

# GeeksforGeeks

A computer science portal for geeks

## GeeksQuiz

- [Home](#)
- [Algorithms](#)
- [DS](#)
- [GATE](#)
- [Interview Corner](#)
- [Q&A](#)
- [C](#)
- [C++](#)
- [Java](#)
- [Books](#)
- [Contribute](#)
- [Ask a Q](#)
- [About](#)

[Array](#)

[Bit Magic](#)

[C/C++](#)

[Articles](#)

[GFactS](#)

[Linked List](#)

[MCQ](#)

[Misc](#)

[Output](#)

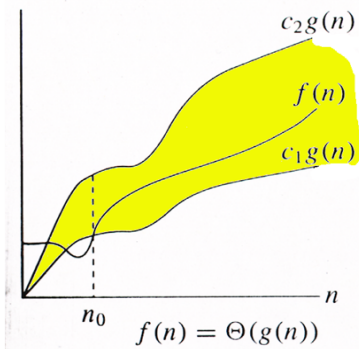
[String](#)

[Tree](#)

[Graph](#)

## Analysis of Algorithms | Set 3 (Asymptotic Notations)

We have discussed [Asymptotic Analysis](#), and [Worst, Average and Best Cases of Algorithms](#). The main idea of asymptotic analysis is to have a measure of efficiency of algorithms that doesn't depend on machine specific constants, and doesn't require algorithms to be implemented and time taken by programs to be compared. Asymptotic notations are mathematical tools to represent time complexity of algorithms for asymptotic analysis. The following 3 asymptotic notations are mostly used to represent time complexity of algorithms.



**1)  $\Theta$  Notation:** The theta notation bounds a functions from above and below, so it defines exact asymptotic behavior.

A simple way to get Theta notation of an expression is to drop low order terms and ignore leading constants. For example, consider the following expression.

$$3n^3 + 6n^2 + 6000 = \Theta(n^3)$$

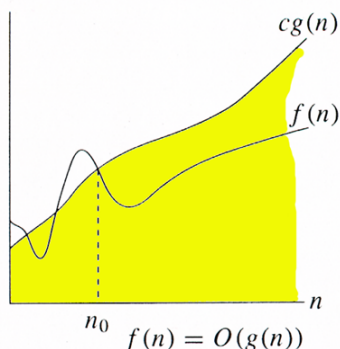
Dropping lower order terms is always fine because there will always be a  $n_0$  after which

$\Theta(n^3)$  beats  $\Theta(n^2)$  irrespective of the constants involved.

For a given function  $g(n)$ , we denote  $\Theta(g(n))$  is following set of functions.

$$\Theta(g(n)) = \{f(n): \text{there exist positive constants } c_1, c_2 \text{ and } n_0 \text{ such that } 0 \leq c_1 \cdot g(n) \leq f(n) \leq c_2 \cdot g(n) \text{ for all } n \geq n_0\}$$

The above definition means, if  $f(n)$  is theta of  $g(n)$ , then the value  $f(n)$  is always between  $c_1 \cdot g(n)$  and  $c_2 \cdot g(n)$  for large values of  $n$  ( $n \geq n_0$ ). The definition of theta also requires that  $f(n)$  must be non-negative for values of  $n$  greater than  $n_0$ .



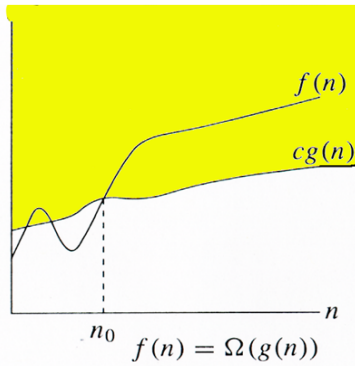
**2) Big O Notation:** The Big O notation defines an upper bound of an algorithm, it bounds a function only from above. For example, consider the case of Insertion Sort. It takes linear time in best case and quadratic time in worst case. We can safely say that the time complexity of Insertion sort is  $O(n^2)$ . Note that  $O(n^2)$  also covers linear time.

