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What Is Big O Notation

Big O notation is a mathematical notation used in computer science to describe the upper bound or worst-case behavior of an algorithm or function. It represents the maximum growth rate of the algorithm's time complexity or space complexity as the input size approaches infinity.

- In Big O notation, we use the symbol "O" followed by a function to express the upper bound of the algorithm's complexity.
- The function typically represents the number of operations performed by the algorithm or the amount of space required.

Some Commonly Used Big O Notations:

1. $O(1)$ - Constant Time:

The algorithm's running time or space requirements remain constant regardless of the input

size.

Example: Accessing an element in an array by index. It takes the same amount of time regardless of the size of the array.

2. $O(\log n)$ – Logarithmic Time:

The algorithm's running time grows logarithmically with the input size.

Example: Binary search on a sorted array. At each step, the algorithm eliminates half of the remaining elements, reducing the search space logarithmically.

3. $O(n)$ – Linear Time:

The algorithm's running time increases linearly with the input size.

Example: Searching for an element in an unsorted array. In the worst case, the algorithm may need to traverse the entire array to find the element.

4. $O(n \log n)$ – Linearithmic Time:

The algorithm's running time grows in a rate that is proportional to n multiplied by the logarithm of n .

Example: Merge sort. It divides the input array into smaller halves recursively and merges them in a sorted order. The time complexity grows in $n \log n$ as each division takes $O(\log n)$ time, and the merging step takes $O(n)$ time.

5. $O(n^2)$ – Quadratic Time:

The algorithm's running time grows quadratically with the input size. It is commonly associated with nested loops or algorithms that involve comparing every element with every other element.

Example: Selection sort. It repeatedly finds the minimum element and swaps it with the current position. The algorithm requires nested loops, resulting in a quadratic time complexity.

6. $O(2^n)$ – Exponential Time:

The algorithm's running time grows exponentially with the input size.

Example: Generating all subsets of a set. As the size of the set grows, the number of subsets doubles, leading to an exponential increase in time complexity.

Analyze The Time Complexity Of The Algorithm, Using Big O Notation.

Example 1:

Let n represent the number of elements in the array.

```
#include <stdio.h>

int find_max(int arr[], int length) {
    int max_value = arr[0]; // Assume the first element is the
    maximum

    for (int i = 1; i < length; i++) {
        if (arr[i] > max_value) {
            max_value = arr[i];
        }
    }

    return max_value;
}

int main() {
    int arr[] = {5, 8, 2, 10, 3};
    int length = sizeof(arr) / sizeof(arr[0]);

    int max = find_max(arr, length);
    printf("Maximum value: %d\n", max);

    return 0;
}
```

1. Initializing `max_value` with `arr[0]` takes constant time and can be considered $O(1)$.
2. The for loop iterates through the array from index 1 to `length - 1`, where `length` is the length of the array. The loop runs `length - 1` times.
3. Within the loop, the comparison `if (arr[i] > max_value)` and the subsequent assignment `max_value = arr[i]` both take constant time and can be considered $O(1)$.
4. The return statement also takes constant time and can be considered $O(1)$.

Thus, find_max's time complexity is:

- The initialization step takes $O(1)$.
- The for loop runs length - 1 times, so it has a time complexity of $O(\text{length})$.
- The remaining constant-time operations also take $O(1)$.

As a result, find_max's time complexity is $O(\text{length})$, where length is the array's length.

Example 2:



```
#include <stdio.h>

int main() {
    int i;
    for (i = 1; i <= 10; i++) {
        printf("%d\n", i);
    }
    return 0;
}
```

The loop runs for a fixed number of iterations, specifically from 1 to 10. Since the loop does not depend on any variable or input size, the time complexity is constant.

Therefore, the time complexity of this code snippet is $O(1)$.

Example 3:



```
#include <stdio.h>

int main() {
    int i,n;
    printf("Enter a number");
    scanf("%d",&n);
    for (i = 1; i <= n; i++) {
        printf("%d\n", i);
    }
    return 0;
}
```

1. The scanf function for getting user input takes constant time and can be considered $O(1)$.
2. The for loop iterates from 1 to n, where n represents the user input. The loop runs n times.
3. Inside the loop, the printf function prints the value of i. The printf function takes constant time as it performs a fixed number of operations. Thus, it can be considered $O(1)$.

Therefore, the time complexity of the for loop is $O(n)$ since the loop iterates n times.

