

# DATA STRUCTURES & ALGORITHMS

ALGORITHM ANALYSIS

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# On the last episode lecture



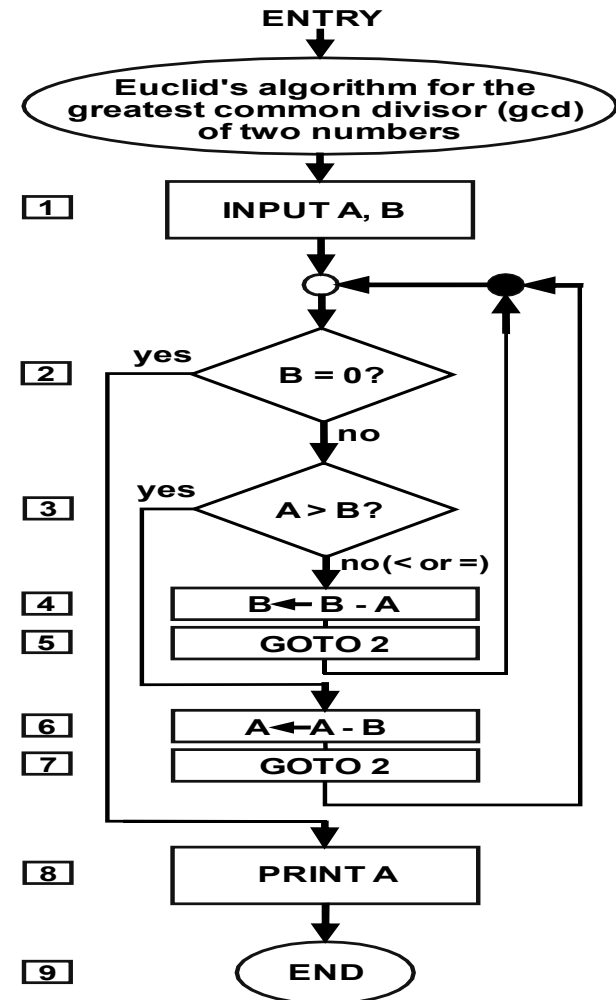
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So, we have **two implementations** for **the same ADT List**. Can we just **ignore Linked List** and **always use Array List** in our coding?

- **Algorithms**
- Efficiency of Algorithms
  - Experiments approach and its issues
  - Number of operations
  - Asymptotic analysis – Big-Oh notation

# Algorithms

An algorithm is a **step-by-step procedure** for **performing some task** in a **finite** amount of **time**



This example shows the Euclid's algorithm depicted using a flow chart

Image by [Wikipedia](#)

# An example Task

Given an integer  $N$ , **compute** the **sum** of all integer **from 1 to  $N$** , including  $N$



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How many different algorithms can you think of?

# 3 algorithms to compute the sum

## Algorithm 1

Add 1, 2, 3... n to sum

```
sum = 0
for i = 1 to N
    sum = sum + i
```

## Algorithm 2

Add 1, (1+1),  
(1+1+1), ...  
(1+1+1... +1) to sum

```
sum = 0
for i = 1 to N
    for j = 1 to i
        sum = sum + 1
```

## Algorithm 3

Use the sum of consecutive numbers formula\*

```
sum =
    N * (N + 1) / 2
```

\* Sum of consecutive numbers



Which algorithm is more efficient?

- Algorithms
- **Efficiency of Algorithms**
  - Experiments approach and its issues
  - Number of operations
  - Asymptotic analysis – Big-Oh notation

# What does “efficiency” mean?

- **Time:** How fast an algorithm runs
- **Space:** How much memory an algorithm needs
- **Usually**, we care more about time than space



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Why?



# Next

How can we  
**analyze** an  
**algorithm** to  
see if it runs  
**faster** than  
another one?



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Hey! Computers are **very fast** nowadays. So who care?