If we use  $\Theta$  notation to represent time complexity of Insertion sort, we have to use two statements for best and worst cases:

1. The worst case time complexity of Insertion Sort is  $\Theta(n^2)$ .
2. The best case time complexity of Insertion Sort is  $\Theta(n)$ .

The Big O notation is useful when we only have upper bound on time complexity of an algorithm. Many times we easily find an upper bound by simply looking at the algorithm.

$$O(g(n)) = \{f(n): \text{there exist positive constants } c \text{ and } n_0 \text{ such that } 0 \leq f(n) \leq c \cdot g(n) \text{ for all } n \geq n_0\}$$



3) **[Tex]\Omega[/Tex] Notation:** Just as Big O notation provides an asymptotic upper bound on a function, [Tex]\Omega[/Tex] notation provides an asymptotic lower bound.

[Tex]\Omega[/Tex] Notation can be useful when we have lower bound on time complexity of an algorithm. As discussed in the previous post, the best case performance of an algorithm is generally not useful, the Omega notation is the least used notation among all three.

For a given function  $g(n)$ , we denote by  $[Tex]\Omega[/Tex](g(n))$  the set of functions.

$[Tex]\Omega[/Tex](g(n)) = \{f(n) : \text{there exist positive constants } c \text{ and } n_0 \text{ such that } 0 \leq cg(n) \leq f(n) \text{ for all } n \geq n_0\}.$

Let us consider the same Insertion sort example here. The time complexity of Insertion Sort can be written as  $[Tex]\Omega[/Tex](n)$ , but it is not a very useful information about insertion sort, as we are generally interested in worst case and sometimes in average case.

### Exercise:

Which of the following statements is/are valid?

1. Time Complexity of QuickSort is  $[Tex]\Theta[/Tex](n^2)$
2. Time Complexity of QuickSort is  $O(n^2)$
3. For any two functions  $f(n)$  and  $g(n)$ , we have  $f(n) = [Tex]\Theta[/Tex](g(n))$  if and only if  $f(n) = O(g(n))$  and  $f(n) = [Tex]\Omega[/Tex](g(n))$ .
4. Time complexity of all computer algorithms can be written as  $[Tex]\Omega[/Tex](1)$

### References:

[Lec 1 | MIT \(Introduction to Algorithms\)](#)

[Introduction to Algorithms 3rd Edition by Clifford Stein, Thomas H. Cormen, Charles E. Leiserson, Ronald L. Rivest](#)

This article is contributed by **Abhay Rathi**. Please write comments if you find anything incorrect, or you want to share more information about the topic discussed above.

### Related Topics:

- [An interesting time complexity question](#)
- [A Problem in Many Binary Search Implementations](#)
- [Analysis of Algorithms | Set 4 \(Analysis of Loops\)](#)
- [NP-Completeness | Set 1 \(Introduction\)](#)
- [Static and Dynamic Libraries | Set 1](#)

- [The Ubiquitous Binary Search | Set 1](#)
- [Reservoir Sampling](#)
- [Analysis of Algorithms | Set 2 \(Worst, Average and Best Cases\)](#)

Like  34  Tweet  4   +1  1

Writing code in comment? Please use [ideone.com](http://ideone.com) and share the link here.

46 Comments

GeeksforGeeks

 1 Login ▾

 Recommend 1  Share

Sort by Newest ▾



Join the discussion...



**Ammar Baig** • 2 months ago

You got a little error Aditya, first you claim that " We write  $f(n) = \Omega(g(n))$  if there exist constants  $c > 0, n_0 > 0$  ( which is right ),, and then u deny yourself by saying, " such that  $0 \leq cg(n) \leq f(n)$  for all  $n \geq n_0$  ",, which is also right except for [  $0 \leq$  ] thing as we agreed earlier in this statement that  $c > 0$ , so in this case true relation that will hold for  $c.g(n)$  and  $f(n)$  is  $0 < cg(n) \leq f(n)$  for  $n \geq n_0$ .="" all=""

^ | ▾ • Reply • Share ›

**Aditya Goel** • 2 months ago

For all those saying 4th statement is false, they are wrong.

