

Introduction to algorithms

What is Algorithm?

Sequence of finite steps used to solve the particular problem.

Properties of Algorithms:

- It should terminate after finite time
- It should produce at-least one output
- It should be unambiguous [DETERMINISTIC]
 - For same input same output will always come

Types of Analysis:

Aposteriori Analysis	Apriori Analysis
Dependent on Language of compiler and type of hardware used	Independent of Language of compiler and hardware.
Exact Answers	Approximate answers

Few problems:

```
main(){
    i = 1
    while i < n{

        i = 2 * i
    }
}
```

$$TimeComplexity = \log_2(n)$$

```

main(){
    i = n
    while i > 1{
        i = i / 2
        i = i / 5
        i = 2 * i
    }
}

```

$$i/5^k = 1n/5^k = 1$$

k is number of times my loop is running.

$$Timecomplexity = \log_5(n)$$

```

main(){
    i = n
    while i > 2 {
        i = sqrt(i) }
}

```

$$i = n = n^{(1/2)} = n^{(1/2)}(1/2)$$

$$n^{(1/2^k)} = 2$$

$$\log_2(n^{(1/2^k)}) = \log_2(2)$$

$$(1/2^k) \log_2(n) = 1$$

$$\log_2(n) = 2^k$$

$$\log_2(\log_2(n)) = k \log_2(2)$$

$$Timecomplexity = O(\log_2(\log_2(n)))$$

```

main(){
    i = n
    while i > 2 {
        i = (i)**(1/25) }
}

```

$$TimeComplexity = O(\log_{25}(\log_2(n)))$$

+/-	$O(n)$
* or /	$\log_k(n)$

Problem 7

```
main() {
  for(i=1; i<=n; i=i+2){
    for(j=n; j>1; j=j/7){
      for(k=19; k<=n; k=k*(17)) {
        x = y + z;
      }
    }
  }
}
```

Handwritten analysis:

- $+/- \Rightarrow \frac{n}{k} = O(n)$
- $* \text{ or } / \Rightarrow \log_k n$
- For the innermost loop: $k = 19$, $\text{while}(k \leq n)$, $k = k^{17}$. This results in $\log_{17}(\log_{19} n)$.
- For the middle loop: $j = n$, $j = j/7$. This results in $\log_7 n$.
- For the outermost loop: $i = 1$, $i = i+2$. This results in $\frac{n}{2}$.
- Overall time complexity: $O(n \cdot \log_7 n \cdot \log_{17}(\log_{19} n))$.
- Time Complexity of above program -

Asymptotic Notations:

Big O Notation

Let $f(n)$ and $g(n)$ be two positive functions. Then,

$$f(n) = O(g(n))$$

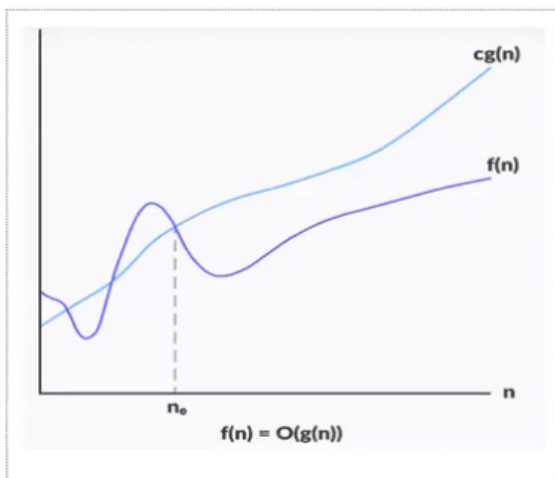
if $f(n) \leq c \cdot g(n)$; for all n , $n \geq n_0$ such that there exists two positive constants.

$$c > 0, n_0 \geq 1$$

For example : $a = O(b)$

Meaning : b is somehow greater than a after taking c help.

Graphical Representation of Big O



- $F(n) = O(g(n))$
- If there exists a positive constant c such that $c \geq 0$, for sufficiently large value of n .
- For any value of n , the running time of an algorithm does not cross the time provided by $O(g(n))$

Omega Notation

Let $f(n)$ and $g(n)$ be two positive functions. Then,

$$f(n) = \Omega(g(n))$$

if $f(n) \geq c \cdot g(n)$; for all n , $n \geq n_0$ such that there exists two positive constants.

$$c > 0, n_0 \geq 1$$

For example : $a = \Omega(b)$

Meaning : b is somehow lesser than a after taking c help.

Problem 2

$f(n) = n^2$
 $g(n) = n^2 + n + 10$
 $f(n) = \Omega(g(n))$
 $n = 1000$

For what value of c above statement holds true?

Handwritten work:

$$f(n) = \Omega(g(n))$$
$$f(n) \geq c \cdot g(n)$$
$$n^2 \geq c \cdot (n^2 + n + 10)$$

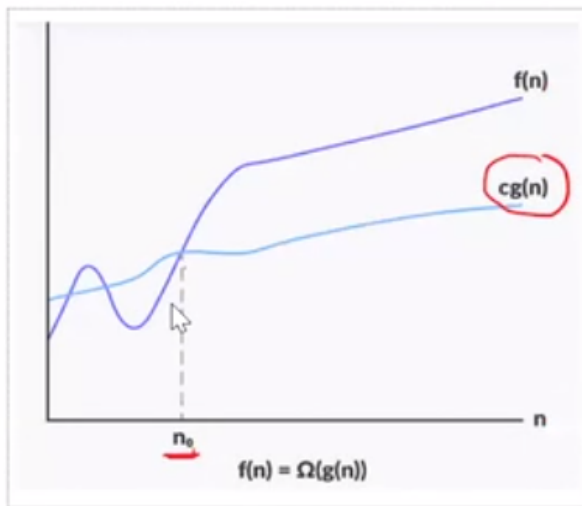
10000

$c \leq \frac{1}{2}$

$\frac{(n^2 + n + 10)}{2}$
 $\frac{(10000 + 1000 + 10)}{2}$

$f(n) = \Omega(g(n))$

Graphical Representation of Omega Notation



- If there exists a constant c such that it lies above $cg(n)$, for sufficiently large value of n .
- For any value of n , the minimum time required by the algorithm is given by $\Omega(g(n))$

