

Session 02 - Algorithm Efficiency

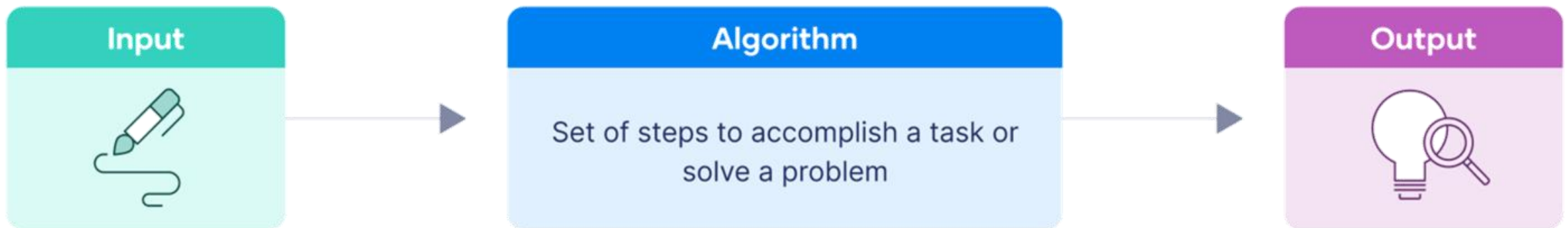
Instructor:

Dr. LE Thanh Tung

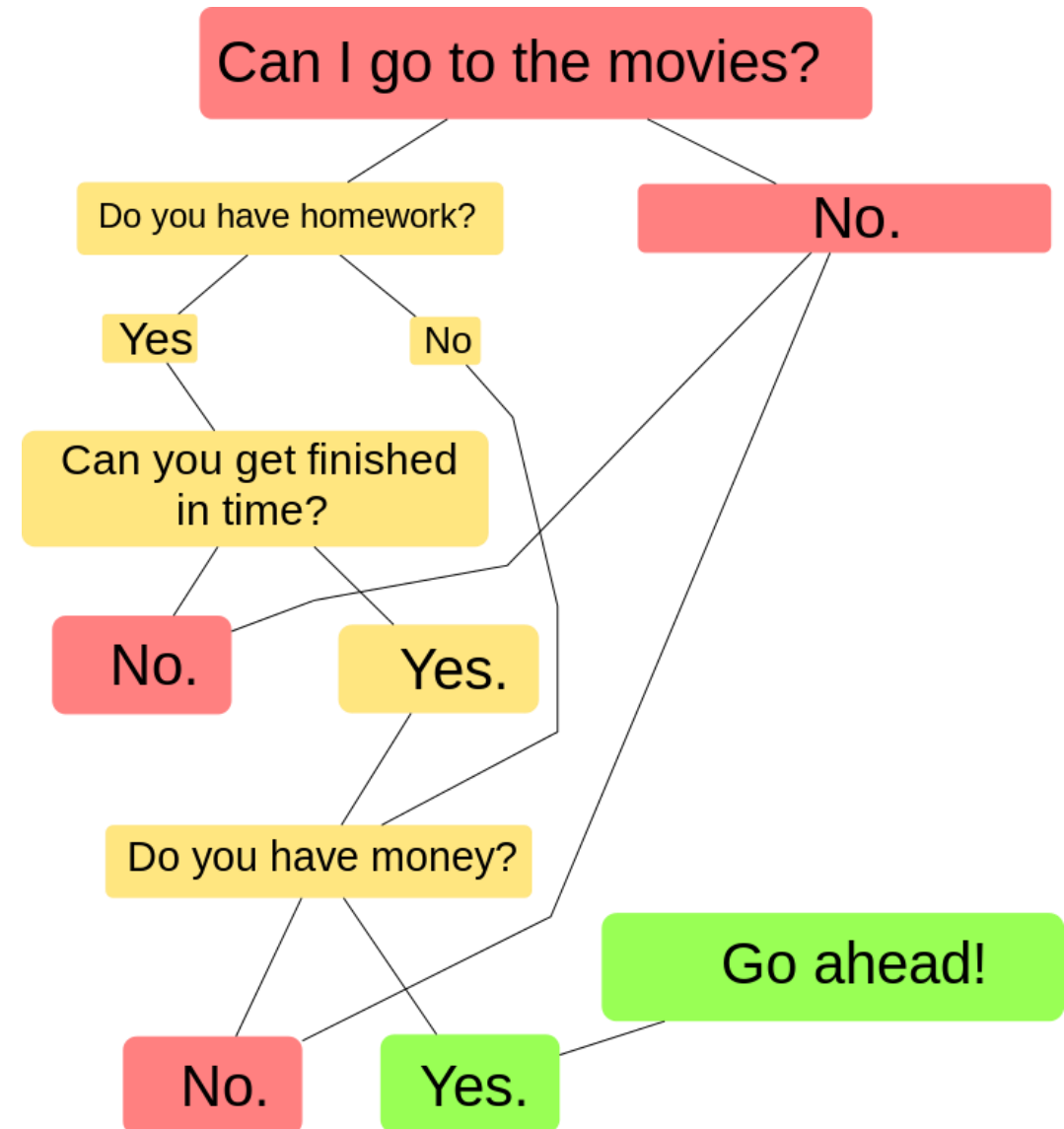
- 1 A review on algorithm
- 2 Analysis and Big-O notation
- 3 Algorithm efficiency

A review on Algorithm

- What is Algorithm?
 - A strictly defined **finite** sequence of **well-defined** steps (statements, often called instructions or commands)
 - that provides the solution to a problem.

