

Best case time complexity, Average case time complexity, Worst case time complexity and Lower bound

We talk about the **Best case time complexity**, the **Average case time complexity** and the **Worst case time complexity** in the context of an algorithm **A** to solve a problem **P**. For example, you know those values for the quick sort.

We talk about lower bound in the context of a problem **P**. What we say applies to all algorithms **A** (present and future) to solve **P**.

For example, we proved that “The problem **P** of sorting n items using comparison” has a lower bound of $\Omega(n \log n)$. What we are saying is “any sorting algorithm **A** to sort n item through comparison (present or future) will have worst case time complexity $O(n \log n)$.”

We have the following relationship.

Let **P** be a problem and **A** be an algorithm to solve **P**. Then

The Worst case time complexity $A \geq$ The Lower Bound of (P).

If Worst case time complexity A is equal to the Lower Bound of (P), then A is an optimal algorithm to solve P

Example:

P is the problem of sorting n items using comparison.

Merge sort (**A**) an algorithm that solves **P**.

Worst case time complexity of **A** (merge sort) is $O(n \log n)$.

Lower bound of **P** (the problem of sorting n items using comparison) is $\Omega(n \log n)$.

Since both time complexity expressions are same (in this case **$n \log n$**), we say Merge Sort is an **optimal algorithm**. Note that Quick Sort is **not** an optimal algorithm!

f is $O(g)$ means $f(n) \leq cg(n)$ for large values of n .

f is $\Omega(g)$ means $f(n) \geq cg(n)$ for large values of n .

f is $\Theta(g)$ means $f(n) = cg(n)$ for large values of n .