

## Documentation

The "Coin Change" problem is a fundamental challenge in computer science and optimization. The task involves determining the minimum number of coins needed to make up a specific amount using coins of given denominations. If it's impossible to form the required amount using the available coins, the result should be -1. The problem assumes an unlimited supply of each type of coin, making it a classic dynamic programming (DP) problem that aims to minimize the number of coins required.

The dynamic programming approach breaks the problem into smaller subproblems, solving each and building upon them to find the solution for the complete situation. To do this, a DP array is created where each index  $i$  represents the minimum number of coins required to make up the amount  $i$ . The key advantage of this approach is that it avoids redundant calculations by storing intermediate results, significantly reducing computational complexity compared to a naive brute-force approach.

The core idea behind the solution lies in the transition formula: for each amount  $i$ , we evaluate all possible coins and check whether using a particular coin reduces the total number of coins needed. If a coin's value is less than or equal to  $i$ , we calculate the minimum coins for  $i$  by comparing the current stored value with 1 plus the value needed for the remaining amount ( $i - \text{coin}$ ). This iterative process ensures that the DP array correctly reflects the minimum number of coins for all amounts up to the target.

The solution begins by initializing the DP array with a large value (infinity) for all indices except  $\text{dp}[0]$ , which is set to 0 because no coins are required to make an amount of 0. This initialization enables the algorithm to differentiate between achievable amounts and those that are not possible. If, after processing all coins, the value at  $\text{dp}[\text{amount}]$  remains infinity, it indicates that the amount cannot be formed, and the function returns -1.

Several edge cases are handled in the solution. For example, if the amount is 0, the result is 0 since no coins are needed. If the coins array contains only one denomination, the algorithm determines whether the target amount is divisible by that denomination. Additionally, cases where the target amount is very large or the coin denominations are sparse are seamlessly managed by the DP logic.

The computational complexity of the approach is efficient for the problem's constraints. The time complexity is  $O(\text{amount} \times \text{len}(\text{coins}))$ , as the algorithm iterates through each amount and evaluates all coins for each amount. The space complexity is  $O(\text{amount})$  since only a single array of size  $\text{amount} + 1$  is used to store results.

Beyond theoretical significance, the "Coin Change" problem has practical applications in various domains. It can be used to optimize resource allocation, such as managing currency exchanges, minimizing transaction fees, or solving logistical problems that involve combining units of different sizes. Moreover, the dynamic programming approach demonstrated here is a versatile tool that can be applied to numerous other optimization problems, making it an essential concept for software engineers, mathematicians, and data scientists alike.