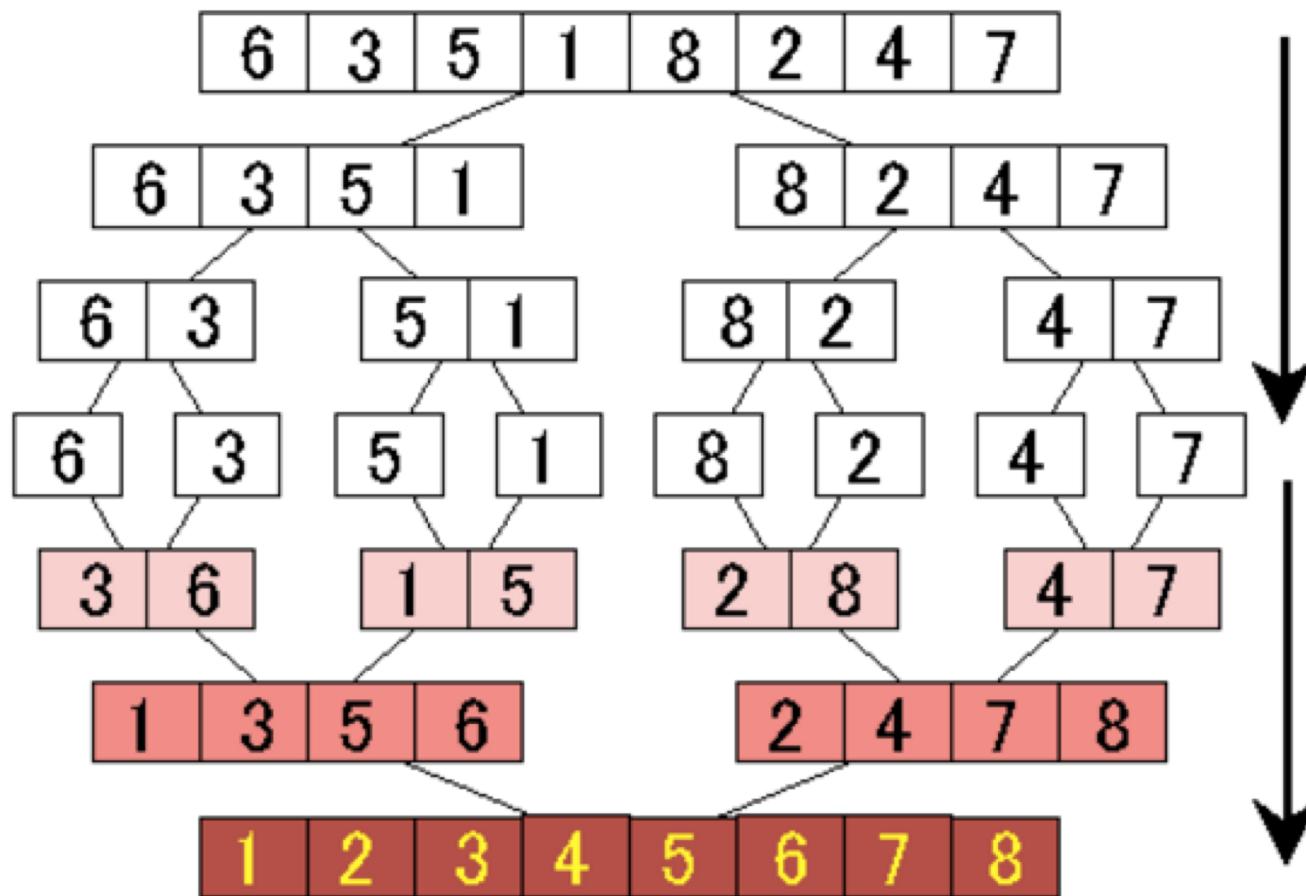
Time complexity?

	0	1	2	3	4	5	6	7	8	9	10	11	12	13
Fib	1	1	2	3	5	8								

Top-Down Recursive Algorithm

- Problem with the recursive Fib algorithm:
 - **Same subproblem was solved for many times!**
- Solution: avoid solving the same subproblem more than once
 - (1) Bottom-up iterative : pre-compute all subproblems that may be needed later
 - (2) Top-Down bottom-up: Compute on demand, but memorize the solution to avoid recomputing

Merge Sort



- Divide-and-conquer
- Recursively sort $\frac{1}{2}$ the set
- Can you always speedup a recursive algorithm by making it a bottom-up iterative algorithm?
 - E.g., merge sort
 - No. since there is no overlap between the two sub-problems

Maximum-Subarray problem (Ex 4.1.5, p75)

Find the contiguous subarray with the largest sum.

- For example, given the array A= [-2, 1, -3,4,-1,2,1,-5,4],
→The contiguous subarray [4,-1,2,1] has the largest sum = 6
- Brute-force solution (Exhaustive Enumerations)?

Example: House Robbery

- You are a professional robber planning to rob houses along a street.
- Each house has a certain amount of money stashed
- **You will trigger alarm if you break into any two adjacent houses.**
- Example:
 - Money in 7 houses: $M=[\$5, \$5, \$6, \$4, \$0, \$1, \$5]$
 - Houses robbed : [1, 0, 1, 0, 0, 0, 1]
 - Money robbed : \$16
- Solution:
 - Brute Force: generate all possible (2^n) scenarios, and pick the best.
 - Can we do better?





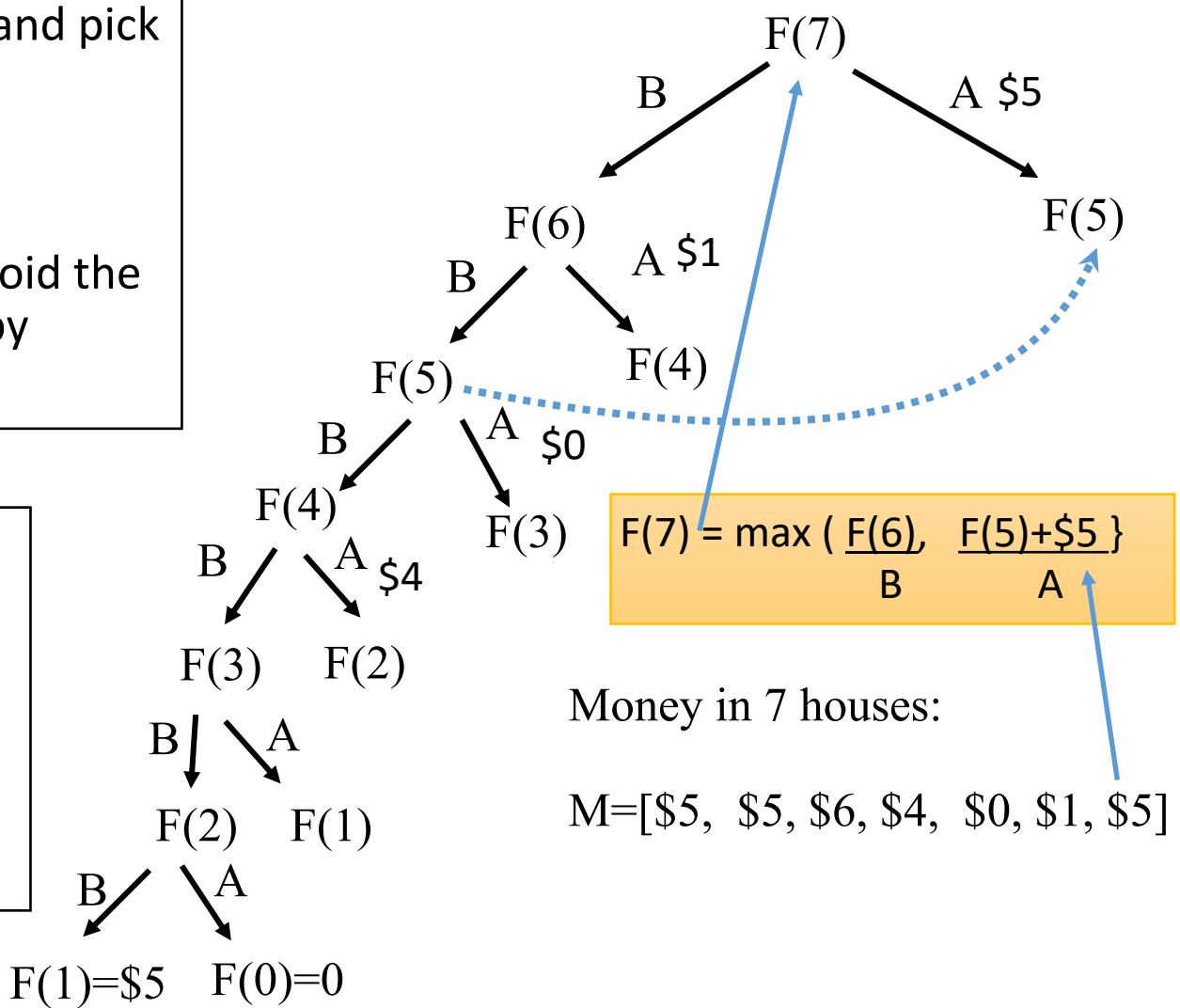
Dynamic Programming for House Robber

- Brute Force: generate all possible (2^n) scenarios, and pick the best.
 - Can we do better? (with less enumerations)
 - Yes. **Dynamic Programming**
- Break problem to overlapping subproblems, but avoid the repeated computation for the same subproblems by “memorization”

Define $F[k]$ = The most money robbed from the first k houses.

