

Two-dimensional Dynamic programming

When there are two parameters which define the state, then we call those problems 2D Dynamic Programming problems.

We use a 2D array to cache the results of the subproblems.

- There are two input strings or arrays
- Involves processing input string or array from both the sides.
- Input is a 2D array or list of strings

1. Minimum cost path

You are given a grid. You start at (0,0), top-left corner and you have to reach bottom left corner. You can only move down or right. Find out the minimum cost to reach the destination.

Hint: start from the last

0	47	8	18	1
43	25	39	36	13
22	8	13	38	46
41	41	40	25	44
29	43	22	50	10

Minimum cost path

0	47	8	18	1
43	25	39	36	13
22	8	13	38	46
41	41	40	25	44
29	43	22	50	10

Minimum cost path

0	47	8	18	1
43	25	39	36	13
22	8	13	38	46
41	41	40	25	44
29	43	22	50	10

Minimum cost path

0	47	8	18	1
↓				
43	25	39	36	13
22	8	13	38	46
41	41	40	25	44
29	43	22	50	10

Minimum cost path

0	47	8	18	1
43	25	39	36	13
↓				
22	8	13	38	46
41	41	40	25	44
29	43	22	50	10

Minimum cost path

0	47	8	18	1
43→25		39	36	13
22	8	13	38	46
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1. State

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Parameters

i - row of the current cell.

j - column of the current cell.

Cost function

$\text{minPath}(i,j,G)$ - minimum cost to reach the cell (i,j) starting from $(0,0)$ by moving only right or down.

Minimum sum path

0	47	8	18	1
43	25	39	36	13
22	8	13	38	46
41	41	40	25	44
29	43	22	50	10



2. Transitions

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Base case

$i = 0, j = 0$, return $G[0][0]$

Choices

$\text{minPath}(i, j, G)$

$\text{minPath}(i-1, j, G) + G[i][j]$, if $i > 0$

$\text{minPath}(i, j-1, G) + G[i][j]$, if $j > 0$

Optimal choice

$\text{minPath}(i, j, G) = \text{MIN}(\text{minPath}(i-1, j, G), \text{minPath}(i, j-1, G)) + G[i][j]$

3. Recursive solution

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Pseudo code

minPath(i,j,G)

 if i==0 and j==0:

 return G[0][0]

 min = INFINITE

 if i>0

 min = minPath(i-1,j,G)+G[i][j]

 if j>0

 min = MIN(min,minPath(i,j-1,G)+G[i][j])

 return min

Java

```
public static int minPath(int i, int j, int[][] grid) {  
    if (i == 0 && j == 0) {  
        return grid[0][0];  
    }  
    int min = Integer.MAX_VALUE;  
    if (i > 0) {  
        min = minPath(i - 1, j, grid) + grid[i][j];  
    }  
    if (j > 0) {  
        min = Math.min(min, minPath(i, j - 1, grid) + grid[i][j]);  
    }  
    return min;  
}
```

Python

```
def min_path(i, j, G):  
    if i == 0 and j == 0:  
        return G[0][0]  
    min_cost = sys.maxsize  
    if i > 0:  
        min_cost = min_path(i - 1, j, G) + G[i][j]  
    if j > 0:  
        min_cost = min(min_cost, min_path(i, j - 1, G) + G[i]  
[j])  
  
    return min_cost
```

Minimum cost path

0	47	8	18	1
43	25	39	36	13
22	8	13	38	46
41	41	40	25	44
29	43	22	50 ← 10	

Minimum cost path

0	47	8	18	1
43	25	39	36	13
22	8	13	38	46
41	41	40	25	44
29	43	22	50	10



Minimum cost path

0	47	8	18	1
43	25	39	36	13
22	8	13	38	46
41	41	40	25	44
29	43	22	50	10



Minimum cost path

0	47	8	18	1
43	25	39	36	13
22	8	13	38	46
41	41	40	25 ← 44	
29	43	22	50	10

4. Memoize

4. Memoization

We can cache the results in a 2D array.

