

# fibonacci\_report

January 31, 2026

## 1 Fibonacci Matrix Exponentiation

### 1.1 Divide and Conquer Algorithm

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**Course:** Análisis y Diseño de Algoritmos

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### 1.2 Abstract

This report presents an efficient algorithm for computing Fibonacci numbers using matrix exponentiation with divide and conquer. The method achieves  $O(\log n)$  time complexity.

### 1.3 1. Mathematical Foundation

The Fibonacci sequence can be expressed using matrix exponentiation:

$$\begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}^n = \begin{bmatrix} F_{n+1} & F_n \\ F_n & F_{n-1} \end{bmatrix}$$

Therefore:  $\mathbf{F}(n) = \mathbf{M}^n[0,1]$

#### 1.3.1 Divide and Conquer Strategy

$$M^n = \begin{cases} M & \text{if } n = 1 \\ (M^{n/2})^2 & \text{if } n \text{ is even} \\ (M^{n/2})^2 \times M & \text{if } n \text{ is odd} \end{cases}$$

```
[ ]: import numpy as np

M = np.array([[1, 1], [1, 0]], dtype=np.int64)
print("Base matrix M:")
print(M)
```

Base matrix M:

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

## 1.4 2. Algorithm Implementation

```
[ ]: def power_matrix(matrix, n):  
    """Recursively compute matrix power using divide and conquer.  
  
    Args:  
        matrix: 2x2 numpy array  
        n: Positive integer exponent  
  
    Returns:  
        matrix^n as numpy array  
    """  
    if n == 1:  
        return matrix  
  
    half = power_matrix(matrix, n // 2)  
    half = np.dot(half, half)  
    return half if n % 2 == 0 else np.dot(half, matrix)  
  
def fibonacci_matrix(n):  
    """Calculate nth Fibonacci number using recursive matrix exponentiation.  
  
    Args:  
        n: Positive integer position in Fibonacci sequence  
  
    Returns:  
        The nth Fibonacci number  
  
    Raises:  
        ValueError: If n is not positive  
    """  
    if n <= 0:  
        raise ValueError("n must be positive")  
  
    base_matrix = np.array([[1, 1], [1, 0]], dtype=np.int64)  
    return power_matrix(base_matrix, n)[0, 1]
```

## 1.5 3. Example Execution

For  $n=21$  (binary: 10101):

$$\begin{aligned}M^{21} &= (M^{10})^2 \times M \\M^{10} &= (M^5)^2 \\M^5 &= (M^2)^2 \times M \\M^2 &= (M^1)^2 \\M^1 &= M\end{aligned}$$

Only 5 recursive calls

```
[ ]: print("Test Results:")
      print("-" * 30)
      for n in [5, 10, 21, 50]:
          print(f"F({n}) = {fibonacci_matrix(n)}")
```

Test Results:

```
-----
F(5) = 5
F(10) = 55
F(21) = 10946
F(50) = 12586269025
```

## 1.6 4. Complexity Analysis

### 1.6.1 Time Complexity

Method	Complexity
Naive Recursion	$O(2^n)$
Iterative	$O(n)$
<b>Matrix (Divide &amp; Conquer)</b>	<b><math>O(\log n)</math></b>

### 1.6.2 Recurrence Relation

$$T(n) = T(n/2) + O(1)$$

**By Master Theorem:**  $T(n) = O(\log n)$

## 1.7 5. Conclusions

1. Matrix exponentiation achieves  **$O(\log n)$**  time complexity
2. Divide and conquer reduces problem size by half at each step
3. Efficient for computing large Fibonacci numbers
4. Technique extends to other linear recurrence relations