- Algorithms
- **Efficiency of Algorithms**
  - **Experimental approach and its issues**
  - Number of operations
  - Asymptotic analysis – Big-Oh notation

# Approach 1 – Experiments

We **implement** each of algorithms, **then measure** how long they run

```
static void sum1(long n)
{
    Stopwatch sw =
        Stopwatch.StartNew();

    long sum = 0;
    for (long i = 1;
        i <= n; i++)
        sum += i;

    Console.WriteLine(
        "Run {1}ns.",
        ElapsedNanoSecond(sw));
}
```

*Same for* sum3(long n)

```
static void sum2(long n) {
    Stopwatch sw =
        Stopwatch.StartNew();

    long sum = 0;
    for (long i = 1;
        i <= n; i++) {
        for (long j = 1;
            j <= i; j++)
            sum += 1;
    }

    Console.WriteLine(
        "Run {1}ns.",
        ElapsedNanoSecond(sw));
}
```

The code is just for illustrative purpose

# Approach 1 – Experiments

<pre>static void Main(string[] args) {     sum1(1000);     sum1(10000);     sum1(100000);      sum2(1000);     sum2(10000);     sum2(100000);      sum3(1000);     sum3(10000);     sum3(100000); }</pre>	<pre>Run 2000ns. Run 13500ns. Run 132200ns.  Run 822100ns. Run 81774400ns. Run 8129553800ns.  Run 100ns. Run 100ns. Run 100ns.</pre>
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What issues may this approach have?

# Looking for improvements



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If only there is a solution that:

1. **Not depend** on the **hardware** and **software**
2. **Not need** to **implement algorithms** first
3. **All possible inputs** are considered
4. Evaluating the **relative efficiency** only

- Algorithms
- **Efficiency of Algorithms**
  - Experimental approach and its issues
  - **Number of operations**
  - Asymptotic analysis – Big-Oh notation

## Approach 2 - Key idea

Instead,  
**count** the  
**number of**  
**operations**  
in the  
algorithm



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# Primitive Operations

A primitive operation has **constant execution time**, including:

1. Assign a value to a variable
2. Follow an object reference
3. Perform an arithmetic operation
4. Compare two numbers
5. Access a single element of an array by index
6. Call a method
7. Return from a method



# Quiz

Count the number of primitive operations for the algorithm below

*Counting sum of number from 1 to N*

*Input: N*

```
sum = 0
```

```
loop i from 1 to N
```

```
    sum = sum + i
```



Do we need to **implement** the algorithm **before** being able to **count**?

# A challenging question

There are one elephant  
and one mouse

- The elephant weighs  
5880 kg
- The mouse weighs  
6.5g

What is their total weight?

