

## CSC10004: Data Structure and Algorithms

### Lecture 2: Asymptotic notations

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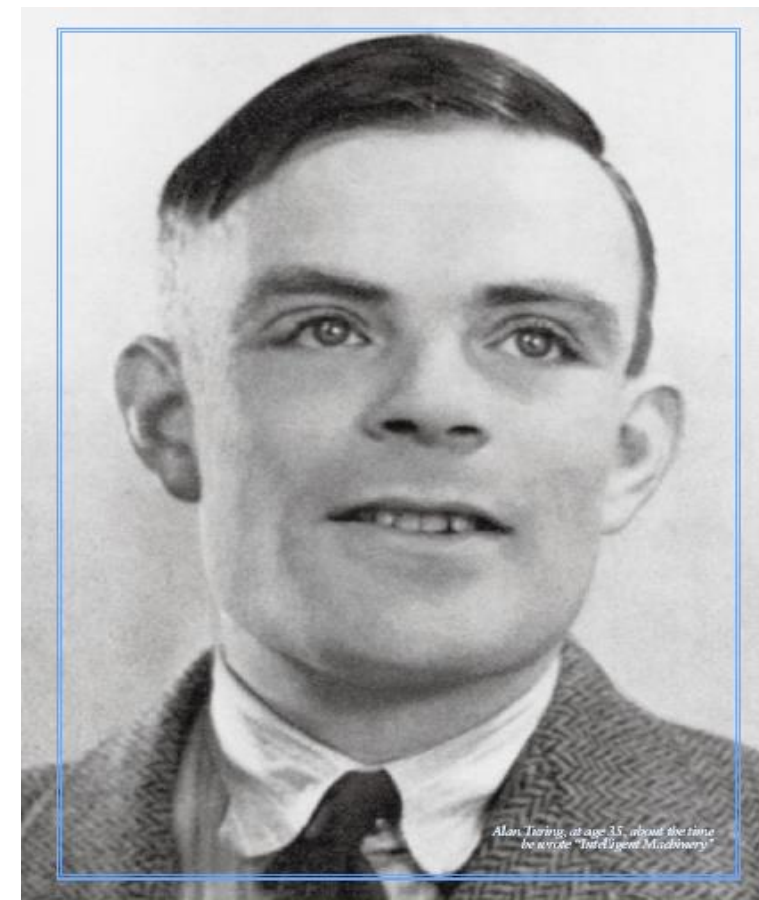
# Outline

- Turing machines
- RAM model
- What is an algorithm?
- Asymptotic notations

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- The **theory of computation** and the practical application it made possible — the computer — was developed by an Englishman called **Alan Turing**.

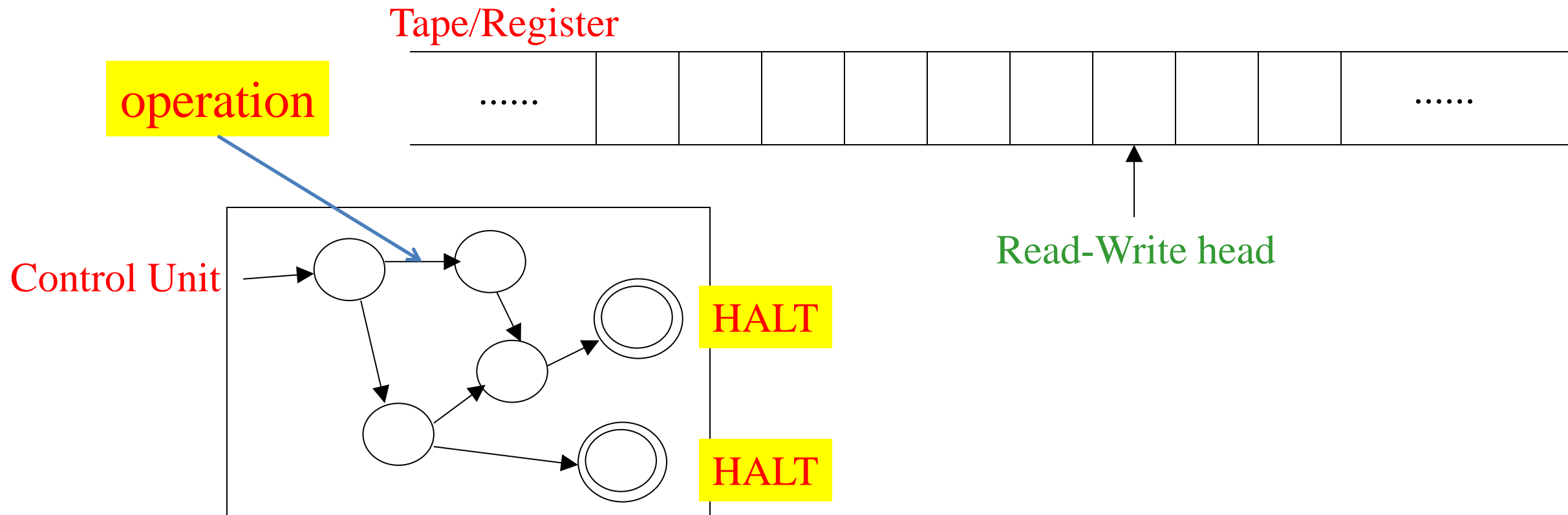


# Definition

Turing's machine — which came to be called the Turing machine — was this:

1. A tape of **infinite** length.
2. **Finitely many squares of the tape** have a single symbol from a finite language.
3. Someone (or something) that can **read** the squares and **write** in them.
4. At any time, the machine is **in one of a finite number of internal states**.
5. The machine has instructions that determine what it does given its internal state and the symbol it encounters on the tape. It can
  - change its internal state;
  - change the symbol on the square;
  - move forward;
  - move backward;
  - halt (i.e. stop).

