



BIG O-NOTATION

CS A250 – C++ Programming Language 2

INTRODUCTION

- Key factors in designing software:
 - **Reliability:** Your program should anticipate and handle all types of exceptional circumstances.
 - **Flexibility:** Your program should be easy to modify to handle circumstances that may change in the future.
 - **Reusability and expandability:** If your program is successful, it will frequently spawn new computing needs; you should be able to incorporate solutions to these new needs into the original system with relative ease.

INTRODUCTION (CONT.)

- Key factors in designing software:
 - **User-friendliness:** Your program should be clearly documented so that it is easy to use—*both* internal and external documentation).
 - **Structured organization:** The system should be divided into compact modules, each of which is responsible for a specific, well-defined task.
 - **Efficiency:** The system should make optimal use of **time** and **space** resources.

INTRODUCTION (CONT.)

- As a **computer scientist** you should:
 - Have a detailed knowledge of **algorithms** and **data storage techniques**
 - Do not re-invent the wheel!
 - Apply these techniques when designing software
 - Choose the most appropriate algorithms and tailor them to the application you are creating

EFFICIENCY

- A system to be **efficient** needs to make optimal use of
 - Storage space
 - Programming effort
 - Computer time
 - *Let's look at each one of them...*

EFFICIENCY – STORAGE SPACE

- **Storage space** is the amount of memory required to store data
 - If the list is too large to be kept in high-speed memory, you need to find other options
 - One approach for sorting
 - Divide lists into sub-lists
 - Sort them internally within high-speed memory
 - Merge the sorted sub-lists externally

EFFICIENCY – PROGRAMMING EFFORT

- Several factors to consider:
 - An **algorithm** is going to be running only a few times
 - No need for a programmer to spend days/weeks investigating sophisticated algorithms
 - Should you use **recursion**?
 - Consider **readability**
 - Consider if easy to implement
 - Some programming languages do **not** have recursion (FORTRAN, COBOL)
 - **Simplicity** and **correctness** are essential

EFFICIENCY – COMPUTER TIME

- One **computer** may be **faster** than another
 - You should use the **same computer** to **test** different algorithms
- Some **languages** are better suited for certain algorithms.
- Some **compilers** generate better machine code than others
- Some **programmers** write better programs than others

TIME EFFICIENCY

- Although the efficient use of both **time** and **space** is important, **inexpensive memory** has **reduced** the significance of **space efficiency**. Thus, we will focus primarily on **time efficiency**.
- How can we measure time efficiency?

COMPUTATIONAL COMPLEXITY

- Same problem can be solve with algorithms that differ in efficiency.
- Differences may be immaterial for processing a small number of data items, but can grow with the amount of data.
- **Computational complexity** indicates how costly it is to apply an algorithm:
 - The cost can be measured in a variety of ways.
 - We will focus on the relationship between **input size** and **execution time**.

EXECUTION TIME

- How do we measure the efficiency of algorithms in terms of execution time?
 - Use a systematic and quantitative way to evaluate comparatively
 - Example: Sorting
 - Determine a function $f(n)$, where n is the size of the data
 - How many comparisons to sort the data?

EFFICIENCY OF ALGORITHMS

- Efficiency can be measured for
 - **Best case**
 - Searching a key in a list:
 - The key is immediately found
 - **Worst case**
 - Searching a key in a list:
 - The key is at the end of the list or there is no key
 - **Average case**
 - Difficult to determine

EFFICIENCY OF ALGORITHMS (CONT.)

- How do we analyze a particular algorithm?
 - We **count the number of operations** the algorithm executes
 - We do **NOT** focus on the actual computer time to execute the algorithm
 - Why not?
 - A particular algorithm can be implemented on a variety of computers and the speed of the computer can affect the execution time
 - **But** the number of operations performed by an algorithm would be the same

EXAMPLE 1

- How many times is each statement executed?

```
cout << "Enter two numbers: ";    // 1 time
cin >> num1 >> num2;

if (num 1 > num2)
    largest = num1;
else
    largest = num2;

cout << "The largest number is"
    << largest << endl;
```

EXAMPLE 1 (CONT.)

- How many times is each statement executed?

```
cout << "Enter two numbers: ";    // 1 time
cin >> num1 >> num2;              // 1 time

if (num 1 > num2)
    largest = num1;
else
    largest = num2;

cout << "The largest number is"
    << largest << endl;
```

EXAMPLE 1 (CONT.)

