

# 1137. N-th Tribonacci Number

Easy   Memoization   Math   Dynamic Programming

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## Problem Description

The Tribonacci sequence is a variation of the Fibonacci sequence where each number is the sum of the three preceding ones. The sequence begins with  $T_0 = 0$ ,  $T_1 = 1$ , and  $T_2 = 1$ . For all integers  $n \geq 0$ , the  $n+3$ rd number of the sequence is calculated as  $T_{n+3} = T_n + T_{n+1} + T_{n+2}$ . The problem asks for the  $T_n$  value of the sequence given an integer  $n$ .

## Intuition

The intuitive way to calculate the  $n$ th Tribonacci number would be to use recursion or iteration, starting from the base cases  $T_0$ ,  $T_1$ , and  $T_2$ . However, this can be computationally expensive for larger  $n$  due to repeated calculations.

To optimize, one can use matrix exponentiation, realizing that each Tribonacci number can be obtained by multiplying a transformation matrix with the vector of the three previous Tribonacci numbers. The transformation matrix is:

```
1 [(1, 1, 0),
2  (1, 0, 1),
3  (1, 0, 0)]
```

By successively squaring this matrix and reducing the power, we apply an exponentiation by squaring technique which significantly reduces the number of multiplications needed. We only multiply the `res` vector by the transformation matrix when the current bit in the binary representation of  $n$  is set (using the bitwise AND operation  $n \& 1$ ). The operation  $n \gg= 1$  is used to right shift the bits of  $n$  by one position to check the next bit in the next iteration. The intuition behind this approach comes from a realization that  $n$ th power of the transformation matrix can be broken down into powers of 2, which allows the sequence to be computed in a time complexity that is logarithmic to  $n$  instead of linear, making it much more efficient for large values of  $n$ .

## Solution Approach

The solution is implemented using a class named `Solution` with a method `tribonacci` that takes an integer  $n$  as its parameter and returns the  $n$ th Tribonacci number.

To calculate the Tribonacci number efficiently, the solution applies matrix exponentiation. Here is a step-by-step explanation of the implementation:

- Base Cases:** For  $n == 0$ , return 0. For  $n < 3$  (which covers  $n == 1$  or  $n == 2$ ), return 1.
- Transformation Matrix:** The `factor` variable is defined as a  $3 \times 3$  transformation matrix using `numpy`:

```
1 factor = np.mat([(1, 1, 0), (1, 0, 1), (1, 0, 0)], np.dtype("0"))
```

Each multiplication of this matrix by a vector  $[T[n], T[n+1], T[n+2]]$  will result in  $[T[n+1], T[n+2], T[n+3]]$ .
- Result Vector:** The `res` variable is defined as a  $1 \times 3$  matrix to keep track of the current values of  $[T[n], T[n+1], T[n+2]]$ :

```
1 res = np.mat([[1, 1, 0]], np.dtype("0"))
```

Initially, it's set to represent  $[T[0], T[1], T[2]]$  (0, 1, 1) before the loop.
- Exponentiation by Squaring:** Before entering the loop,  $n$  is decremented by 3 as the initial `res` already accounts for the first three numbers. The loop then iterates, applying the following steps until  $n$  becomes 0:
  - Conditional Multiplication:** If the least significant bit of  $n$  is 1 ( $n \& 1$ ), `res` is multiplied by the `factor` matrix.
  - Matrix Squaring:** The `factor` matrix is squared which is equivalent to doubling the exponent.
  - Right Shift:** The bits of  $n$  are shifted to the right ( $n \gg= 1$ ) to process the next bit in the next iteration of the loop.The loop effectively multiplies `res` by `factor` raised to the power of  $n$ , but it does so in logarithmic time relative to the value of  $n$ .

- Final Sum:** After the loop ends, the sum of the `res` array will contain the value of  $T[n]$ , and this is returned as the final result:

```
1 return res.sum()
```

In terms of data structures, the solution utilizes `numpy` matrices for efficient manipulation and multiplication of large arrays, which is pivotal in matrix exponentiation. The use of bit manipulation to check individual bits of the integer  $n$  is an efficient pattern often used in algorithms involving repetitive multiplication, as it greatly reduces the number of calculations required.

## Example Walkthrough

Let's use  $n = 4$  as a small example to illustrate the solution approach.

