

Constraints	Worst time complexity	Algorithmic solution	Examples
$n \leq 12$	$O(n!)$	Recursion & Backtracking	Permutation 1....n
$n \leq 25$	$O(2^n)$	Recursion , Backtracking & Bit Manipulation	All subsets of an array of size n
$n \leq 100$	$O(n^4)$	Dynamic Programming	4Sum
$n \leq 500$	$O(n^3)$	Dynamic Programming	All triangles with side length less than n
$n \leq 10^4$	$O(n^2)$	Dynamic Programming, Graphs , Trees	Bubble Sort (Slow comparison-based sorting)
$n \leq 10^6$	$O(n \log n)$	Sorting , Binary Search , Divide and Conquer	Merge Sort (Fast comparison-based sorting)
$n \leq 10^8$	$O(n)$	Mathematical , Greedy	Min and max of element
$n > 10^8$	$O(\log n)$ or $O(1)$	Mathematical, Greedy	Binary Search

max ops = 10^7

Solve LeetCode Questions with This One Pattern (Constraints to Time Complexity)

$n < 20$	$2^n, n!$ e.g. brute force, backtracking
$n < 3000$	n^2 e.g. dynamic programming
$3000 < n < 10^6$	$O(n), O(n \log n)$ e.g. 2 pointers, greedy, heap, sorting
$n > 10^6$	$O(\log n), O(1)$ e.g. binary search, math

Common time complexities

Let n be the main variable in the problem.

- If $n \leq 12$, the time complexity can be $O(n!)$.
- If $n \leq 25$, the time complexity can be $O(2^n)$.
- If $n \leq 100$, the time complexity can be $O(n^4)$.
- If $n \leq 500$, the time complexity can be $O(n^3)$.
- If $n \leq 10^4$, the time complexity can be $O(n^2)$.
- If $n \leq 10^6$, the time complexity can be $O(n \log n)$.
- If $n \leq 10^8$, the time complexity can be $O(n)$.
- If $n > 10^8$, the time complexity can be $O(\log n)$ or $O(1)$.

Examples of each common time complexity

- $O(n!)$ [Factorial time]: Permutations of $1 \dots n$
- $O(2^n)$ [Exponential time]: Exhaust all subsets of an array of size n
- $O(n^3)$ [Cubic time]: Exhaust all triangles with side length less than n
- $O(n^2)$ [Quadratic time]: Slow comparison-based sorting (eg. Bubble Sort, Insertion Sort, Selection Sort)
- $O(n \log n)$ [Linearithmic time]: Fast comparison-based sorting (eg. Merge Sort)
- $O(n)$ [Linear time]: Linear Search (Finding maximum/minimum element in a 1D array), Counting Sort
- $O(\log n)$ [Logarithmic time]: Binary Search, finding GCD (Greatest Common Divisor) using Euclidean Algorithm

- $O(1)$ [Constant time]: Calculation (eg. Solving linear equations in one unknown)

you get a clearer picture of how quickly each type of complexity increases with larger inputs. This is useful for understanding and comparing the efficiency of different algorithms.

