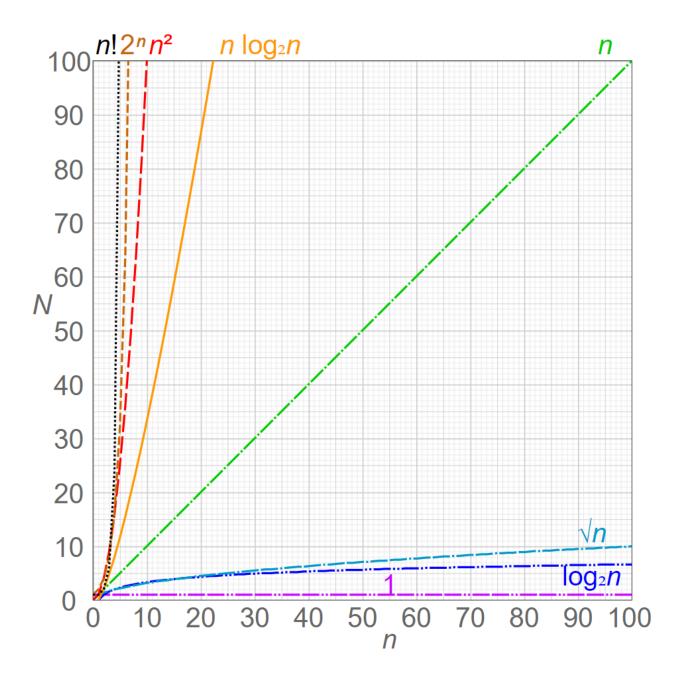
Running Time Time Complexity Space Complexity

In computer science, runtime, run time or execution time is the time when the cpu is executing the machine code.

In computer science, the time complexity is the computational complexity that describes the amount of time it takes to run an algorithm.

In computer science, the space complexity of an algorithm or a computer program is the amount of memory space required to solve an instance of the computational problem as a function of the size of the input.



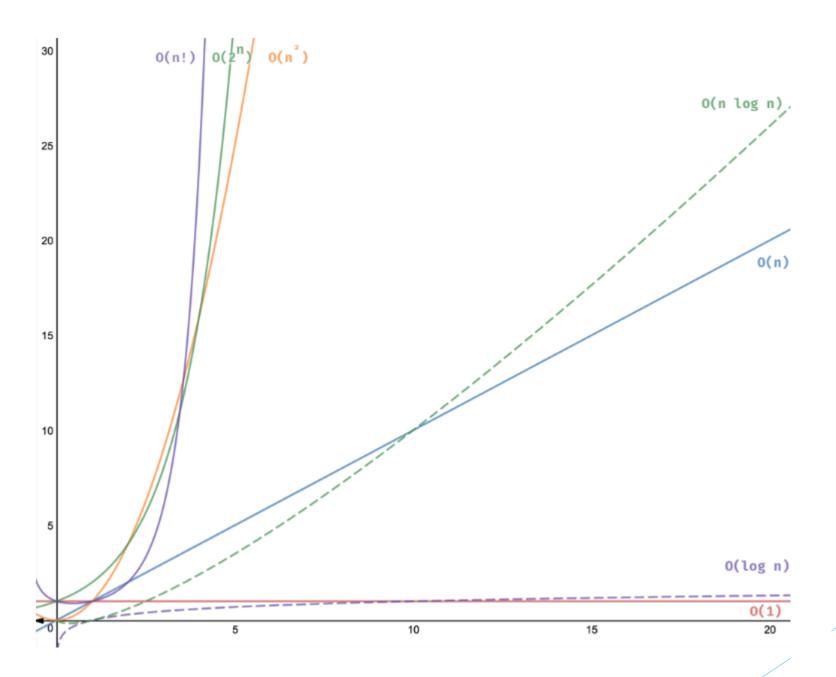


Table of common time complexities

Further information: Computational complexity of mathematical operations

The following table summarizes some classes of commonly encountered time complexities. In the table, $poly(x) = x^{O(1)}$, i.e., polynomial in x.

