

Unique Paths

Solution 

★★★★★

There is a robot on an $m \times n$ grid. The robot is initially located at the **top-left corner** (i.e., `grid[0][0]`). The robot tries to move to the **bottom-right corner** (i.e., `grid[m - 1][n - 1]`). The robot can only move either down or right at any point in time.

Given the two integers m and n , return *the number of possible unique paths that the robot can take to reach the bottom-right corner*.

The test cases are generated so that the answer will be less than or equal to $2 * 10^9$.

Example 1:



Input: $m = 3, n = 7$

Output: 28

Example 2:

Input: $m = 3, n = 2$

Output: 3

Explanation: From the top-left corner, there are a total of 3 ways to reach the bottom-right corner:

1. Right -> Down -> Down
2. Down -> Down -> Right
3. Down -> Right -> Down

Constraints:

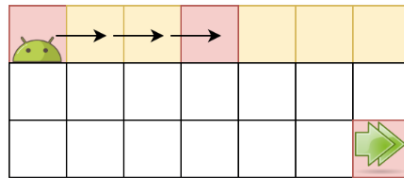
- $1 \leq m, n \leq 100$

Solution

Overview

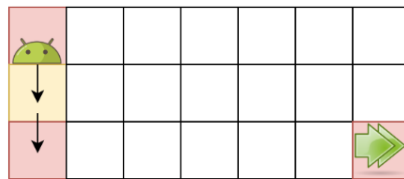
Since robot can move either down or right, there is only one path to reach the cells in the first row: right->right->...->right.

There is only one path to reach the cell in the first row:
right -> right -> ... -> right



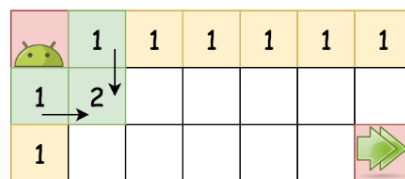
The same is valid for the first column, though the path here is down->down->...->down.

There is only one path to reach the cell in the first column:
down -> down -> ... -> down



What about the "inner" cells (m, n) ? To such cell one could move either from the cell on the left $(m, n - 1)$, or from the cell above $(m - 1, n)$. That means that the total number of paths to move into (m, n) cell is $\text{uniquePaths}(m - 1, n) + \text{uniquePaths}(m, n - 1)$.

$$\text{uniquePaths}(1, 1) = \text{uniquePaths}(0, 1) + \text{uniquePaths}(1, 0)$$



Now, one could transform these ideas into 3-liner recursive solution:

```
class Solution {  
    public int uniquePaths(int m, int n) {  
        if (m == 1 || n == 1) {  
            return 1;  
        }  
    }  
}
```

```

    return uniquePaths(m - 1, n) + uniquePaths(m, n - 1);
}
}

```

This solution is not fast enough to pass all the testcases, though it could be used as a starting point for the DP solution.

Approach 1: Dynamic Programming

One could rewrite recursive approach into dynamic programming one.

Algorithm

- Initiate 2D array `d[m][n] = number of paths`. To start, put number of paths equal to 1 for the first row and the first column. For the simplicity, one could initiate the whole 2D array by ones.
- Iterate over all "inner" cells: `d[col][row] = d[col - 1][row] + d[col][row - 1]`.
- Return `d[m - 1][n - 1]`.

Implementation: video

```

class Solution {
public int uniquePaths(int m, int n) {
    int[][] d = new int[m][n];

    for(int[] arr : d) {
        Arrays.fill(arr, 1);
    }
    for(int col = 1; col < m; ++col) {
        for(int row = 1; row < n; ++row) {
            d[col][row] = d[col - 1][row] + d[col][row - 1];
        }
    }
    return d[m - 1][n - 1];
}
}

```

Complexity Analysis

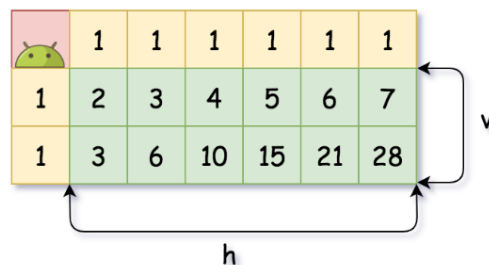
- Time complexity: $\mathcal{O}(N \times M)$.
- Space complexity: $\mathcal{O}(N \times M)$.

Approach 2: Math (Python3 only)

Could one do better than $\mathcal{O}(N \times M)$? The answer is yes.

The problem is a classical combinatorial problem: there are $h + v$ moves to do from start to finish, $h = m - 1$ horizontal moves, and $v = n - 1$ vertical ones. One could choose when to move to the right, i.e. to define h horizontal moves, and that will fix vertical ones. Or, one could choose when to move down, i.e. to define v vertical moves, and that will fix horizontal ones.

**$h + v$ moves from start to finish.
One could choose when to move to the right,
i.e. h horizontal moves.**



In other words, we're asked to compute in how many ways one could choose p elements from $p + k$ elements. In mathematics, that's called [binomial coefficients](#)

$$C_{h+v}^h = C_{h+v}^v = \frac{(h+v)!}{h!v!}$$

The number of horizontal moves to do is $h = m - 1$, the number of vertical moves is $v = n - 1$. That results in a simple formula

$$C_{h+v}^h = \frac{(m+n-2)!}{(m-1)!(n-1)!}$$

The job is done. Now time complexity will depend on the algorithm to compute factorial function $(m + n - 2)!$. In short, standard computation for $k!$ using the definition requires $\mathcal{O}(k^2 \log k)$ time, and that will be not as good as DP algorithm.

[The best known algorithm to compute factorial function is done by Peter Borwein](#). The idea is to express the factorial as a product of prime powers, so that $k!$ can be computed in $\mathcal{O}(k(\log k \log \log k)^2)$ time. That's better than $\mathcal{O}(k^2)$ and hence beats DP algorithm.

The authors prefer not to discuss here various factorial function implementations, and hence provide Python3 solution only, with built-in [divide and conquer factorial algorithm](#). If you're interested in factorial algorithms, please check out good review on [this page](#).

Implementation

Code:

```
from math import factorial
```

```
class Solution:
```

```
    def uniquePaths(self, m: int, n: int) -> int:
```

```
return factorial(m + n - 2) // factorial(n - 1) // factorial(m - 1)
```

Complexity Analysis

- Time complexity: $\mathcal{O}((M + N)(\log(M + N) \log \log(M + N))^2)$.
- Space complexity: $\mathcal{O}(1)$.

Comment:

In Python 3.8, there's a new function that calculates n-choose-k directly:

```
def uniquePaths(self, m: int, n: int) -> int:  
    return math.comb(m + n - 2, n - 1)
```