We write  $f(n) = \Omega(g(n))$  if there exist constants  $c > 0, n_0 > 0$  such that  $0 \leq cg(n) \leq f(n)$  for all  $n \geq n_0$

Here  $g(n)$  is 1.

$0 \leq c \leq f(n)$  holds true for any positive constant  $c$  and and algorithm in world(which takes  $f(n)$  time). Of couse we can provide more precise lower bound but that doesn't make this statement false.

^ | ▾ • Reply • Share ›



**Ritz** • 2 months ago

Guys what is the lower bound for binary search algorithm ? Upper Bound is  $O(\log n)$  .. Is lower bound big-Omega of a constant ? Plz explain

^ | ▾ • Reply • Share ›



**karunakar** → Ritz • 2 months ago

its  $\omega(1)$  because , for binary search elements are in sorted order we are search an element it can be found in  $\text{mid} = (\text{low} + \text{high}) / 2$  takes constant time so, it is  $\omega(1)$  or  $\theta(1)$

^ | ▾ • Reply • Share ›



**calvin** • 3 months ago



Hi,

I liked the way you analyzed the asymptotic notations, I was reading them in certain book but I couldn't get the difference well but you made it simple and clear . Thank you.

^ | v • Reply • Share ›

**DPS** • 3 months ago

2 nd 3 are valid option !!

^ | v • Reply • Share ›



**po** • 4 months ago

1,2,3 are valid 4 in invalid

1--->time complexity of quick sort best case  $\theta(n \log n)$  and worst case is  $O(n^2)$  we can say that  $\theta(n^2)$

^ | v • Reply • Share ›



**anshika** • 4 months ago

2. and 3. are valid!!

^ | v • Reply • Share ›

**abc** • 7 months ago

Suppose  $B(n)$  and  $W(n)$  are respectively the best case and worst case asymptotic running times for sorting an array of size  $n$  using Quick Sort. Consider the two statements:(i)  $B(n)$  is  $O(W(n))$ (ii)  $B(n)$  is  $\Theta(W(n))$ . (Select ONE answer)

A. Both (i) and (ii) are true

B. (i) is true but (ii) is false

C. (i) is false but (ii) is true

D. Both (i) and (ii) are false

6 ^ | v • Reply • Share ›



**still learning** → **abc** • 5 months ago

it should be B

^ | v • Reply • Share ›

**dk** → **abc** • 5 months ago

C.

^ | v • Reply • Share ›

**Chandresh Singh** → **abc** • 5 months ago

C.

^ | v • Reply • Share ›

**aswinii** • 7 months ago

can anyone provide a link to learn enter concepts of COMPLEXITY?

3 ^ | v • Reply • Share ›

**dk** → aswinii • 5 months ago<http://www.geeksforgeeks.org/>

2 ^ | v • Reply • Share ›

**guest** • 8 months ago4) Suppose  $T(n) = 2T(n/2) + n$ ,  $T(0) = T(1) = 1$ 

Which one of the following is false.

- a)  $T(n) = O(n^2)$
- b)  $T(n) = (n \log n)$
- c)  $T(n) = (n^2)$
- d)  $T(n) = O(n \log n)$

Answer (c)

i think c and D... what you say

3 ^ | v • Reply • Share ›

**Adauta Garcia Ariel** → guest • 4 months ago

I think c is false, b may be false, but D is true.