- It is essential to the idea of a **Turing machine** that it is not a physical machine, but an **abstract one** — a set of procedures.



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## Random-access machine (RAM) model

- Simple operations (arithmetic, comparison, conditional, etc.) each take the **same, constant amount of time**.
- Data stored in an infinite array of registers  $(0, 1, 2, \dots)$ , each of which can hold  $c \log x$  bits, where
  - $x$ : problem size
  - $c$ : some constant independent of  $x$



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## Definition

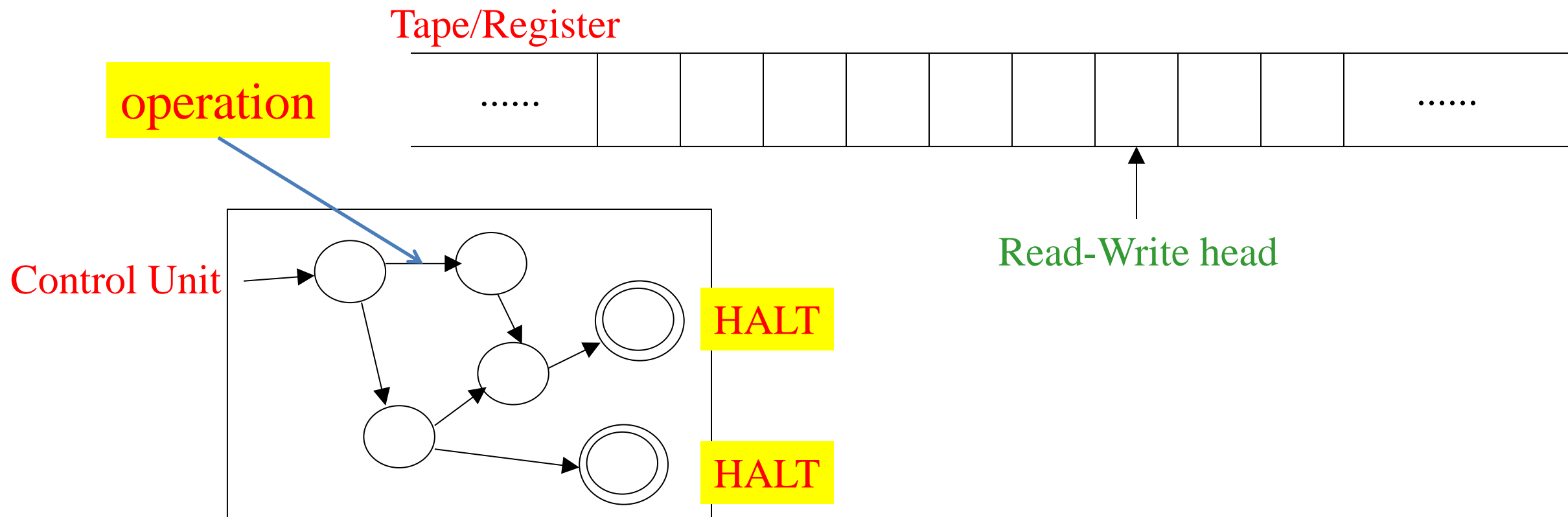
- An algorithm is a sequence of **unambiguous instructions** for solving a problem, that is, for obtaining a required output for any legitimate input in a finite amount of time.



- Analysis of algorithms is the quantitative study of the performance of algorithms, in terms of their run time, memory usage, or other properties.