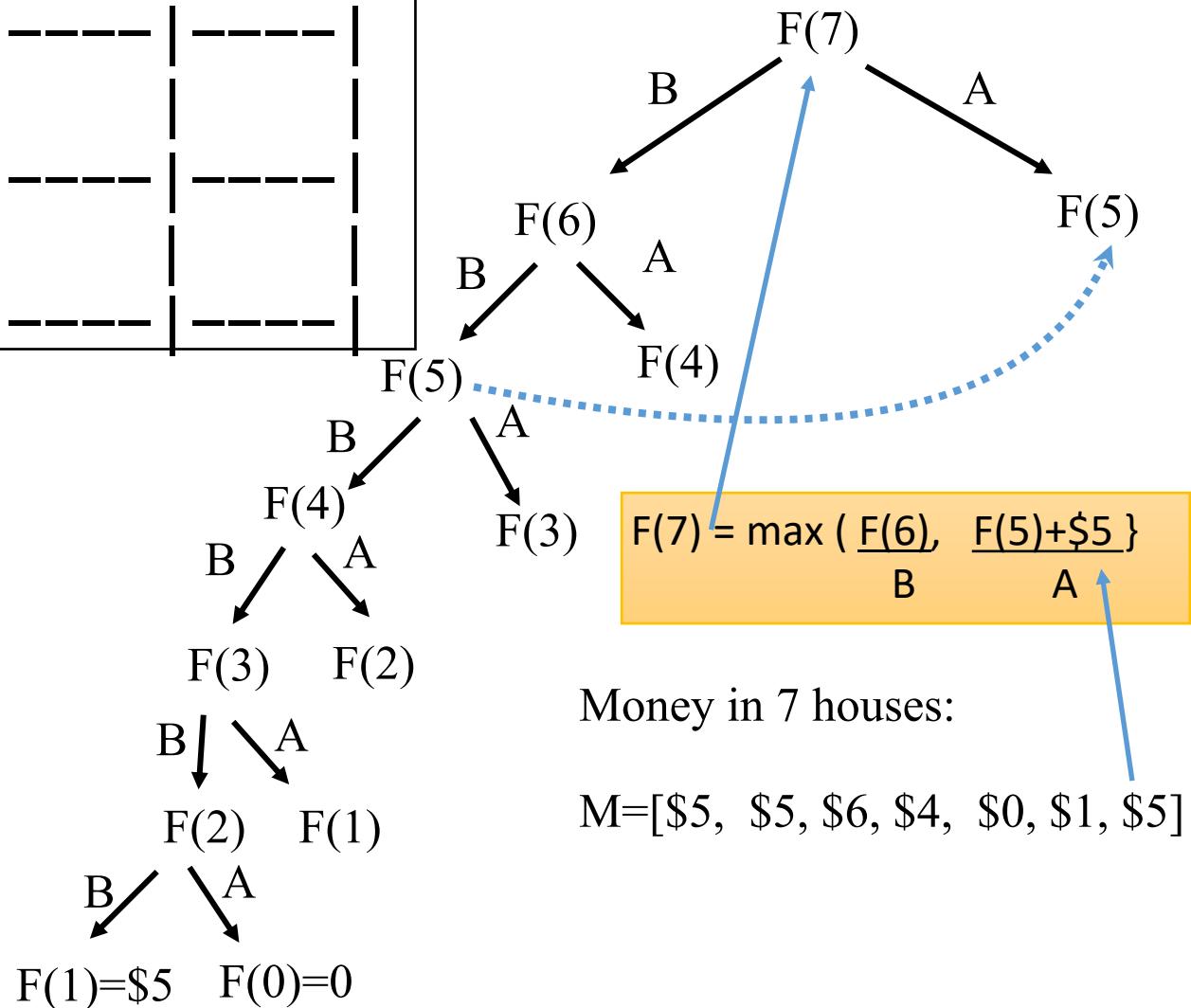
$$F[k] = \max \{ M[k]+F[k-2], \begin{array}{l} \text{Case A: rob house } k \\ \text{Case B: skip house } k \end{array} \}, \quad k \geq 2$$

$$T(N) = ?$$

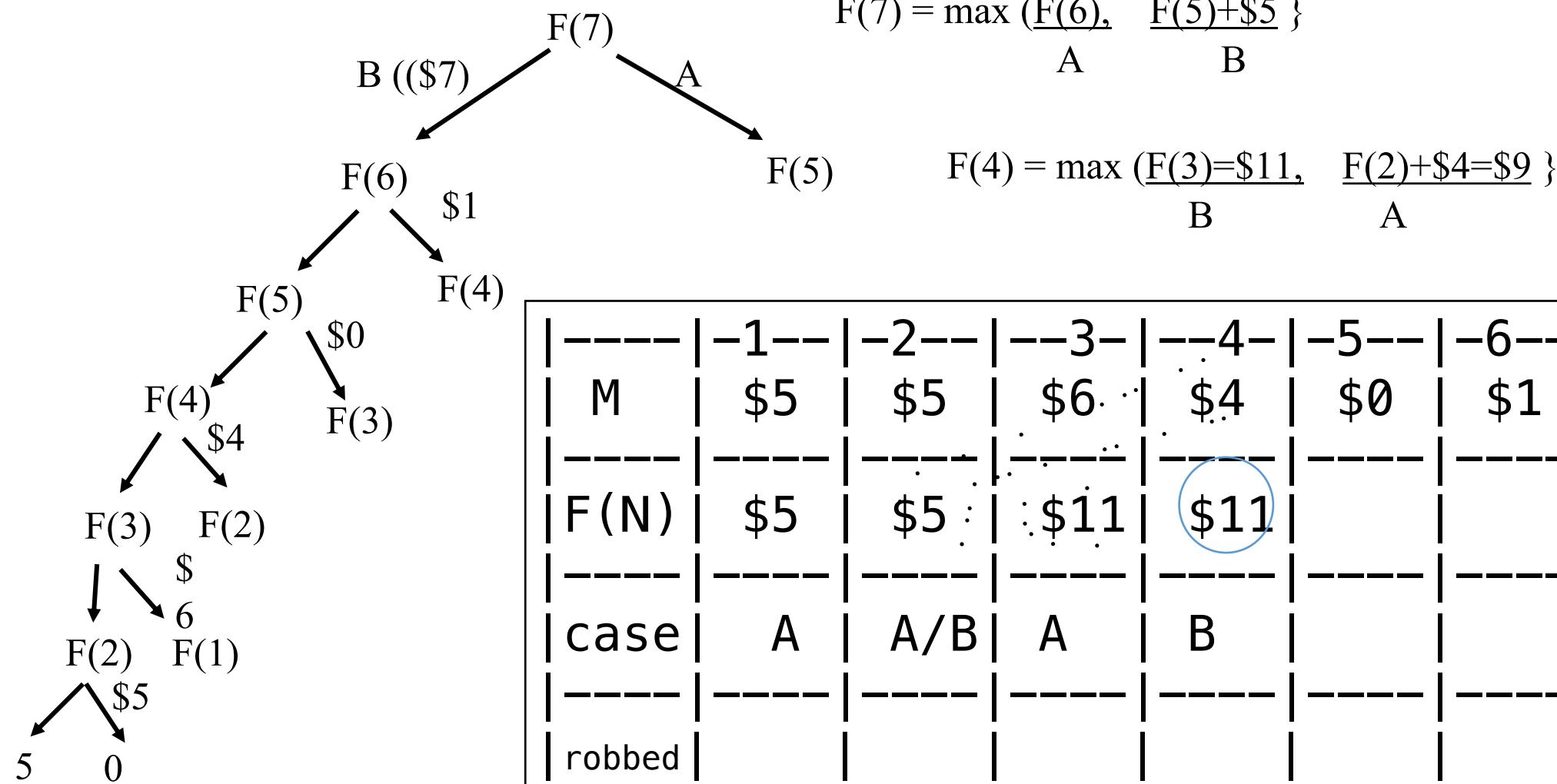




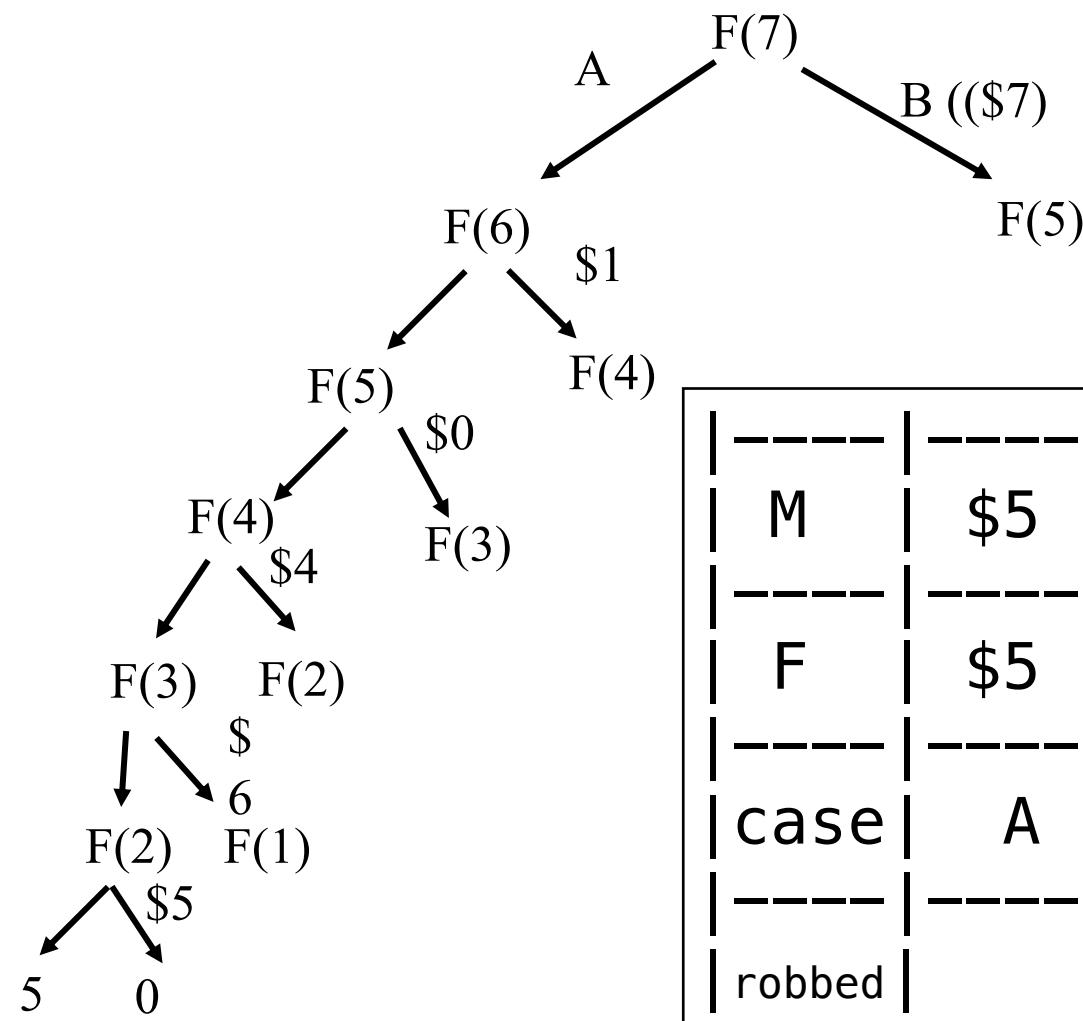
	-1--	-2--	-3--	-4--	-5--	-6--	-7--
M	\$5	\$5	\$6	\$4	\$0	\$1	\$5
F	\$5	\$5	\$11				
case	A	A	A				
robbed							