- How many times is each statement executed?

```
cout << "Enter two numbers: ";    // 1 time
cin >> num1 >> num2;              // 1 time

if (num1 > num2)                   // 1 time
    largest = num1;
else
    largest = num2;

cout << "The largest number is"
    << largest << endl;
```


EXAMPLE 1 (CONT.)

- How many times is each statement executed?

```
cout << "Enter two numbers: ";    // 1 time
cin >> num1 >> num2;              // 1 time

if (num1 > num2)                   // 1 time
    largest = num1;               // 0 to 1 time
else
    largest = num2;

cout << "The largest number is"
    << largest << endl;
```

EXAMPLE 1 (CONT.)

- How many times is each statement executed?

```
cout << "Enter two numbers: ";    // 1 time
cin >> num1 >> num2;              // 1 time

if (num1 > num2)                   // 1 time
    largest = num1;               // 0 to 1 time
else
    largest = num2;               // 0 to 1 time

cout << "The largest number is"
    << largest << endl;
```

EXAMPLE 1 (CONT.)

- How many times is each statement executed?

```
cout << "Enter two numbers: ";      // 1 time
cin >> num1 >> num2;                // 1 time

if (num1 > num2)                     // 1 time
    largest = num1;                  // 0 to 1 time
else
    largest = num2;                  // 0 to 1 time

cout << "The largest number is"
    << largest << endl;             // 1 time
```

EXAMPLE 1 (CONT.)

- How many times is each statement executed?

```
cout << "Enter two numbers: ";    // 1 time
cin >> num1 >> num2;              // 1 time

if (num1 > num2)                   // 1 time
    largest = num1;               // 0 to 1 time
else
    largest = num2;               // 0 to 1 time

cout << "The largest number is"
    << largest << endl;          // 1 time
```

What is the total?

EXAMPLE 1 (CONT.)

- How many times is each statement executed?

```
cout << "Enter two numbers: ";    // 1 time
cin >> num1 >> num2;              // 1 time

if (num1 > num2)                   // 1 time
    largest = num1;               // 0 to 1 time
else
    largest = num2;               // 0 to 1 time

cout << "The largest number is"
    << largest << endl;          // 1 time
```

Total = 1 + 1 + 1 + 1 + 1 = 5

EXAMPLE 2

- How many times is each statement executed?

```
int num = 0;
int count = 0;
int sum = 0;
cout << "Enter 5 integers: ";
while (count < 3)
{
    cin >> num;
    sum += num;
    ++count;
}
cout << "The sum is"
    << sum << endl;
```

EXAMPLE 2 (CONT.)

- How many times is each statement executed?

```
int num = 0;           // 1 time
int count = 0;         // 1 time
int sum = 0;           // 1 time
cout << "Enter 5 integers: "; // 1 time
while (count < 3)
{
    cin >> num;
    sum += num;
    ++count;
}
cout << "The sum is"
    << sum << endl;
```

EXAMPLE 2 (CONT.)

- How many times is each statement executed?

```
int num = 0;           // 1 time
int count = 0;         // 1 time
int sum = 0;           // 1 time
cout << "Enter 5 integers: "; // 1 time
while (count < 3)      // 4 times
{
    cin >> num;
    sum += num;
    ++count;
}
cout << "The sum is"
    << sum << endl;
```


EXAMPLE 2 (CONT.)

- How many times is each statement executed?

```
int num = 0;           // 1 time
int count = 0;         // 1 time
int sum = 0;           // 1 time
cout << "Enter 5 integers: "; // 1 time
while (count < 3)      // 4 times
{
    cin >> num;         // 3 times
    sum += num;         // 3 times
    ++count;           // 3 times
}
cout << "The sum is"
    << sum << endl;
```

EXAMPLE 2 (CONT.)

- How many times is each statement executed?

```
int num = 0;           // 1 time
int count = 0;         // 1 time
int sum = 0;           // 1 time
cout << "Enter 5 integers: "; // 1 time
while (count < 3)      // 4 times
{
    cin >> num;         // 3 times
    sum += num;         // 3 times
    ++count;           // 3 times
}
cout << "The sum is"
    << sum << endl;    // 1 time
```

EXAMPLE 2 (CONT.)

- How many times is each statement executed?

```
int num = 0;           // 1 time
int count = 0;         // 1 time
int sum = 0;           // 1 time
cout << "Enter 5 integers: "; // 1 time
while (count < 3)      // 4 times
{
    cin >> num;         // 3 times
    sum += num;         // 3 times
    ++count;           // 3 times
}
cout << "The sum is"
    << sum << endl;    // 1 time
```

Total = 18

EXAMPLE 2 (CONT.)

