



Lecture 06

- Algorithmic Complexity Analysis
- Binary Search Complexity
- Sequential Search Complexity
- Bubble Sort Complexity
- Insertion Sort Complexity
- LinkedList<T>



Algorithmic Complexity

- Time Complexity running time
- Space Complexity memory required to work
- Whenever we talk about the Algorithmic Complexity, it means Time Complexity



- T(n) where n is size of the problem
- Define time taken without depending on the implementation details
- An algorithm will take different amounts of time on the same inputs depending on:
 - Processor speed, Instruction set, Disk speed, Brand of compiler, ...

Estimate efficiency of each algorithm Asymptotically



- T(n) where n is size of the problem
- Define in terms of number of steps each step take constant time to complete
- Example: Addition of two Integers
 - Add two integers digit by digit (or bit by bit) a step in our computational model
 - Adding of two n digit/bit integers takes n steps



- \bullet $T(n) = c \times n$
 - where c is time taken to add two digits or bits
- \bullet On different computers, c can be different
- Computer 1: $T(n) = c_1 \times n$
- Computer 1: $T(n) = c_2 \times n$

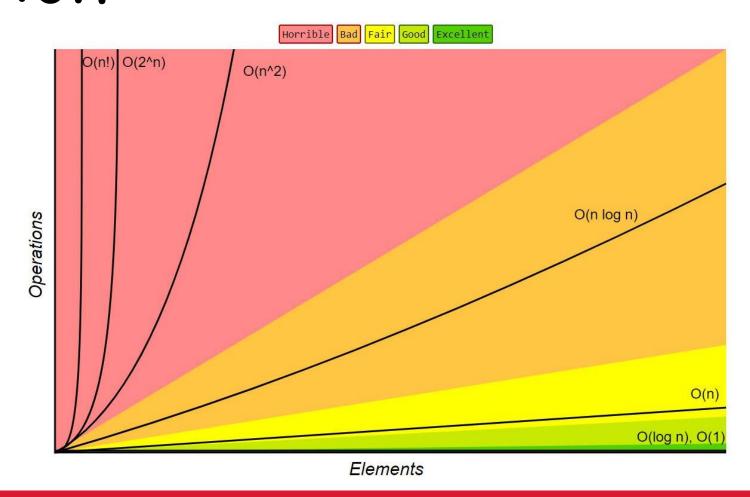


Asymptotic Notation

- Goal of computational complexity is to classify algorithms according to their performances
- Big-O notation upper bound
- Big- Ω notation lower bound
- Big-⊕ notation upper and lower bound



Order of growth in Big-O notation





Big - O Notation

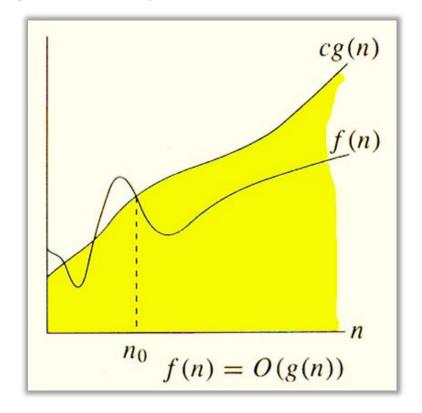
An upper bound of complexity of running time

```
For f(n) and g(n),

f(n) = O(g(n)), if there

exists c and n_0 such that

f(n) \le c \times g(n) \ \forall \ n \ge n_0
```





Big - Ω Notation

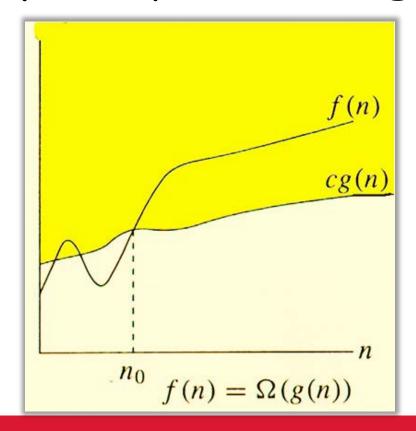
An Lower bound of complexity of running time

```
For f(n) and g(n),

f(n) = \Omega(g(n)), if there

exists c and n_0 such that

f(n) \ge c \times g(n) \ \forall \ n \ge n_0
```





Big - @ Notation

An upper and Lower bound of complexity of

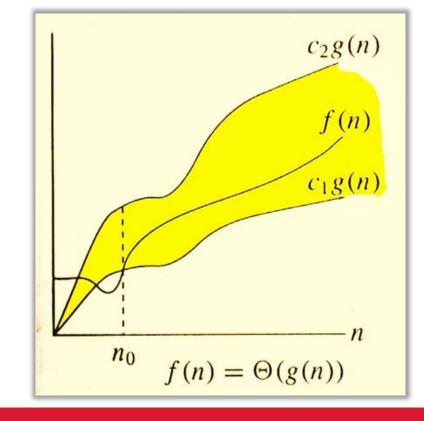
running time

```
For f(n) and g(n),

f(n) = \Theta(g(n)), if there

exists c_1, c_2 and n_0 such that

c_1 \times g(n) \le f(n) \le c_2 \times g(n) \ \forall \ n \ge n_0
```