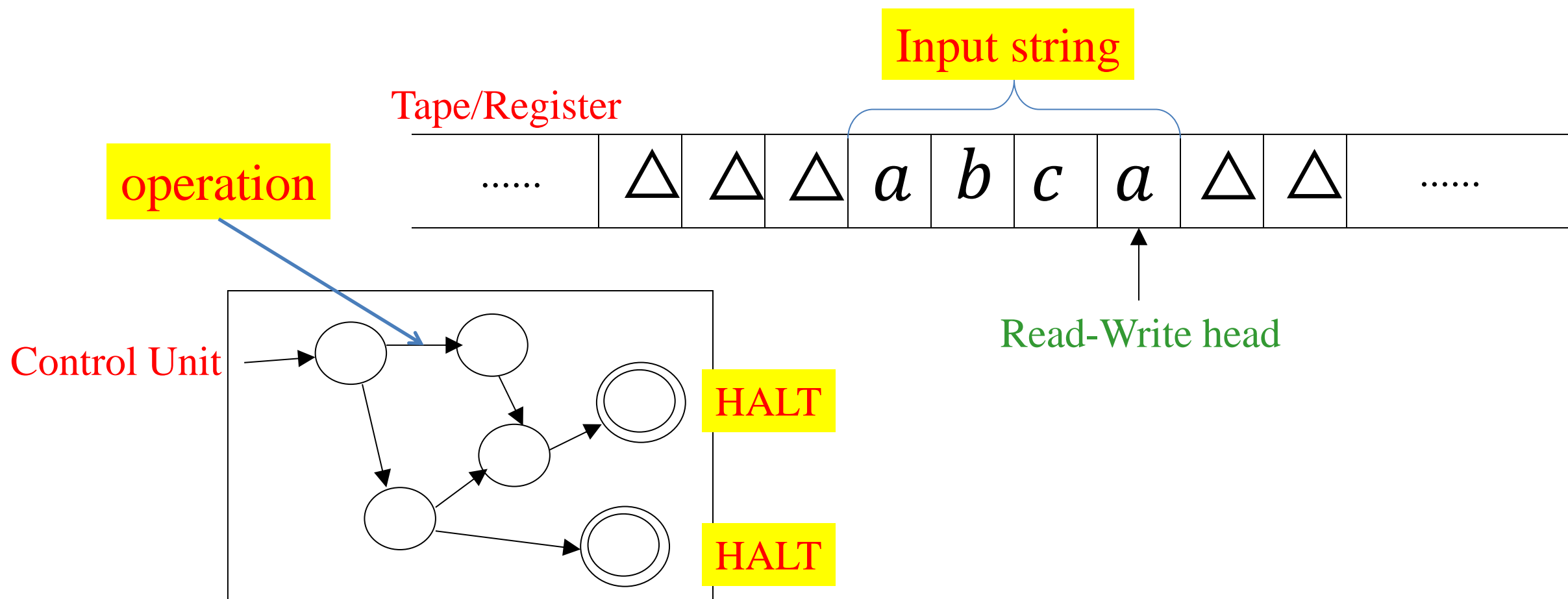
# Informal descriptions

- Run time = #operations
- Memory usage = tape length



# Informal descriptions

- The problem size = the length of the input string in the tape



## Question (1/4)

Do these two algorithms have the same asymptotic running time?

```
int AlgA(int n) {  
    int sum = 0;  
    for(int i=0; i<n; i++) {  
        for(int j=0; j<i; j++) {  
            sum++;  
        }  
    }  
    return sum;  
}
```

```
int AlgB(int n) {  
    int sum = 0;  
    for(int i=0; i<n; i++) {  
        for(int j=0; j<n; j++) {  
            sum++;  
        }  
    }  
    return sum;  
}
```

## Question (2/4)

- In an iterative algorithm, let  $a_i$  be the number of operations, e.g, **comparisons and assignments**, at iteration  $i$ .
- A common tool for analyzing the iterative algorithms is the summation:

$$\sum_{i=\ell}^n a_i = a_\ell + a_{\ell+1} + \cdots + a_{n-1} + a_n$$

- If the upper limit is infinite, we interpret this as an implicit limit:

$$\sum_{i=\ell}^{\infty} a_i = \lim_{n \rightarrow \infty} \sum_{i=\ell}^n a_i$$

## Question (3/4)

$$A(n) = \sum_{i=0}^{n-1} \sum_{j=0}^{i-1} 1 = \sum_{i=0}^{n-1} i = \frac{n(n-1)}{2} \\ = \Theta(n^2)$$

$$B(n) = \sum_{i=0}^{n-1} \sum_{j=0}^{n-1} 1 = \sum_{i=0}^{n-1} n = n^2 \\ = \Theta(n^2)$$

```
int AlgA(int n) {  
    int sum = 0;  
    for(int i=0; i<n; i++) {  
        for(int j=0; j<i; j++) {  
            sum++;  
        }  
    }  
    return sum;  
}
```

```
int AlgB(int n) {  
    int sum = 0;  
    for(int i=0; i<n; i++) {  
        for(int j=0; j<n; j++) {  
            sum++;  
        }  
    }  
    return sum;  
}
```

## Question (4/4)

Do these two algorithms have the same asymptotic running time?

→ Yes, the run times are both  $\Theta(n^2)$ .

```
int AlgA(int n) {  
    int sum = 0;  
    for(int i=0; i<n; i++) {  
        for(int j=0; j<i; j++) {  
            sum++;  
        }  
    }  
    return sum;  
}
```

```
int AlgB(int n) {  
    int sum = 0;  
    for(int i=0; i<n; i++) {  
        for(int j=0; j<n; j++) {  
            sum++;  
        }  
    }  
    return sum;  
}
```



# Exercise

Do these two algorithms have the same asymptotic running time?

```
int AlgC(int n) {  
    int sum = 0;  
    for(int i=0; i<n; i*=2) {  
        for(int j=0; j<i; j++) {  
            sum++;  
        }  
    }  
    return sum;  
}
```

```
int AlgD(int n) {  
    int sum = 0;  
    for(int i=0; i<n; i*=2) {  
        for(int j=0; j<n; j++) {  
            sum++;  
        }  
    }  
    return sum;  
}
```