LeetCode 198. House Robber



LeetCode 198. House Robber

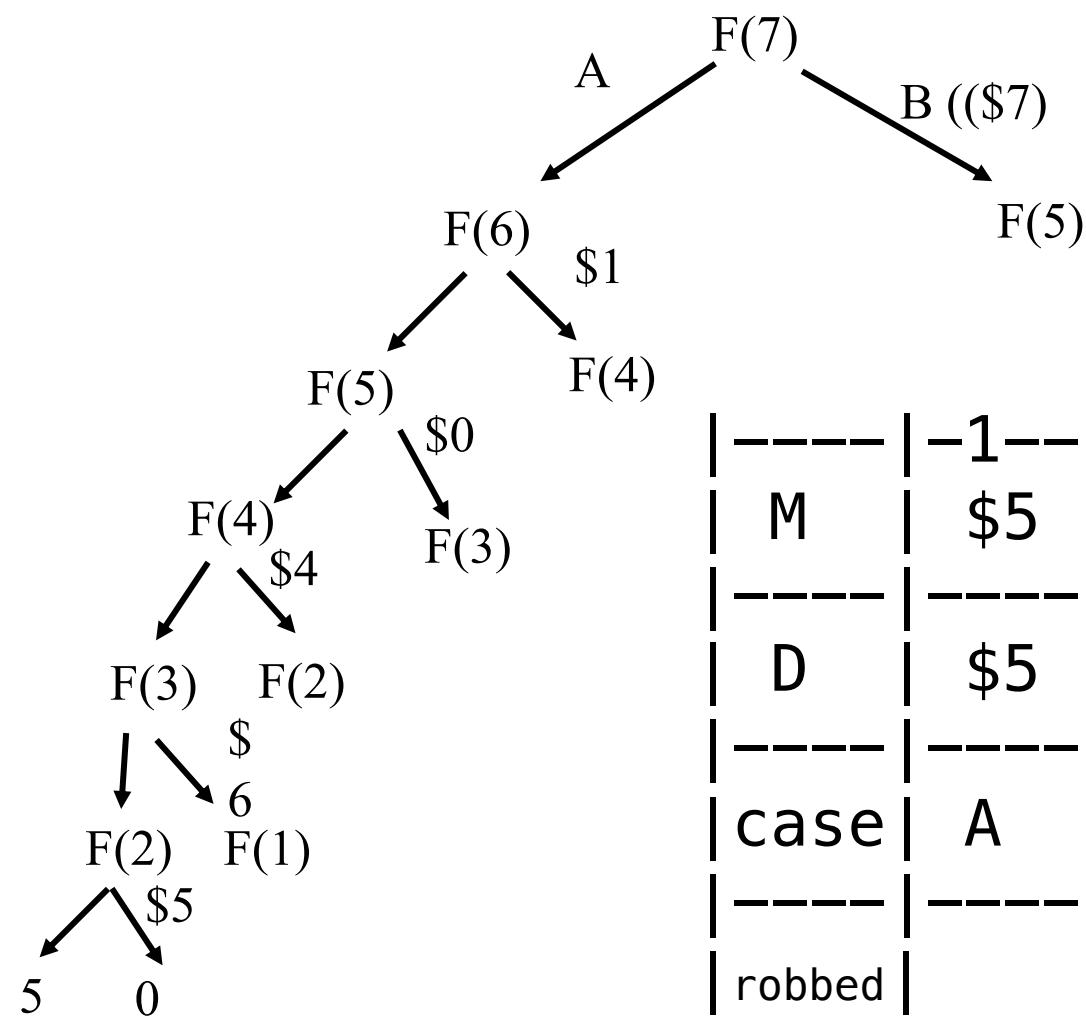


$$F(7) = \max (\underline{F(6)}, \underline{F(5)+\$5} \}$$

M	\$5	\$5	\$6	\$4	\$0	\$1	\$5
F	\$5	\$5	\$11	\$11	\$11		
case	A	A/B	A	B	A/B		
robbed							

The table shows the results of the recursion tree. The first row contains the values from the recursion tree. The second row contains the corresponding cases: A, A/B, A, B, A/B. The third row contains the labels: case, A, A/B, A, B, A/B. The fourth row contains the label: robbed. The fifth row contains empty cells. A blue circle highlights the value \$11 in the row F, column 6.

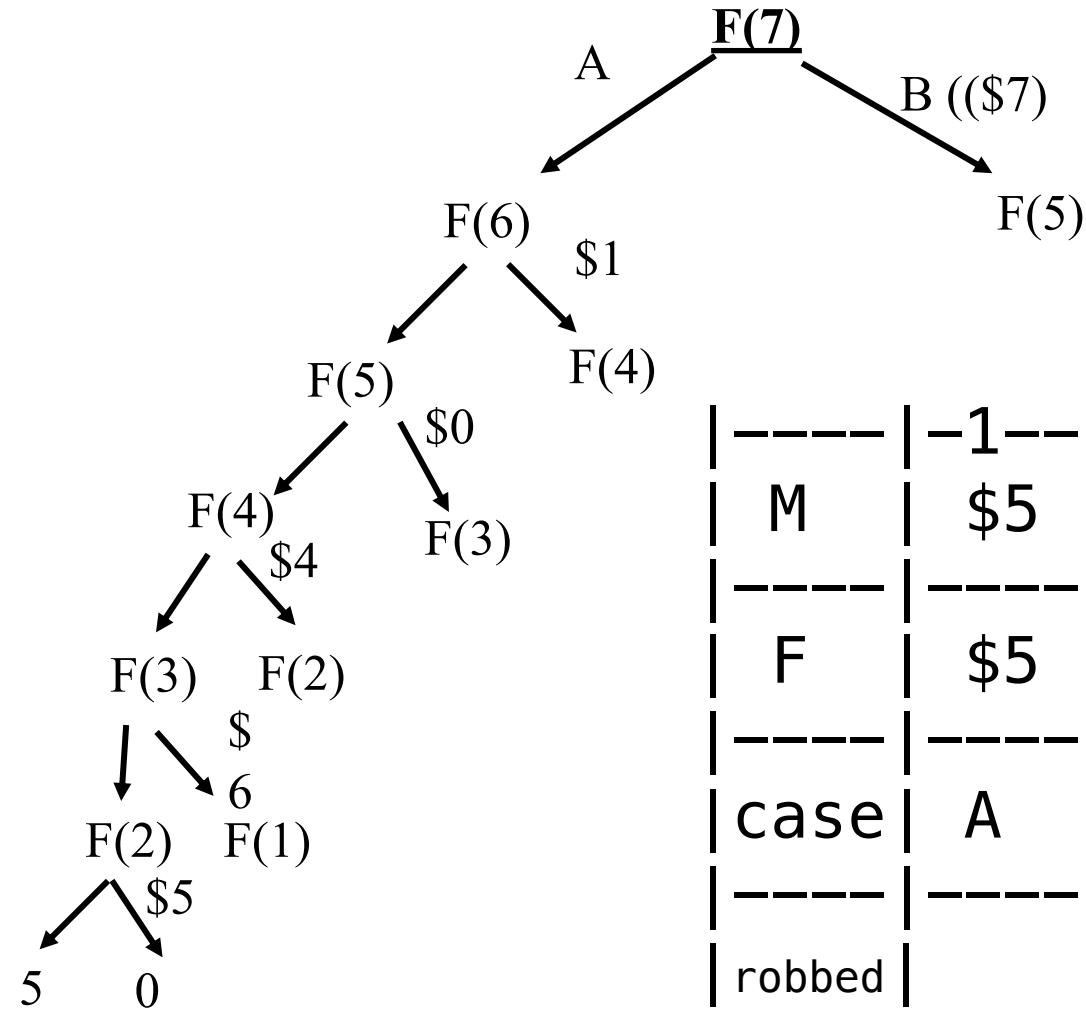
LeetCode 198. House Robber



$$F(7) = \max \{ F(6), \frac{F(5)+\$5}{A} \}$$

	-1	-2	-3	-4	-5	-6	-7
M	\$5	\$5	\$6	\$4	\$0	\$1	\$5
D	\$5	\$5	\$11	\$11	\$11	\$12	
case	A	A	A	B	B	A	
robbed							

