

# **Documentation**

## **Problem Overview**

The Burst Balloons problem is a dynamic programming challenge that involves finding the optimal sequence of operations to maximize the number of coins collected. Each balloon is associated with a numerical value, and bursting a balloon yields coins equal to the product of its value and the values of its adjacent balloons. If no adjacent balloons exist (because they have already been burst), the multiplier defaults to 1. This problem is a classic example of optimization over permutations, requiring careful selection of subproblems to solve efficiently.

## **Key Challenges**

The primary difficulty in solving this problem lies in determining the optimal sequence of balloon bursts. Each decision impacts the values of adjacent balloons, making the problem non-trivial when approached through brute force. With  $n$  balloons, there are factorial ( $n!$ ) sequences to evaluate, which quickly becomes computationally expensive as  $n$  increases. Moreover, the decision of which balloon to burst last in any subrange is critical to simplify computations and ensure accuracy.

## **Dynamic Programming Approach**

Dynamic programming (DP) is used to break the problem into overlapping subproblems. By representing the problem in terms of ranges of balloons and storing the solutions to these subproblems, the DP approach avoids redundant calculations. This involves defining a table  $dp[left][right]$  that holds the maximum coins obtainable by bursting all balloons in the range  $[left, right]$ . The key insight is to treat each balloon in the range as the last one to burst, which simplifies the range into two independent subranges.

## **Array Transformation and Edge Handling**

To simplify calculations and handle edge cases, the input array is augmented with virtual balloons of value 1 at both ends. These virtual balloons ensure that calculations for edge balloons (e.g., the first or last in the original array) do not require special handling. This transformation allows the algorithm to operate uniformly over all balloons and ensures that the product calculation is always well-defined, even when the range of real balloons reduces to zero.

## **Transition and Recursion**

The transition formula in the DP solution calculates the maximum coins obtainable for a given range  $[left, right]$  by iterating over all possible choices of the last balloon to burst within the range. For a chosen balloon  $i$ , the total coins are computed as the sum of coins obtained from bursting  $i$  (using its adjacent balloons) and the coins from solving the subranges  $[left, i]$  and  $[i, right]$ . This recursive structure ensures that each subproblem is solved independently and optimally.

## **Complexity Analysis**

The DP solution has a time complexity of  $O(n^3)$ , where  $n$  is the length of the augmented array (original length plus two). This is due to three nested loops: one for the length of the subrange, one for the starting point of the range, and one for the choice of the last balloon to burst. The space complexity is  $O(n^2)$  to store the DP table. While the algorithm is computationally intensive for larger arrays, it is efficient enough for the given constraint of  $n \leq 300$ .

## **Applications and Insights**

The Burst Balloons problem highlights several key insights into optimization problems, particularly in dynamic programming. It demonstrates the importance of transforming a problem into manageable subproblems, the value of recursion and memoization, and how to effectively model a problem using data structures like DP tables. Beyond algorithmic problem-solving, this challenge can serve as a stepping stone for understanding real-world optimization scenarios, such as resource allocation, scheduling, or decision-making under constraints.