- How many times is each statement executed?

```
int num = 0;           // 1 time
int count = 0;         // 1 time
int sum = 0;           // 1 time
cout << "Enter 5 integers: "; // 1 time
while (count < 3)      // 4 times
{
    cin >> num;         // 3 times
    sum += num;         // 3 times
    ++count;           // 3 times
}
cout << "The sum is"
    << sum << endl;    // 1 time
```

Execution time = 1 + 1 + 1 + 1 + (3 + 1) + (3 + 3 + 3) + 1

EXAMPLE 2 (CONT.)

Assume we do not know that the loop will go around 3 times, and we know that it will go around n times:

```
while (count <  $n$ )    // executes  $n + 1$  times
{
    cin >> num;
    sum += num;    // each executes  $n$  times
    ++count;
}
```

Execution time = 1 + 1 + 1 + 1 + ($n + 1$) + ($n + n + n$) + 1

EXAMPLE 2 (CONT.)

We can simplify our expression:

$$1 + 1 + 1 + 1 + (\textcolor{red}{n} + 1) + (\textcolor{red}{n} + \textcolor{red}{n} + \textcolor{red}{n}) + 1$$

$$5 + (\textcolor{red}{n} + 1) + (\textcolor{red}{n} + \textcolor{red}{n} + \textcolor{red}{n})$$

$$5 + \textcolor{red}{n} + 1 + (\textcolor{red}{n} + \textcolor{red}{n} + \textcolor{red}{n})$$

$$6 + \textcolor{red}{n} + (\textcolor{red}{n} + \textcolor{red}{n} + \textcolor{red}{n}) = 6 + 4\textcolor{red}{n}$$

The execution time for this algorithm is $= 6 + 4\textcolor{red}{n}$

PRACTICAL SHORTCUTS

- Ignore **lesser terms** (find the **dominant term**)

- $n^3 + 5n^2 + n \rightarrow n^3$

- Ignore **coefficients**

- $5n^2 \rightarrow n^2$

- Ignore **bases of logarithms**

- $\log_{10} n \rightarrow \log n$

- $\log_2 n \rightarrow \log n$

(same)

BIG O-NOTATION

○ Big O-notation:

- Mathematical measuring tool to quantitatively evaluate algorithms
- Also called **Big Oh** notation, **Landau** notation, **Bachmann-Landau** notation, and **asymptotic** notation.
- Formal definition:

$$f(x) = O(g(x)) \text{ as } x \rightarrow \infty$$

- Categorizes algorithms with respect to **execution time**
→ in terms of *growth rate*, not speed

BACK TO EXAMPLE 2 (CONT.)

Going back to example 2, we can simplify as follows:

Ignore lesser terms $\rightarrow 6 + 4n \equiv 4n$

Ignore coefficients $\rightarrow 4n \equiv n$

The algorithm's **complexity** is **$O(n)$**

COMMON GROWTH-RATE FUNCTIONS

- **$O(1)$** - Constant
 - Time required is constant
 - Independent of problem size
- **$O(\log n)$** - Logarithmic
 - Time requirement increases slowly as the size increases
 - The base does not affect the growth rate
- **$O(n)$** - Linear
 - Increases directly with size
- **$O(n \log n)$** - Log-linear
 - Time requirement increases more rapidly than a linear algorithm
 - Typical algorithms divide the problem into smaller problem that are solved separately

COMMON GROWTH-RATE FUNCTIONS

- $O(n^2)$ - Quadratic

- Time requirement increases rapidly with the size of the problem
- Practical for small problems only