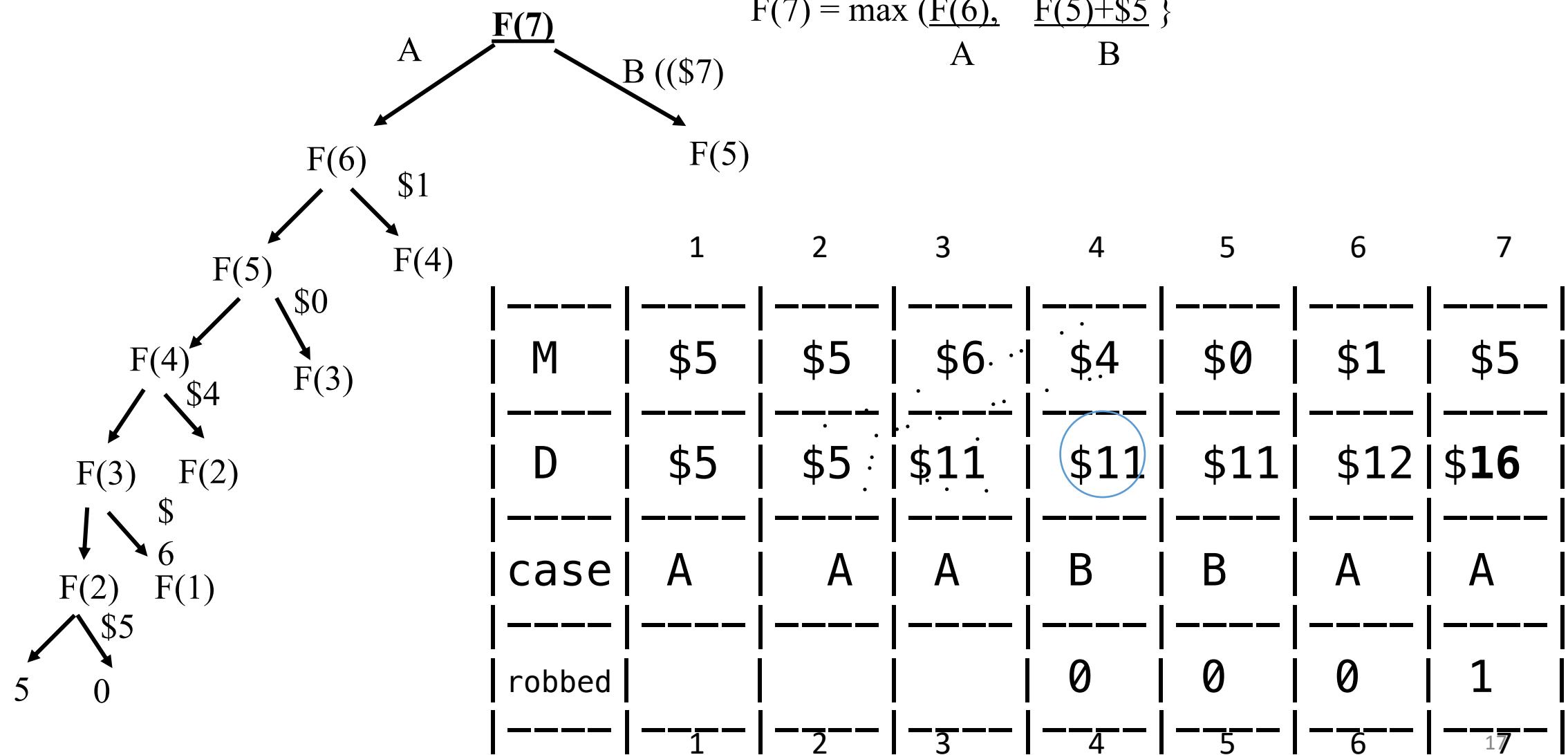
LeetCode 198. House Robber



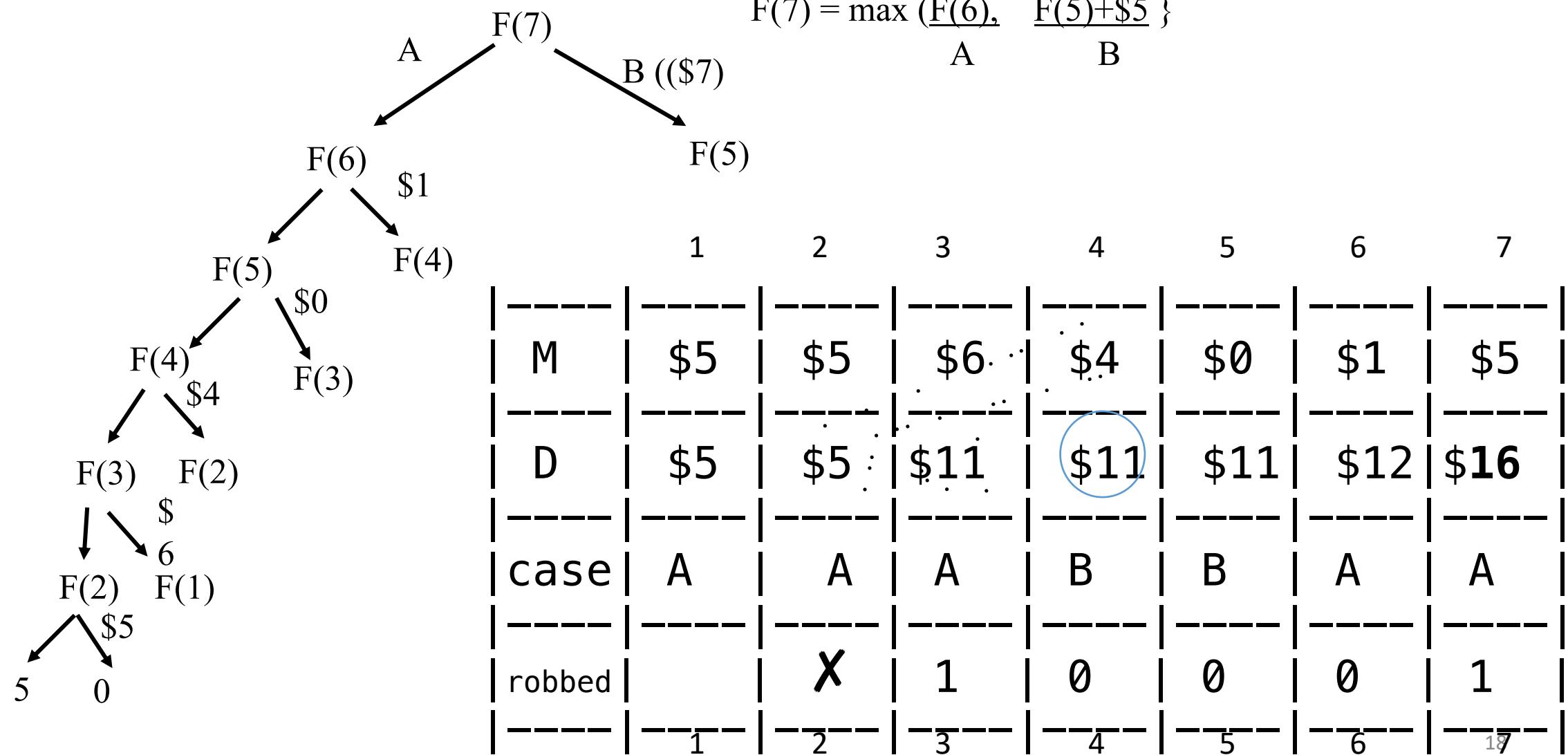
$$F(7) = \max \begin{cases} F(6), \\ A \\ F(5) + \$5 \end{cases} \quad \begin{cases} B \\ \end{cases}$$

	-1	-2	-3	-4	-5	-6	-7
M	\$5	\$5	\$6	\$4	\$0	\$1	\$5
F	\$5	\$5	\$11	\$11	\$11	\$12	\$16
case	A	A	A	B	B	A	A
robbed					X		1
							16

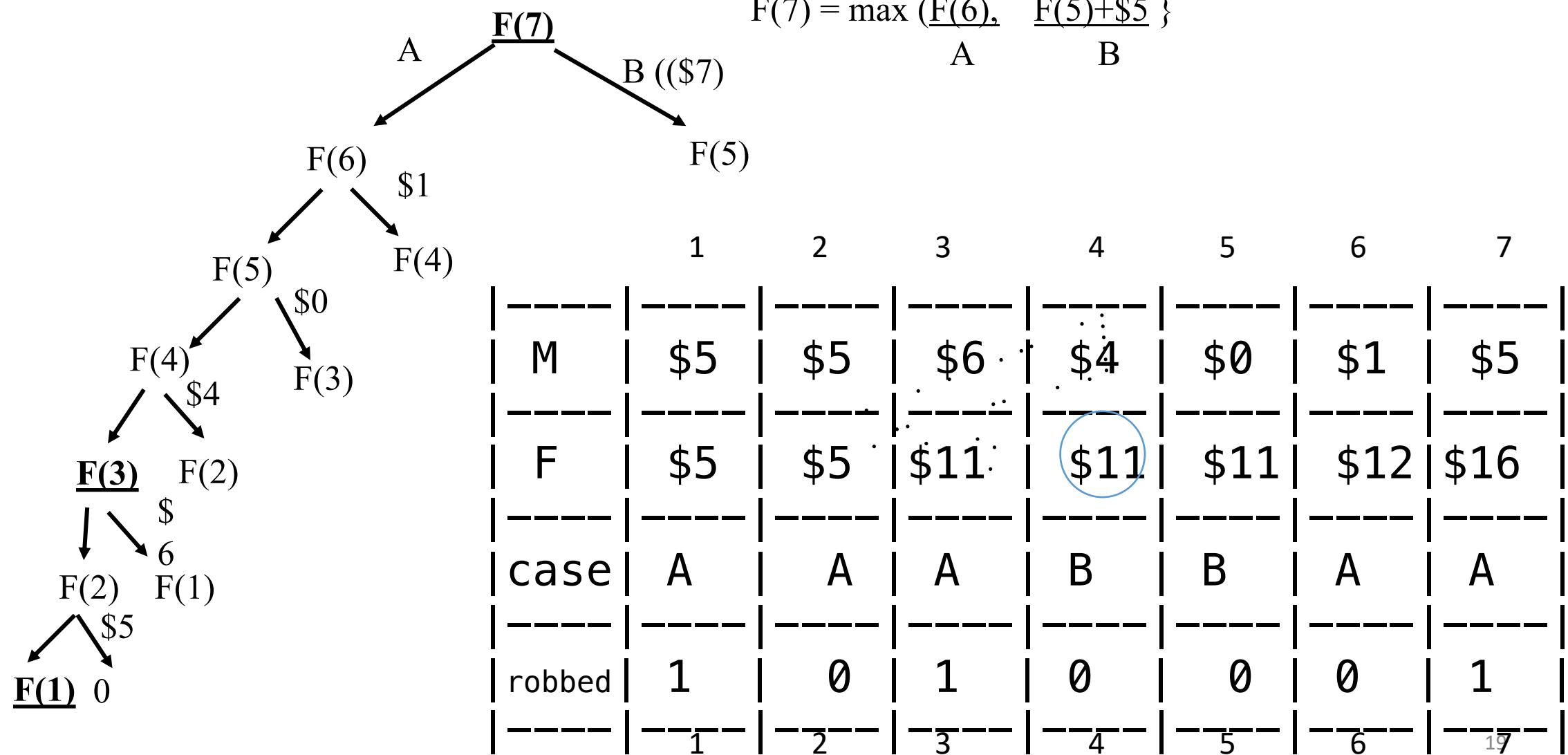
LeetCode 198. House Robber



LeetCode 198. House Robber



LeetCode 198. House Robber



LeetCode 198. House Robber

- How about
 - You will trigger alarm if you break into any two houses that have less than n houses between them, where $n=1,2,3, \dots$

