

Algorithms basics

What is an Algorithm?

An **algorithm** is a finite sequence of well-defined instructions that can be used to solve a computational problem. It provides a step-by-step procedure that convert an input into a desired output.

Algorithms typically follow a logical structure:

- **Input:** The algorithm receives input data.
- **Processing:** The algorithm performs a series of operations on the input data.
- **Output:** The algorithm produces the desired output.

What is the Need for Algorithms?

Algorithms are essential for solving complex computational problems efficiently and effectively. They provide a systematic approach to:

- **Solving problems:** Algorithms break down problems into smaller, manageable steps.
 - **Optimizing solutions:** Algorithms find the best or near-optimal solutions to problems.
 - **Automating tasks:** Algorithms can automate repetitive or complex tasks, saving time and effort.
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Algorithms characteristics

- Inputs and output
 - What does algorithm accept, and what are the result?
 - Algorithm complexity
 - Space complexity - How much memory does it require?
 - Time complexity – How much time does it require to complete?
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Big-O notation

Big O notation is a way to characterize the time or resources needed to solve a computing problem. It's particularly useful in comparing various computing algorithms and approaches under consideration, [such as those used in Machine Learning](#).

Below is a table summarizing Big O functions. The four most commonly referenced and important to remember are:

- **$O(1)$** - *Constant* access time such as the use of a hash table.
 - **$O(\log n)$** - *Logarithmic* access time such as a binary search of a sorted table.
 - **$O(n)$** - *Linear* access time such as the search of an unsorted list.
 - **$O(n^2)$** - Nested iterations.
 - **$O(n \log(n))$** - *Multiple of $\log(n)$* access time such as using Quicksort or Merge sort.
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Big-O notation table

Notation	Type	Examples	Description
$O(1)$	Constant	Hash table access	Remains constant regardless of the size of the data set
$O(\log n)$	Logarithmic	Binary search of a sorted table	Increases by a constant. If n doubles, the time to perform increases by a constant, smaller than n amount
$O(<n)$	Sublinear	Search using parallel processing	Performs at less than linear and more than logarithmic levels
$O(n)$	Linear	Finding an item in an unsorted list	Increases in proportion to n . If n doubles, the time to perform doubles
$O(n \log(n))$	$n \log(n)$	Quicksort, Merge Sort	Increases at a multiple of a constant
$O(n^2)$	Quadratic	Bubble sort	Increases in proportion to the product of $n*n$
$O(c^n)$	Exponential	Travelling salesman problem solved using dynamic programming	Increases based on the exponent n of a constant c
$O(n!)$	Factorial	Travelling salesman problem solved using brute force	Increases in proportion to the product of all numbers included (e.g., $1*2*3*4...$)