^ | v • Reply • Share ›

**Anurag Panwar** → guest • 5 months ago

I think the answer is b and c because the answer is not in asymptotic notation. That means that they are the exact answer which is not true.  $T(n) = O(n \log n)$  is right because running time of above recursion is  $\Theta(n \log n)$  which can be written as  $O(n \log n)$

1 ^ | v • Reply • Share ›

**esjeh** • 8 months ago

everyone here is concerned with the last question. i want to clarify my doubt that big omega always helps in finding the greatest lower bound of any [function.so](#) i think last question is invalid.

correct me if im wrong?

2 ^ | v • Reply • Share ›

**Praveen Kumar Tomar** • 9 months ago

Should we be more interestedi in worst case analysis or average case analysis ?

^ | v • Reply • Share ›

**Techie Me** → Praveen Kumar Tomar • 2 months ago

Worst Case analysis is the best thing to look for. At least it gives an upper bound on the

worst case analysis is the best thing to look for. At least it gives an upper bound on the running time.

However, there is a slight disagreement when we talk about real world problems. Most of the time the inputs on which we run an algorithm will not be the worst possible input. So, it will always complete before the worst running time.

If you are looking for an order in which we should be interested in, there is none specified. It all depends upon your need.

To understand more about this watch my video <https://www.youtube.com/watch?...>

^ | v • Reply • Share ›



**Kartik** → Praveen Kumar Tomar • 9 months ago

Worst case analysis is done usually, because it is not feasible to do average case analysis most of the time. See <http://www.geeksforgeeks.org/a...> for details.

^ | v • Reply • Share ›

**Praveen Kumar Tomar** • 9 months ago

I have observed most of the times we use big O notation, why shouldn't we use big theta notation more often because it seems to represent the average case situations

^ | v • Reply • Share ›



**Kartik** → Praveen Kumar Tomar • 9 months ago

Because Big O notation covers all complexities. We can always write worst case complexity in big O. For example, it is technically correct to say that the time complexity of insertion sort is  $O(n^2)$ , but not correct to say that the time complexity of insertion sort is  $\theta(n^2)$

2 ^ | v • Reply • Share ›



**Kartik** → Kartik • 9 months ago

Please read the first line as "Because Big O notation covers complexities of all cases"

1 ^ | v • Reply • Share ›

**Vaibhav Jain** • 10 months ago

Can anybody tell me that what is " $3n^3 + 6n^2 + 6000$ " or  $f(n)$  in the above article and how it is calculated for any algorithm (if there is any proposal to do so)?

^ | v • Reply • Share ›

**Vaibhav Bajpai** → Vaibhav Jain • 10 days ago

The time complexity of an algorithm is a measure of its efficiency in terms of time and space, where time comes as a result of number of key operations/instructions executed by the code.

Now what is a key operation ?

A key operation in a algorithm/code will be executed repeatedly in a single execution/run of that code;

$f(n)$  or  $T(n)$  or  $3(n^3) + 6(n^2) + 6000$ , all are same, and these can be called as cost functions, because they majorly contribute to the total time cost for the algorithm.

Every term in this function represents an instruction, like  $3(n^3)$  represents an instruction which will run the number of times equivalent to the value of  $3(n^3)$ , where  $n$  is the size of the input value. Similarly,  $6(n^2)$  again represents an instruction which will run the number of times equivalent to the value of  $6(n^2)$ , and 6000 represents one or more instructions( whose execution doesn't depend on the value of input ) and these will be executed 6000 times in total in a single run of that code.

Bundling these terms together as  $3(n^3) + 6(n^2) + 6000$  tells the total no of key operations that the code will perform. And hence we calculate the time complexity from this cost function.

^ | v • Reply • Share ›

ak • a year ago

Which of the following statements is/are valid?

1. Time Complexity of QuickSort is  $(n^2)$

Ans : INVALID because it is  $\Theta(n \log n)$

2. Time Complexity of QuickSort is  $O(n^2)$

Ans: Valid

3. For any two functions  $f(n)$  and  $g(n)$ , we have  $f(n) = (g(n))$  if and only if  $f(n) = O(g(n))$  and  $f(n) = \Omega(g(n))$ .

