

Lab 2

1. Big O Notation

1. Let
$$f(n) = 2n + 6$$
. Is $f(n) = O(n)$

We say f(n) = O(n) if there exist constants c > 0 and $n_0 \ge 0$ such that:

$$f(n) \le c \cdot n$$
 for all $n \ge n_0$

Given:

$$f(n) = 2n + 6$$

We want:

$$2n + 6 \le c \cdot n \Rightarrow 6 \le (c - 2)n \Rightarrow n \ge \frac{6}{c - 2}$$

Choose c = 3:

$$n \ge \frac{6}{1} = 6$$

So,

$$f(n) = 2n + 6 = O(n)$$

2. Let
$$f(n) = 3n^2 + 4n - 8$$
. Is $f(n) = O(n^2)$

We want:

$$3n^2+4n-8 \leq c \cdot n^2 \Rightarrow 4n-8 \leq (c-3)n^2$$

Try c = 5:

$$4n - 8 < 2n^2 \Rightarrow 0 < 2n^2 - 4n + 8$$

This is always true for $n \ge 1$.

So,

$$f(n) = 3n^2 + 4n - 8 = O(n^2)$$



3. Prove or disprove that log(n!) is O(nlog n)

We use the fact that:

$$4^4 = 4 \times 4 \times 4 \times 4 = 256$$

$$4! = 1 \times 2 \times 3 \times 4 = 24$$

Prove

$$f(n) \le c \cdot g(n)$$

$$log(1 \times 2 \times 3 \dots) \le log(n \times n \times n \dots)$$

So:

$$\log(n!) \le n \log n$$

This shows:

$$\log(n!) = O(n \log n)$$

4. Prove that 2n + 3 is $O(n^2)$

We want to prove:

$$2n + 3 \le c \cdot n^2$$
 for large n

Divide both sides:

$$\frac{2n+3}{n^2} \le c \Rightarrow \frac{2}{n} + \frac{3}{n^2} \le c$$

As $n \to \infty$, both terms on the left go to 0.

So there exists some c (e.g., c = 1) such that the inequality holds for $n \ge 3$.

So,

$$2n + 3 = O(n^2)$$

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5. Prove that 2^{n+2} is $O(2^n)$

$$f(n) = 2^{n+2} = 4 \cdot 2^n \Rightarrow f(n) = 4 \cdot 2^n$$

We want:

$$f(n) \le c \cdot 2^n \Rightarrow 4 \cdot 2^n \le c \cdot 2^n \Rightarrow 4 \le c$$

Let c=4, inequality holds for all $n\geq 0$

So,

$$2^{n+2} = O(2^n)$$

2. Write the recurrence relation of the following algorithms

1.

$$T(n) = T(n-1) + 4$$
$$T(0) = 1$$

$$T(n) = T(n - 1) + 4$$
, $T(0) = 1$

This is a **linear recurrence relation** with a constant added at each step.

$$T(n) = 4n + 1$$

2.

$$T(n) = T(n-1) + 8$$
$$T(1) = 8$$

$$T(n) = T(n-1) + 8$$
, $T(1) = 8$

This is also linear, similar to the first one but with a different base and increment.

Unrolling the recurrence:

$$T(n) = 8(n-1) + 8 = 8n$$

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3.

$$T(n) = 4T(\frac{n}{2}) + n$$

$$T(n) = 4T\left(\frac{n}{2}\right) + n$$

This is a divide and conquer recurrence.

Use the **Master Theorem**:

$$-a = 4, b = 2, d = 1, f(n) = n$$

- Compare
$$f(n) = n$$
 to $n^{\log_b a} = n^{\log_2 4} = n^2$ and $n^d = n^1 = n$

So,
$$n^2 > n^1$$

Solution:

$$T(n) = O(n^2)$$

4.

$$T(n) = 4T\left(\frac{n}{2}\right) + n^2$$

Use Master Theorem:

$$-(a = 4), (b = 2), (d = 2), (f(n) = n^2)$$

$$-(n^{\log_b a}=n^2), and (n^2)$$

So,
$$n^2 = n^2$$

Solution:

$$T(n) = O(n^2 log_2 n)$$

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5.

$$T(n) = 2T\left(\frac{n}{8}\right) + \sqrt[3]{n}$$

Use Master Theorem:

$$-(a=2), (b=8), \left(d=\frac{1}{3}\right), (f(n)=n^{1/3})$$
 - Compare with $(n^{\log_b a}=n^{\log_8 2}=n^{1/3})$ So, $n^{1/3}=n^{1/3}$

Solution:

$$T(n) = O\left(n^{\frac{1}{3}}log_8n\right) = O\left(\sqrt[3]{n} \cdot log_8n\right)$$

```
6.
int factorial(int n){
   if (n==1)
      return 1;
   return n * factorial(n-1);
}
```



Find time complexity of the following code

```
1.
void foo(int n){
    int i = 1; int s = 1;
    while (s <= n){
        i++;
        s = s + i;
        System.out.print("*");
    }
}</pre>
```

Analysis:

This loop continues while:

$$s = 1 + 2 + 3 + \dots + i \le n \Rightarrow \frac{i(i+1)}{2} \le n \Rightarrow i = O(\sqrt{n})$$

Time Complexity:

 $O(\sqrt{n})$

```
2.
void foo(int n){
    int count = 0;
    for (int i = n / 2; i <= n ; i++) {
        for(int j = 1; j + (n/2) <= n; j++){
            for(int k = 1; k <= n; k*=2){
                 count++;
            }
        }
    }
}</pre>
```

Analysis:

- Outer loop: runs from n/2 to $n \to O(n)$
- Middle loop: $j + n/2 \le n \Rightarrow j \le n/2 \rightarrow O(n)$
- Inner loop: $k *= 2 \rightarrow O(\log n)$



Total:

Analysis:

- Outer loop: runs from n/2 to $n \to O(n)$
- Middle loop: O(logn)
- Inner loop: $O(\log n)$

Total:

$$O(n \log n \log n) = \boxed{O(n \log^2 n)}$$



```
4.
void foo(int n){
    if(n == 1) return;
    for (int i = 1; i <= n; i++) {
        for(int j = 1; j <= n; j++){
            System.out.print("*");
            break;
        }
    }
}</pre>
```

Analysis:

- If condition: constant time O(1)
- Outer loop: O(n)
- Inner loop breaks after 1st iteration → constant

Total:

$$O(n) \cdot O(1) = \boxed{O(n)}$$

```
5.
void foo(int n){
   int a = 0; int i = n;
   while (i > 0){
        a += i;
        i /= 2;
   }
}
```

Analysis:

Loop halves i each time: n, n/2, n/4, ...

Number of iterations: $O(\log n)$

Time Complexity:

 $O(\log n)$



```
6.
void foo(int n){
    int count = 0;
    for(int i = n; i > 0; i /= 2){
        for(int j = 0; j < i; j++){
            count++;
        }
    }
}</pre>
```

Analysis:

Outer loop: $i = n, n/2, n/4, ... \rightarrow O(\log n)$

Inner loop:

• First: *n*

• Second: n/2

• Third: n/4

• ...

Total work: geometric sum

$$n + n/2 + n/4 + \dots = n(1 + 1/2 + 1/4 + \dots) = O(n)$$

Time Complexity:

O(n)