Name ◆	Complexity class +	Running time (T(n)) ♦	Examples of running times \$	Example algorithms \$
				Finding the median value in a sorted array of numbers
constant time		O(1)	10	Calculating (-1) ⁿ
inverse Ackermann time		O(α(n))		Amortized time per operation using a disjoint set
iterated logarithmic time		O(log* n)		Distributed coloring of cycles
log-logarithmic		O(log log n)		Amortized time per operation using a bounded priority queue ^[2]
logarithmic time	DLOGTIME	O(log n)	$\log n$, $\log(n^2)$	Binary search
polylogarithmic time		poly(log n)	$(\log n)^2$	
fractional power		$O(n^c)$ where $0 < c < 1$	n ^{1/2} , n ^{2/3}	Searching in a kd-tree
linear time		O(n)	n, 2n + 5	Finding the smallest or largest item in an unsorted array, Kadane's algorithm
"n log-star n" time		O(n log* n)		Seidel's polygon triangulation algorithm.
linearithmic time		O(n log n)	n log n, log n!	Fastest possible comparison sort; Fast Fourier transform.
quasilinear time		n poly(log n)		
quadratic time		O(n ²)	n ²	Bubble sort; Insertion sort; Direct convolution
cubic time		O(n ³)	n ³	Naive multiplication of two <i>n</i> × <i>n</i> matrices. Calculating partial correlation.
polynomial time	Р	$2^{O(\log n)} = poly(n)$	$n^2 + n, n^{10}$	Karmarkar's algorithm for linear programming; AKS primality test ^{[3][4]}
quasi-polynomial time	QP	2 ^{poly(log n)}	nlog log n, nlog n	Best-known O(log ² n)-approximation algorithm for the directed Steiner tree problem.
sub-exponential time (first definition)	SUBEXP	$O(2^{n^{\epsilon}})$ for all $\epsilon > 0$	$O(2^{\log n^{\log \log n}})$	Contains BPP unless EXPTIME (see below) equals MA. ^[5]
sub-exponential time (second definition)		2 ^{o(n)}	2 ^{n1/3}	Best-known algorithm for integer factorization; formerly-best algorithm for graph isomorphism
exponential time (with linear exponent)	Е	2 ^{O(n)}	1.1 ⁿ , 10 ⁿ	Solving the traveling salesman problem using dynamic programming
exponential time	EXPTIME	2 ^{poly(n)}	2 ⁿ , 2 ^{n²}	Solving matrix chain multiplication via brute-force search
factorial time		O(n!)	n!	Solving the traveling salesman problem via brute-force search
double exponential time	2-EXPTIME	2 ^{2poly(n)}	2 ^{2ⁿ}	Deciding the truth of a given statement in Presburger arithmetic

An algorithm is said to be constant time (also written as O(1) time) if the value of T(n) is bounded by a value that does not depend on the size of the input. For example, accessing any single element in an array takes constant time as only one operation has to be performed to locate it.

- An algorithm is said to take logarithmic time when $T(n) = O(\log n)$.
- Algorithms taking logarithmic time are commonly found in operations on binary trees or when using binary search.

An algorithm is said to run in polylogarithmic time if $T(n) = O((\log n)^k)$, for some constant k.

An algorithm is said to take linear time, or O(n) time, if its time complexity is O(n).

- An algorithm is said to be subquadratic time if $T(n) = o(n^2)$.
- For example, simple, comparison-based sorting algorithms are quadratic (e.g. insertion sort)

An algorithm is said to be of polynomial time if its running time is upper bounded by a polynomial expression in the size of the input for the algorithm, i.e., $T(n) = O(n^k)$ for some positive constant k.

Big O Notation	Name	Example(s)	
O(1)	Constant	# Odd or Even number, # Look-up table (on average)	
O(log n)	Logarithmic	# Finding element on sorted array with binary search	
O(n)	Linear	# Find max element in unsorted array, # Duplicate elements in array with Hash Map	
O(n log n)	Linearithmic	# Sorting elements in array with merge sort	
O(n ²)	Quadratic	# Duplicate elements in array **(naïve)**, # Sorting array with bubble sort	
O(n ³)	Cubic	# 3 variables equation solver	
O(2 ⁿ)	Exponential	# Find all subsets	
O(n!)	Factorial	# Find all permutations of a given set/string	

References

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