Ans: Valid

4. Time complexity of all computer algorithms can be written as  $\Omega(1)$ .

Ans:VALID

It depends on how complex the algorithm is

$O(1)$  represents a constant running time which is not possible.

Linear Search :  $O(n)$

Binary Search  $O(n \log n)$

QuickSort  $O(n^2)$

Matrix Multiplication  $O(n^3)$

BFS/DFS  $V + E \log(V)$

3 ^ | v • Reply • Share ›

DS+Algo=Placement ➔ ak • 9 months ago

I think 4th one is invalid



^ | v • Reply • Share ›



**kar** → DS+Algo=Placement • 9 months ago

yes 4th is invalid .  $\omega(1)$  means in best case , any algorithm will perform in constant time , which is not true .

^ | v • Reply • Share ›



**Aish** → ak • 9 months ago

How is the 4th one correct ?

^ | v • Reply • Share ›

**Rahul Ramesh** → ak • a year ago

You have said that Time complexity of Quick sort is  $\Theta(n \log n)$ ??? Isn't it wrong?

^ | v • Reply • Share ›



**Albut** → Rahul Ramesh • a year ago

It should be  $\theta(n \log n)$  and  $\theta(n^2)$  both same as in insertion sort.

^ | v • Reply • Share ›



**Marsha Donna** • a year ago

so does this mean that

big O notation corresponds to worst case time complexity,

big omega notation corresponds to best case time complexity,

big theta notation corresponds to average case time complexity???

12 ^ | v • Reply • Share ›

**ak** → Marsha Donna • a year ago

YES

^ | v • Reply • Share ›

**darubramha** → Marsha Donna • a year ago

Somewhat correct.

big O notation corresponds to worst case time complexity, generally it means "less than or equal to", i.e if  $T(n)$  [time complexity] =  $O(n)$  then it means,  $T(n)$  can have maximum  $cn$  value,  $\rightarrow T(n) \leq cn$ .

big omega notation corresponds to best case complexity, it is similar to " $\geq$ " as explained in the previous case.

But, big theta notation, tells us that the time complexity of the algorithm is asymptotically bounded by the given function, which means if  $T(n) = \theta(n)$ , the time complexity can never be worse than  $c_1n$  or can never be better than  $c_2n$ , where  $c_1, c_2$  are constants. It

does not mean the average case, it corresponds more towards "=", i.e  $c_1n \leq f(n) \leq c_2n$

3 ^ | v • Reply • Share ›

**Ankur Teotia** → Marsha Donna • a year ago

i have the same doubt , someone please clarify.

it seems to be the case but no where i see this explicitly mentioned.

2 ^ | v • Reply • Share ›



**noob** • 2 years ago

1. Time Complexity of QuickSort is  $(n^2)$  -----> NO
2. Time Complexity of QuickSort is  $O(n^2)$  -----> YES
3. For any two functions  $f(n)$  and  $g(n)$ , we have  $f(n) = (g(n))$  if and only if  $f(n) = O(g(n))$  and  $f(n) = (g(n))$ . -----> YES
4. Time complexity of all computer algorithms can be written as (1) -----> YES

11 ^ | v • Reply • Share ›



**Suvarna** → noob • a year ago

Can You please tell us that How are you calculating time complexity ?

I mean is it based upon worst case or any of the rest one ?

Can you please brief it ?

^ | v • Reply • Share ›

**gansai** → Suvarna • 5 months ago

In the exercise it is mentioned whether it is theta or BigO.

1. Theta( $n^2$ ) -- average case complexity.
2.  $O(n^2)$  -- worst case complexity

^ | v • Reply • Share ›

**Chaithanya Kanumolu** → noob • a year ago

I am confused with 4th point. Since best case for insertion sort is Theta( $n$ ) right? In that case how can we write it as Theta (1)?

3 ^ | v • Reply • Share ›



**ibn** → noob • a year ago

Why was question 4 Yes?

