

### General idea

- Dynamic programming is a technique for solving problems with overlapping subproblems.
- Typically, these subproblems arise from a recurrence relating a given problem's solution to solutions of its smaller subproblems.
- Rather than solving overlapping subproblems again and again, dynamic programming suggests solving each of the smaller subproblems only once and recording the results in a table from which a solution to the original problem can then be obtained.

# Review (Fibonacci numbers)

- F(n) = F(n-1) + F(n-2) for n > 1
- F(0) = 0, F(1) = 1.

## Review Knapsack Problem

$$F(i, j) = \begin{cases} \max\{F(i-1, j), v_i + F(i-1, j-w_i)\} & \text{if } j-w_i \geq 0, \\ F(i-1, j) & \text{if } j-w_i < 0. \end{cases}$$

$$F(0, j) = 0 \text{ for } j \ge 0 \text{ and } F(i, 0) = 0 \text{ for } i \ge 0.$$

### Coin Row Problem

- There is a row of n coins whose values are some positive integers c1, c2, . . . , cn, not necessarily distinct.
- The goal is to pick up the maximum amount of money subject to the constraint that no two coins adjacent in the initial row can be picked up.

# Example

• 5, 1, 2, 10, 6, 2.

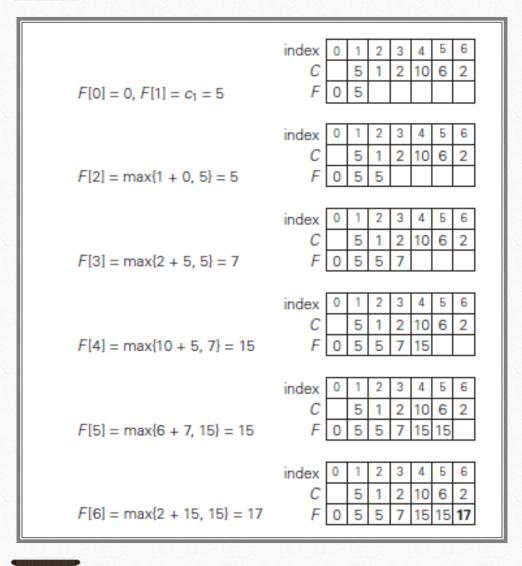
### General Idea

- Let F(n) be the maximum amount that can be picked up from the row of n coins.
- To derive a recurrence for F(n), we partition all the allowed coin selections into two groups:
  - those that include the last coin and
  - those without it.
- The largest amount we can get from the first group is equal to cn + F(n 2)—the value of the nth coin plus the maximum amount we can pick up from the first n 2 coins.
- The maximum amount we can get from the second group is equal to F(n 1) by the definition of F(n).

#### Recurrence

$$F(n) = \max\{c_n + F(n-2), F(n-1)\}$$
 for  $n > 1$ ,  
 $F(0) = 0$ ,  $F(1) = c_1$ .

#### Solve example with DP



# Coin-collecting problem

- Several coins are placed in cells of an  $n \times m$  board, no more than one coin per cell.
- A robot, located in the upper left cell of the board, needs to collect as many of the coins as possible and bring them to the bottom right cell.
- On each step, the robot can move either one cell to the right or one cell down from its current location.
- When the robot visits a cell with a coin, it always picks up that coin.
- Design an algorithm to find the maximum number of coins the robot can collect and a path it needs to follow to do this.

### General Idea

- Let F(i, j) be the largest number of coins the robot can collect and bring to the cell (i, j) in the ith row and jth column of the board.
- It can reach this cell either from the adjacent cell (i 1, j) above it or from the adjacent cell (i, j 1) to the left of it.
- The largest numbers of coins that can be brought to these cells are F(i 1, j) and F(i, j 1), respectively

#### recourence

$$F(i, j) = \max\{F(i - 1, j), F(i, j - 1)\} + c_{ij} \text{ for } 1 \le i \le n, 1 \le j \le m$$
  
 $F(0, j) = 0 \text{ for } 1 \le j \le m \text{ and } F(i, 0) = 0 \text{ for } 1 \le i \le n,$ 

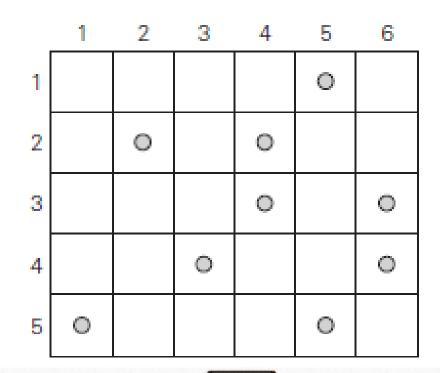
where  $c_{ij} = 1$  if there is a coin in cell (i, j), and  $c_{ij} = 0$  otherwise.