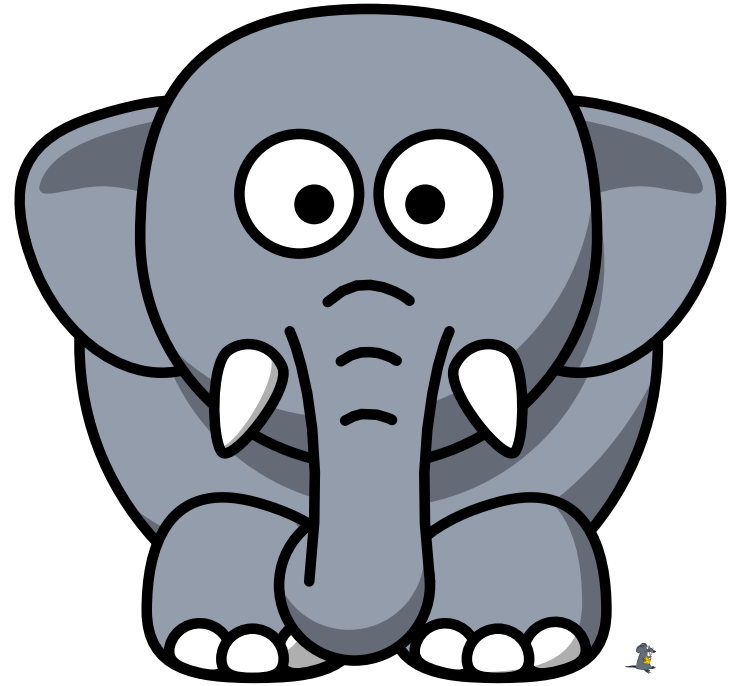


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# Basic Operations

Basic operations are the **most significant contributor** to its total time. The following operations are ~~*not basic*~~:

- ~~1. *Assign a value to a variable*~~
- ~~2. *Follow an object reference*~~
3. Perform an arithmetic operation
  - ~~• *Operations that control the loop*~~
4. Compare two numbers
5. Access a single element of an array by index
- ~~6. *Call a method*~~
- ~~7. *Return from a method*~~

Count **basic operations only!**

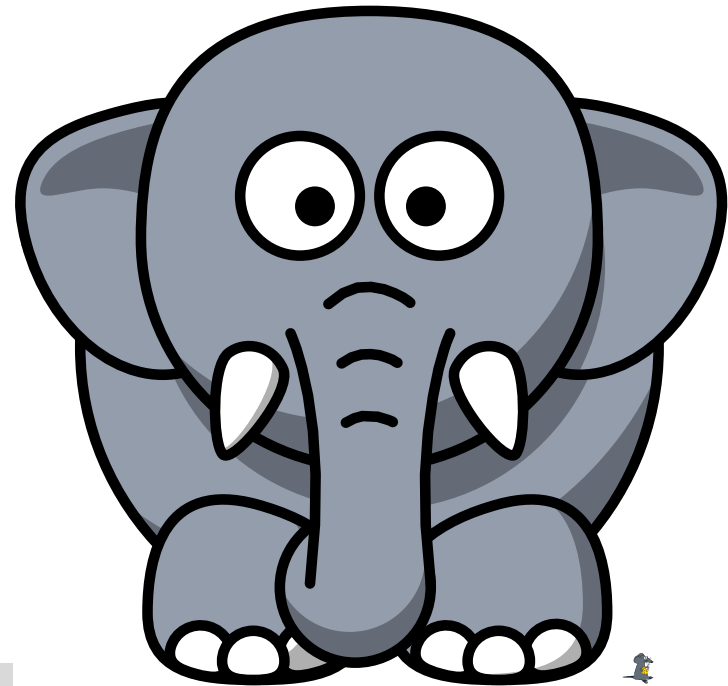


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# Basic Operations

Let's re-count only the number of basic operations for the algorithm below

*Counting sum of number from 1 to N*

*Input: N*

```
sum = 0
```

```
for i = 1 to N
```

```
    sum = sum + i
```



Have we considered **all** the **possible inputs** in our counting?

# Quiz

Count the number of basic operations for the algorithm below

*Find a value in an array*

*Input: an array and a value*

```
loop i from 0 to array's length - 1
    if (arr[i] == value)
        return i
return -1
```



Does the number of operations **depend on** the **input values**?

# Best Case and Worst Case

For some algorithms, execution **time depends not only on the size** of the data set, **but also** the data **values**

**Best case** – the number of basic operations **never less than**

- Last example: the best case is 1

**Worst case** – the number of basic operations **never more than**

- Last example: the worst case is  $N$

**Average cases** – the number of basic operations **averaged over all possible inputs**



Can you guess:  
which one is more  
of our interest?

The **number** of basic **operations** in the **worst cases** of two algorithms solving the **same problem** are

Algorithm 1

$$7n + n \log n + 2$$

Algorithm 2

$$2n + 1 + n^2$$



In general,  
which one is  
faster?