Theta Notation

Let $f(n)$ and $g(n)$ be two positive functions. Then,

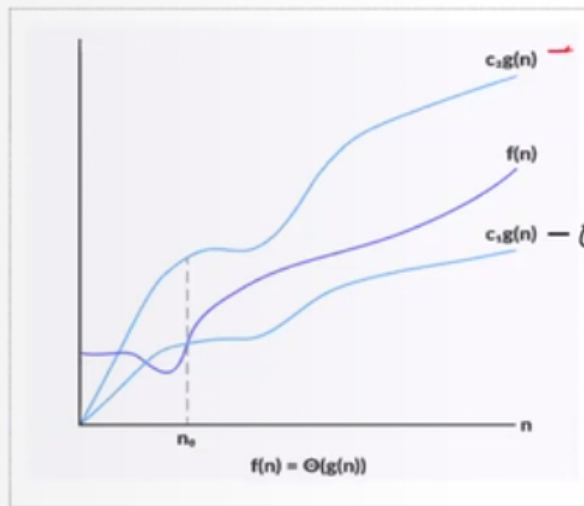
$$f(n) = \Theta(g(n))$$

if $f(n) \geq c_1 \cdot g(n)$ and $f(n) \leq c_2 \cdot g(n)$;

for all n , $n \geq n_0$ such that there exists two positive constants.

$$c_1 > 0, c_2 > 0, n_0 \geq 1$$

Graphical Representation of Theta Notation



• If there exists positive constants c_1 and c_2 such that it can be sandwiched between $c_1g(n)$ and $c_2g(n)$, for sufficiently large value of n .

• If a function lies anywhere in between $c_1g(n)$ and $c_2g(n)$ for all $n \geq n_0$, then $f(n)$ is said to be asymptotically tight bound.

Theta Notation
Problem 1

$$f(n) = n$$

$$g(n) = 5n$$

$$f(n) = \Theta(g(n))$$

$$f(n) \geq c_1 g(n) \text{ (Omega)} \leftarrow$$

$$n \geq c_1 \cdot 5n$$

$$c_1 = \frac{1}{5} \text{ - constant}$$

$$\text{constant } f(n) \leq c_2 g(n)$$

For what value of c_1 and c_2 above statement holds true?

$$f(n) = \Theta(g)$$

$$n \leq c_2 \cdot (5n)$$

$$c_2 = \frac{1}{5} \text{ - constant}$$

● Click to add text

$$\begin{aligned}
 & n^9 \\
 & n^9 \geq c_1 n^9 \quad \text{LHS} \\
 & \quad c_1 = 1 \\
 & n^9 \leq c_2 n^9 \\
 & \quad c_2 = 1
 \end{aligned}$$

$$\begin{aligned}
 & 5) \quad (128^{\log_2 n} \cdot n^2) = \Theta(n^9) \\
 & \quad 2^7 \log_2 n \cdot n^2 \\
 & \quad 2^{\log_2 n^7} \\
 & \quad \log_2 n^7 \cdot n^2 \\
 & \quad n^9
 \end{aligned}$$

Complexity Classes

1. $O(1)$: Constant time complexity
2. $O(\log(\log(n)))$
3. $O(\log n)$: Logarithmic time complexity
4. $O(\text{root}(n))$
5. $O(n)$: Linear time complexity
6. $O(n^2)$: Quadratic Time complexity
7. $O(n^3)$: Cubic Time Complexity
8. $O(c^n)$: Exponential Time complexity

Some points: **Important:**

1. $n! < n^n$
2. $2^n < n^n$
3. $n! > 2^n$

$$2^n < n! < n^n$$

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

$$n! = O(n^n)$$

$$\log(n) < n$$

$$(\log(n))^2 < n$$

$$\log(n)^{1000} < n$$

$$(\log n)^{(\log n)} > n$$

$$11) \log_2 n > \log_3 n$$

$$\begin{array}{l} \text{LHS} \\ 10) 2^n < 3^n \quad \text{RHS} \\ \downarrow \\ (2 \times 1.5)^n \\ \underline{2^n \times (1.5)^n} \end{array}$$

8) n^n

COMPLEXITY CLASSES

Relationship b/w these 3 classes

$\left\{ \begin{array}{l} n! < n^n \\ 2^n < n^n \\ n! > 2^n \end{array} \right\}$

\downarrow
 $(2^n < n! < n^n)$
 $2^n = O(n!)$ - ①
 $n! = O(n^n)$ - ②
 $2^n = O(n^n)$

$\log \log n$ - ① $O(1)$ - constant time complexity
 $O(\sqrt{n})$ - ② $O(\log n)$ - Logarithmic time complexity
 $O(n)$ - ③ $O(n)$ - Linear time complexity
 $O(n^2)$ - ④ $O(n^2)$ - Quadratic time complexity
 $O(n^3)$ - ⑤ $O(n^3)$ - Cubic time complexity
 $O(2^n)$ - ⑥ $O(2^n)$ - Exponential time complexity

$c > 1$
 $O(2^n)$

9) $\log n < n$
 $(\log n)^2 < n$
 $(\log n)^{1000} < n$ - True
 $((\log n)^{\log n} > n)$
 $\log n (\log \log n) > 1 \log n$