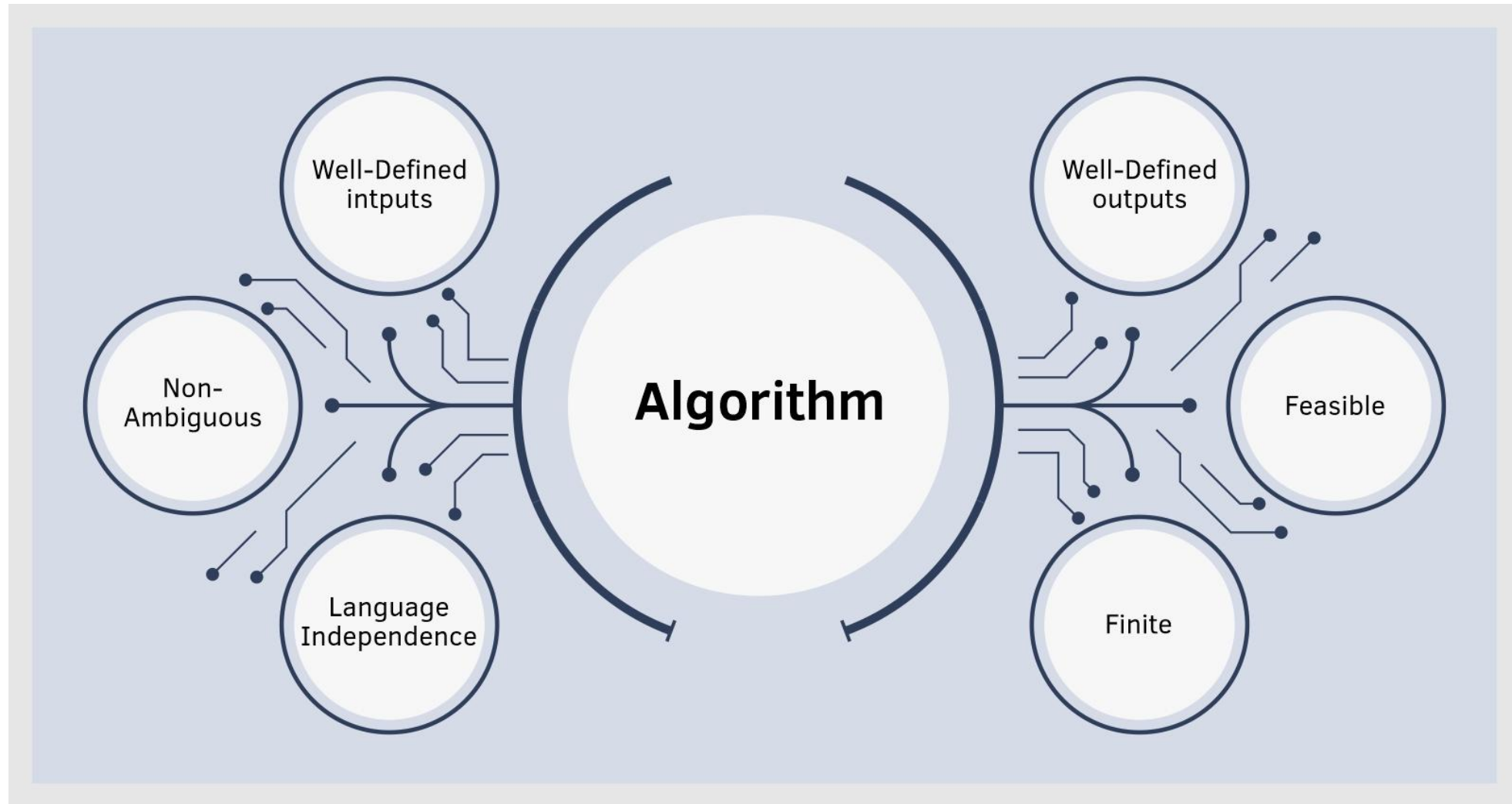


- Example:
 - Problem: a child wonder whether they can go to the movies in case of having a homework today.



- Why should we study algorithm?
 - To understand the basic idea of the problem.
 - To find an approach to solve the problem.
 - To improve the efficiency of existing techniques.
 - To understand the basic principles of designing the algorithms.
 - To break down problems and conceptualize solutions in terms of discrete steps

- Algorithm's Characteristics:



- Algorithm's Characteristics:
 - **Finiteness:** for any input, the algorithm must terminate after a finite number of steps.
 - **Correctness:** always correct. Give the same result for different run time.
 - **Definiteness:** all steps of the algorithm must be precisely defined.
 - **Effectiveness:** It must be possible to perform each step of the algorithm correctly and in a finite amount of time.