- Define $F[k]$ = The most money you can rob from the first k houses.
$$F[k] = \max \{ \begin{array}{ll} F[k-n-1]+M[k], & \text{Case A: rob house } k \\ F[k-1], & \text{Case B: not rob house } k \text{ (Same as the previous solution)} \end{array} \}, \quad k \geq n+1$$
- $F[k] = \max \{ M[1], \dots, M[n] \}, \quad k \leq n$

Elements of dynamic programming

- Optimal sub-structures
 - Optimal solutions to the original problem contains optimal solutions to sub-problems
- Overlapping sub-problems
 - Some sub-problems appear repeatedly
 - Memorization and reuse
 - Carefully choose the order that sub-problems are solved, such that each sub-problem will be solved for at most once and the solution can be reused

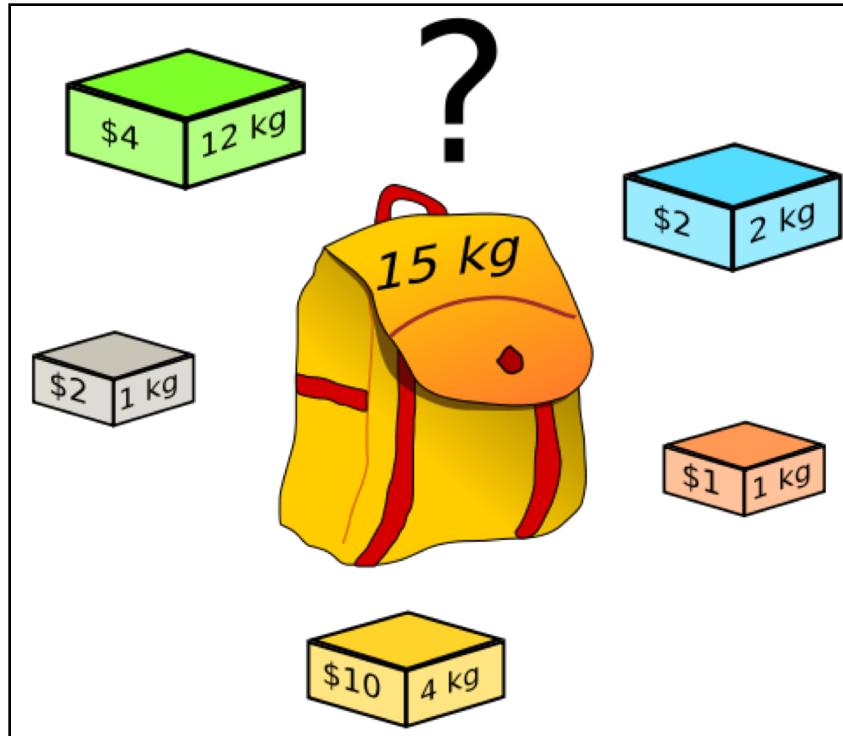
Two steps to dynamic programming

- Formulate the solution as a recurrence relation of solutions to subproblems.
- Specify an order to solve the subproblems so you always have what you need.
 - Bottom-up
 - Tabulate the solutions to all subproblems before they are used
 - Top-down
 - Compute when needed
 - Remember the ones you've computed

Knapsack problem



- Each item has a **value** and a **weight**
- **Objective:** maximize value
- **Constraint:** knapsack has a **weight limitation**



Three versions:

0-1 knapsack problem: take each item or leave it

Unbounded knapsack problem: unlimited supplies of each item

Fractional knapsack problem: items are divisible

Which one is easiest to solve?

Which one is most difficult to solve?

Fractional Knapsack

Greedy

- Consider following items (weight, value):

#1(10oz, \$5)

#2(15oz, \$6)

#3(20,oz \$5)

#4(18oz, \$6)

- Weight limit 35oz

What will be the most money?
What items to be picked up?

Sort items by their ratio \$/oz

item #1: $\$5/10\text{oz} = \mathbf{\$0.5/\text{oz}}$

item #2: $\$6/15\text{oz} = \mathbf{\$0.4/\text{oz}}$

item #4: $\$6/18\text{oz} = \mathbf{\$.333/\text{oz}}$

item #3: $\$5/20\text{oz} = \mathbf{.25/\text{oz}}$

#1(10oz) ==> \$5

#2(15oz) ==> \$6

#4 (10oz only) ==> \$3.33

total \$14.333

Unbounded Knapsack

Top-down Recursion

Consider following items

(weight, value):

A(10oz, \$5),

B(15oz, \$6),

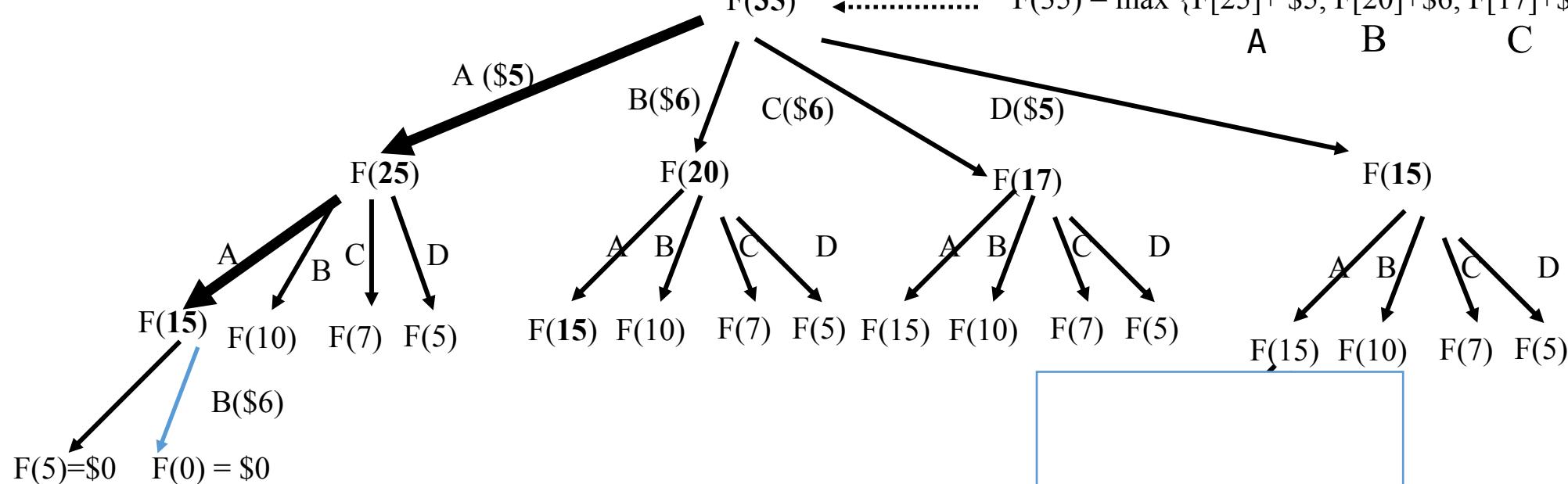
C(18oz, \$6),

D(20oz \$5),

Weight limit 35

$$\begin{aligned} F(35) &= \max \{F[25] + \$5, F[20] + \$6, F[17] + \$6, F[15] + \$5\} \\ F[25] &= \max \{F[15] + \$5, F[10] + \$6, \cancel{F[7]} + \$6, \cancel{F[5]} + \$5\} \\ F[20] &= \max \{F[10] + \$5, \cancel{F[5]} + \$6, \cancel{F[2]} + \$6, \cancel{F[0]} + \$5\} \\ F[17] &= \max \{F[7] + \$5, F[2] + \$6, \dots\} \\ F[15] &= \max \{F[5] + \$5, F[0] + \$6, \dots\} \\ F[10] &= \max \{F[0] + \$5, \dots\} \end{aligned}$$

$F(35) = \max \{F[25] + \$5, F[20] + \$6, F[17] + \$6, F[15] + \$5\}$



What will be the most money? **\\$16**

What items to pick? **_____ items A,A,B,_____**

Number of Coin Change

- <http://www.geeksforgeeks.org/dynamic-programming-set-7-coin-change/>
- Given some denomination of coins (e.g., \$1,\$2,\$5), Find the total number of coins to make change for a value (e.g, \$11)
- Unbounded knapsack: Unlimited number of coins for each denomination
 - Coins = [\$1, \$2, \$5], amount = \$11 \Rightarrow return 3 ($\$11 = \$5 + \$5 + \1)
- 0-1 Knapsack problem: Use each denomination at most once
 - Coins = [\$1, \$2, \$5], amount = \$11 \Rightarrow no solution

Unbounded Knapsack

LeetCode#322. Fewest Coin Change

- Define $D[i] = \text{fewest coins to make up } \j .
- $D[i][j] = \min\{ D[j - \$i] + 1 \}$, } for all possible coin with value $\$i$
- Example:
 - Coins = $[\$2, \$5, \$7, \$10]$, amount = $\$13$, fewest coins=? **Answer = 4(1 pt)** (using one $\$7$ and three $\$2$)

```
F[13] = 1 + min {F[11], F[8], F[6].F[3]} = 1 + 3 = 4
F[11] = 1 + min {F[ 9], F[6], F[4].F[1]} = 1 + 2 = 3
F[ 9] = 1 + min {F[ 7], F[4], F[2]} = 2
F[ 8] = 1 + min {F[ 6], F[3], F[1]} = 4
F[ 7] = 1
F[ 6] = 1 + min {F[ 4], F[1]} = 3
F[ 4] = 1 + min {F[ 2]} = 2
```


0-1 Knapsack

Formal definition (0-1 problem)

- Knapsack has weight limit W
- Items labeled $1, 2, \dots, n$ (arbitrarily)
- Items have weights w_1, w_2, \dots, w_n
 - Assume all weights are integers
 - For practical reason, only consider $w_i < W$
- Items have values v_1, v_2, \dots, v_n
- Objective: find a subset of items, S , such that $\sum_{i \in S} w_i \leq W$ and $\sum_{i \in S} v_i$ is maximal among all such (*feasible*) subsets

Naïve algorithms

- Enumerate all subsets.
 - Optimal. But exponential time
- Greedy 1: take the item with the largest value
 - Not optimal
 - Give an example
- Greedy 2: take the item with the largest value/weight ratio
 - Not optimal
 - Give an example

Brute force!

