



Sunbeam Institute of Information Technology

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Module – Data Structures and Algorithms

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Algorithm analysis

- it is done for efficiency measurement and also known as time/space complexity
- It is done to finding time and space requirements of the algorithm
 1. Time - time required to execute the algorithm (ns, μ s, ms, s)
 2. Space - space required to execute the algorithm inside memory (bytes, kb, mb, gb)
- finding exact time and space of the algorithm is not possible because it depends on few external factors like
 - time is dependent on type of machine (CPU), number of processes running at that time
 - space is dependent on type of machine (architecture), data types
- Approximate time and space analysis of the algorithm is always done
- Mathematical approach is used to find time and space requirements of the algorithm and it is known as "Asymptotic analysis"
- Asymptotic analysis also tells about behaviour of the algorithm for different input or for change in sequence of input
- This behaviour of the algorithm is observed in three different cases
 1. Best case
 2. Average case
 3. Worst case

To denote time and space complexity, we use 'Big O' / O() notation is used

Time complexity

- time is directly proportional to number of iterations of the loops used in an algorithm
- To find time complexity/requirement of the algorithm count number of iterations of the loops

1. Print 1D array on console

```
void print1DArray(arr[], n) {
    for(i=0; i < n; i++)
        cout << arr[i];
}
```

No. of iterations = n

time \propto iterations

time $\propto n$

$$T(n) = O(n)$$

2. Print 2D array on console

```
void print2DArray(arr[][], m, n) {
    for(i=0; i < m; i++) {
        for(j=0; j < n; j++) {
            cout << arr[i][j];
        }
    }
}
```

iterations of outer loop = m
 iterations of inner loop = n
 total no. of iterations = $m * n$

Time \propto iterations

Time $\propto m * n$

$$T(m, n) = O(m, n)$$

$\because m \approx n$
 total iterations = $n * n$
 time \propto itr
 time $\propto n^2$

$$T(n) = O(n^2)$$

3. Add two numbers

```
int addition(n1, n2) {  
    res = n1 + n2;  
    return res;  
}
```

- irrespective of values of $n1$ & $n2$, this algorithm will be completed in constant/ fixed time.
- constant time requirement, & it is denoted as

$$T(n) = O(1)$$

4. Print table of given number

```
void printTable(int n) {  
    for( i = 1; i <= 10; i++)  
        cout << i * n << " ";  
}
```

- loop is going to iterate fix number of times
- it will constant time, means constant time requirement

$$T(n) = O(1)$$

5. Print binary of decimal number

2	9	
	4	1
	2	0
	1	0
		1

$$(9)_{10} = (1001)_2$$

```

void printBinary( n) {
    while( n > 0) {
        cout << (n % 2);
        n = n / 2;
    }
}

```

n	n > 0	n % 2
9	T	1
4	T	0
2	T	0
1	T	1
0	F	

$$\begin{aligned}
 n &= 9, 4, 2, 1 \\
 &= \frac{9}{1}, \frac{9}{2}, \frac{9}{4}, \frac{9}{8} \\
 &= \frac{n}{2^0}, \frac{n}{2^1}, \frac{n}{2^2} \dots \frac{n}{2^i}
 \end{aligned}$$

no. of itr

$$\frac{n}{2^i} = 1$$

$$n = 2^i$$

$$\log 2^i = \log n$$

$$i \log 2 = \log n$$

$$i = \frac{\log n}{\log 2}$$

$$\begin{aligned}
 \text{time} &\propto \text{itr} \\
 \text{time} &\propto \frac{\log n}{\log 2}
 \end{aligned}$$

$$T(n) = O(\log n)$$

Time complexity

Time complexities : $O(1)$, $O(\log n)$, $O(n)$, $O(n \log n)$, $O(n^2)$, $O(n^3)$, ..., $O(2^n)$,

Modification : + or - : time complexity is in terms of n

Modification : * or / : time complexity is in terms of $\log n$

$\text{for}(i=0; i < n; i++) \rightarrow O(n)$

$\text{for}(i=n; i > 0; i--) \rightarrow O(n)$

$\text{for}(i=0; i < n; i+=2) \rightarrow O(n)$

$\text{for}(i=1; i \leq 20; i++) \rightarrow O(1)$

$\text{for}(i=n; i > 0; i/=2) \rightarrow O(\log n)$

$\text{for}(i=1; i < n; i*=2) \rightarrow O(\log n)$

$n=9$
 $i = 9, 4, 2, 1, \cancel{0}$
 $i = 1, 2, 4, 8, \cancel{16}$

$\text{for}(i=0; i < n; i++) \rightarrow n = n^2 \rightarrow O(n^2)$
 $\text{for}(j=0; j < n; j++) \rightarrow n$

