

Algorithms

Formally :

A tool for solving a well-specified computational problem.



Algorithms: Is a finite sequence of unambiguous instructions for solving a well-specified computational problem.

Important Features :

1-Finiteness

2- Definiteness

3- input

4- Output

5- Effectiveness

Turing Machine : used to measure the efficiency of the algorithm.

Algorithm analysis :

- Determining performance characteristics
 - Time
 - Memory
 - Communication bandwidth
- Why analyze Algorithms ?
 - Choose the most efficient of several possible Algorithm for the same problem.
 - Choose the Algorithm with best running time

Running Time :

Run time expression should be machine independent

Use a model of computation or “hypothetical” computer

- Model should be :
 - Simple
 - Applicable

RAM Model :

- Generic single processor model
 - Supports simple constant-time instructions
 - Run time (cost) is uniform (1 time unit)
 - Memory is unlimited
 - Flat memory model → no hierarchy
 - Access to a word of memory takes 1 time unit
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Running time : is the number of steps executed by the Algorithm on that input often referred to as the **Complexity**.

Complexity of an Algorithm **depends on** :

- Size of input
 - Other characteristics of the input data
 - Data is Sorted
 - Data cycles in the graph
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Complexity

Worst case complexity :

Maximum numbers of steps the Algorithm takes for any possible input.

Average case complexity :

Average of the running times of all possible input.

Best case complexity :

Minimum numbers of steps the Algorithm takes for any possible input.

Pseudo code conventions :

- Indentation (for block structure)
- Value of loop counter variable upon loop termination
- Call by value **not** reference
- Local variable
- Error handling are omitted

<i>LinearSearch(A, key)</i>	<i>cost</i>	<i>times</i>
1 $i \leftarrow 1$	c_1	1
2 while $i \leq n$ and $A[i] \neq key$	c_2	x
3 do $i++$	c_3	$x-1$
4 if $i \leq n$	c_4	1
5 then return <i>true</i>	c_5	1
6 else return <i>false</i>	c_6	1

x ranges between 1 and $n+1$.

So, the running time ranges between

$c_1 + c_2 + c_4 + c_5$ – **best case**

and

$c_1 + c_2(n+1) + c_3n + c_4 + c_6$ – **worst case**

Worst Case :

$$1+n+1+n+1+1 = 2n + 4$$

$O(n)$

Average case :

$$1 + n/2 + n/2 + 1 + 1 = n + 3$$

$\Theta(n)$

Best Case :

$$1+1+1+1 = 4$$

$\Omega(1)$

Complexity of an algorithm : is denoted by the highest order term in the expression for running time

Order of growth : how running time grows with input size

Asymptotic complexity : The running time for large inputs