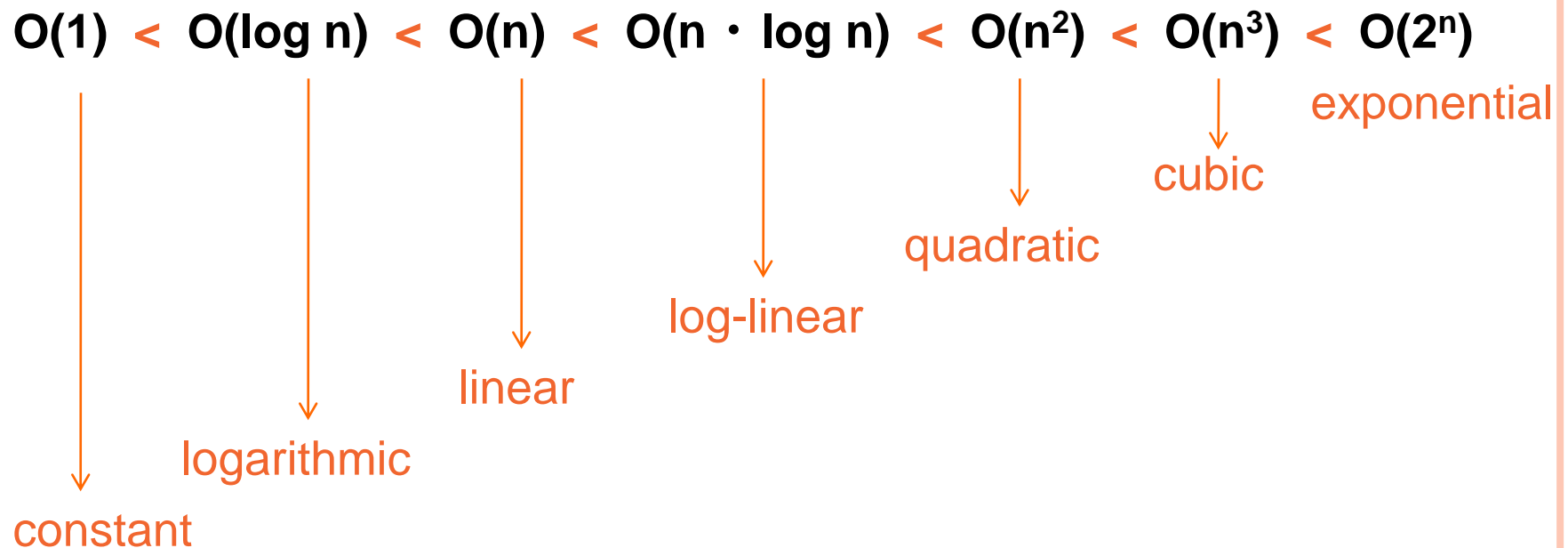
- $O(n^3)$ - Cubic

- Time requirement increases more rapidly than the quadratic algorithm
- Practical for small problems only

- $O(2^n)$ - Exponential

- Time requirement increases too rapidly to be practical

ORDER OF GROWTH



REDUCING AN EXPRESSION TO BIG-O

- The number of logical operations in an algorithm as a function of n can be reduced to a simplified **O-notation**:

$$n + 4n^2 + 4n$$

What is the dominant term? $\rightarrow 4n^2$

What is the O-notation? $\rightarrow O(n^2)$

EFFICIENCY OF BINARY SEARCH

- We have seen that **Binary Search** cuts the list in **half** after each **comparison**.
- Assume you have a list of 32 items.
- After comparing the **middle element**, if the middle element is **not** the item we are looking for, only one half of the list will be considered.

EFFICIENCY OF BINARY SEARCH (CONT.)

- Start with 32 items in the list.
- Roughly, we cut the list in half after each comparison
 - Assume the item we are looking for is not in the list.

32

16

8

4

2

1

EFFICIENCY OF BINARY SEARCH (CONT.)

- Start with 32 items in the list.
- Roughly, we cut the list in half after each comparison
 - Assume the item we are looking for is not in the list.

32		2^5
16		2^4
8	equivalent to	2^3
4		2^2
2		2^1
1		2^0

EFFICIENCY OF BINARY SEARCH (CONT.)

- Start with 32 items in the list.
- Roughly, we cut the list in half after each comparison
 - Assume the item we are looking for is not in the list.

32	2^5		$\log_2 32 = 5$
16	2^4		$\log_2 16 = 4$
8	2^3	in logarithmic notation	$\log_2 8 = 3$
4	2^2		$\log_2 4 = 2$
2	2^1		$\log_2 2 = 1$
1	2^0		$\log_2 1 = 0$

EFFICIENCY OF BINARY SEARCH (CONT.)

- Start with 32 items in the list.
- Roughly, we cut the list in half after each comparison
 - Assume the item we are looking for is not in the list.

$$\log_2 32 = 5$$

$$\log_2 16 = 4$$

$$\log_2 8 = 3$$

$$\log_2 4 = 2$$

$$\log_2 2 = 1$$

$$\log_2 1 = 0$$

Therefore, the **growth rate** is **logarithmic**:

$$O(\log_2 n)$$

But we **omit the base** and we have:

$$O(\log n)$$

EFFICIENCY OF BINARY SEARCH (CONT.)

- The **running time** of **binary search** for worst and average case is **$O(\log n)$**
- This is better than **linear search**, which has worst and average case **$O(n)$**
- BUT, for Binary search to work, we need to assume that all **elements are in order**.



O-NOTATION (END)