### PseudoCode

```
ALGORITHM RobotCoinCollection(C[1..n, 1..m])

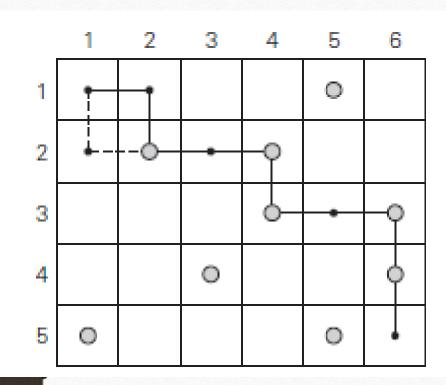
//Applies dynamic programming to compute the largest number of
//coins a robot can collect on an n \times m board by starting at (1, 1)
//and moving right and down from upper left to down right corner
//Input: Matrix C[1..n, 1..m] whose elements are equal to 1 and 0
//for cells with and without a coin, respectively
//Output: Largest number of coins the robot can bring to cell (n, m)
F[1, 1] \leftarrow C[1, 1]; for j \leftarrow 2 to m do F[1, j] \leftarrow F[1, j - 1] + C[1, j]
for i \leftarrow 2 to n do
F[i, 1] \leftarrow F[i - 1, 1] + C[i, 1]
for j \leftarrow 2 to m do
F[i, j] \leftarrow \max(F[i - 1, j], F[i, j - 1]) + C[i, j]
return F[n, m]
```

# Example



## Solution

·	1	2	3	4	5	6
1	0	0	0	0	1	1
2	0	1	1	2	2	2
3	0	1	1	3	3	4
4	0	1	2	3	3	5
5	1	1	2	3	4	5



# Change-making problem

- Consider the general instance of the following well-known problem. Give change for amount n using the minimum number of coins of denominations d1<d2 < . . . < dm.
- Here, we consider a dynamic programming algorithm for the general case, assuming availability of unlimited quantities of coins for each of the m denominations
- d1 < d2 < ... < dm where <math>d1 = 1.
- Let F(n) be the minimum number of coins whose values add up to n; it is convenient to define F(0) = 0. The amount n can only be obtained by adding one coin of denomination dj to the amount n dj for j = 1, 2, ..., m such that  $n \ge dj$ .
- Therefore, we can consider all such denominations and select the one minimizing F(n dj) + 1. Since 1 is a constant, we can, of course, find the smallest F(n dj) first and then add 1 to it.

#### recurrence

$$F(n) = \min_{j: n \ge d_j} \{F(n - d_j)\} + 1 \quad \text{for } n > 0,$$
  
$$F(0) = 0.$$

## pseudocode

```
ALGORITHM ChangeMaking(D[1..m], n)

//Applies dynamic programming to find the minimum number of coins
//of denominations d_1 < d_2 < \cdots < d_m where d_1 = 1 that add up to a
//given amount n

//Input: Positive integer n and array D[1..m] of increasing positive
// integers indicating the coin denominations where D[1] = 1
//Output: The minimum number of coins that add up to n

F[0] \leftarrow 0

for i \leftarrow 1 to n do

temp \leftarrow \infty; j \leftarrow 1

while j \leq m and i \geq D[j] do

temp \leftarrow \min(F[i - D[j]], temp)

j \leftarrow j + 1

F[i] \leftarrow temp + 1

return F[n]
```

# example

$$F[0] = 0$$

$$F[1] = \min\{F[1-1]\} + 1 = 1$$

$$F[2] = \min\{F[2-1]\} + 1 = 2$$

$$F[3] = \min\{F[3-1], F[3-3]\} + 1 = 1$$

$$F[4] = \min\{F[4-1], F[4-3], F[4-4]\} + 1 = 1$$

$$n = 0 + 1 + 2 + 3 + 4 + 5 = 6$$
 $F[5] = min\{F[5-1], F[5-3], F[5-4]\} + 1 = 2$ 
 $F[5] = min\{F[5-1], F[5-3], F[5-4]\} + 1 = 2$ 

$$F[6] = \min\{F[6-1], F[6-3], F[6-4]\} + 1 = 2$$

n	0	1	2	3	4	5	6
F	0	1					

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n	U		4	3	4	ъ	0
F	0	1	2				

n	0	1	2	3	4	5	6
F	0	1	2	1			

n	0	1	2	3	4	5	6
F	0	1	2	1	1		

n	0	1	2	3	4	5	6
F	0	1	2	1	1	2	2