# Solution

- **NO**, the asymptotic run times are different. Observe that  $i$  is always a power of 2. Let  $i = 2^k$ , so that  $k = \log_2 n$ .
- $C(n) = \sum_{k=0}^{\lceil \log_2 n \rceil - 1} \sum_{j=0}^{2^k - 1} 1 = \sum_{k=0}^{\lceil \log_2 n \rceil - 1} 2^k = 2^{\lceil \log_2 n \rceil} - 1 = \Theta(n)$
- $D(n) = \sum_{k=0}^{\lceil \log_2 n \rceil - 1} \sum_{j=0}^{n-1} 1 = \sum_{k=0}^{\lceil \log_2 n \rceil - 1} n = n \lceil \log_2 n \rceil = \Theta(n \ln n)$

# Outline

- Turing machines
- RAM model
- What is an algorithm?
- Asymptotic notations
  - Big-Oh ( $O$ )
  - Big-Omega ( $\Omega$ )
  - Big-Theta ( $\Theta$ )
  - Little-Oh ( $o$ )
  - Little-Omega ( $\omega$ )

- In the analysis of algorithms, we are usually interested in **how** the performance of our algorithm **changes** as the problem input size **increases**.
- The primary tools for measuring the growth rate of a function that describes the run time of an algorithm are the asymptotic notations.
- This provides a way of studying the algorithms themselves, independent of any specific hardware, operating system, compiler, programmer, etc.

# Overview

• Let  $L = \lim_{x \rightarrow \infty} \frac{f(x)}{g(x)}$

• We have:

$$f < g$$

$$f \geq g$$

$$f = g$$

$$f \leq g$$

$$f > g$$

	$f(x) =$	$o(g(x))$	$\Omega(g(x))$	$\Theta(g(x))$	$O(g(x))$	$\omega(g(x))$
$L$						
$0$		$\times$			$\times$	
$(0, \infty)$			$\times$	$\times$	$\times$	
$\infty$			$\times$			$\times$

## Big-Oh notation (1/3): definition

- $f$  is asymptotically bounded **ABOVE** by  $g$  **up to** constant factor  $c$ .
- Write:  $f(x) = O(g(x))$  or  $f(x) \in O(g(x))$ .
- Mathematically,  $\exists c > 0$  and  $\exists x_0 > 0: \forall x \geq x_0, |f(x)| \leq cg(x)$ .
- To prove that  $f(x) \in O(g(x))$ , we need to provide the existence of a pair  $(c, x_0)$ .

## Big-Oh notation (2/3): examples

- $x^2 + 2x + 5 = O(x^2)$ .

Proof: Select  $(x_0 = 1, c = 10)$  then  $x^2 + 2x + 5 \leq 10x^2 \quad \forall x \geq x_0$ .

- $x^2 + 2x + 5 \notin O(x)$ .

Proof: Assume there exists a pair  $(x_0, c > 0)$  such that  $x^2 + 2x + 5 \leq cx \quad \forall x \geq x_0$ . Then for all  $x \geq x_0$ :

$$x^2 + (2 - c)x + 5 = \left(x - \left(1 - \frac{c}{2}\right)\right)^2 + \left(5 - \left(1 - \frac{c}{2}\right)^2\right) \leq 0$$

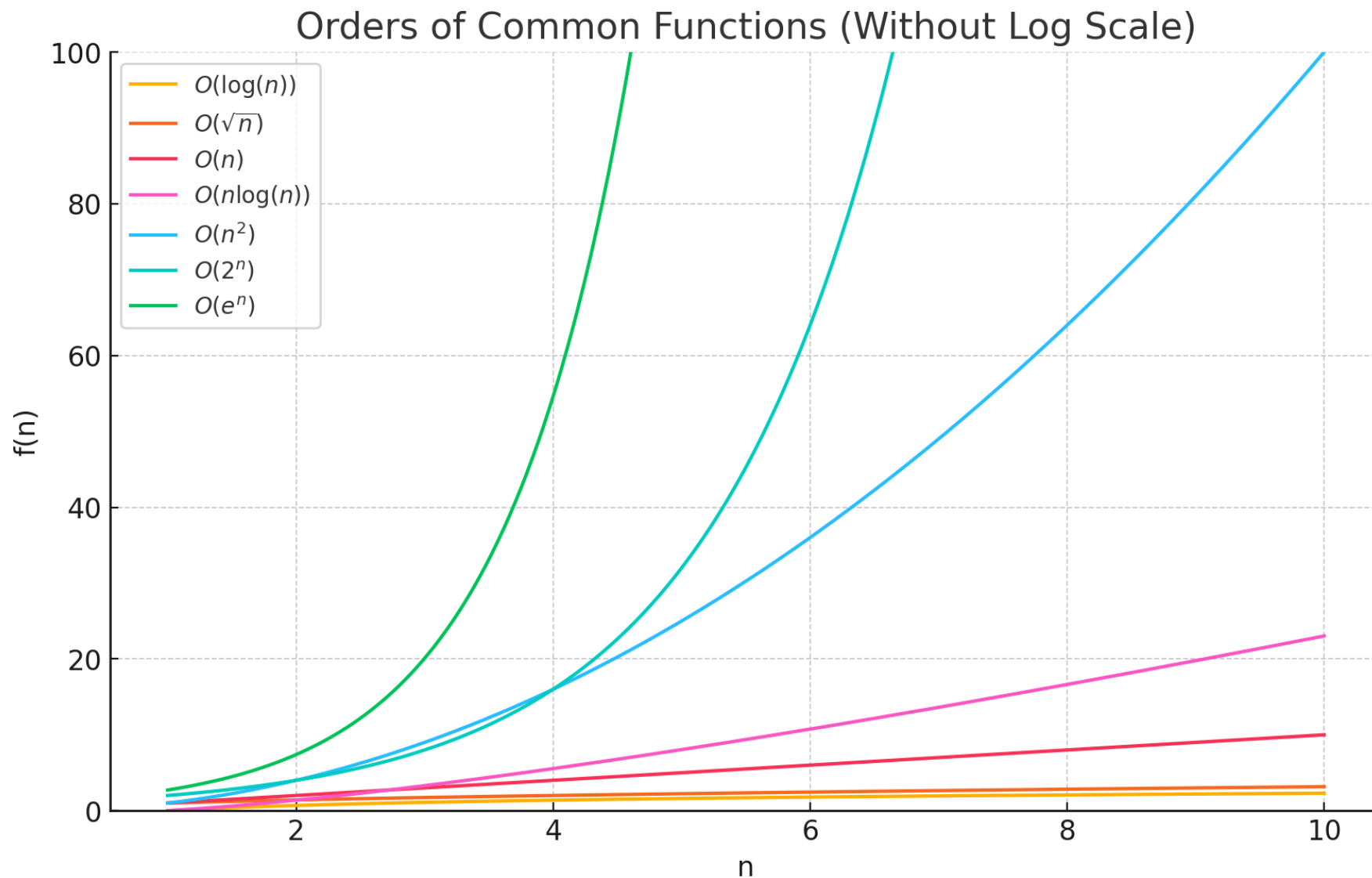
This inequality does not hold when  $x$  goes to infinity.