**Brute Force
Algorithm**



**Divide and
Conquer
Algorithm**



**Dynamic
Programming
Algorithm**



**Greedy
Approach**



**Backtracking
Algorithm**

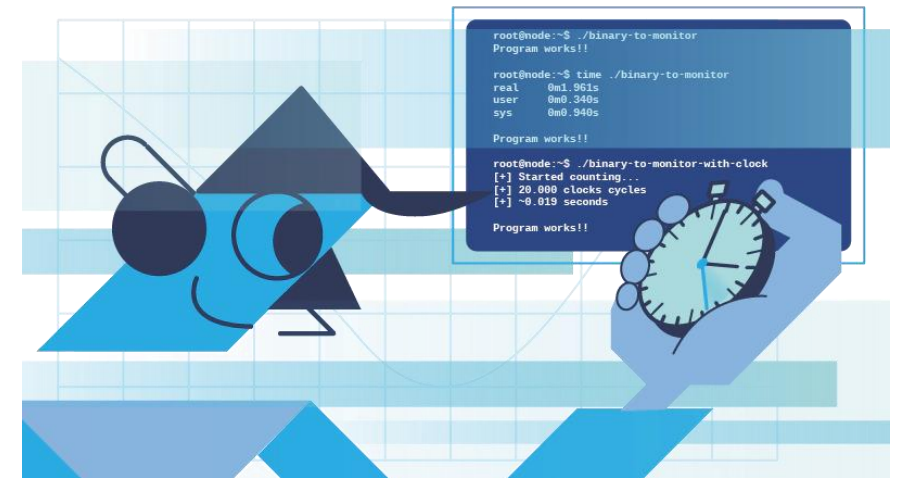
- What kind of problems are solved by Algorithm?
 - Sorting
 - Searching
 - String matching
 - Graph problems
 - Combinatorial problems
 - Geometric problems
 - Numerical problems

- The two factors of **Algorithm Efficiency** are:
 - **Time Factor:** Time is measured by counting the number of key operations.
 - **Space Factor:** Space is measured by counting the maximum memory space required by the algorithm.

- Nowadays, the amount of extra space required by an algorithm is typically not of as much concern
 - In most problems, we can achieve much more spectacular progress in speed than in space
- Concentrate on **time efficiency**, but analytical framework in this course is applicable to analyzing space efficiency as well

- Can we compare two algorithms (in time factor) like this?
 - Implement those algorithms (into programs)
 - Calculate the execution time of those programs
 - Compare those two values of time measurement.

Is it fair in this measuring process?



- Comparison of algorithms should focus on **significant differences** in efficiency
 - Difficulties with comparing programs instead of algorithms
 - How are the algorithms coded?
 - What computer should you use?
 - What data should the programs use?
- Employ mathematical techniques that **analyze algorithms independently** of specific implementations, computers, or data

- Time complexity is measured by counting the **primitive operations** for the computation that the algorithm needs to perform.
- Key operation is to contribute the most to the total running time of an algorithm
 - Comparisons
 - Assignments
- Derive an algorithm's time requirement as a function of the **problem size**

- Almost all algorithms run **longer on larger inputs**
- For example:
 - Sorting arrays: $A1 = \{12, 1, 3\}$
 - Sorting arrays: $A2 = \{88, 12, 3, 19, 32, 9, 1, 3, 45, 17, 89, 12, 34, 52, 61, 41, 24, 98, 19, 38\}$
- Algorithm's efficiency is investigated as a function of some parameter **n** indicating the algorithm's **input size**

- **Straightforward:** problems dealing with lists (e.g., sorting, searching, min, max, ...)
 - n is the size of the list
- **Not straightforward:**
 - Computing the product of two matrix
 - Checking primality of a positive integer n
 - Spell-checking a document

- Traversal of linked nodes – example:

```
void printList(Node* pHead){  
    Node* pCur = pHead;           ← 1      assignment  
    while (pCur != nullptr){      ← n + 1    comparisons  
        cout << pCur->key << endl; ← n + 1    writes  
        pCur = pCur->next;        ← n + 1    assignments  
    }  
}
```

- Assignment: a time units. Comparison: c time units.
- Write: w time units.
- Displaying data in linked list of **n nodes** requires **time proportional to n**

- Nested loops

```
for (i = 1 through n)  
  for (j = 1 through i)  
    for (k = 1 through 5)  
      Task T
```

- Task T requires t times units
- How to find the relationship between t times and the problem sizes n

Step 1. Assign `sum = 0`. Assign `i = 0`.

Step 2.

Assign `i = i + 1`

Assign `sum = sum + i`

Step 3. Compare `i` with `10`

`if i < 10`, back to step 2.

otherwise, `if i ≥ 10`, go to step 4.

Step 4. Return `sum`

How many
Assignments?
Comparisons?

Step 1. Assign `sum = 0`. Assign `i = 0`.

Step 2.

Assign `i = i + 1`

Assign `sum = sum + i`

Step 3. Compare `i` with `n`

`if i < n`, back to step 2.

otherwise, `if i ≥ n`, go to step 4.

Step 4. Return `sum`

How many
Assignments?
Comparisons?

