Big O notation is used in Computer Science to describe the performance or complexity of an algorithm. Big O specifically describes the worst-case scenario, and can be used to describe the execution time required or the space used (e.g. in memory or on disk) by an algorithm.

#### 0(1)

O(1) describes an algorithm that will always execute in the same time (or space) regardless of the size of the input data set.

```
bool IsFirstElementNull(IList<string> elements)
{
   return elements[0] == null;
}
```

#### O(N)

O(N) describes an algorithm whose performance will grow linearly and in direct proportion to the size of the input data set. The example below also demonstrates how Big O favours the worst-case performance scenario; a matching string could be found during any iteration of the for loop and the function would return early, but Big O notation will always assume the upper limit where the algorithm will perform the maximum number of iterations.

```
bool ContainsValue(IList<string> elements, string value)
{
   foreach (var element in elements)
   {
     if (element == value) return true;
```

```
}
  return false;
O(N^2)
O(N<sup>2</sup>) represents an algorithm whose performance is directly proportional to the square of the size of the input data set. This is
common with algorithms that involve nested iterations over the data set. Deeper nested iterations will result in O(N3), O(N4)
etc.
bool ContainsDuplicates(IList<string> elements)
  for (var outer = 0; outer < elements.Count; outer++)</pre>
  {
     for (var inner = 0; inner < elements.Count; inner++)</pre>
     {
        // Don't compare with self
        if (outer == inner) continue;
        if (elements[outer] == elements[inner]) return true;
```

```
}
   }
  return false;
O(2<sup>N</sup>)
O(2<sup>N</sup>) denotes an algorithm whose growth doubles with each addition to the input data set. The growth curve of an O(2<sup>N</sup>)
function is exponential - starting off very shallow, then rising meteorically. An example of an O(2^N) function is the recursive
calculation of Fibonacci numbers:
int Fibonacci(int number)
  if (number <= 1) return number;
  return Fibonacci(number - 2) + Fibonacci(number - 1);
```

### Legend



### **Data Structure Operations**

Data Structure	Time Comp	Time Complexity								
	Average				Worst				Worst	
	Access	Search	Insertion	Deletion	Access	Search	Insertion	Deletion		
Array	0(1)	O(n)	O(n)	O(n)	O(1)	O(n)	O(n)	O(n)	O(n)	
Stack	O(n)	O(n)	0(1)	0(1)	O(n)	O(n)	O(1)	0(1)	O(n)	
Singly-Linked List	O(n)	O(n)	0(1)	0(1)	O(n)	O(n)	0(1)	0(1)	O(n)	
Doubly- Linked List	O(n)	O(n)	O(1)	0(1)	O(n)	O(n)	0(1)	0(1)	O(n)	

Space Complexity

**Data Structure** Time Complexity

	Average				Worst				Worst
	Access	Search	Insertion	Deletion	Access	Search	Insertion	Deletion	
Skip List	O(log(n))	O(log(n))	O(log(n))	O(log(n))	O(n)	O(n)	O(n)	O(n)	O(n log(n))
Hash Table	-	0(1)	0(1)	0(1)	-	O(n)	O(n)	O(n)	O(n)
Binary Search Tree	O(log(n))	O(log(n))	O(log(n))	O(log(n))	O(n)	O(n)	O(n)	O(n)	O(n)
Cartesian Tree	-	O(log(n))	O(log(n))	O(log(n))	-	O(n)	O(n)	O(n)	O(n)
B-Tree	O(log(n))	O(n)							
Red-Black Tree	O(log(n))	O(n)							

Data Structure	Time Complexity								
	Average				Worst				Worst
	Access	Search	Insertion	Deletion	Access	Search	Insertion	Deletion	
Splay Tree	-	O(log(n))	O(log(n))	O(log(n))	-	O(log(n))	O(log(n))	O(log(n))	O(n)
AVL Tree	O(log(n))	O(log(n))	O(log(n))	O(log(n))	O(log(n))	O(log(n))	O(log(n))	O(log(n))	O(n)

Space

### **Array Sorting Algorithms**

Time Complexity

Algorithm

Algorithm	Time Complexi	ty		Space Complexity
	Best	Average	Worst	Worst
Quicksort	O(n log(n))	O(n log(n))	O(n^2)	O(log(n))
Mergesort	O(n log(n))	O(n log(n))	O(n log(n))	O(n)
Timsort	O(n)	O(n log(n))	O(n log(n))	O(n)
Heapsort	O(n log(n))	O(n log(n))	O(n log(n))	0(1)
Bubble Sort	O(n)	O(n^2)	O(n^2)	0(1)
Insertion Sort	O(n)	O(n^2)	O(n^2)	0(1)
Selection Sort	O(n^2)	O(n^2)	O(n^2)	0(1)

Snace Complexity

Algorithm	Time Comple	exity		Space Complexity
	Best	Average	Worst	Worst
Shell Sort	O(n)	O((nlog(n))^2)	O((nlog(n))^2)	O(1)
Bucket Sort	O(n+k)	O(n+k)	O(n^2)	O(n)
Radix Sort	O(nk)	O(nk)	O(nk)	O(n+k)

# **Graph Operations**

Node / Edge Management	Storage	Add Vertex	Add Edge	Remove Vertex	Remove Edge	Query
Adjacency list	O( V + E )	0(1)	0(1)	O( V  +  E )	O( E )	0( V )
Incidence list	O( V + E )	0(1)	0(1)	O( E )	O( E )	O( E )
Adjacency matrix	O( V ^2)	0( V ^2)	0(1)	0( V ^2)	0(1)	0(1)
Incidence matrix	O( V  ·  E )	$O( V  \cdot  E )$	$O( V  \cdot  E )$	O( V  ·  E )	$O( V  \cdot  E )$	O( E )

# **Heap Operations**

Type

**Time Complexity** 

	Heapify	Find Max	Extract Max	Increase Key	Insert	Delete	Merge
Linked List (sorted)	_	O(1)	0(1)	O(n)	O(n)	O(1)	O(m+n)
Linked List (unsorted)	_	O(n)	O(n)	0(1)	0(1)	0(1)	0(1)
Binary Heap	O(n)	0(1)	O(log(n))	O(log(n))	O(log(n))	O(log(n))	O(m+n)
Binomial Heap		0(1)	O(log(n))	O(log(n))	0(1)	O(log(n))	O(log(n))
Fibonacci Heap	_	0(1)	O(log(n))	0(1)	0(1)	O(log(n))	0(1)

## **Big-O Complexity Chart**