# Big-Oh notation (3/3): common functions

Notation	Name (+ function)	Example
$O(1)$	Constant	
$O(\log \log n)$	double logarithmic	
$O(\log n)$	logarithmic	
$O(\log^c n), c > 1$	polylogarithmic	
$O(n^\alpha), 0 < \alpha < 1$	fractional power	
$O(n)$	Linear	
$O(n \log n)$	Quasilinear	
$O(n^2)$	Quadratic	
$O(n^c), 1 < c$	Polynomial	
$O(c^n), c > 1$	Exponential	
$O(n!)$	factorial	



# Big-Oh notation (3/3): plot of common function orders



## Big-Omega notation (1/2): definition

- $f$  is asymptotically bounded **BELOW** by  $g$ .
- Write:  $f(x) = \Omega(g(x))$  or  $f(x) \in \Omega(g(x))$ .
- Mathematically,  $\exists c > 0$  and  $\exists x_0 > 0: \forall x \geq x_0, f(x) \geq cg(x)$ .
- To prove that  $f(x) \in \Omega(g(x))$ , we need to provide the existence of a pair  $(c, x_0)$ .
- Note that:  $f(x) = \Omega(g(x)) \Leftrightarrow g(x) = O(f(x))$ .

## Big-Omega notation (2/2): examples (1/2)

- $x^2 + 2x + 5 = \Omega(x)$ .

Proof: Select  $(x_0 = 1, c = 1)$  then  $x^2 + 2x + 5 \geq 1 \cdot x \ \forall x \geq x_0$ .

- $x^2 + 2x + 5 = \Omega(x^2)$ .

Proof: Select  $(x_0 = 1, c = 1)$  then  $x^2 + 2x + 5 \geq 1 \cdot x^2 \ \forall x \geq x_0$ .

## Big-Omega notation (2/2): examples (2/2)

- $x^2 + 2x + 5 \notin \Omega(x^3)$ .

Proof: Assume there exists a pair  $(x_0 > 0, c > 0)$  such that  $x^2 + 2x + 5 \geq cx^3 \forall x \geq x_0$ . Then for all  $x \geq x_0$ :

$$-x^2(cx - 1) + x + 5 \geq 0$$

This inequality does not hold because when  $x$  goes to infinity, the right inequality goes to negative infinity.

## Big-Theta notation (1/2): definition

- $f$  is asymptotically bounded by  $g$  both **ABOVE** (with constant factor  $c_2$ ) and **BELOW** (with constant factor  $c_1$ ).
- Write:  $f(x) = \Theta(g(x))$  or  $f(x) \in \Theta(g(x))$ .
- Mathematically,  $\exists c_1, c_2 > 0$  and  $\exists x_0 > 0: \forall x \geq x_0, c_1 g(x) \leq f(x) \leq c_2 g(x)$ .
- To prove that  $f(x) \in \Omega(g(x))$ , we need to provide the existence of a triple  $(c_1, c_2, x_0)$ .
- Note that:  $f(x) = \Theta(g(x)) \Leftrightarrow g(x) = O(f(x))$  and  $f(x) = O(g(x))$ .

## Big-Theta notation (2/2): examples (1/2)

- $x^2 + 2x + 5 = \Theta(x^2)$ .

Proof: Select  $(c_1 = 1, c_2 = 10, x_0 = 1)$  then  $1 \cdot x^2 \leq x^2 + 2x + 5 \leq 10x^2 \ \forall x \geq x_0$ .