- Sum of n integer $S(n) = 0 + 1 + 2 + \dots + n - 1$

```
int sum = 0;  
for (int i = 0; i < n; i++)  
    sum = sum + i;
```

Assignment: $2n + 2$

```
int sum = 0;  
for (int i = 0; i < n ; i++)  
    sum = sum + i;
```

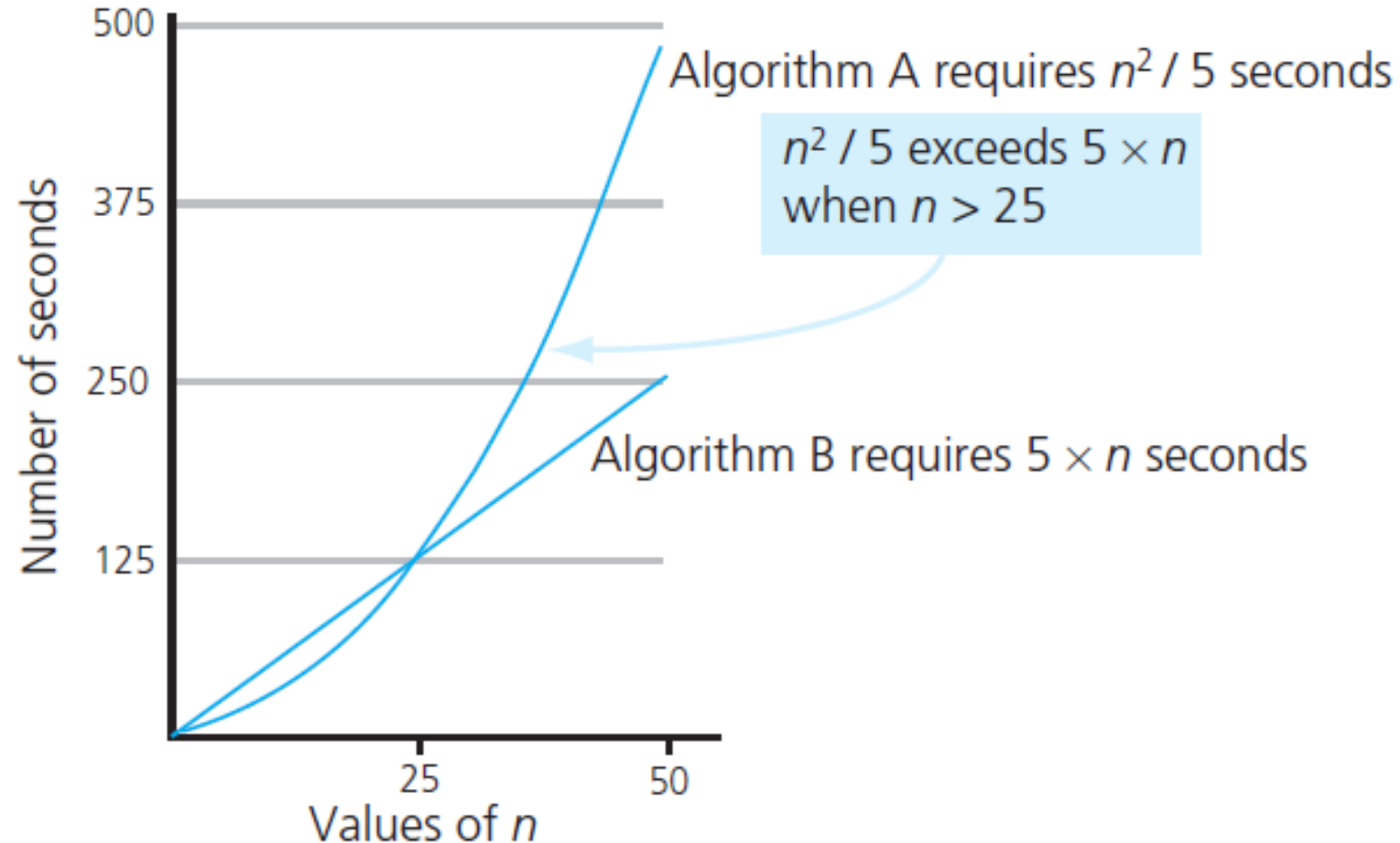
Comparison: $n + 1$

- Running Time: $T(n) = 3n + 3$