- Initial Setup:** We are given the Tribonacci sequence with  $T_0 = 0$ ,  $T_1 = 1$ , and  $T_2 = 1$ .
- Base Cases Handling:** For  $n = 4$ , it's not less than 3, so we don't return the base cases directly.
- Transformation Matrix:** We have a transformation matrix `factor`, which when multiplied by  $[T_n, T_{n+1}, T_{n+2}]$ , gives  $[T_{n+1}, T_{n+2}, T_{n+3}]$ .
- Result Vector Initialization:** We define the initial `res` vector as  $[T_0, T_1, T_2]$ , which is  $[0, 1, 1]$ .
- Exponentiation by Squaring Process:** Since  $n = 4$  and `res` covers the first three numbers we deduct 3 from  $n$ . Now  $n$  is 1. The iteration would proceed as follows:
  - Start the loop with  $n = 1$ .
  - Since  $n \& 1$  is 1 ( $4 \& 1$  equals 1), we multiply our `res` vector  $[0, 1, 1]$  by our `factor` matrix to get  $[1, 1, 0]$ .
  - We square our `factor` matrix, which doesn't affect the `res` vector for this small value of  $n$ .
  - Right-shift  $n$  by 1 ( $n \gg= 1$ ). Now  $n$  is 0, which means our loop will terminate.
- Final Result:** The loop has ended, and `res` vector is  $[1, 1, 0]$ . The sum of `res` is  $T_4 = 1 + 1 + 0 = 2$ .

Therefore, the 4th Tribonacci number  $T_4$  is 2.

## Python Solution

```
1 import numpy as np
2
3 class Solution:
4     def tribonacci(self, n: int) -> int:
5         # Base case: return 0 for the zero-th element in the sequence.
6         if n == 0:
7             return 0
8
9         # Base cases: return 1 for the first or second element in the sequence.
10        if n < 3:
11            return 1
12
13        # Define the transition matrix for the Tribonacci sequence.
14        # This matrix represents the relationship between successive elements.
15        transition_matrix = np.matrix([(1, 1, 0), [1, 0, 1], [1, 0, 0]], np.dtype("0"))
16
17        # Initialize the result matrix as the identity matrix aligned with the sequence order.
18        result_matrix = np.matrix([(1, 1, 0)], np.dtype("0"))
19        n -= 3
20
21        # Use the Fast Exponentiation algorithm to raise the matrix to the power of (n-3).
22        while n:
23            # When the current power is odd, multiply result_matrix with transition_matrix.
24            if n & 1:
25                result_matrix *= transition_matrix
26
27            # Square the transition_matrix to get the next higher power of 2.
28            transition_matrix *= transition_matrix
29
30            # Shift right to divide n by 2, flooring it.
31            n >>= 1
32
33        # Return the sum of the elements in the resulting matrix as the answer.
34        return result_matrix.sum()
35
```

## Java Solution

```
1 class Solution {
2     // Calculates the n-th Tribonacci number
3     public int tribonacci(int n) {
4         // Base cases for n = 0, 1, 2
5         if (n == 0) {
6             return 0;
7         }
8         if (n < 3) {
9             return 1;
10        }
11        // Transformation matrix for Tribonacci sequence
12        int[][] transformationMatrix = {{1, 1, 0}, {1, 0, 1}, {1, 0, 0}};
13        // Calculate the power of the matrix to (n-3), since we know the first three values
14        int[][] resultingMatrix = matrixPower(transformationMatrix, n - 3);
15        // Initialize answer
16        int answer = 0;
17        // Add the top row of the matrix to get the answer
18        for (int element : resultingMatrix[0]) {
19            answer += element;
20        }
21        return answer;
22    }
23
24    // Multiplies two matrices and returns the result
25    private int[][] matrixMultiply(int[][] a, int[][] b) {
26        int rows = a.length, cols = b[0].length;
27        int[][] resultMatrix = new int[rows][cols];
28        // Perform matrix multiplication
29        for (int i = 0; i < rows; ++i) {
30            for (int j = 0; j < cols; ++j) {
31                for (int k = 0; k < b.length; ++k) {
32                    resultMatrix[i][j] += a[i][k] * b[k][j];
33                }
34            }
35        }
36        return resultMatrix;
37    }
38
39    // Calculates the matrix exponentiation of matrix 'a' raised to the power of 'n'
40    private int[][] matrixPower(int[][] a, int n) {
41        // Create an identity matrix for initial result
42        int[][] result = {{1, 1, 0}};
43        // Loop to do binary exponentiation
44        while (n > 0) {
45            // Multiply with 'a' when the least significant bit is 1
46            if ((n & 1) == 1) {
47                result = matrixMultiply(result, a);
48            }
49            // Square the matrix 'a'
50            a = matrixMultiply(a, a);
51            // Right shift 'n' to process the next bit
52            n >>= 1;
53        }
54        return result;
55    }
56 }
57
```