- $x^2 + 2x + 5 \neq \Theta(x)$ .

Proof: Assume there exists a triple  $(c_1 > 0, c_2 > 0, x_0)$  such that  $c_1 x \leq x^2 + 2x + 5 \leq c_2 x \ \forall x \geq x_0$ . Then for all  $x \geq x_0$ :

$$x^2 + (2 - c_2)x + 5 = \left(x - \left(1 - \frac{c_2}{2}\right)\right)^2 + \left(5 - \left(1 - \frac{c_2}{2}\right)^2\right) \leq 0$$

This inequality does not hold when  $x$  goes to infinity.

## Big-Theta notation (2/2): examples (2/2)

- $x^2 + 2x + 5 \notin \Theta(x^3)$ .

Proof: Assume there exists a pair  $(x_0, c > 0)$  such that  $x^2 + 2x + 5 \geq cx^3 \forall x \geq x_0$ . Then for all  $x \geq x_0$ :

$$-x^2(cx - 1) + x + 5 \geq 0$$

This inequality does not hold because when  $x$  goes to infinity, the right inequality goes to negative infinity.

## Little-Oh notation (1/2): definition

- $f$  is asymptotically dominated by  $g$  (for **ANY** constant factor  $c$ ).
- Write:  $f(x) = o(g(x))$  or  $f(x) \in o(g(x))$ .
- Mathematically,  $\forall c > 0$  and  $\exists x_0 > 0: \forall x \geq x_0, f(x) \leq cg(x)$ .
- To prove that  $f(x) \in o(g(x))$ , we need to provide the existence of  $x_0$  for every  $c > 0$ .
- Note that:  $f(x) = o(g(x)) \Leftrightarrow \lim_{x \rightarrow \infty} \frac{f(x)}{g(x)} = 0$ .



## Little-Oh notation (2/2): example

- $x^2 + 2x + 5 = o(x^3)$ .

Proof: 
$$\lim_{x \rightarrow \infty} \frac{f(x)}{g(x)} = \lim_{x \rightarrow \infty} \frac{x^2 + 2x + 5}{x^3} = \lim_{x \rightarrow \infty} \left( \frac{1}{x} + \frac{2}{x^2} + \frac{5}{x^3} \right) = 0.$$

- $x^2 + 2x + 5 \notin o(x)$ .

Proof: 
$$\lim_{x \rightarrow \infty} \frac{f(x)}{g(x)} = \lim_{x \rightarrow \infty} \frac{x^2 + 2x + 5}{x} = \lim_{x \rightarrow \infty} \left( x + 2 + \frac{5}{x} \right) = \infty.$$

## Little-Omega notation (1/2): definition

- $f$  asymptotically dominates  $g$ .
- Write:  $f(x) = \omega(g(x))$  or  $f(x) \in \omega(g(x))$ .
- Mathematically,  $\forall c > 0$  and  $\exists x_0 > 0: \forall x \geq x_0, f(x) \geq cg(x)$ .
- To prove that  $f(x) \in o(g(x))$ , we need to provide the existence of  $x_0$  for every  $c > 0$ .
- Note that:  $f(x) = \omega(g(x)) \Leftrightarrow \lim_{x \rightarrow \infty} \frac{f(x)}{g(x)} = \infty$ .

## Little-Oh notation (2/2): example

- $x^2 + 2x + 5 = \omega(x)$ .

- Proof:  $\lim_{x \rightarrow \infty} \frac{f(x)}{g(x)} = \lim_{x \rightarrow \infty} \frac{x^2 + 2x + 5}{x} = \lim_{x \rightarrow \infty} \left( x + 2 + \frac{5}{x} \right) = \infty$ .

- $x^2 + 2x + 5 \notin o(x^3)$ .

Proof:  $\lim_{x \rightarrow \infty} \frac{f(x)}{g(x)} = \lim_{x \rightarrow \infty} \frac{x^2 + 2x + 5}{x^3} = \lim_{x \rightarrow \infty} \left( \frac{1}{x} + \frac{2}{x^2} + \frac{5}{x^3} \right) = 0$ .

## Exercises (1/5)

Prove the following statements:

1.  $n^3 + 1000n^2 = O(n^4)$ .
2.  $\log n = O(n)$ .
3.  $\log n = O(\sqrt{n})$ .
4.  $n! \notin O(n^c)$  for any positive constant  $c$ .
5.  $n^a = O(b^n)$  for any positive constants  $a$  and  $b > 1$ .
6.  $\log n! = O(n \log n)$  and  $\log n! \geq \frac{n}{2} \log \frac{n}{2}$  to get  $\log n! = \Theta(n \log n)$ .
7.  $1000x^3 - x^2 + 79 = \Theta(x^3)$ .

## Exercises (2/5)

8. Let  $f(x) = O(h(x))$  and  $g(x) = O(h(x))$ . Let  $a, b > 0$ . Prove that  $af(x) + bg(x) = O(h(x))$ .
9. Let  $f_1(x) = O(g_1(x))$  and  $f_2(x) = O(g_2(x))$ . Prove that  $f_1(x)f_2(x) = O(g_1(x)g_2(x))$ .
10. Prove that  $x^n + a_{n-1}x^{n-1} + \dots + a_0 = O(x^n)$ .
11. Prove that  $x^n + a_{n-1}x^{n-1} + \dots + a_0 = \Theta(x^n)$ .
12.  $\log n = O(n^c)$  where  $0 < c < 1$ .