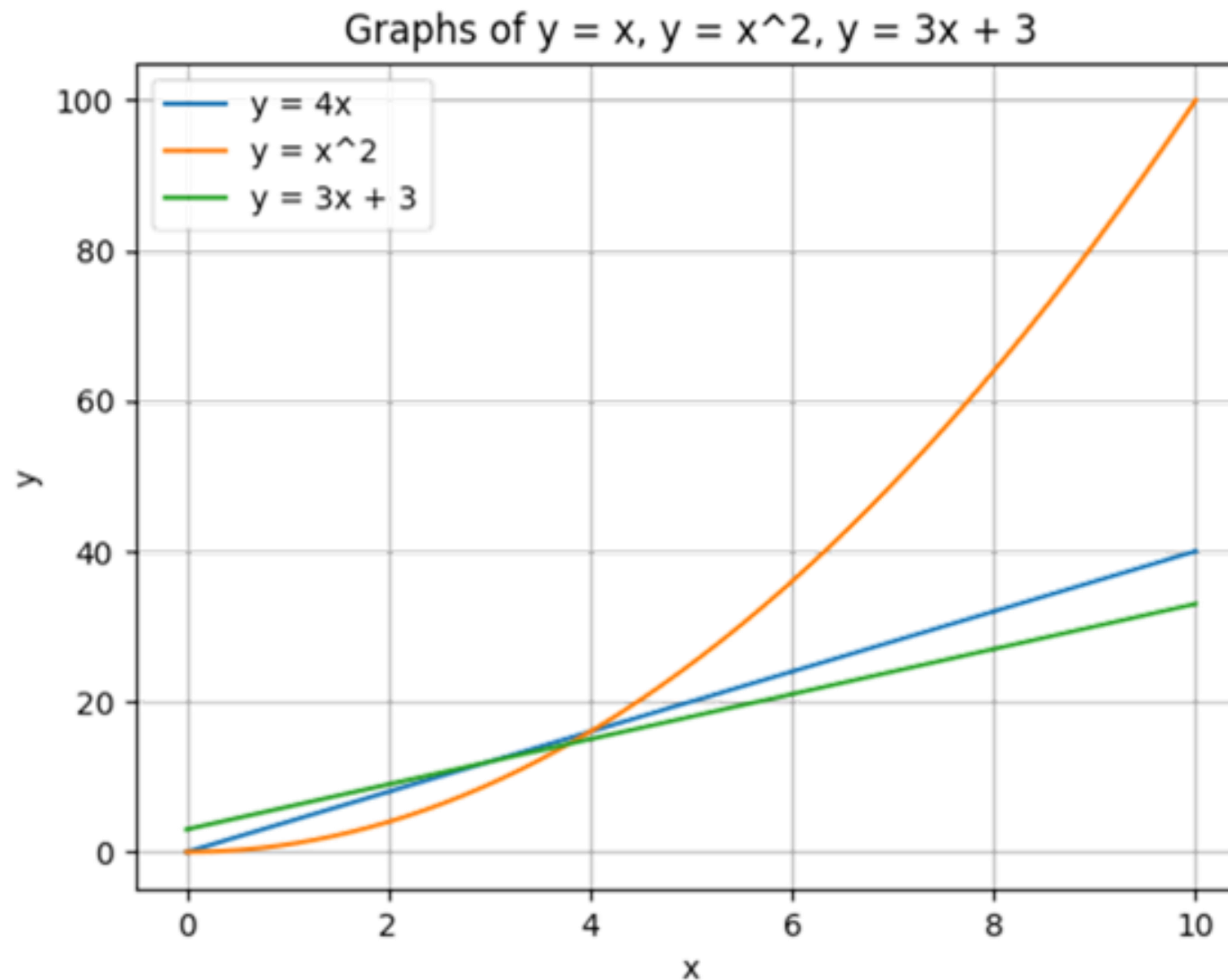
- Measure algorithm's time requirement as a **function of problem size**
- Compare algorithm efficiencies for **large problems**
- Look only at **significant differences**.

Analysis & Big O Notation

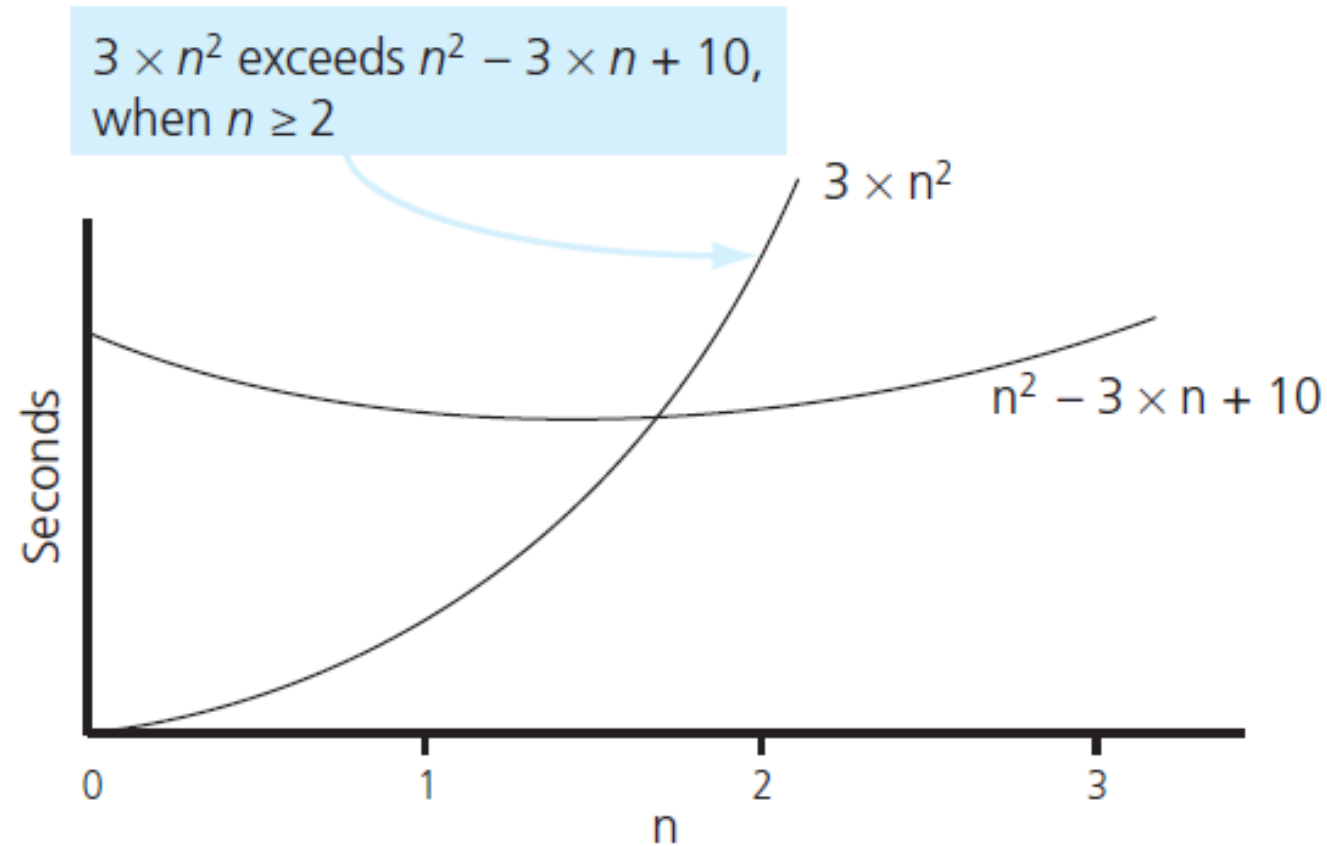
- Time requirements as a function of the problem size n