## C++ Solution

```
1 #include <vector>
2 #include <numeric>
3
4 class Solution {
5 public:
6     // Computes the n-th Tribonacci number using matrix exponentiation
7     int tribonacci(int n) {
8         // Base case when n is 0
9         if (n == 0) {
10            return 0;
11        }
12        // Base cases for n being 1 or 2
13        if (n < 3) {
14            return 1;
15        }
16        // Initial transformation matrix
17        std::vector<std::vector<long long>> transformationMatrix = {
18            {1, 1, 0},
19            {1, 0, 1},
20            {1, 0, 0}
21        };
22        // Raise the transformation matrix to the power (n - 3)
23        std::vector<std::vector<long long>> resultMatrix = matrixPower(transformationMatrix, n - 3);
24        // The sum of the first row of the result matrix gives us the nth Tribonacci number
25        return std::accumulate(resultMatrix[0].begin(), resultMatrix[0].end(), 0);
26    }
27
28    private:
29        using ll = long long; // Alias for long long type for easier reading
30
31        // Multiplies two matrices and returns the result
32        std::vector<std::vector<ll>> matrixMultiply(std::vector<std::vector<ll>>& a, std::vector<std::vector<ll>>& b) {
33            int rows = a.size(), columns = b[0].size();
34            std::vector<std::vector<ll>> product(rows, std::vector<ll>(columns));
35
36            for (int i = 0; i < rows; ++i) {
37                for (int j = 0; j < columns; ++j) {
38                    for (int k = 0; k < b.size(); ++k) {
39                        product[i][j] += a[i][k] * b[k][j];
40                    }
41                }
42            }
43            return product;
44        }
45
46        // Calculates the power of a matrix to the given exponent n
47        std::vector<std::vector<ll>> matrixPower(std::vector<std::vector<ll>>& a, int n) {
48            // Starting with the identity matrix of size 3x3
49            std::vector<std::vector<ll>> result = {
50                {1, 0, 0},
51                {0, 1, 0},
52                {0, 0, 1}
53            };
54
55            while (n) {
56                if (n & 1) { // If n is odd, multiply the result by the current matrix a
57                    result = matrixMultiply(result, a);
58                }
59                a = matrixMultiply(a, a); // Square the matrix a
60                n >>= 1; // Divide n by 2
61            }
62            return result;
63        }
64    };
65 }
```

## Typescript Solution

```
1 // Calculates the n-th Tribonacci number using matrix exponentiation.
2 function tribonacci(n: number): number {
3     // Base cases for n = 0, 1, and 2.
4     if (n === 0) {
5         return 0;
6     }
7     if (n < 3) {
8         return 1;
9     }
10
11    // Matrix representation of the Tribonacci relation.
12    const tribonacciMatrix = [
13        [1, 1, 0],
14        [1, 0, 1],
15        [1, 0, 0],
16    ];
17
18    // Raise the matrix to the (n-3)-th power and sum the top row for the result.
19    return matrixPower(tribonacciMatrix, n - 3)[0].reduce((sum, element) => sum + element);
20 }
21
22 // Multiplies two matrices and returns the result.
23 function multiplyMatrices(matrix1: number[][], matrix2: number[][]): number[][] {
24     const rows = matrix1.length;
25     const cols = matrix2[0].length;
26     // Create an empty matrix for the result.
27     const result = Array.from({ length: rows }, () => Array.from({ length: cols }, () => 0));
28
29     // Calculate the product of the two matrices.
30     for (let i = 0; i < rows; ++i) {
31         for (let j = 0; j < cols; ++j) {
32             for (let k = 0; k < matrix2.length; ++k) {
33                 result[i][j] += matrix1[i][k] * matrix2[k][j];
34             }
35         }
36     }
37
38     return result;
39 }
40
41 // Raises a matrix to the power of n and returns the result.
42 function matrixPower(matrix: number[][], n: number): number[][] {
43     // Initialize the result as an identity matrix in the shape of a 1 x 3 matrix.
44     let result = [
45         [1, 1, 0],
46     ];
47
48     while (n > 0) {
49         // If the exponent is odd, multiply the result by the matrix.
50         if (n & 1) {
51             result = multiplyMatrices(result, matrix);
52         }
53         // Square the matrix.
54         matrix = multiplyMatrices(matrix, matrix);
55         // Halve the exponent by right-shifting it.
56         n >>= 1;
57     }
58
59     return result;
60 }
61
```

## Time and Space Complexity

The given Python code implements a fast matrix exponentiation approach to find the  $n$ -th number in the Tribonacci sequence. The Tribonacci sequence is a generalization of the Fibonacci sequence, where each number is the sum of the preceding three numbers in the sequence.

### Time Complexity

The time complexity of this solution mainly depends on the matrix exponentiation part of the algorithm. The algorithm uses a loop that runs in  $O(\log n)$  time, which accounts for the repeated squaring of the matrix `factor`. In each iteration of the loop, the matrix is either multiplied by itself (squaring) which takes  $O(1)$  time since the matrix size is fixed at  $3 \times 3$ , or the result matrix `res` is multiplied by `factor`, which also takes  $O(1)$  time for the same reason. Thus, the overall time complexity of the algorithm is  $O(\log n)$ .

### Space Complexity

The space complexity of the algorithm is determined by the size of the matrices used for the calculations. The matrices `factor` and `res` have a constant size of  $3 \times 3$  and  $1 \times 3$ , respectively, so the space used does not depend on  $n$ . Consequently, the space complexity is  $O(1)$ , which means it uses constant space.