- Algorithms
- **Efficiency of Algorithms**
  - Experimental approach and its issues
  - Number of operations
  - **Asymptotic analysis – Big-Oh notation**



# Asymptotic Analysis

Self study

Taking a “**big-picture**” approach

**Estimate** the number of basic operations executed, rather than the **exact** number

Focus on the **growth rate**

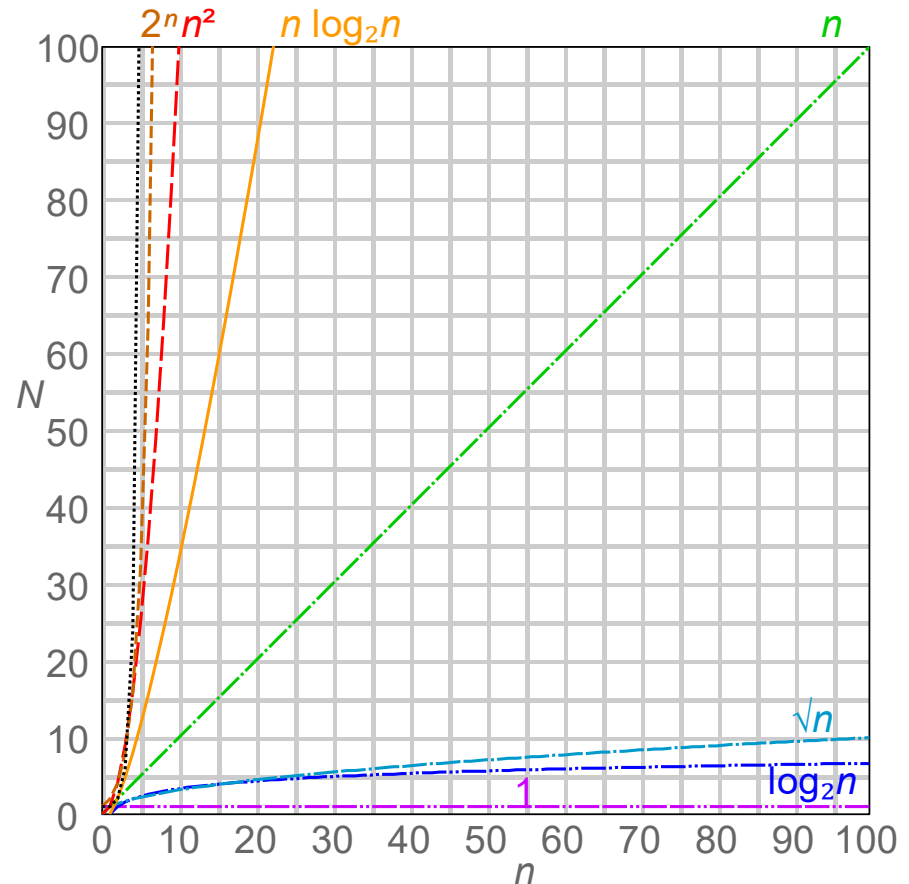


Image by Cmglee - Own work, CC BY-SA 4.0, [Wikipedia](https://en.wikipedia.org/wiki/File:Growth_of_functions.png)

# Big-Oh Notation

Self study

- Let  $f(n)$  and  $g(n)$  be 2 functions mapping positive integers to positive real numbers
- A function  $f(n)$  is of order at most  $g(n)$ , i.e.,  $f(n)$  is  $O(g(n))$  if:
  - There is a real constant  $c > 0$ , and a positive integer  $N$
  - Such that  $f(n) \leq c \cdot g(n)$  for all  $n \geq N$
  - It means,  $c \cdot g(n)$  is an **upper bound** on  $f(n)$  when  $n$  is **large enough**

