

1. Time Complexity

Time complexity is the measure of the number of operations an algorithm performs as a function of input size n .

Detailed Steps to Analyze Time Complexity

1. Break Down the Code:

- Divide your program into individual operations (loops, function calls, conditions, etc.)
- Assign complexity to each operation.

2. Consider Iterations of Loops:

- Single loop over n elements $\rightarrow O(n)$
- Nested loops \rightarrow Multiply complexities of each loop.

3. Analyze Recursion:

- Determine the number of recursive calls.
- Derive the recurrence relation and solve it (e.g., using the Master Theorem or substitution).

4. Drop Constants and Non-Dominant Terms:

- Focus on the fastest-growing term as n increases.
 - Example: If complexity is $5n^2 + 3n + 7$, simplify to $O(n^2)$.
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Common Scenarios and Their Complexities

Operation	Time Complexity
Accessing an array element	$O(1)$
Traversing a list with a loop	$O(n)$
Searching in a sorted array (binary search)	$O(\log n)$
Nested loops (double traversal)	$O(n^2)$
Sorting (merge sort, quick sort)	$O(n \log n)$
Recursive Fibonacci calculation	$O(2^n)$

Examples

Example 1: Single Loop

```
def sum_of_elements(arr):
```

```
    total = 0
```

```
    for num in arr:
```

```
        total += num
```

```
    return total
```

- **Steps to analyze:**
 - for num in arr → Loop runs n times (where n is the size of arr).
 - Inside loop: total += num is $O(1)$.
 - **Total Complexity:** $O(n)$
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Example 2: Nested Loop

```
def find_duplicates(arr):
```

```
    duplicates = []
```

```
    for i in range(len(arr)):
```

```
        for j in range(i + 1, len(arr)):
```

```
            if arr[i] == arr[j]:
```

```
                duplicates.append(arr[i])
```

```
    return duplicates
```

- **Steps to analyze:**
 - Outer loop runs n times.
 - Inner loop runs $n-1, n-2, \dots, 1$ times.
 - Total iterations: $n(n-1)/2 = O(n^2)$.
 - **Total Complexity:** $O(n^2)$
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Example 3: Recursion

```
def factorial(n):
```

```
    if n == 0:
```

```
        return 1
```

```
    return n * factorial(n - 1)
```

- **Steps to analyze:**
 - Function calls itself n times until $n = 0$.
 - Each call takes $O(1)$.
 - **Total Complexity:** $O(n)$
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Example 4: Divide-and-Conquer

```
def merge_sort(arr):
```

```
    if len(arr) <= 1:
```

```
        return arr
```

```
    mid = len(arr) // 2
```

```
    left = merge_sort(arr[:mid])
```

```
    right = merge_sort(arr[mid:])
```

```
    return merge(left, right)
```

- **Steps to analyze:**
 - Divide array into two halves $\rightarrow O(\log n)$ levels of recursion.
 - Merging arrays $\rightarrow O(n)$ work at each level.
 - **Total Complexity:** $O(n \log n)$
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Advanced Time Complexity Tools

1. Recurrence Relations (for recursive functions):

- Example: $T(n) = 2T(n/2) + O(n)$
 - This simplifies to $O(n \log n)$ using the Master Theorem.

2. Master Theorem:

- For recurrence relations of the form $T(n) = aT(n/b) + O(n^d)$:

- a: Number of subproblems.
 - b: Factor by which input size is reduced.
 - d: Complexity of the division/merge step.
 - Use $\log_b(a)$ to compare with d.
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2. Space Complexity

Space complexity measures the memory consumed by your program.

Detailed Steps to Analyze Space Complexity

1. Account for Variables and Data Structures:

- Scalar variables (e.g., x, count) $\rightarrow O(1)$.
- Arrays/lists $\rightarrow O(n)$, where n is the size of the list.
- Nested data structures (e.g., lists of lists) \rightarrow Multiply dimensions.

2. Account for Recursion:

- Recursive calls take stack space.
- Depth of recursion determines space complexity.

3. Separate Input vs. Auxiliary Space:

- Input space is usually not included in the calculation.
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Examples

Example 1: Iterative Function

```
def sum_array(arr):
```

```
    total = 0
```

```
    for num in arr:
```

```
        total += num
```

```
    return total
```

- **Space Complexity:**

- Variables: total $\rightarrow O(1)$
- Input: arr is not counted.

- Total: $O(1)$
-

Example 2: Recursive Function

```
def factorial(n):
```

```
    if n == 0:
```

```
        return 1
```

```
    return n * factorial(n - 1)
```

- **Space Complexity:**
 - **Base case:** When $n=0$, the recursion stops.
 - **Recursive depth:** The function keeps calling itself until n reaches 0.
 - For $n=5$:
 - Recursive calls: `factorial(5), factorial(4), factorial(3), factorial(2), factorial(1)`
 - Depth: 5.
 - **Space Complexity of Depth:** $O(n)$, because the call stack grows linearly with n .
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Example 3: Data Structures

```
def reverse_array(arr):
```

```
    return arr[::-1]
```

- **Space Complexity:**
 - Input: `arr` is not counted.
 - Output: New array of size $n \rightarrow O(n)$.
 - Total: $O(n)$
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Summary of Space Complexity

Scenario	Space Complexity
Single scalar variable	$O(1)$
Single loop (no additional array)	$O(1)$
Recursive function	$O(\text{depth})$
Additional list of size n	$O(n)$