$$\log_{10}(n!) = 8.68 \text{ for } n = 12$$

$$\log_{10}(2^n) = 7.52 \text{ for } n = 25$$

$$\log_{10}(n^4) = 8.00 \text{ for } n = 100$$

$$\log_{10}(n^3) = 8.09 \text{ for } n = 500$$

$$\log_{10}(n^2) = 8.00 \text{ for } n = 10^4$$

$$\log_{10}(n \log n) = 7.29 \text{ for } n = 10^6$$

$$\log_{10}(n) = 8.00 \text{ for } n = 10^8$$

Data Types and their ranges

Data Type	Actual Range	Approximate Range (10^x)
short int	-32,768 to 32,767	$\pm 3.2 \times 10^4$
unsigned short int	0 to 65,535	0 to 6.5×10^4
int	-2,147,483,648 to 2,147,483,647	$\pm 2.1 \times 10^9$
unsigned int	0 to 4,294,967,295	0 to 4.3×10^9
long int	-2,147,483,648 to 2,147,483,647	$\pm 2.1 \times 10^9$
unsigned long int	0 to 4,294,967,295	0 to 4.3×10^9
long long int	-9,223,372,036,854,775,808 to 9,223,372,036,854,775,807	$\pm 9.2 \times 10^{18}$
unsigned long long int	0 to 18,446,744,073,709,551,615	0 to 1.8×10^{19}
float	$\sim 1.2\text{E-}38$ to $\sim 3.4\text{E}+38$	$\sim 10^{-38}$ to $\sim 10^{38}$
double	$\sim 2.3\text{E-}308$ to $\sim 1.7\text{E}+308$	$\sim 10^{-308}$ to $\sim 10^{308}$
long double	Greater than double (varies by implementation, e.g., $1.2\text{E-}4932$ to $1.2\text{E}+4932$)	$\sim 10^{-4932}$ to $\sim 10^{4932}$ (varies)

Data Type	Actual Range	Approximate Range (10^x)
char	-128 to 127 (signed) or 0 to 255 (unsigned)	$\pm 1.2 \times 10^2$ (signed) or 0 to 2.5×10^2 (unsigned)
unsigned char	0 to 255	0 to 2.5×10^2
signed char	-128 to 127	$\pm 1.2 \times 10^2$
bool	true or false	N/A
wchar_t	Implementation-defined (e.g., 0 to 4,294,967,295 for 4-byte wchar_t)	Varies

Types of Solutions Feasible for Upper Limits

Given the constraints, the following table suggests feasible solutions for different input size constraints and time complexity:

Input Size Constraint	Time Complexity	Feasible Solutions
$N \leq 10$	$O(N!)$	Brute force permutations, backtracking.
$N \leq 20$	$O(2^N)$	Dynamic programming with bitmasks, exhaustive search.
$N \leq 100$	$O(N^3)$	Floyd-Warshall for shortest paths, cubic algorithms.
$N \leq 1,000$	$O(N^2)$	Dynamic programming, graph algorithms (Floyd-Warshall).
$N \leq 10,000$	$O(N \log N)$	Sorting algorithms, binary search, segment trees.
$N \leq 100,000$	$O(N)$	Linear time algorithms, counting sort, prefix sums.
$N \leq 1,000,000$	$O(N \log N)$	Efficient sorting, divide and conquer, balanced trees.
$N \leq 10,000,000$	$O(N)$	Linear algorithms with efficient data structures (e.g., counting sort).

Special Constraints

- **Definition:** Unique conditions specific to the problem, like distinct values or specific properties of data (e.g., sorted array, connected graph).
- **Impact:** Adds complexity and requires consideration of additional factors in the solution.

Special Constraints and Their Solutions

1. Distinct Elements

- **Example:** All array elements are distinct.

- **Feasible Solutions:** Sorting algorithms, unique combinations, or permutations can be leveraged effectively.

2. Graph Connectivity

- **Example:** The graph is connected.
- **Feasible Solutions:** DFS, BFS, and union-find algorithms for efficient traversal and connectivity checks.

3. Sorted Input

- **Example:** The input array is already sorted.
- **Feasible Solutions:** Binary search, two-pointer techniques, and efficient merge operations.

a comprehensive table that suggests feasible solutions based on different input size constraints and their corresponding time complexities:

Detailed Feasible Solutions for Different Input Sizes

Input Size Constraint	Time Complexity	Space Complexity	Feasible Solutions
$N \leq 10$	$O(N!)$	$O(N)$	Brute force permutations, backtracking, exhaustive search.
$N \leq 20$	$O(2^N)$	$O(2^N)$	Dynamic programming with bitmasks, exhaustive search, subset generation.
$N \leq 100$	$O(N^3)$	$O(N^2)$	Floyd-Warshall for shortest paths, cubic algorithms, dynamic programming with three nested loops.
$N \leq 10^3$	$O(N^2)$	$O(N^2)$	Dynamic programming (e.g., longest common subsequence), graph algorithms (e.g., Floyd-Warshall, matrix multiplication).
$N \leq 10^4$	$O(N \log N)$	$O(N)$	Efficient sorting algorithms (e.g., merge sort, quicksort), binary search, line sweeping algorithms, segment trees.
$N \leq 10^5$	$O(N)$	$O(N)$	Linear time algorithms (e.g., counting sort, radix sort, prefix sums, two-pointer techniques, linear scan).
$N \leq 10^6$	$O(N \log N)$	$O(N)$	Balanced tree algorithms, divide and conquer strategies, efficient data structures (e.g., Fenwick trees, AVL trees).

Input Size Constraint	Time Complexity	Space Complexity	Feasible Solutions
$N \leq 10^7$	$O(N)$	$O(N)$	Linear algorithms with efficient implementation (e.g., counting sort, radix sort, linear scan for specific problems).
$N \leq 10^8$	$O(N \log \log N)$	$O(N)$	Sieve of Eratosthenes for prime number generation, certain amortized linear algorithms.
$N \leq 10^9$	$O(N)$	$O(N)$	Specific linear algorithms, data streaming algorithms, counting frequencies.
$N \leq 10^{18}$	$O(\log N)$	$O(1)$	Binary search on ranges, logarithmic algorithms, number theory algorithms (e.g., modular arithmetic, logarithmic exponentiation).

Small Input Size ($N \leq 10$)

- $O(N!)$
 - Brute force solutions.
 - Permutations and combinations.
 - Backtracking algorithms.

Moderate Input Size ($N \leq 20$)

- $O(2^N)$
 - Subset generation.
 - Dynamic programming with bitmasks.
 - Recursive algorithms with memoization.

Larger Input Size ($N \leq 100$)

- $O(N^3)$
 - Floyd-Warshall algorithm for all pairs shortest paths.
 - Matrix multiplication.
 - Dynamic programming for problems like the traveling salesman problem with small N .

Significant Input Size ($N \leq 1,000$)

- **$O(N^2)$**
 - Dynamic programming for problems like the longest common subsequence.
 - Graph algorithms like Floyd-Warshall.
 - Matrix chain multiplication.

Large Input Size ($N \leq 10,000$)

- **$O(N \log N)$**
 - Efficient sorting algorithms like merge sort and quicksort.
 - Binary search algorithms.
 - Data structures like segment trees for range queries.

Very Large Input Size ($N \leq 100,000$)

- **$O(N)$**
 - Linear algorithms like counting sort, radix sort.
 - Prefix sums for range sum queries.
 - Two-pointer techniques for problems involving subarrays or subsequences.

Extremely Large Input Size ($N \leq 1,000,000$)

- **$O(N \log N)$**
 - Balanced tree algorithms like AVL trees.
 - Efficient divide and conquer strategies.
 - Data structures like Fenwick trees for range queries.

Ultra Large Input Size ($N \leq 10,000,000$)

- **$O(N)$**
 - Highly optimized linear algorithms.
 - Counting sort for specific ranges.
 - Linear scan algorithms for specific types of problems.

Massive Input Size ($N \leq 100,000,000$)

- **$O(N \log \log N)$**
 - Sieve of Eratosthenes for prime number generation.
 - Amortized linear time algorithms for specific problem domains.

Giga Input Size ($N \leq 1,000,000,000$)

- **$O(N)$**
 - Data streaming algorithms.

- Counting frequencies in large data sets.
- Highly specialized linear algorithms.

$N \leq 10^{18}$

- **Time Complexity: $O(\log N)$**
- **Space Complexity: $O(1)$**
 - Binary search on ranges, logarithmic algorithms like modular exponentiation, and number theory algorithms.

#note

Standard CPU operation 10^8 ops/s