## Exercises (3/5)

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

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

## Exercises (4/5)

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];  
    }
```

## Exercises (5/5)

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

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



Q & A

# Solutions: Problem 1

- For all  $n \geq 1$ :
- $|n^3| = n^3 \leq n^4$
- $|1000n^2| = 1000n^2 \leq 1000n^4$
- It implies that  $|n^3 + 1000n^2| \leq |n^3| + |1000n^2| \leq 1001n^4 \Rightarrow n^3 + 1000n^2 = O(n^4)$

## Solutions: Problem 2

- For all  $x \geq 1$ , let  $f(x) = x - \log x$ , then  $f'(x) = 1 - \frac{1}{x} \geq 0$ .
- Therefore,  $f(x)$  is monotonically increasing.
- It implies that  $x - \log x = f(x) \geq f(1) = 1 > 0 \Rightarrow x > \log x$
- Therefore, for all  $n \geq 1$ ,  $0 \leq \log n < n \Rightarrow |\log n| < n \Rightarrow \log n = O(n)$

## Solutions: Problem 3

- For all  $x \geq 1$ , let  $f(x) = x - \log x$ , then  $f'(x) = 1 - \frac{1}{x} \geq 0$ .
- Therefore,  $f(x)$  is monotonically increasing.
- It implies that  $x - \log x = f(x) \geq f(1) = 1 > 0 \Rightarrow x > \log x$
- Then, replacing  $x$  with  $\sqrt{x}$ , it implies that  $\sqrt{x} > \log \sqrt{x} \Rightarrow 2\sqrt{x} > \log x$
- Therefore, for all  $n \geq 1$ ,  $0 \leq \log n < 2\sqrt{n} \Rightarrow |\log n| < 2\sqrt{n} \Rightarrow \log n = O(\sqrt{n})$

# Solutions: Problem 4

- Assume that  $\exists n_0 > 0, k > 0, \forall n \geq n_0$ :
- $n! \leq kn^c$
- Let  $m_0 = \min\{x \in \mathbb{Z} | x \geq n_0\}$ ,  $h = \min\{x \in \mathbb{Z} | x \geq k\}$ , and  $d = \min\{x \in \mathbb{Z} | x \geq c\}$ .
- Then, let  $m = m_0 + h + 2d + 1 \in \mathbb{Z}$
- $m \geq 2d + 1 \Rightarrow m! \geq (m - 2d) \dots (m - d - 1)(m - d)(m - d + 1) \dots m$

$$\Rightarrow m! \geq (m - d) \prod_{i=1}^d (m - d - i)(m - d + i)$$

$$\Rightarrow m! \geq (m - d) \prod_{i=1}^d ((m - d)^2 - i^2)$$

- For all  $1 \leq i \leq d$  and  $m \geq 2d + 1$ :
- $(m - d)^2 - i^2 \geq (m - d)^2 - d^2 = m^2 - 2md = m(m - 2d) \geq m$
- It implies that  $m! \geq (m - d)m^d$
- However,  $m - d > h$ , then  $m! > hm^d \geq km^c$ . That contradicts the assumption.

## Solutions: Problem 5

- Let  $c = \frac{a}{\log b} > 0$  because  $a > 0$  and  $b > 1$
- For all  $x \geq c > 0$ , let  $f(x) = x - c \log x$ , then  $f'(x) = 1 - \frac{c}{x} \geq 0$ .
- Therefore,  $f(x)$  is monotonically increasing.
- It implies that  $x - c \log x = f(x) \geq f(c) = -\frac{\log k}{\log b}$  where  $k = b^{-f(c)} > 0$
- $c \log x \leq x + \frac{\log k}{\log b} \Rightarrow a \log x \leq x \log b + \log k$   
 $\Rightarrow \log x^a \leq \log kb^x$   
 $\Rightarrow x^a \leq kb^x$
- So, with  $c = \frac{a}{\log b}$  and  $k = b^{-f(c)}$ , for all  $n \geq n_0 = \min\{x \in \mathbb{Z} | x \geq c\}$ :  $n^a \leq kb^n \Rightarrow n^a = O(b^n)$

## Solutions: Problem 6

- Prove that  $\log n! = \Theta(n \log n)$
- For all  $n \geq 4$ :
- $\log n! = \sum_{k=1}^n \log k < \sum_{k=1}^n \log n = n \log n \Rightarrow \log n! = O(n \log n)$
- $\log n! = \sum_{k=1}^n \log k > \sum_{k=\lceil \frac{n}{2} \rceil}^{2\lceil \frac{n}{2} \rceil - 1} \log k > \sum_{k=\lceil \frac{n}{2} \rceil}^{2\lceil \frac{n}{2} \rceil - 1} \log \left\lceil \frac{n}{2} \right\rceil = \left\lceil \frac{n}{2} \right\rceil \log \left\lceil \frac{n}{2} \right\rceil \geq \frac{n}{2} \log \frac{n}{2}$