- Definition:
 - Algorithm A with time unit function $g(n)$ is order $f(n)$
 - Denoted $g(n)$ is $O(f(n))$
 - If constants k and n_0 exist such that A requires no more than $k \times f(n)$ time units to solve a problem of size $n \geq n_0$.
 - It means that for all $n \geq n_0$, $g(n) \leq c \cdot f(n)$



- An algorithm requires $g(n) = n^2 - 3n + 10$ (time units).
- What is the order of algorithm?
 - Hint: Find the values k and n_0 .



- With $c = 3$ and $n_0 = 2$, $g(n) = O(n^2)$

- Another algorithm requires $n^2 + 3n + 10$ time units. What is the order of this algorithm?

- How about the order of an algorithm requiring
$$(n + 1) \times (a + c) + n \times w$$
time units?

Where n is the problem size

- **$f(n)$** =
 - 1: Constant
 - $\log_2 n$: Logarithmic
 - n : Linear
 - $n \log_2 n$: Linearithmic
 - n^2 : Quadratic
 - n^3 : Cubic
 - 2^n : Exponential

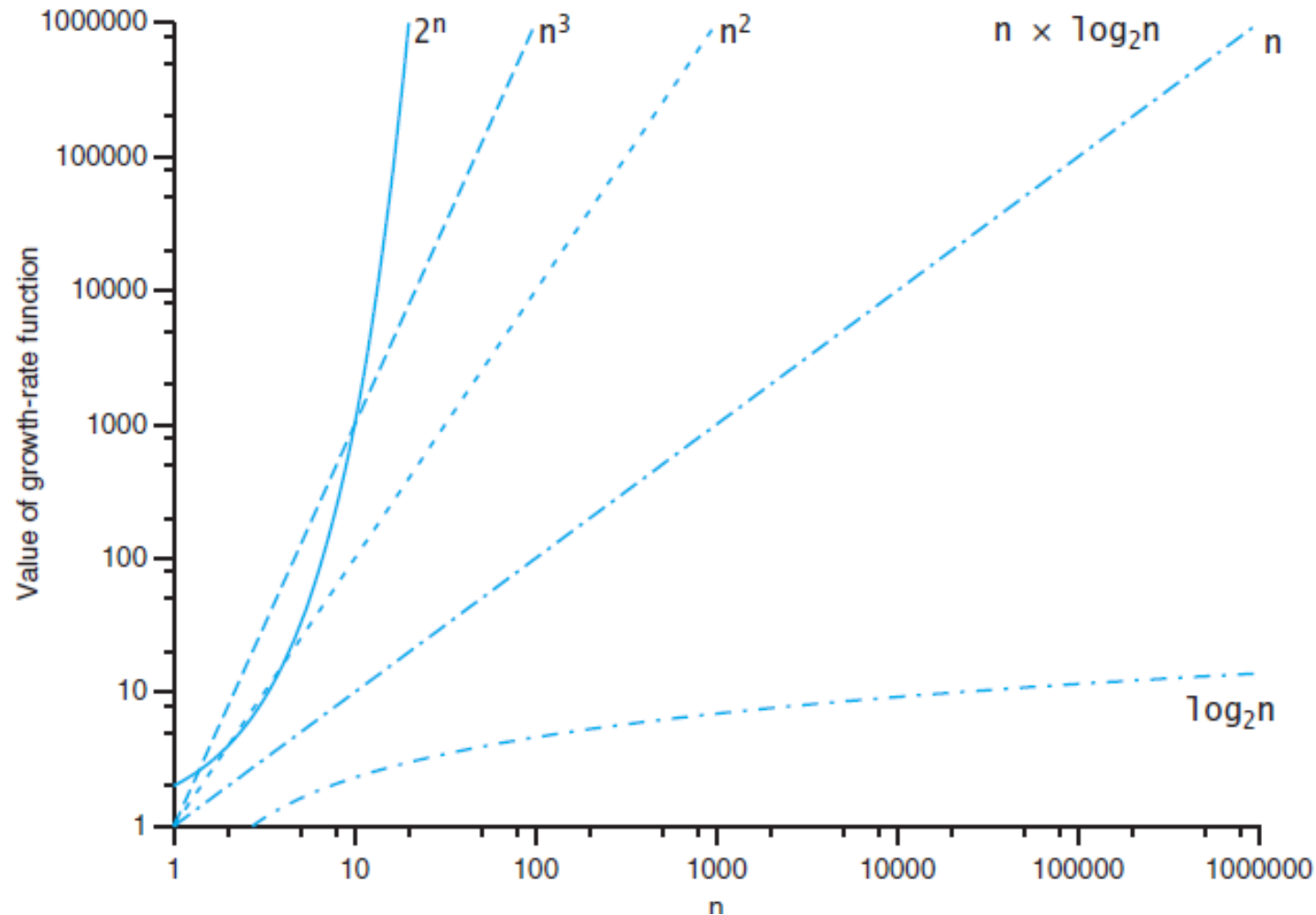
- Order of growth of some common functions

$$O(1) < O(\log_2 n) < O(n) < O(n \times \log_2 n) < \\ O(n^2) < O(n^3) < O(2^n)$$