$\text{for}(i=0; i < n; i++) ; \rightarrow n = 2n \rightarrow O(n)$
 $\text{for}(j=0; j < n; j++) ; \rightarrow n$

$\text{for}(i=0; i < n; i++) \rightarrow n = n * \log n$
 $\text{for}(j=n; j > 0; j/=2) \rightarrow \log n$
 \downarrow
 $O(n \log n)$

Time complexity

$\text{for}(i=n/2; i \leq n; i++) \rightarrow n$
 $\text{for}(j=1; j+n/2 \leq n; j++) \rightarrow n$
 $\text{for}(k=2; k \leq n; k=k*2) \rightarrow \log n$
Total itr = $n * n * \log n$
 $= n^2 \log n$

$\text{for}(i=n/2; i \leq n; i++) \rightarrow n$
 $\text{for}(j=1; j \leq n; j=2*j) \rightarrow \log n$
 $\text{for}(k=1; k \leq n, k=k*2) \rightarrow \log n$
total itr = $n * \log n * \log n$
 $= n \log^2 n$

Space complexity

- Finding approximate space requirement of the algorithm to execute inside memory

Total space

=

Input space

+

Auxiliary space

↓
space required
to store input

↓
space required to
process the input

e.g. linear search

```
int linearSearch (arr[], key, n) {
    for(i=0; i<n; i++)
        if (key == arr[i])
            return i;
    return -1;
}
```

}

input variables = arr

processing variables = key, n, i

input space = n

Auxiliary space = 3

Total space $\propto n+3$

$\because n \gg 3$

$S(n) = O(n)$

Auxiliary space Complexity

processing vars = k, n, i

Auxiliary space $\propto 3.1$

$AS(n) = O(1)$

Algorithm analysis

Iterative

- loops are used

```
int fact(int num) {  
    int f=1;  
    for(int i=1; i<=num; i++)  
        f *= i;  
    return f;  
}
```

Time \propto no. of iterations of the loop

Time $\propto n$

$$T(n) = O(n)$$

$$AS(n) = O(1)$$

Recursive

- recursion is used

```
int rfact(int num) {  
    if(num == 1)  
        return 1;  
    return num * rfact(num-1);  
}
```

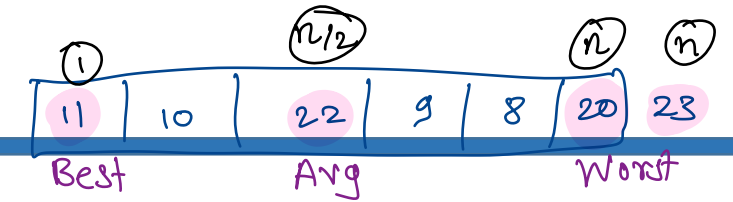
Time \propto no. of recursive calls

Time $\propto n$

$$T(n) = O(n)$$

$$AS(n) = O(n)$$

Searching algorithms analysis



- Time is directly proportional to number of comparisons
- For searching and sorting algorithms, count number of comparisons done

1. Linear search

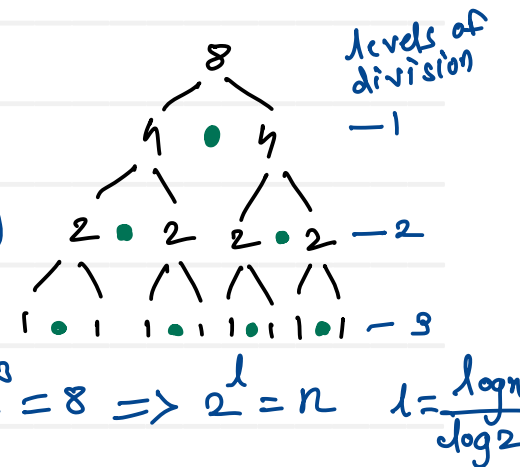
- Best case - if key is found at few initial locations $\rightarrow O(1)$
- Average case - if key is found at middle locations $\rightarrow O(n)$
- Worst case - if key is found at last few locations/
key is not found $\rightarrow O(n)$

$$S(n) = O(1)$$

2. Binary search

- Best case - if key is found at first few levels $\rightarrow O(1)$
- Average case - if key is found at middle levels $\rightarrow O(\log n)$
- Worst case - if key is found at last level/
not found $\rightarrow O(\log n)$

$$S(n) = O(1)$$



Missing Number

Given an array `nums` containing n distinct numbers in the range $[0, n]$, return the only number in the range that is missing from the array.