Key -> (i,j)

Value -> Minimum cost to reach the cell, (i,j) from (0,0)

Default value -> -1

Java

```
public static int minPathMemo(int i, int j, int[][] grid, int[][] cache) {  
    if (i == 0 && j == 0) {  
        return 0;  
    }  
    if (cache[i][j] != -1) {  
        return cache[i][j];  
    }  
    int min = Integer.MAX_VALUE;  
    if (i > 0) {  
        min = minPathMemo(i - 1, j, grid, cache) + grid[i][j];  
    }  
    if (j > 0) {  
        min = Math.min(min, minPathMemo(i, j - 1, grid, cache) + grid[i][j]);  
    }  
    cache[i][j] = min;  
    return min;  
}
```

Python

```
def min_path_memo(i, j, G, cache):  
    if i == 0 and j == 0:  
        return 0  
    if cache[i][j] != -1:  
        return cache[i][j]  
  
    min_cost = sys.maxsize  
    if i > 0:  
        min_cost = min_path(i - 1, j, G) + G[i][j]  
    if j > 0:  
        min_cost = min(min_cost, min_path(i, j - 1, G) + G[i]  
[j])  
    cache[i][j] = min_cost  
    return min_cost
```

5. Bottom up approach

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$\text{minPath}(i,j,G) = \text{MIN}(\text{minPath}(i-1,j,G), \text{minPath}(i,j-1,G)) + G[i][j]$

$\text{dp}[i][j] = \text{MIN}(\text{dp}[i-1][j], \text{dp}[i][j-1]) + G[i][j]$

$\text{dp}[5][5]$ depends on $\text{dp}[4][5]$ and $\text{dp}[5][4]$,

so our for loops should go from

$i=0, 1 \dots M$

and

$j=0, 1, \dots M$

Java

```
public static int minPathDP(int[][] G){
    int M = G.length;
    int[][] dp = new int[M][M];
    dp[0][0] = G[0][0];
    for(int i=0;i<M;i++){
        for(int j=0;j<M;j++){
            if(i==0 && j == 0){
                continue;
            }
            dp[i][j] = Integer.MAX_VALUE;
            if(i>0){
                dp[i][j] = dp[i-1][j]+G[i][j];
            }
            if(j>0){
                dp[i][j] = Math.min(dp[i][j],dp[i][j-1]+G[i][j]);
            }
        }
    }
    return dp[M-1][M-1];
}
```

Python

```
def min_path_dp(G):
    M = len(G)
    dp = [[0 for _ in range(0, M)] for _ in range(0, M)]

    dp[0][0] = G[0][0];
    for i in range(0, M):
        for j in range(0, M):
            if i == 0 and j == 0:
                continue
            dp[i][j] = sys.maxsize
            if i > 0:
                dp[i][j] = dp[i-1][j] + G[i][j]
            if j > 0:
                dp[i][j] = min(dp[i][j], dp[i][j-1] + G[i][j])
    return dp[M-1][M-1]
```

Reconstruct the path

6. Reconstruct the path

We need to print the path.

Along with the minimum cost, we will record the direction in a 2D array 'dir'

0 - Down

1 - Right

We start from the last position $(N-1, N-1)$, $i=N-1, j=N-1$

if its 0, then we assign $i=i-1$, we print (i, j)

if its 1, then we assign $j=j-1$, we print (i, j)

We repeat the process until $i=0$ and $j=0$

Minimum cost path

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Time and space complexity

Recursive solution

We draw recursion tree.

Its a binary tree.

Height of the tree is $2N$.

Number of nodes in the tree = $2^{2N} = 4^N$

Time complexity is , $O(4^N)$, Exponential

Space complexity $O(1)$

Dynamic programming solution

We have two forloops iterating through every cell in the grid of length N once.

There are $N \times N$ cells = N^2

Time complexity is $O(N^2)$

Space complexity is $O(N^2)$