



Minimum Coin Change

We'll cover the following

- Introduction
- Problem Statement
- Basic Solution
 - Code
- Top-down Dynamic Programming with Memoization
- Bottom-up Dynamic Programming
 - Code

Introduction

Given an infinite supply of 'n' coin denominations and a total money amount, we are asked to find the minimum number of coins needed to make up that amount.

Example 1:

Denominations: {1,2,3}

Total amount: 5

Output: 2

Explanation: We need minimum of two coins {2,3} to make a total of '5'

Example 2:

Denominations: {1,2,3}

Total amount: 11

Output: 4

Explanation: We need minimum four coins {2,3,3,3} to make a total of '11'

Problem Statement

Given a number array to represent different coin denominations and a total amount 'T', we need to find the minimum number of coins needed to make change for 'T'. We can assume an infinite supply of coins, therefore, each coin can be chosen multiple times.

This problem follows the Unbounded Knapsack

(https://www.educative.io/collection/page/5668639101419520/5633779737559040/574586549908 2752/) pattern.

Basic Solution #

A basic brute-force solution could be to try all combinations of the given coins to select the ones that give a total sum of 'T'. This is what our algorithm will look like:

```
1 for each coin 'c'
2    create a new set which includes one quantity of coin 'c' if it does not exceed 'T', and
3    recursively call to process all coins
4    create a new set without coin 'c', and recursively call to process the remaining coins
5    return the count of coins from the above two sets with a smaller number of coins
```

Code

Here is the code for the brute-force solution:

```
(§) JS
                        Python3
                                     ⊘ C++
👙 Java
 1 import math
                                                                                              Ψ,
 2
 3
 4
    def count_change(denominations, total):
 5
      result = count_change_recursive(denominations, total, 0)
       return -1 if result == math.inf else result
 6
 7
 8
 9 def count_change_recursive(denominations, total, currentIndex):
10
      # base check
11
      if total == 0:
12
        return 0
13
      n = len(denominations)
14
15
      if n == 0 or currentIndex >= n:
16
        return math.inf
17
      # recursive call after selecting the coin at the currentIndex
18
19
      # if the coin at currentIndex exceeds the total, we shouldn't process this
      count1 = math.inf
20
21
      if denominations[currentIndex] <= total:</pre>
22
         res = count_change_recursive(
          denominations, total - denominations[currentIndex], currentIndex)
23
        if res != math.inf:
24
25
           count1 = res + 1
26
27
      # recursive call after excluding the coin at the currentIndex
      count2 = count_change_recursive(denominations, total, currentIndex + 1)
28
29
      return min(count1, count2)
30
31
32
33 def main():
      print(count_change([1, 2, 3], 5))
34
      print(count_change([1, 2, 3], 11))
35
      print(count_change([1, 2, 3], 7))
36
37
      print(count_change([3, 5], 7))
38
39
40 main()
41
                                                                                              :3
\triangleright
```

The time complexity of the above algorithm is exponential $O(2^{C+T})$, where 'C' represents total coin denominations and 'T' is the total amount that we want to make change. The space complexity will be O(C+T).

Let's try to find a better solution.

Top-down Dynamic Programming with Memoization

We can use memoization to overcome the overlapping sub-problems. We will be using a twodimensional array to store the results of solved sub-problems. As mentioned above, we need to store results for every coin combination and for every possible sum:

```
Js JS
                        Python3
                                     G C++
👙 Java
 1 import math
                                                                                             Ψ,
 2
 3
 4
   def count_change(denominations, total):
 5
      dp = [[-1 for _ in range(total+1)] for _ in range(len(denominations))]
       result = count_change_recursive(dp, denominations, total, 0)
 6
 7
       return -1 if result == math.inf else result
 8
 9
10 def count_change_recursive(dp, denominations, total, currentIndex):
11
      # base check
12
      if total == 0:
13
        return 0
14
      n = len(denominations)
15
      if n == 0 or currentIndex >= n:
16
        return math.inf
17
      # check if we have not already processed a similar sub-problem
18
19
      if dp[currentIndex][total] == -1:
        # recursive call after selecting the coin at the currentIndex
20
21
        # if the coin at currentIndex exceeds the total, we shouldn't process this
        count1 = math.inf
22
         if denominations[currentIndex] <= total:</pre>
23
24
           res = count_change_recursive(
             dp, denominations, total - denominations[currentIndex], currentIndex)
25
26
           if res != math.inf:
27
             count1 = res + 1
28
29
        # recursive call after excluding the coin at the currentIndex
30
        count2 = count_change_recursive(
31
           dp, denominations, total, currentIndex + 1)
        dp[currentIndex][total] = min(count1, count2)
32
33
       return dp[currentIndex][total]
34
35
36
37
    def main():
      print(count_change([1, 2, 3], 5))
38
39
      print(count_change([1, 2, 3], 11))
40
      print(count_change([1, 2, 3], 7))
      print(count_change([3, 5], 7))
41
42
43
44 main()
```





Bottom-up Dynamic Programming #

Let's try to populate our array dp[TotalDenominations][Total+1] for every possible total with a minimum number of coins needed.