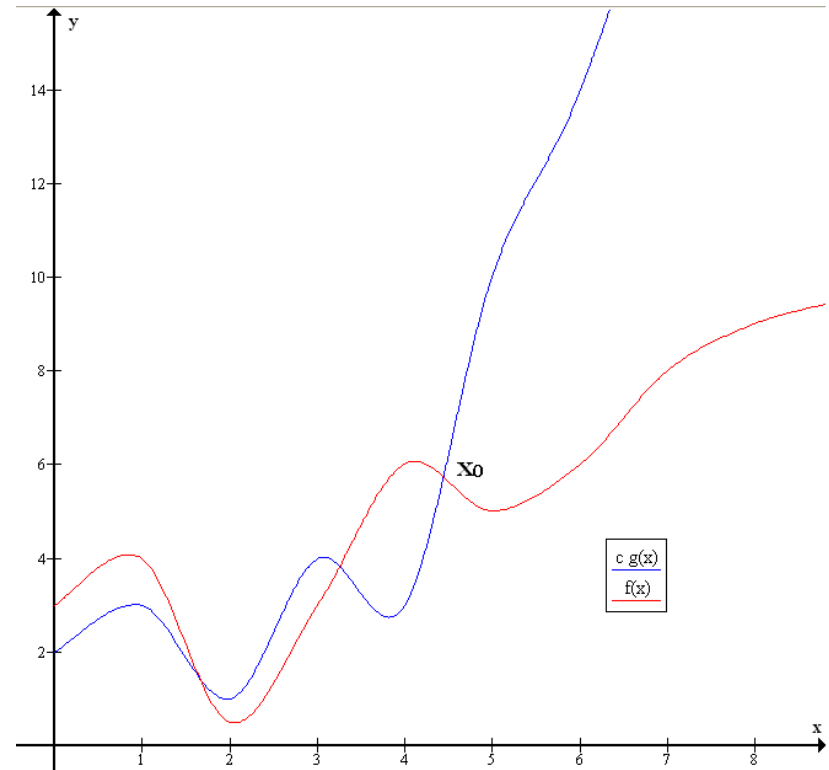


Image by Fede\_Reghe, [Wikipedia](#),

# Properties of Big-Oh Notation

Self study

$$O(k(g(n))) = O(g(n)) \text{ for a constant } k$$

$$O(g_1(n)) + O(g_2(n)) = O(g_1(n) + g_2(n))$$

$$O(g_1(n)) \times O(g_2(n)) = O(g_1(n) \times g_2(n))$$

$$O(g_1(n) + g_2(n) + g_3(n)) = O(\max(g_1(n), g_2(n), g_3(n)))$$

$$O(\max(g_1(n), g_2(n), g_3(n))) = \max(O(g_1(n)), O(g_2(n)), O(g_3(n)))$$

# Properties of the Big-Oh

Allows us to **ignore** the **constant factors**

$$O(k(g(n))) = O(g(n)) \text{ for a constant } k$$

For example:

$$t(n) = n \text{ is } O(n)$$

$$t(n) = 2n \text{ is also } O(n)$$

$$t(n) = 5n \text{ is also } O(n)$$

$$t(n) = 8n \text{ is also } O(n)$$

*Hint:* What is the elephant? What are mouses?

# Properties of the Big-Oh

Allows us to **ignore** the **lower-order terms**

$$\begin{aligned} O(g_1(n) + g_2(n) + g_3(n)) &= O(\max(g_1(n), g_2(n), g_3(n))) \\ &= \max(O(g_1(n)), O(g_2(n)), O(g_3(n))) \end{aligned}$$

For example:

$$t(n) = n^4 + 2n^3 + 1 \text{ is } O(n^4)$$

$$t(n) = 2n^3 + 3n + 4 \text{ is } O(n^3)$$

$$t(n) = n^2 + 2\log n \text{ is } O(n^2)$$

$$t(n) = 2n + 3\log n \text{ is } O(n)$$

$$t(n) = 4\log n + 4 \text{ is } O(\log n)$$

*Hint:* What is the elephant? What are mice?

# Quiz

What is the time efficiency of the following code?