- Best Case
- Worst Case
- Average Case



Binary Search Analysis

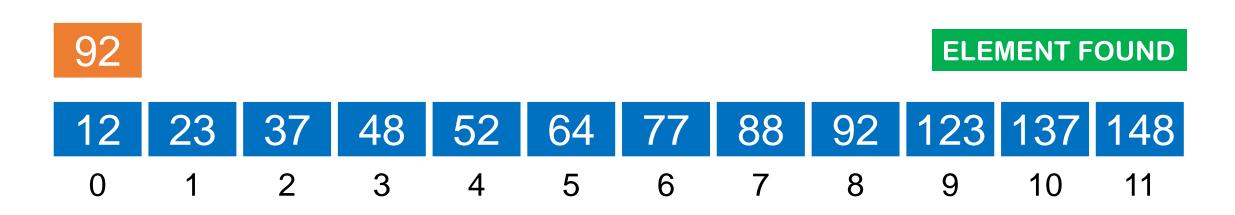
 $O(\log_2 n)$

```
1. lower_{Index} \leftarrow 0; upper_{index} = array.Length - 1; 1 time
                                                                                        ? time
2. while (lower_{index} < upper_{index}) log_2 n times
      middle = \left| \frac{(lower_{index} + upper_{index})}{2} \right| \log_2 n \text{ times}
     if(array[middle] > element) then upper_{index} = middle - 1
                                                                                        log<sub>2</sub> n times
     else\ if(array[middle] < element) lower_{index} = middle + 1
                                                                                        log<sub>2</sub> n times
      else return middle log<sub>2</sub> n times
7. return - 1 1 time
         Time Analysis = 1 + \log_2 n + 1
         Time Analysis = 2 + 5 \times \log_2 n
```



Sequential Search

 Start searching from the first element and go till the end of list or array one by one





Sequential Search Input: an integer array

Input: an integer array search integer element
Output: Index of element

ELEMENT FOUND

```
1. for(int \ i = 0; i < array. Length; i + +) n times

2. if(array[i] == element) n times

3. return \ i; 0 or 1 times

4. return \ -1 0 or 1 times
```

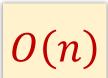
 12
 23
 37
 48
 52
 64
 77
 88
 92
 123
 137
 148

 0
 1
 2
 3
 4
 5
 6
 7
 8
 9
 10
 11



Sequential Search Analysis

```
Time Analysis = n + n + 1
Time Analysis = 2n + 1
```



Input: an integer array



Bubble Sort Analysis

```
Output: Sorted array
1. temp \leftarrow 0
2. for (i \leftarrow array. Length - 1; i > 0; i - -) (n-1) times
      for (j \leftarrow 0; j < i; j + +) (n-1) + (n-2) + (n-3) + \dots + 3 + 2 + 1 times
          if(array[j] > array[j+1]) (n-1) + (n-2) + (n-3) + \dots + 3 + 2 + 1 times
             temp \leftarrow array[j+1](n-1) + (n-2) + (n-3) + \dots + 3 + 2 + 1 times
             array[j+1] \leftarrow array[j](n-1) + (n-2) + (n-3) + \dots + 3 + 2 + 1 times
6.
             array | j | \leftarrow temp (n-1) + (n-2) + (n-3) + \dots + 3 + 2 + 1 times
```



Bubble Sort Analysis

$$O(n^2)$$

$$1 + 2 + 3 + \dots + n = \frac{n(n+1)}{2}$$

$$1 + 2 + 3 + \dots + (n-1) = \frac{(n-1)(n-1+1)}{2} = \frac{n(n-1)}{2}$$

$$Time\ Analysis = 1 + (n-1) + 5 \times \frac{n(n-1)}{2} = n + \frac{5}{2}n^2 - \frac{5}{2}n$$

$$Time\ Analysis = \frac{5}{2}n^2 - \frac{3}{2}n$$



Insertion Sort Analysis

```
Input: an integer array
1. i \leftarrow 1
                                                       Output: Sorted array
2. while (i < array. Length) (n-1) times
      temp \leftarrow array[i] (n-1) times
     j \leftarrow i - 1 \quad (n-1) \text{ times}
      while (j \ge 0 \text{ and } array[j] > temp) 1 + 2 + \cdots + (n-1) times
5.
         array[j+1] \leftarrow array[j] \ 1+2+\cdots+(n-1) \ times
         j \leftarrow j - 1 \ 1 + 2 + \dots + (n-1) \ times
      array[j+1] \leftarrow temp (n-1) times
8.
9.
      i \leftarrow i + 1 (n-1) times
```



Insertion Sort Analysis

$$1 + 2 + 3 + \dots + (n - 1) = \frac{n(n - 1)}{2}$$

$$Time\ Analysis = 1 + 5 \times (n - 1) + 3 \times \frac{n(n - 1)}{2}$$

$$Time\ Analysis = 5n - 4 + \frac{3}{2}n^2 - \frac{3}{2}n$$

$$Time\ Analysis = \frac{3}{2}n^2 + \frac{7}{2}n - 4$$

 $O(n^2)$



LinkedList<T>

- A general-purpose generic linked list
- Nodes of type <u>LinkedListNode<T></u>
 - Properties: Previous, Next and Value
- Properties: count, first, last



LinkedList<T>

Methods

- AddAfter
- AddBefore
- AddFirst
- AddLast
- Clear
- Contains

- Find
- FindLast
- Remove
- RemoveFirst
- RemoveLast



Reference and Reading Material

- Algorithmic Complexity <u>Link</u>, <u>Link</u>
- LinkedList<T>: Link, Link