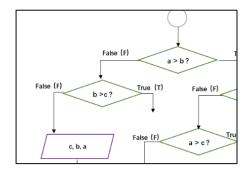
1. Flowchart

[10 points] Write a function Print_values with arguments a, b, and c to reflect the following flowchart. Here the purple parallelogram operator on a list [x, y, z] is to compute and print x+y-10z. Try your output with some random a, b, and c values. Report your output when a = 5, b = 15, c = 10.

Result:

当 a = 5, b = 15, c = 10, 不执行运算, 输出: none, 运行结果如图:



2. Continuous celing function

[10 points] Given a list with N positive integers. For every element x of the list, find the value of continuous ceiling function defined as F(x) = F(ceil(x/3)) + 2x, where F(1) = 1.

Result:

首先看到 ceil 考虑使用 math 模块,观察函数为递归,定义边界情况后实现即可(最简单思路,若提高效率可考虑使用哈希表等记忆函数值),运行结果如图:

3. Dice rolling

3.1 [15 points] Given 10 dice each with 6 faces, numbered from 1 to 6. Write a function Find_number_of_ways to find the number of ways to get sum x, defined as the sum of values

on each face when all the dice are thrown.

Result:

来自于 Leecode 1155 题 (https://leetcode.com/problems/number-of-dice-rolls-with-target-sum/description/),是原题的简版,即设定了 n=10、k=6,最合适算法无疑是 dynamic programming



这里尝试更容易理解的递归,即从后往前推当最后一个骰子的值为 1-6 时,前 9 个 骰子应该为多少,递归解决剩余问题。基准情况:如果没有骰子(n=0)且目标为 0,说明成功找到一种方法,返回 1,如果没有骰子但目标不为 0,或者目标变成负数,说明这种方法不可行,返回 0。(递归思路: 1.https://www.deepseek.com/zh)

```
定义递归函数: ways(n, target) 表示用 n 个骰子掷出总和 target 的方法数基准情况:

1. 如果 n == 0 且 target == 0: 返回 1 (成功)

2. 如果 n == 0 且 target != 0: 返回 0 (失败)

3. 如果 target < 0: 返回 0 (不可能)

递归关系:
最后一个骰子可能掷出 1 到 6:
ways(n, target) = ways(n-1, target-1) // 最后一个骰子为1
+ ways(n-1, target-2) // 最后一个骰子为2
+ ways(n-1, target-3) // 最后一个骰子为3
+ ways(n-1, target-4) // 最后一个骰子为4
+ ways(n-1, target-5) // 最后一个骰子为5
+ ways(n-1, target-6) // 最后一个骰子为6
```

传统递归会出现重复计算的问题,大量情况被重复计算浪费大量时间,考虑使用记忆化 ,所有值只计算一次。(记忆存储思路: B 站 up 主 flyingchow123)

Python 中的 functools.lru_cache 装饰器, functools.lru_cache 是 Python 标准库中 functools 模块的一部分。lru_cache 可以用来为一个函数添加一个缓存系统。它可以帮助 我们优化递归函数, 避免重复计算已经计算过的值。(@lru_cache 装饰器理解: https://zhuanlan.zhihu.com/p/640954732), 运行结果如图:

总和10的方法数: 1

总和15的方法数: 2002

总和30的方法数: 2930455 总和45的方法数: 831204

总和**60**的方法数: 1

3.2 [5 points] Count the number of ways for any x from 10 to 60, assign the number of ways to a list called Number_of_ways, so which x yields the maximum of Number_of_ways?

Result:

3.2 计算所有可能总和的方式数:

方式数列表长度: 51

产生最大方式数的x值: 35 最大方式数: 4395456

4. Dynamic programming

4.1 [5 points] Write a function Random_integer to fill an array of N elements by randomly selecting integers from 0 to 10.

Result:

使用 random 模块的 random.randint 生成随机整数,使用 append 添加到列表中。

```
import random

# 4.1

def Random_integer(N):
    result = []
    for i in range(N):
        num = random.randint(0, 10)
        result.append(num)
    return result

arr = Random_integer(5)
print(arr)

[4, 5, 9, 5, 10]
```

4.2 [15 points] Write a function Sum_averages to compute the sum of the average of all subsets of the array. For example, given an array of [1, 2, 3], you Sum_averages function should compute the sum of: average of [1], average of [2], average of [3], average of [1, 2], average of [1, 3], average of [2, 3], and average of [1, 2, 3].

Result:

动态规划思路描述: (来源: 1.https://chat.deepseek.com/ 2.) (1)状态定义:

dp sum[k]:存储所有长度为 k 的子集的和的总和

dp count[k]:存储所有长度为k的子集的个数

(2)初始化:

dp count[0] = 1 表示空集,这是动态规划的起点

(3)核心动态规划过程:

遍历数组中的每个数字,对于每个数字,从后往前更新状态(避免重复计算)

 $dp_sum[k] = dp_sum[k] + dp_sum[k-1] + num * dp_count[k-1]$:

dp sum[k]: 保持已有的长度为 k 的子集和

dp sum[k-1]: 之前长度为 k-1 的子集和

num * dp_count[k-1]: 当前数字添加到所有 k-1 长度子集中产生的新和

dp count[k] = dp count[k] + dp count[k-1]: 更新子集数量

(4)计算最终结果:

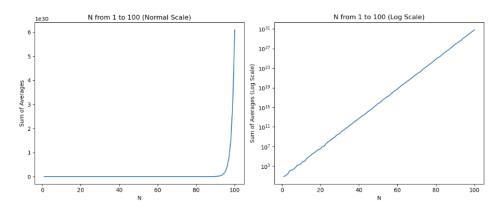
遍历所有可能的子集长度(1 到 n)对于每个长度 k,将 $dp_sum[k]/k$ 加到总和中这样就得到了所有子集平均值的总和。

数组 [1, 2, 3] 的结果: 14.0 数组 [1, 2, 3, 4, 5] 的结果: 93.0

4.3 [5 points] Call Sum_averages with N increasing from 1 to 100, assign the output to a list called Total_sum_averages. Plot Total_sum_averages, describe what do you see.

Result:

从图中可以观察到,随着数组长度N的增大,所有子集平均值之和(Sum of Averages) 呈现快速增长的指数趋势(左图为正常图,右图对y取了对数以便观察趋势),子集数量 的指数级增长主导了整个求和过程,即使单个子集的平均值很小,但数量庞大的子集使 得总和呈现爆炸式增长。



5. Path counting

5.1 [5 points] Create a matrix with N rows and M columns, fill the right-bottom corner and top-left corner cells with 1, and randomly fill the rest of matrix with integer 0 or 1.

Result:

使用 matrix = [[0 for _ in range(M)] for _ in range(N)]的写法来代替传统写法: [0 for _ in range(M)]: 创建一个长度为 M 的列表,每个元素都是 0 for _ in range(M): 循环 M 次,_是约定表示不关心循环变量值的变量名 0: 每次循环都放入数字 0 for _ in range(N): 外层循环 N 次 每次循环都创建一个新的长度为 M 的零列表 测试结果如下:

```
# Nid

matrix = create_matrix(3, 4)

for row in matrix:

    print(row)

[1, 0, 1, 0]

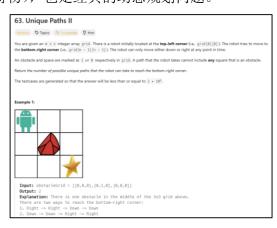
[0, 0, 1, 1]

[0, 1, 0, 1]
```

5.2 [25 points] Consider a cell marked with 0 as a blockage or dead-end, and a cell marked with 1 is good to go. Write a function Count_path to count total number of paths to reach the right-bottom corner cell from the top-left corner cell.

Notice: for a given cell, you are **only allowed** to move either rightward or downward.

来自于 Leecode 63 题(https://leetcode.com/problems/unique-paths-ii/),改动了有无障碍物的状态(0 代表障碍物),也是经典的动态规划问题。



(算法思路: 1.B 站 up 代码随想录 2. https://chat.deepseek.com/)

(1) 状态定义

创建 dp[i][j] 表示从起点 (0,0) 到达 (i,j) 的路径数量

(2) 初始化

起点 (0,0): 如果可通过,路径数为1;如果是障碍物,路径数为0

(3) 状态转移

对于每个单元格 (i,j):

如果是障碍物: dp[i][j] = 0 (无法到达)

如果不是障碍物: 路径数 = 从上面来的路径数 + 从左边来的路径数 dp[i][j] = dp[i-1][j] + dp[i][j-1]

(4) 边界处理

第一行 (i=0): 只能从左边来(不能从上面来)

第一列 (j=0): 只能从上面来(不能从左边来)

时间复杂度: O(N×M), 遍历整个网格一次 空间复杂度: O(N×M), 存储 DP 表格测试结果如下:

```
测试矩阵:
[1, 1, 0, 0, 1, 1]
[1, 0, 1, 0, 0, 0]
[1, 1, 1, 1, 0, 0]
[1, 1, 1, 1, 1, 1]
[0, 1, 1, 1, 1, 1]
[1, 0, 1, 1, 0, 1]
路径数量: 5
```

```
测试矩阵:
[1, 1, 0]
[1, 1, 1]
[1, 1, 1]
路径数量: 5
```

5.3 [5 points] Let N = 10, M = 8, run Count_path for 1000 times, each time the matrix (except the right-bottom corner and top-left corner cells, which remain being 1) is re-filled with integer 0 or 1 randomly, report the mean of total number of paths from the 1000 runs.

Result:

```
[57]: #5. Path counting
#5.3

def run_simulation():
    N = 10
    M = 8
    total_paths = 0
    runs = 1000

for i in range(runs):
    matrix = create_matrix(N, M)
    paths = Count_path(matrix)
    total_paths += paths

mean_paths = total_paths / runs
    return mean_paths

mean_result = run_simulation()
print(f*1000次运行的平均路径数: (mean_result)*)

1000次运行的平均路径数: 0.338
```

```
[58]: #5. Path counting
#5.3

def run_simulation():
    N = 10
    M = 38
    total_paths = 0
    runs = 1000

for i in range(runs):
    matrix = create_matrix(N, M)
    paths = Count_path(matrix)
    total_paths += paths

mean_paths = total_paths / runs
    return mean_paths

mean_result = run_simulation()
print(f*1000次运行的平均路径数: 0.426
```