*Calculate the sum from 1 to n*

```
sum = 0
loop i from 1 to n
  loop j from 1 to i
    sum = sum + 1
```

# Properties of the Big-Oh

Allows us to **ignore** the **less complex segments** among the **consecutive ones** in a program

```
for (int i = 0; i < n; i++)  
    arr[i] = i;
```

Segment S1

```
for (int i = 0; i < n; i++)  
    for (int j = 0; j < n; j++)  
        arr[j] += arr[j] + i + j;
```

Segment S2

# Properties of the Big-Oh

Allows us to **ignore** the **less complex branches** in an *if/else* statement

```
if (condition)
    S1
else
    S2
```

Complexity of the program is:  
 $O(\text{condition}) + \max(O(S1), O(S2))$



# Quiz

What is the time efficiency of the following code?

```
if (n % 2 == 0)
    for (i = 1; i <= n; i++)
        for (j = 1; j <= 10; j++)
            sum = sum + j;
    for (i = 1; i <= n/2; i++)
        sum = sum + i;

else
    for (i = 1; i <= n; i++)
        for (j = 1; j <= n; j++)
            sum = sum + j;
```

# Question

An algorithm takes 0.5 ms for input size 100. How long will it take for input size 500 if the running time is the following (assume low-order terms are negligible):

- a)  $O(N)$
- b)  $O(N \log N)$
- c)  $O(N^2)$
- d)  $O(N^3)$

# 7 most common functions

The following table lists 7 most growth-rate functions and the estimated time required to process **one million** items

Growth-Rate function	Time
$O(1)$	Extremely fast (even with more items)
$O(\log n)$	0.0000199 seconds
$O(n)$	1 second
$O(n \log n)$	19.9 seconds
$O(n^2)$	11.6 days
$O(n^3)$	31,709.8 years
$O(2^n)$	$10^{301,016}$ years

# Example

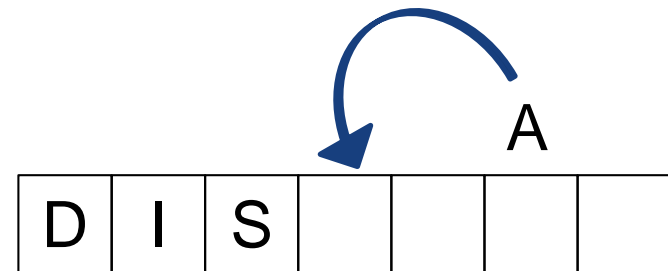
Self study

What is the complexity of implement ADT List method *Add()* using an array?

```
public void Add(string newElement) {
    arr[numElements] = newElement;
    numElements++;
```

**EnsureCapacity();**

```
}
private void EnsureCapacity() {
    int capacity = arr.Length - 1;
    if (numElements >= capacity) {
        // Replace with a new bigger array
        int newCapacity = capacity * 2;
        string[] newArr = new string[newCapacity];
        arr.CopyTo(newArr, 0);
        arr = newArr;
    }
}
```



Most of the time  $O(1)$   
Rarely worst-case  $O(n)$

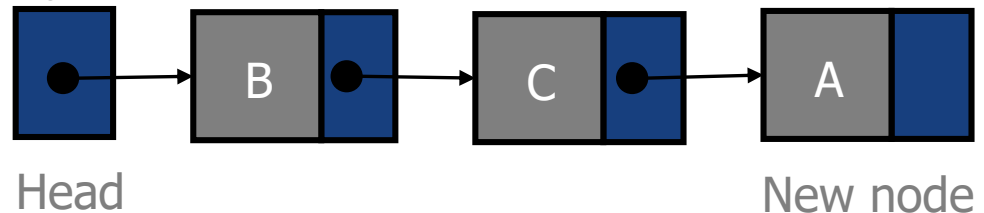
# Example

What is the complexity of implement ADT List method *Add()* using a Linked List?

```
public void Add(string element)
{
    Node newNode = new Node(element);

    if (numElements == 0) {
        Head = newNode;
    }
    else
    {
        Node lastNode = GetLastNode();
        lastNode.Next = newNode;
    }

    numElements++;
}
```