- Generate all 2^n subsets
- Discard all subsets whose sum of the weights exceed W (*not feasible*)
- Select the maximum total benefit of the remaining (feasible) subsets
- What is the run time? $O(n 2^n)$
- Lets try the obvious greedy strategy .

Example with “brute force”

$$S = \{ (A, 5, \$70), (B, 10, \$90), (C, 25, \$140) \}, W=25$$

Subsets:

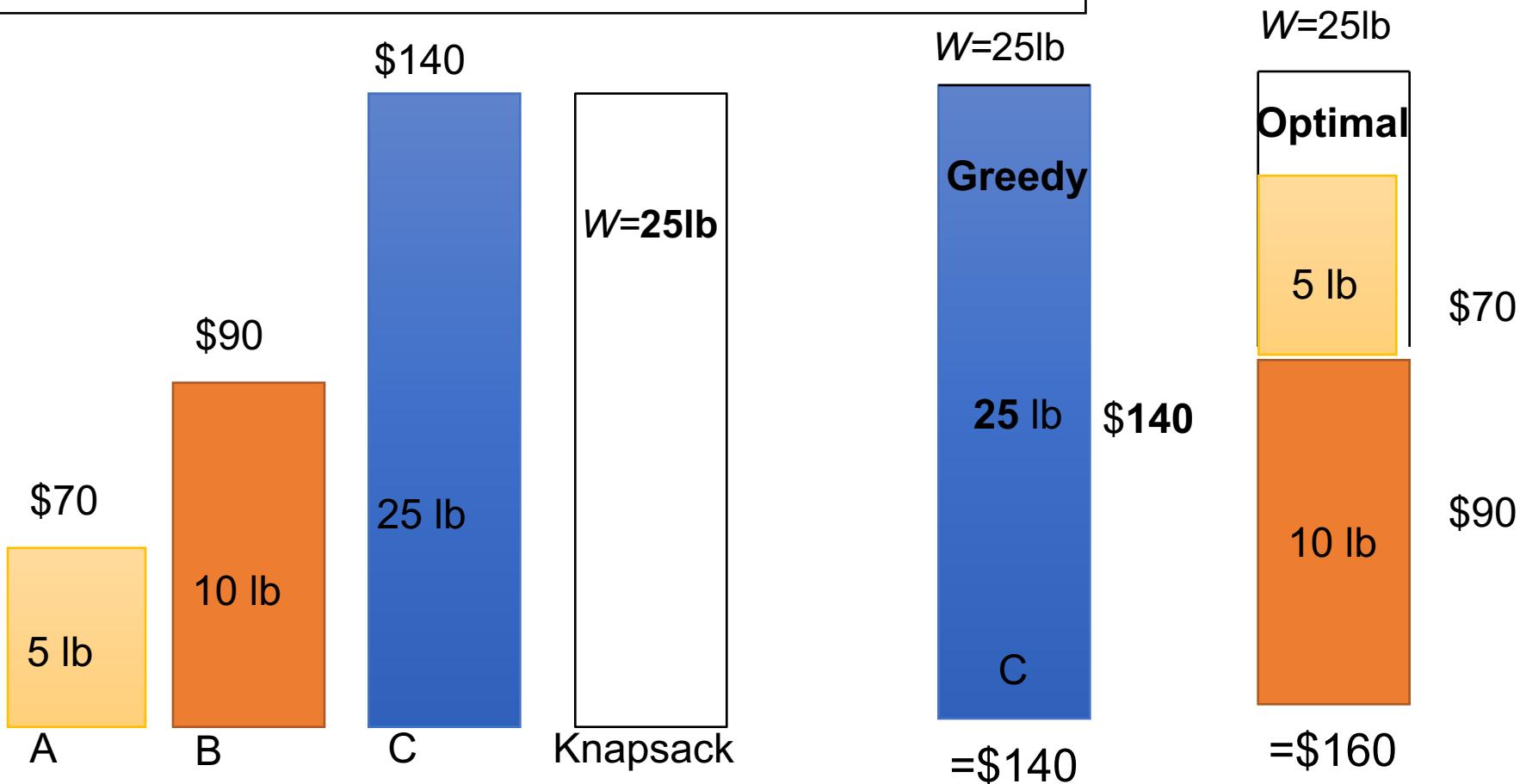
- | | |
|--|--------------------------|
| 1. {} | Profit=\$0 |
| 2. (A, 5, \$70) | Profit=\$70 |
| 3. (B, 10, \$90) | Profit=\$90 |
| 4. (C, 25, \$140) | Profit=\$140 |
| 5. (A, 5, \$70), (B, 10, \$90) | Profit=\$160 **** |
| 6. (B, 10, \$90), (C, 25, \$140) | exceeds W |
| 7. (A, 5, \$70), (C, 25, \$140) | exceeds W |
| 8. (A, 5, \$70), (B, 10, \$90), (C, 25, \$140) | exceeds W |

0-1 Knapsack

Greedy 1: **Maximum beneficial** item.

Counter Example:

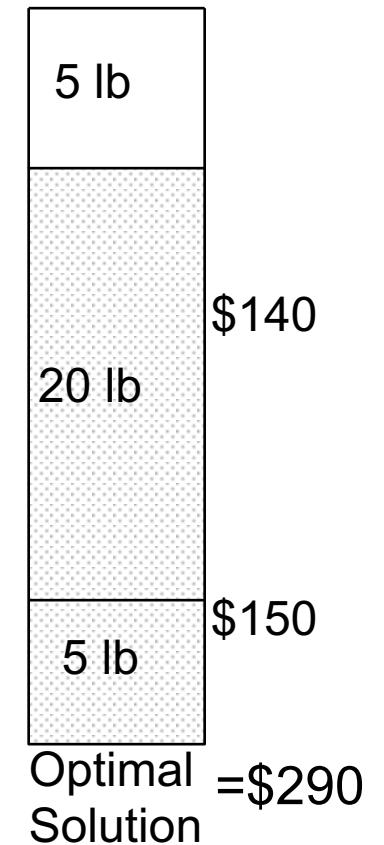
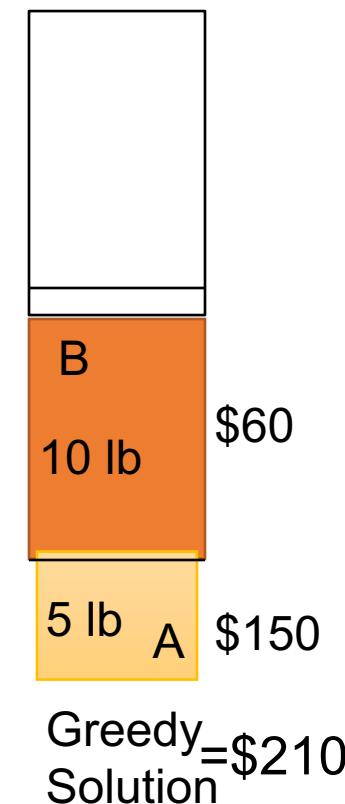
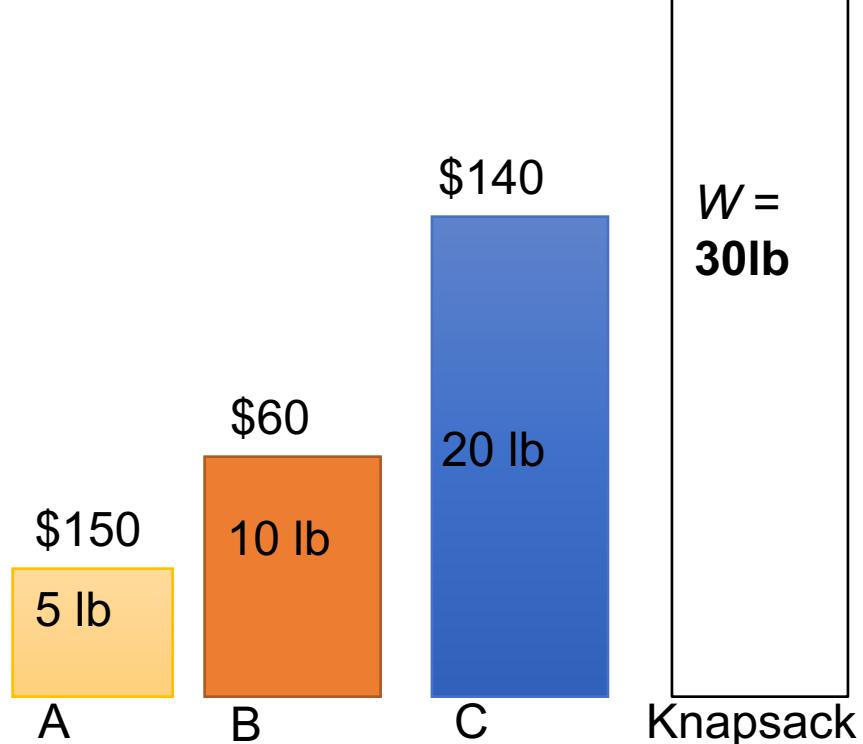
Counter Example: (A , 5, \$70), (B , 10, \$90), (C, 25, \$140)