- A comparison of growth-rate functions in tabular form

Function	n					
	10	100	1,000	10,000	100,000	1,000,000
1	1	1	1	1	1	1
$\log_2 n$	3	6	9	13	16	19
n	10	10^2	10^3	10^4	10^5	10^6
$n \times \log_2 n$	30	664	9,965	10^5	10^6	10^7
n^2	10^2	10^4	10^6	10^8	10^{10}	10^{12}
n^3	10^3	10^6	10^9	10^{12}	10^{15}	10^{18}
2^n	10^3	10^{30}	10^{301}	$10^{3,010}$	$10^{30,103}$	$10^{301,030}$

- A comparison of growth-rate functions in graphical form



- Ignore low-order terms

$$O(n^3 + 4n^2 + 3n) == O(n^3)$$

- Ignore a multiplicative constant in the high-order term

$$O(5n^3) == O(n^3)$$

- Can combine growth rate functions

$$O(f(n)) + O(g(n)) = O(f(n) + g(n))$$

- Constant Multiplication:

If $f(n)$ is $O(g(n))$

then $c * f(n)$ is $O(g(n))$, where c is a constant.

- Polynomial Function:

$$f(x) = a_n x^n + a_{n-1} x^{n-1} + \dots + a_1 x + a_0 \text{ is } O(x^n)$$

- **Summation Function:**

- If $f_1(n)$ is $O(g_1(n))$ and $f_2(n)$ is $O(g_2(n))$
- Then $f_1(n) + f_2(n)$ is **$O(\max(g_1(n), g_2(n)))$**

- **Multiplication Function:**

- If $f_1(n)$ is $O(g_1(n))$ and $f_2(n)$ is $O(g_2(n))$
- Then $f_1(n) \times f_2(n)$ is **$O(g_1(n) \times g_2(n))$**

- Use like this:
 - $f(x)$ **is** $O(g(x))$, or
 - $f(x)$ **is of order** $g(x)$, or
 - $f(x)$ **has order** $g(x)$

- **Are these functions of order $O(x)$?**

- $f(x) = 10$

- $f(x) = 3x + 7$

- $f(x) = 2x^2 + 2$

- **Give the order of growth (as a function of n) of the running time of the following function?**
 - $f(n) = (2 + n) * (3 + \log_2 n)$
 - $f(n) = 11 * \log_2 n + n/2 - 3542$
 - $f(n) = n * (3 + n) - 7 * n$
 - $f(n) = \log_2(n^2) + n$

- **Give the order of growth (as a function of n) of the running time of the following function?**
 - $f(n) = (2 + n) * (3 + \log_2 n)$
 - $f(n) = 11 * \log_2 n + n/2 - 3542$
 - $f(n) = n * (3 + n) - 7 * n$
 - $f(n) = \log_2(n^2) + n$

$$\log_a xy = \log_a x + \log_a y$$

$$\log_a x^y = y \log_a x$$

Algorithm Efficiency

- Best case scenario
- Worst case scenario
- Average case scenario

- Input: ???
- Output: ???
- **Step 1.** Set the first integer the temporary maximum value (temp_max).
- **Step 2.** Compare the current value with the temp_max .
 - If it is greater than, assign the current value to temp_max.
- **Step 3.** If there is other integer in the list, move to next value. Back to step 2.
- **Step 4.** If there is no more integer in the list, stop.
- **Step 5.** return temp_max (the maximum value of the list).

- Input:
- Output:
- Step 1. Assign $i = 0$
- Step 2. `while` ($i < n$ and $x \neq a_i$)
 $i = i + 1$
- Step 3.
 - `if` $i < n$, `return` i
 - Otherwise ($i \geq n$), `return` -1 to tell that x does not exist in list a .

- Use comparisons for counting.
- Worst case:
 - When it occurs?
 - How many operations?
- Best case:
 - When it occurs?
 - How many operations?