^ | v • Reply • Share ›

**Utkarsh Gupta** → ibn • a year ago

If  $f(n) = O(n)$  then it means  $f(n)$  can be less than or equal to order of  $n$ .

If  $f(n) = \Theta(n)$  then it means  $f(n)$  is equal to the order of  $n$ .

If  $f(n) = \Omega(n)$  then it means  $f(n)$  is greater than or equal to order of  $n$ .

So if you see  $\Omega(1)$ , it means that this is the smallest lower bound for any

algorithm. Therefore for any algorithm  $f(n) = \Omega(1)$ . But it can be even more

algorithm. Therefore for any algorithm  $f(n) = \Omega(1)$ . But it can be even more precise that is  $f(n) = \Omega(n)$  or  $\Omega(n^2)$  etc for different different conditions.

3 ^ | v • Reply • Share ›



**mog** → ibn • a year ago

bcoz oemga always provide a lower bound on running time of **algorithm**.so minimum running time would be  $\omega(1)$ ,it cant be less than  $\omega(1)$ .....so we can write time complexity of all algorithms as  $\omega(1)$ .

^ | v • Reply • Share ›

**Sourabh** → mog • a year ago

For a sorting algorithm, best case can not be better than  $\omega(n)$  as you have to check all  $n$  elements before saying that it is sorted, so I guess answer is NO!

2 ^ | v • Reply • Share ›

**rohit** → Sourabh • a year ago

$1 < n == ""$  > yes

^ | v • Reply • Share ›

**Sourabh** → rohit • a year ago

Well you got me there, I saw this answer again today and thought about it again and IMO  $\omega(1)$  is true for any algorithm (its a lower bound after, not a tight one though)

^ | v • Reply • Share ›

---

Subscribe

Add Disqus to your site

Privacy



GeeksforGeeks

Like

97,207 people like GeeksforGeeks.



Facebook social plugins

- [Interview Experiences](#)
- [Advanced Data Structures](#)
- [Dynamic Programming](#)
- [Greedy Algorithms](#)
- [Backtracking](#)
- [Pattern Searching](#)
- [Divide & Conquer](#)
- [Mathematical Algorithms](#)
- [Recursion](#)
- [Geometric Algorithms](#)

## • Popular Posts

- [All permutations of a given string](#)
- [Memory Layout of C Programs](#)
- [Understanding “extern” keyword in C](#)
- [Median of two sorted arrays](#)
- [Tree traversal without recursion and without stack!](#)
- [Structure Member Alignment, Padding and Data Packing](#)
- [Intersection point of two Linked Lists](#)
- [Lowest Common Ancestor in a BST.](#)
- [Check if a binary tree is BST or not](#)
- [Sorted Linked List to Balanced BST](#)

Follow @GeeksforGeeks

## • Recent Comments

◦ [It\\_k](#)

i need help for coding this function in java...

[Java Programming Language](#) · [1 hour ago](#)

◦ [Piyush](#)

What is the purpose of else if (recStack[\*i])...

[Detect Cycle in a Directed Graph](#) · [1 hour ago](#)

◦ [Andy Toh](#)

My compile-time solution, which agrees with the...

[Dynamic Programming | Set 16 \(Floyd Warshall Algorithm\)](#) · [1 hour ago](#)

◦ [lucy](#)

because we first fill zero in first col and...

[Dynamic Programming | Set 29 \(Longest Common Substring\)](#) · [1 hour ago](#)

◦ [lucy](#)

@GeeksforGeeks i don't n know what is this long...

[Dynamic Programming | Set 28 \(Minimum insertions to form a palindrome\)](#) · [2 hours ago](#)

◦ [manish](#)

Because TAN is not a subsequence of RANT. ANT...

[Given two strings, find if first string is a subsequence of second](#) · [2 hours ago](#)

•  
@geeksforgeeks, [Some rights reserved](#) \_\_\_\_ [Contact Us!](#)

Powered by [WordPress](#) & [MooTools](#), customized by geeksforgeeks team