0-1 Knapsack

Greedy 2: *Minimum weight item*

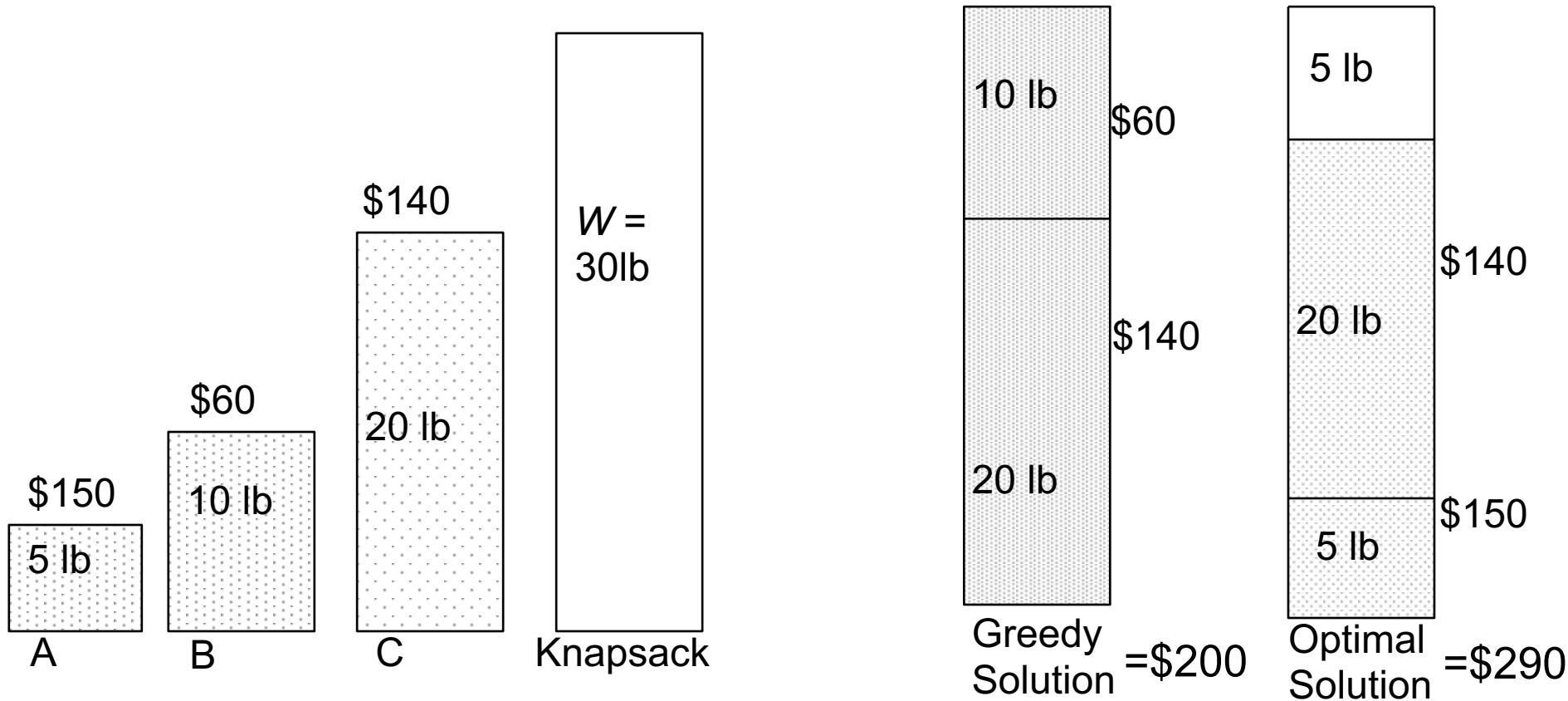
Counter Example: $S = \{ (A, 5\text{lb}, \$150), (B, 10, \$60), (C, 20, \$140) \}$



0-1 Knapsack

Greedy 3: *Maximum weight* item

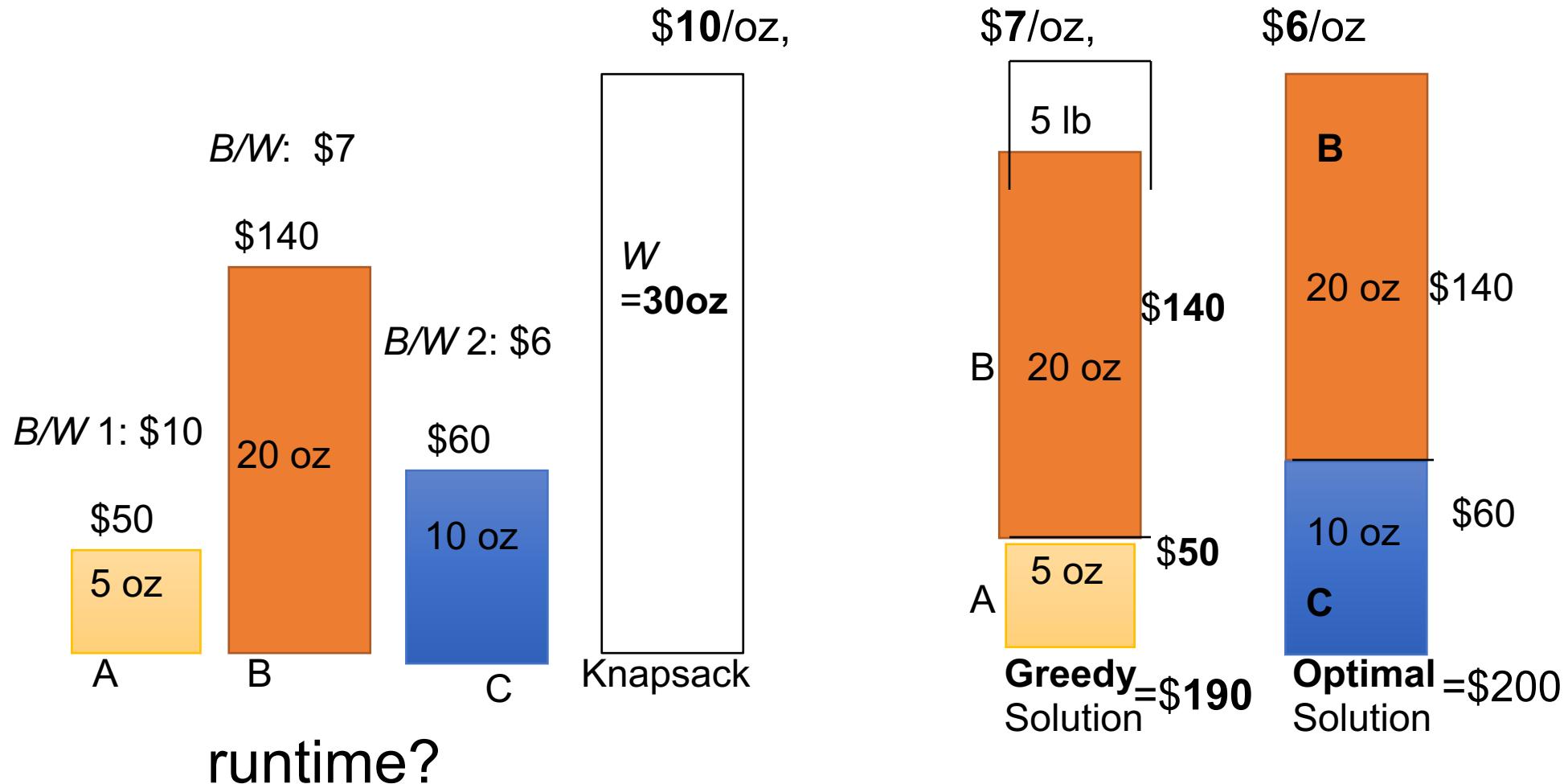
Counter Example: $S = \{ (A, 5, \$150), (B, 10, \$60), (C, 20, \$140) \}$



0-1 Knapsack

Greedy 4: *Maximum benefit per unit item* item

Counter Example $S = \{ (A, 5\text{oz}, \$50), (B, 20\text{oz}, \$140), (C, 10\text{oz}, \$60) \}$

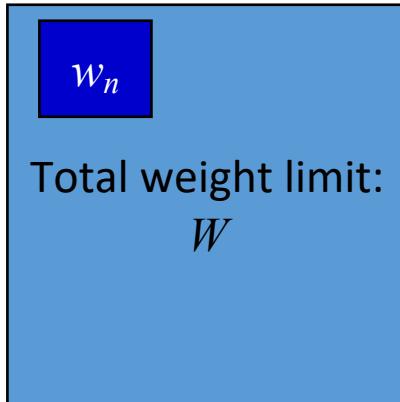


0-1 Knapsack

A DP algorithm

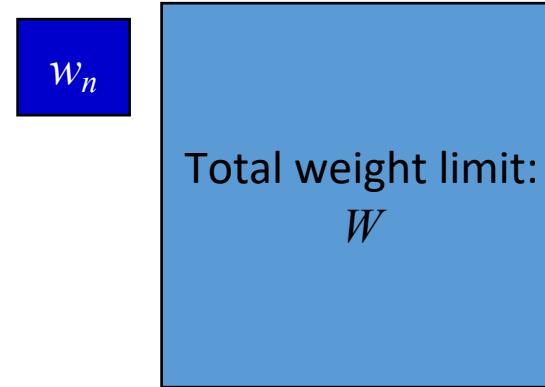
Suppose you've find the optimal solution S

Case 1: item n is included



Find an optimal solution using items
 $1, 2, \dots, n-1$ with weight limit $W - w_n$

Case 2: item n is not included



Find an optimal solution using items $1, 2, \dots, n-1$ with weight limit W

$$F[i, w] = \max \begin{cases} F[i-1, w-w_i] + v_i & \text{Case I: item } i \text{ is taken, and } w \geq w_i \\ F[i-1, w] & \text{Case II: item } i \text{ not taken} \end{cases}$$