- Use comparisons for counting.
- Average case:
 - If x is found at position i^{th} , the number of comparisons is
 $2i + 1$
 - The average number of comparisons is:

$$\frac{3 + 5 + 7 + \dots + (2n + 1)}{n} = \frac{2(1 + 2 + 3 + \dots + n) + n}{n} = \frac{2 \frac{n(n+1)}{2} + n}{n} = n + 2$$

- Decide n – the input size
- Identify the algorithm's **basic operation** (as a rule, it is in the innermost loop)
- Check whether the number of times the basic operation is executed depends only on n
 - If it depends on some additional property, specify the **worst-case** for Big-Oh
- Set up a **sum** expressing the number of times the algorithm's basic operation is executed.
- Find a closed-form formula for the count and **establish its order of growth.**

- **Example:** Check whether all the elements in a given array of n elements are distinct.

```
UniqueElements(A[0..n - 1])
```

```
//Determines whether all the elements in a given array are distinct
```

```
//Input: An array A[0..n - 1]
```

```
//Output: Returns "true" if all the elements in A are distinct
```

```
// and "false" otherwise
```

```
for i ← 0 to n - 2 do
```

```
    for j ← i + 1 to n - 1 do
```

```
        if A[i] = A[j]
```

```
            return false
```

```
return true
```

Basic operation

□ Worst-case:

$$\begin{aligned}C_{worst}(n) &= \sum_{i=0}^{n-2} \sum_{j=i+1}^{n-1} 1 = \sum_{i=0}^{n-2} [(n-1) - (i+1) + 1] = \sum_{i=0}^{n-2} (n-1-i) \\&= \sum_{i=0}^{n-2} (n-1) - \sum_{i=0}^{n-2} i = (n-1) \sum_{i=0}^{n-2} 1 - \frac{(n-2)(n-1)}{2} \\&= (n-1)^2 - \frac{(n-2)(n-1)}{2} = \frac{(n-1)n}{2} \approx \frac{1}{2}n^2\end{aligned}$$

Exercise

Propose an algorithm to calculate the value of S defined below. What order does the algorithm have?

$$S = 1 + \frac{1}{2} + \frac{1}{6} + \dots + \frac{1}{n!}$$

How many comparisons, assignments are there in the following code fragment with the size n ?

```
sum = 0;
for (i = 0; i < n; i++)
{
    std::cin >> x;
    sum = sum + x;
}
```

How many assignments are there in the following code fragment with the size n ?

```
for (i = 0; i < n ; i++)  
    for (j = 0; j < n; j++){  
        C[i][j] = 0;  
        for (k = 0; k < n; k++)  
            C[i][j] = C[i][j] + A[i][k] * B[k][j];  
    }
```

Give the order of growth (as a function of n) of the running time of the following code fragment:

```
int sum = 0;
for (int i = n; i > 0; i /= 2)
    for (int j = 0; j < n; j++)
        sum++;
```


Give the order of growth (as a function of N) of the running time of the following code fragment:

```
int sum = 0;
for (int i = 1; i < N; i *= 2)
    for (int j = 0; j < N; j++)
        sum++;
```

Give the order of growth (as a function of N) of the running time of the following code fragment:

```
int sum = 0;
for (int i = 1; i < n; i *= 2)
    for (int j = 0; j < i; j++)
        sum++;
```

1. $\sum_{i=l}^u 1 = \underbrace{1 + 1 + \cdots + 1}_{u-l+1 \text{ times}} = u - l + 1$ (l, u are integer limits, $l \leq u$); $\sum_{i=1}^n 1 = n$

2. $\sum_{i=1}^n i = 1 + 2 + \cdots + n = \frac{n(n+1)}{2} \approx \frac{1}{2}n^2$

3. $\sum_{i=1}^n i^2 = 1^2 + 2^2 + \cdots + n^2 = \frac{n(n+1)(2n+1)}{6} \approx \frac{1}{3}n^3$

4. $\sum_{i=1}^n i^k = 1^k + 2^k + \cdots + n^k \approx \frac{1}{k+1}n^{k+1}$

5. $\sum_{i=0}^n a^i = 1 + a + \cdots + a^n = \frac{a^{n+1} - 1}{a - 1}$ ($a \neq 1$); $\sum_{i=0}^n 2^i = 2^{n+1} - 1$

6. $\sum_{i=1}^n i2^i = 1 \cdot 2 + 2 \cdot 2^2 + \cdots + n2^n = (n-1)2^{n+1} + 2$

Give the order of growth (as a function of n) of the running time of the following code fragment:

```
int sum = 0;
for (int k = 1; k < n; k = k*2)
    for (int i = 0; i < k; i++)
        sum++;
```

- A geometric sequence, or geometric progression, is a sequence of numbers where each successive number is the product of the previous number and some constant r

$$a_n = r a_{n-1}$$

- common ratio: $r = \frac{a_n}{a_{n-1}}$

- A geometric series is the sum of the terms of a geometric sequence
- Consider the sequence as follows:

$$S_n = a_1 + a_1 r + a_1 r^2 + \dots + a_1 r^{n-1}$$

- the formula for the n-th partial sum of a geometric sequence

$$S_n = \frac{a_1(1-r^n)}{1-r} \quad (r \neq 1)$$

What is the goal of the following code and give the order of growth function of its running time

```
void mystery(int*& arr, int n, int a[], int k) {
    arr = new int[n]{0};
    for (int i = 0; i < k; i++) arr[i] = a[i];
    for (int i = k; i < n; ++i) {
        int j = 0;
        while (j < k){
            *(arr + i) += *(arr + i - j - 1);
            j++;
        }
    }
}
```

- Hint: this is the main function for the above fragment of code

```
int main() {  
    int *arr = NULL;  
    int start[3] = {-2, 0, 3};  
    mystery(arr, 7, start, 3);  
    for (int i = 0; i < 7; i++)  
        std::cout << arr[i] << std::endl;  
    if (arr != NULL) delete[] arr;  
    return 0;  
}
```

- What is the output of the function

THANK YOU
for YOUR ATTENTION