Example 1:

Input: `nums = [3,0,1]`

Output: 2

Example 2:

Input: `nums = [0,1]`

Output: 2

Example 3:

Input: `nums = [9,6,4,2,3,5,7,0,1]`

Output: 8

- ① find sum of n numbers
- ② find sum of array element
- ③ find diff of both \leftarrow missing number

```
int missingNumber(int nums[]) {  
    int n = nums.length;  
    int nSum = n * (n + 1) / 2;  
    int numsSum = 0;  
    for (int i = 0; i < n; i++)  
        numsSum += nums[i];  
    return nSum - numsSum;  
}
```

Find smallest letter greater than target

You are given an array of characters letters that is sorted in non-decreasing order, and a character target. There are at least two different characters in letters.

Return the smallest character in letters that is lexicographically greater than target. If such a character does not exist, return the first character in letters.

Example 1:

Input: letters = ["c","f","j"], target = "a"

Output: "c"

Example 2:

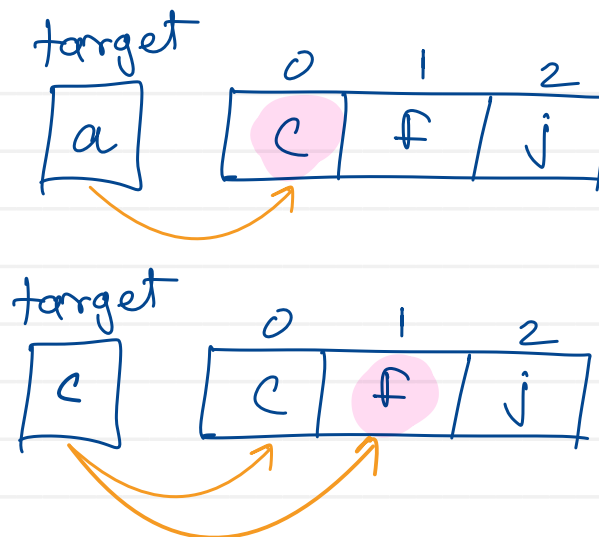
Input: letters = ["c","f","j"], target = "c"

Output: "f"

Example 3:

Input: letters = ["x","x","y","y"], target = "z"

Output: "x"



```
char nextGreatestLetter(char[] letters, char target){
    int n = letters.length;
    for(i=0; i<n; i++){
        if(target < letters[i])
            return letters[i];
    }
    return letters[0];
}
```

Time Complexity
 $T(n) = O(n)$

Find first and last position of element in sorted array

Given an array of integers `nums` sorted in non-decreasing order, find the starting and ending position of a given target value.

If target is not found in the array, return `[-1, -1]`.

You must write an algorithm with $O(\log n)$ runtime complexity.

- ① find key into array
- ② if key is found, find first index of key
- ③ if key is not found, return `[-1, -1]`
- ④ find last index of key

Example 1:

Input: `nums = [5,7,7,8,8,10]`, target = 8

Output: `[3,4]`

Example 2:

Input: `nums = [5,7,7,8,8,10]`, target = 6

Output: `[-1,-1]`

Example 3:

Input: `nums = []`, target = 0

Output: `[-1,-1]`

find first position

```
left = 0, right = nums.length - 1
```

```
first = -1;
```

```
while (left <= right) {
```

```
    mid = (left + right) / 2;
```

```
    if (target == nums[mid]) {
```

```
        first = mid;
```

```
        right = mid - 1;
```

```
    } else if (target < nums[mid]) {
```

```
        right = mid - 1;
```

```
    } else {
```

```
        left = mid + 1;
```

```
    }
```

find last position

```
left = 0, right = nums.length - 1
```

```
last = -1;
```

```
while (left <= right) {
```

```
    mid = (left + right) / 2;
```

```
    if (target == nums[mid]) {
```

```
        last = mid;
```

```
        left = mid + 1;
```

```
    } else if (target < nums[mid]) {
```

```
        right = mid - 1;
```

```
    } else {
```

```
        left = mid + 1;
```

```
    }
```



Thank you!!!

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