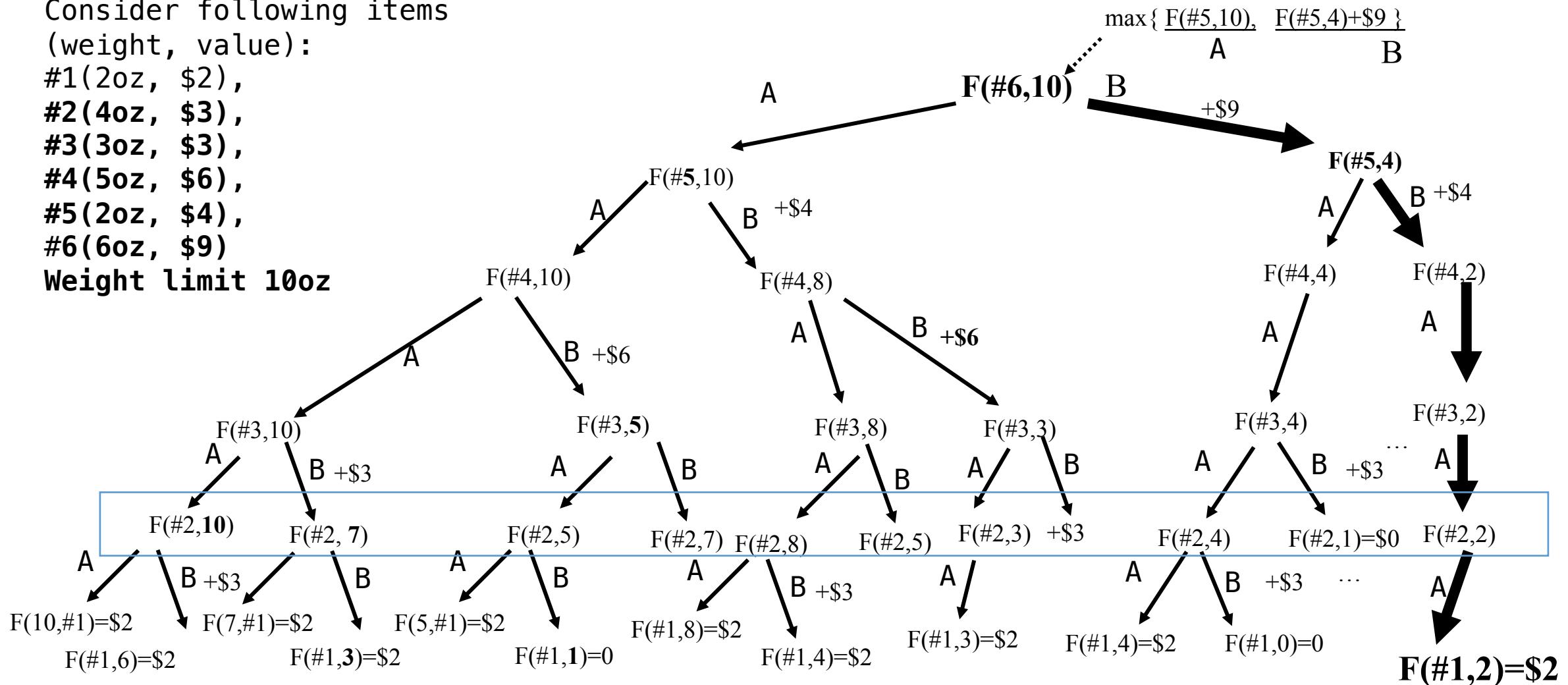
0-1 Knapsack

$$F[i, w] = \max \begin{cases} F[i-1, w-w_i] + v_i & \text{Case I: item } i \text{ is taken, and } w \geq w_i \\ F[i-1, w] & \text{Case II: item } i \text{ not taken} \end{cases}$$

Consider following items
(weight, value):

- #1(2oz, \$2),
- #2(4oz, \$3),
- #3(3oz, \$3),
- #4(5oz, \$6),
- #5(2oz, \$4),
- #6(6oz, \$9)

Weight limit 10oz



0-1 Knapsack

	w	0	1	2	3	4	5	6	7	8	9	10
i	w_i	v_i	0	0	0	0	0	0	0	0	0	0
#1	2lb	\$2	0	0	\$2	\$2	\$2	\$2	\$2	\$2	\$2	\$2
#2	4lb	\$3	0	0								
#3	3lb	\$3	0	0	$F[i-1, w-w_i]$				$F[i-1, w]$			
#4	5lb	\$6	0	0			\$6		$F[i, w]$			
#5	2lb	\$4	0	0								
#6	6lb	\$9	0	0								

- $V[i, w]$ = optimal total value when only considering items 1, 2, ..., i with weight limit w

$$F[i, w] = \max \begin{cases} F[i-1, w-w_i] + v_i & \text{Case I: item } i \text{ is taken, and } w \geq w_i \\ F[i-1, w] & \text{Case II: item } i \text{ not taken} \end{cases}$$

Unbounded Knapsack

Knapsack

	w	0	1	2	3	4	5	6	7	8	9	10
i	w_i	v_i	0	0	0	0	0	0	0	0	0	0
1	2lb	\$2	0									
2	4lb	\$3	0									
3	3lb	\$3	0									
4	5lb	\$6	0									
5	2lb	\$4	0									
6	6lb	\$9	0									

- $V[i, w]$ = optimal total value when only considering items $1, 2, \dots, i$ with weight limit w

$$V[i, w] = \max \begin{cases} V[i, w-w_i] + v_i & \text{Case I: item } i \text{ is taken, and } w \geq w_i \\ V[i-1, w] & \text{Case II: item } i \text{ not taken any more} \end{cases}$$

0-1 Knapsack

	w	0	1	2	3	4	5	6	7	8	9	10
	w_i	v_i	0	0	0	0	0	0	0	0	0	0
#1	2lb	\$2	0	0	\$2	2	2	2	\$2	2	2	2
#2	4lb	\$3	0	0	2	2	3	3	5	5	5	5
#3	3lb	\$3	0	5	0	2	3	+\$3	+\$0	6		
4	5lb	\$6	0									
5	2lb	\$4	0									
6	6lb	\$9	0									

- $V[i, w]$ = optimal total value when only considering items $1, 2, \dots, i$ with weight limit w

$$V[i, w] = \max \begin{cases} V[i-1, w-w_i] + v_i & \text{item } i \text{ is taken, and } w \geq w_i \\ V[i-1, w] & \text{item } i \text{ not taken} \end{cases}$$

0-1 Knapsack

	w	0	1	2	3	4	5	6	7	8	9	10
w_i	v_i	0	0	0	0	0	0	0	0	0	0	0
2lb	\$2	0	0	2	2	2	2	2	2	2	2	2
4lb	\$3	0	0	2	2	3	3	5	5	5	5	5
3lb	\$3	0	0	2	3	3	5	5	6	6	8	8
5lb	\$6	0	0	2	3	3	6	6	8	9	9	11
2lb	\$4	0	0	4	4	6	7	7	10	10	12	13
6lb	\$9	0	0	4	4	6	7	9	10	13	13	15

$$V[i, w] = \max \begin{cases} V[i-1, w-w_i] + v_i & \text{item } i \text{ is taken, and } w \geq w_i \\ V[i-1, w] & \text{item } i \text{ not taken} \end{cases}$$

0-1 Knapsack

	w	0	1	2	3	4	5	6	7	8	9	10
w_i	v_i	0	0	0	0	0	0	0	0	0	0	0
2	\$2	0	0	2	2	2	2	2	2	2	2	2
4	\$3	0	0	2	2	3	3	5	5	5	5	5
3	\$3	0	0	2	3	3	5	5	6	6	8	8
5	\$6	0	0	2	3	3	6	6	8	9	9	11
2	\$4	0	0	4	4	6	7	7	10	10	12	13
6	\$9	0	0	4	4	6	7	9	10	13	13	15

Item: 6, 5, 1

Weight: $6 + 2 + 2 = 10$

Value: $9 + 4 + 2 = 15$

Optimal value: 15

0-1 Knapsack

	w	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14
w_i	v_i	0	0	0	0	0	0	0	0	0	0	0				
2	\$2	0	0	2	2	2	2	2	2	2	2	2				
4	\$3	0	0	2	2	3	3	5	5	5	5	5				
3	\$3	0	0	2	3	3	5	5	6	6	8	8				
5	\$6	0	0	2	3	3	6	6	8	9	9	11				
2	\$4	0	0	4	4	6	7	7	10	10	12	13				
6	\$9	0	0	4	4	6	7	9	10	13	13	15				

0-1 Knapsack

	w	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14
w_i	v_i	0	0	0	0	0	0	0	0	0	0	0	Q	Q	Q	Q
2	\$2	0	0	2	2	2	2	2	2	2	2	2	2	2	2	2
4	\$3	0	0	2	2	3	3	5	5	5	5	5	5	5	5	5
3	\$3	0	0	2	3	3	5	5	6	6	8	8	8	8	8	8
5	\$6	0	0	2	3	3	6	6	8	9	9	11	11	11	11	11
2	\$4	0	0	4	4	6	7	7	10	10	12	12	13	13	13	13
6	\$9	0	0	4	4	6	7	9	10	13	13	13	15			

0-1 Knapsack

	w	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14
w_i	v_i	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
2	\$2	0	0	2	2	2	2	2	2	2	2	2	2	2	2	2
4	\$3	0	0	2	2	3	3	5	5	5	5	5	5	5	5	5
3	\$3	0	0	2	3	3	5	5	6	6	8	8	8	8	8	8
5	\$6	0	0	2	3	3	6	6	8	9	9	11	11	12	12	14
2	\$4	0	0	4	4	6	7	7	10	10	12	13	13	15	15	16
6	\$9	0	0	4	4	6	7	9	10	13	13	15	16	16	19	19

Time complexity

- $\Theta(nW)$
- Polynomial?
 - Pseudo-polynomial
 - Works well if W is small
- Consider following items (weight, value):
 $(10\text{oz}, \$5), (15\text{oz}, \$6), (20\text{oz}, \$5), (18\text{oz}, \$6)$
- Weight limit 35
 - Brute-Force: $2^4 = 16$ subsets
 - Dynamic programming: fill up a $4 \times 35\text{lb} = 140$ table entries
- What's the problem?
 - Many entries are unused: no such weight combination
 - **Top-down may be better**

0-1 Knapsack

0-1 Knapsack: Top-down

Consider following items
(weight, value):
#1(10oz, \$5),
#2(15oz, \$6),
#3(20oz, \$5),
#4(18oz, \$6),
Weight limit 35