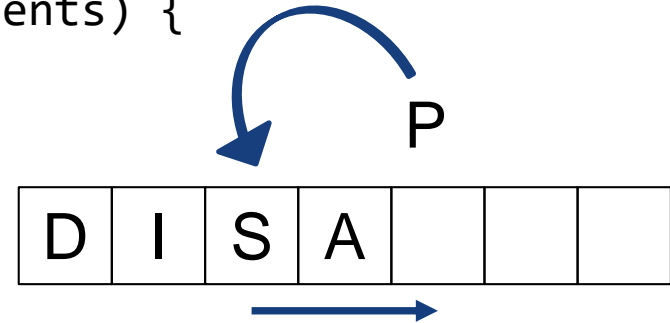


Most of the time  $O(n)$   
Rarely best-case  $O(1)$

# Quiz

What is the complexity of implement List method *Insert()* to a given position using an array?

```
public void Insert(int index, string newElement) {  
    // Allow inserting to the end  
    if (index >= 0 && index <= numElements) {  
        if (index < numElements)  
            MakeRoom(index);  
        arr[index] = newElement;  
        numElements++;  
        EnsureCapacity();  
    } // else Invalid index  
}
```



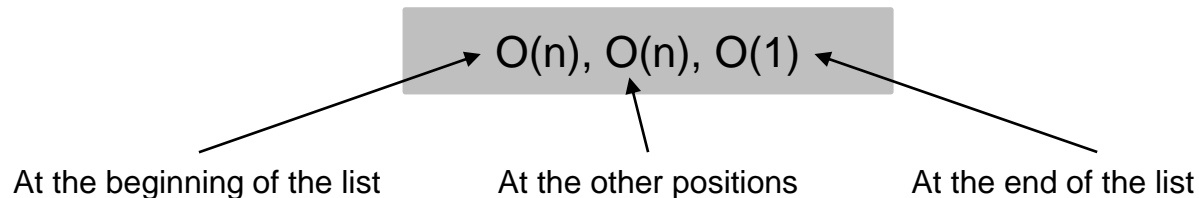
```
// Shift entries toward the end of the array  
private void MakeRoom(int index) {  
    for (int i = numElements;  
         i >= index; i--)  
        arr[i + 1] = arr[i];  
}
```

Most of the time:  
Insert to pos other than end  $O(n)$   
Insert to the end  $O(1)$

# ADT List time efficiency

Self study

Operation	Using Array	Using Linked List	Using Linked List with Tail *
Add(e)	$O(1)$	$O(n)$	$O(1)$
Insert(pos, e)	$O(n)$ , $O(n)$ , $O(1)$	$O(1)$ , $O(n)$	$O(1)$ , $O(n)$ , $O(1)$
RemoveAt(pos)	$O(n)$ , $O(n)$ , $O(1)$	$O(1)$ , $O(n)$	$O(1)$ , $O(n)$ , $O(1)$
Replace(pos, e)	$O(1)$	$O(1)$ , $O(n)$	$O(n)$
GetAt(pos)	$O(1)$	$O(1)$ , $O(n)$	$O(1)$ , $O(n)$ , $O(1)$
Contains(e)	$O(n)$	$O(n)$	$O(n)$



\* Advanced students may read more about Linked List with Tail in sections 14.20-14.24 [Carrano 2016]

# Question

When implementing apps **adding / removing** to / from the **beginning** of the list **frequently**, which List implementation should we choose?

In practice, at the beginning we just **pick an implementation**, e.g., Array List, and **optimize later** when really needed



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- Data structures and abstractions with Java, 4ed – Chapter 4, The efficiency of Algorithms, *Frank M. Carrano and Timothy M. Henry*