# Time Complexity

#### Goal:

- Understand time complexity
- Understand Big-O notation for time complexity.
- Evaluate time complexity of an algorithm.
- Evaluate expected time complexity based on the given constraints of a problem.
- Evaluating space complexity of a program.
- Common verdicts of submissions.

# What is an Elementary Operation?

An operation that takes constant time is called elementary operation.

#### Example:

- Arithmetic operations
- Comparison of primitive types
- Input and output of primitive types

10<sup>8</sup> operations ≈ 1 second

1. Is the following an elementary operation?

```
int a, b, c, d;
cin >> a >> b >> c >> d;

cout << (a + b * c) / d << endl;</pre>
```

2. Is the following an elementary operation?

```
string s, t;
cin >> s >> t;

if (s < t)
    cout << "s is less than t" << endl;</pre>
```

# What is Time Complexity?

Time complexity is a function to describe the approximate amount of operations an algorithm requires for the given input.

We can calculate approximate execution time of code using time complexity and constraints.

# **Big-O** notation

Big-O of an algorithm is a function to calculate the worst case time complexity of the algorithm.

It is written as O(worst case time complexity)

Big-O is used to calculate the approximate *upper* bound of the algorithm. It expresses how the run time of the algorithm grows relative to the input.

More convenient and useful than other notations.

# Rules for Big-O notation

- Should not have constants.
- Should not have constant factors.
- Only include the *fastest growing* function *for each variable*.
- Can never be 0. Has to be atleast O(1)

Example function:  $2(N^2) + 4N + 4(M^3 + 5) + 10$ 

2. 
$$N(N+1)/2$$

3. 
$$N^2 + M(N^2) + M^2(N) + NM$$

4. 
$$N^3/64 + 20N + (32NM)^2$$

# Calculate Time Complexity of an Algorithm

Time complexity usually depends on:

- Loops
- Recursion

Time complexity of recursive algorithms will not be covered.

Note: Usage of STL counts for time complexity

# Calculate Time Complexity of an Algorithm

If there are nested loops, multiply the expected number of iterations of the loops

#### Example:

```
for (int i = 0; i < n; i++) {
    for (int j = 0; j < n; j++) {
        for (int k = 0; k < 4; k++) {
            // Elementary operations
        }
    }
}</pre>
```

Find the time complexity of the following code snippets in Big-O notation:

```
for (int i = 0; i < n; i++) {
    for (int j = 0; j < n/2; j++) {
1.
    }
}</pre>
```

```
for (int i = 0; i < n; i++) {
    for (int j = 0; j*j < n; j++) {
    }
}</pre>
```

```
3. for (int i = 1; i < n; i *= 2) {
}
```

```
4. for (int i = 12; i <= n-123; i += 5) {
    for (int j = 6; j <= m*2; j += 321) {
        for (int k = 4023; k > 23; k -= 16) {
        }
    }
}
```

# Time Complexity based on Constraints

Feasible Big-O Function	Maximum N	Example Algorithms
O( <i>N</i> !)	10	All permutations of a list
$O(N^3)$	400	Multiplication of two matrices
O(N <sup>2</sup> )	5000	Square grid, bubble sort, insertion sort
$O(N\sqrt{N})$	10 <sup>5</sup>	Usually related to factoring
O(NlogN)	10 <sup>6</sup>	Merge sort, binary search for N times
O( <i>N</i> )	10 <sup>7</sup>	Linear search, reversing an array, string comparison
$O(\sqrt{N})$	10 <sup>12</sup>	Factors of a number
O(logN), O(1)	10 <sup>18</sup>	Binary search, Constant time formulas

# **Space Complexity**

Space complexity is similar to time complexity, except it measures the amount of memory.

Any datatype that has constant memory takes O(1) space.

Example: int, char, long long int, double, etc.

# **Space Complexity**

# Most problems have a memory limit of 256MB or ~2e8 bytes.

Datatype	No. of bytes
char	1
int	4
float	4
long long int	8
double	8
long double	16

#### Verdict of a solution

- AC: Accepted
- WA: Wrong Answer
- TLE: Time Limit Exceeded
- MLE: Memory Limit Exceeded
- RE: Runtime Error
  - NZEC: Non Zero Exit Code
  - SIGSEGV: Usually due to out of bounds
  - SIGFPE: Usually division or modulo by 0
  - SIGABRT: Due to assert statements

#### Points to note:

- Identify the variables that contribute to time complexity.
- Just because constraints allow slower solutions, doesn't mean there's not a fast solution.
   For example, if N <= 1000, then both O(N²) and O(N) can pass.
- Testcases matter, unless there's a limit explicitly imposed in the constraints.
- The constants and constant factors removed when calculating Big-O still matter.

# Problems to test understanding

- https://codeforces.com/contest/1647/problem/A
- https://codeforces.com/problemset/problem/1538/C
- https://www.codechef.com/MARCH221D/problems/DISCUS
- https://www.codechef.com/MARCH221D/problems/WORDLE
- https://www.codechef.com/MARCH221D/problems/CHFDBT
- https://codeforces.com/contest/1651/problem/B
- https://codeforces.com/contest/1651/problem/A
- https://codeforces.com/problemset/problem/919/B

For more practice, try to figure out the time complexity for any random problem.

# Further Reading:

- https://towardsdatascience.com/essential-programmingtime-complexity-a95bb2608cac
- https://www.youtube.com/watch?v=9TlHvipP5yA https://www.youtube.com/watch?v=9SgLBjXqwd4 https://www.youtube.com/watch?v=I0DTkS1LJ2k
- https://adrianmejia.com/most-popular-algorithms-timecomplexity-every-programmer-should-know-free-onlinetutorial-course/ (advanced)