- 
- $\frac{n}{2} \log \frac{n}{2} = \frac{1}{2} n (\log n - \log 2) = \frac{1}{4} n (\log n + (\log n - \log 4)) \geq \frac{1}{4} n \log n$
- It implies that  $\log n! > \frac{1}{4} n \log n \Rightarrow \log n! = \Omega(n \log n) \Rightarrow \log n! = \Theta(n \log n)$

## Solutions: Problem 7

- For all  $x \geq 1$
- $|1000x^3 - x^2 + 79| \leq 1000x^3 + x^2 + 79 \leq 1080x^3 \Rightarrow 1000x^3 - x^2 + 79 = O(x^3)$
- $1000x^3 - x^2 + 79 \geq 999x^3 + (x^3 - x^2) + 79 > 999x^3 > 0$   
 $\Rightarrow 1000x^3 - x^2 + 79 = \Omega(x^3)$   
 $\Rightarrow 1000x^3 - x^2 + 79 = \Theta(x^3)$



## Solutions: Problem 8

- $f(x) = O(h(x)) \Leftrightarrow \exists c_1 > 0, x_1 > 0, \forall x \geq x_1: |f(x)| \leq c_1 h(x)$
- $g(x) = O(h(x)) \Leftrightarrow \exists c_2 > 0, x_2 > 0, \forall x \geq x_2: |g(x)| \leq c_2 h(x)$
- Then, it implies that  $\exists c = ac_1 + bc_2 > 0, x_0 = \max\{x_1, x_2\} > 0, \forall x \geq x_0$ :
- $|af(x) + bg(x)| \leq a|f(x)| + b|g(x)| \leq (ac_1 + bc_2)h(x) = ch(x)$
- Therefore,  $af(x) + bg(x) = O(h(x))$ .

## Solutions: Problem 9

- $f_1(x) = O(g_1(x)) \Leftrightarrow \exists c_1 > 0, x_1 > 0, \forall x \geq x_1: |f_1(x)| \leq c_1 g_1(x)$
- $f_2(x) = O(g_2(x)) \Leftrightarrow \exists c_2 > 0, x_2 > 0, \forall x \geq x_2: |f_2(x)| \leq c_2 g_2(x)$
- Then, it implies that  $\exists c = c_1 c_2 > 0, x_0 = \max\{x_1, x_2\} > 0, \forall x \geq x_0:$
- $|f_1(x)f_2(x)| \leq c_1 c_2 g_1(x)g_2(x) = c g_1(x)g_2(x)$
- Therefore,  $f_1(x)f_2(x) = O(g_1(x)g_2(x))$ .

## Solutions: Problem 10

- Prove by induction:
- $+ n = 0: 1 = O(1)$ .
- $+ Assume that the statement holds with  $n = k$ :$
- $x^k + a_k x^{k-1} + \dots + a_1 = O(x^k)$
- And  $x = O(x)$ . Then, it implies that  $x^{k+1} + a_k x^k + \dots + a_1 x = O(x^{k+1})$
- And  $a_0 = O(x^{k+1})$ .
- So,  $x^{k+1} + a_k x^k + \dots + a_1 x + a_0 = O(x^{k+1})$ .

# Solutions: Problem 11

- Let  $c = \frac{1}{n+1} > 0$  and  $x_0 = \max\{(n+1)|a_{n-1}|, \sqrt{(n+1)|a_{n-2}|}, \dots, \sqrt[n]{(n+1)|a_0|}\}$ . Then, for all  $x \geq x_0$ , we have:

$$\begin{aligned} x &\geq (n+1)|a_{n-1}| \\ x^2 &\geq (n+1)|a_{n-2}| \end{aligned}$$

$$\dots \\ x^n \geq (n+1)|a_0|$$

- Therefore, it implies that:

$$\frac{1}{n+1}x^n \geq |a_{n-1}|x^{n-1} \geq -a_{n-1}x^{n-1}$$

$$\frac{1}{n+1}x^n \geq |a_{n-2}|x^{n-2} \geq -a_{n-2}x^{n-2}$$

$$\dots \\ \frac{1}{n+1}x^n \geq |a_0| \geq -a_0$$

- $\Rightarrow x^n + a_{n-1}x^{n-1} + \dots + a_0 \geq \frac{1}{n+1}x^n = cx^n$
- So,  $x^{k+1} + a_kx^k + \dots + a_1x + a_0 = \Omega(x^{k+1})$ .
- Therefore,  $x^{k+1} + a_kx^k + \dots + a_1x + a_0 = \Theta(x^{k+1})$ .

## Solutions: Problem 12

- For all  $x \geq 1$ , let  $f(x) = x - \log x$ , then  $f'(x) = 1 - \frac{1}{x} \geq 0$ .
- Therefore,  $f(x)$  is monotonically increasing.
- It implies that  $x - \log x = f(x) \geq f(1) = 1 > 0 \Rightarrow x > \log x$
- Then, replacing  $x$  with  $x^c$ , it implies that  $x^c > \log x^c \Rightarrow \frac{1}{c}x^c > \log x$
- Therefore, for all  $n \geq 1$ ,  $0 \leq \log n < \frac{1}{c}n^c \Rightarrow |\log n| < \frac{1}{c}n^c \Rightarrow \log n = O(n^c)$