So for every possible total 't' ($0 \le t \le T$) and for every possible coin index ($0 \le t \le T$), we have two options:

- Exclude the coin: In this case, we will take the minimum coin count from the previous set
 => dp[index-1][t]
- Include the coin if its value is not more than 't': In this case, we will take the minimum count needed to get the remaining total, plus include '1' for the current coin => dp[index] [t-denominations[index]] + 1

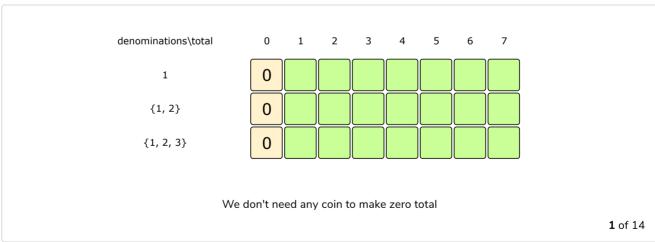
Finally, we will take the minimum of the above two values for our solution:

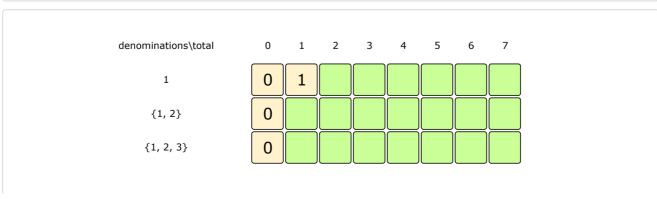
```
dp[index][t] = min(dp[index-1][t], dp[index][t-denominations[index]] + 1)
```

Let's draw this visually with the following example:

```
Denominations: [1, 2, 3]
Total: 7
```

Let's start with our base case of zero total:







denominations\total {1, 2} {1, 2, 3}

Total: 2, Index: 0 => dp[Index][Total-denominations[Index] + 1, we didn't consider dp[Index-1][Total] as Index = 0

of 14

Total:3-7, Index:0 => dp[Index][Total-denominations[Index] + 1, we didn't consider dp[Index-1][Total] as Index = 0

of 14

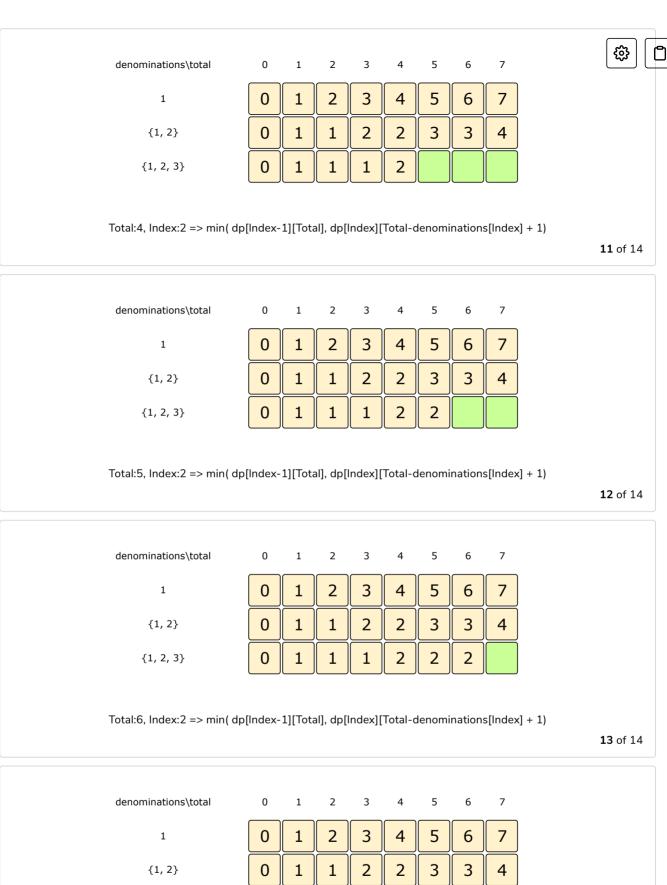
denominations\total {1, 2} {1, 2, 3}

Total:1, Index:1 => dp[Index-1][t], we didn't consider dp[Index][Total-denominations[Index] as Total < denominations[Index]

of 14

Total: 2, Index: 1 => min(dp[Index-1][Total], dp[Index][Total-denominations[Index] + 1)





— : · ·

14 of 14



Here is the code for our bottom-up dynamic programming approach:

```
(S) JS
                        G C++
👙 Java
 1
    import math
                                                                                               Ψ,
 2
 3
    def count_change(denominations, total):
 4
 5
       n = len(denominations)
       dp = [[math.inf for _ in range(total+1)] for _ in range(n)]
 6
 7
       # populate the total=0 columns, as we don't need any coin to make zero total
 8
 g
       for i in range(n):
         dp[i][0] = 0
10
11
       for i in range(n):
12
13
         for t in range(1, total+1):
           if i > 0:
14
15
             dp[i][t] = dp[i - 1][t] # exclude the coin
           if t >= denominations[i]:
16
17
             if dp[i][t - denominations[i]] != math.inf:
18
               # include the coin
19
               dp[i][t] = min(dp[i][t], dp[i][t - denominations[i]] + 1)
20
       # total combinations will be at the bottom-right corner.
21
       return -1 if dp[n - 1][total] == math.inf else <math>dp[n - 1][total]
22
23
24
25
    def main():
26
       print(count_change([1, 2, 3], 5))
27
       print(count_change([1, 2, 3], 11))
       print(count_change([1, 2, 3], 7))
28
29
       print(count_change([3, 5], 7))
30
31
32
    main()
33
                                                                                          \leftarrow
                                                                                                []
\triangleright
```

The above solution has time and space complexity of O(C * T), where 'C' represents total coin denominations and 'T' is the total amount that we want to make change.

