# Unit 02: Pseudocode and Counting

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CSC 225: Algorithms and Data Structures I

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#### Unit 02 Overview

- ► Supplemental Reading:
  - ► Algorithm Design and Analysis. Michael Goodrich and Roberto Tamassia
    - ► Pages 1 9
- ► Learning Objectives: (You should be able to...)
  - understand why we will use pseudocode to support or analysis of algorithms and data structures
  - ▶ understand the syntax of pseudocode, and how it maps to operations in a programming language like Java, C, or C++
  - ▶ understand the methodology we will use in this course to analyze algorithms and data structures, based on the size of the input data, *n*
  - determine the number of operations required to execute an algorithm through an analysis of pseudocode

# Algorithms and Data structures

- ➤ An algorithm is a step-by-step procedure for performing some task in a finite amount of time
- ► A data structure is a systematic way of organizing and accessing data

#### ► Analysis:

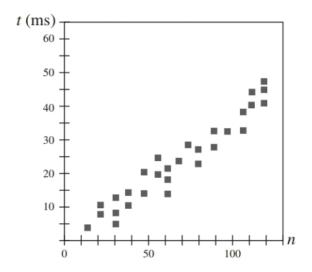
- ▶ In the world of IT, *scalability* refers to the ability of a system to gracefully accommodate growing sizes of inputs or amounts of workload
- ▶ The main focus of our analysis will be to character *running time*,
- but it is also important to consider *space usage* as well

# **Analysis**

- ► What is the "right" way to analyze an algorithm?
  - ▶ Idea: We could just execute it and see how long it takes to complete
  - ► Problem: It is difficult to setup a controlled environment to accurately compare two different algorithms
- ▶ What is the best input data to test with?
  - ► Probably best to perform several experiments on many different test inputs and with test inputs of various sizes
- ▶ What is the best way to interpret / visualize the results?
  - ▶ Plot on a graph:
    - $\triangleright$  x-coordinate represents the input size, n
    - $\triangleright$  y-coordinate represents the running time, t

# **Analysis**

- ► Assume our approach is plot multiple input sets on a graph that illustrates the relationship between input size and execution time
  - ► The hardware environment will affect the results (processor, memory, etc)
  - ▶ and the software environment will too (OS, programming language, compiler)
  - ➤ So we can't get you to compare results!
- ▶ If we can control for these variables, we may see something like:



- ▶ In a perfect world, our analysis can:
  - ► Take into account all possible inputs
  - ► Evaluate relative efficiency of any two algorithms in a way that is not dependent on the hardware and software environment

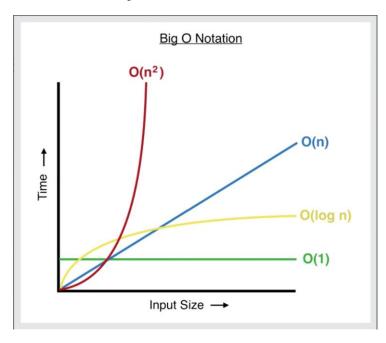
# **Analysis**

- ► How can we achieve these goals?
  - ➤ We want an environment where analysis can be performed that illustrates the efficiency of an algorithm that accounts for all input sizes, without having to implement and execute multiple controlled experiments

▶ The rest of this unit will focus on an approach that aims to do this

#### Methodology

- Proposed methodology:
  - lacktriangle Associate each algorithm a function f(n) that characterizes the running time of the algorithm in terms of the input size n
- Methodology requirements:
  - ► A language for describing the algorithms we want to analyze
    - ► Pseudocode
  - ► A metric for measuring algorithm runtime
    - ► Counting primitive operations
  - ► An approach for characterizing the runtimes
    - ► Worse-case analysis (big-Oh)
- ► Result:



#### Pseudocode

- ➤ Programmers are often asked to describe algorithms in a way that is intended to be read by humans (as opposed to executed by a machine)
  - ► Compared to code, these descriptions should be easier to read and understand
- ➤ These descriptions facilitate high-level analysis of a data structure or algorithm we call such descriptions *pseudocode*

#### Example

#### Java code:

```
public int arrayMax(int[] A, int n) {
   int currentMax = A[0];
   for (int k=1; k < n; k++) {
       if (currentMax < A[k]) {</pre>
          currentMax = A[k];
   return currentMax;
```

#### Pseudocode:

```
Algorithm arrayMax(A, n)
    Input: An array A storing n \ge 1 integers
    Output: The maximum element in A
    currentMax \leftarrow A[0]
    for k \leftarrow 1 to n-1 do
        if currentMax < A[k] then
            currentMax \leftarrow A[k]
        end
    end
    return currentMax
```

#### **Expressions**

- Standard mathematical symbols are used to express numeric and Boolean expressions
- ▶ The left arrow sign ← is used as the assignment operator
  - equivalent to the = operator used for assignment in Java, C, C++, etc
- ► The equal sign = is used to test equality
  - equivalent to using == in Java, C, C++, etc

#### **Method Declarations**

► General form:

```
Algorithm name(param1, param2, ...):
    Input: <Description>
    Output : <Description>
```

➤ Specific example:

**Algorithm** arrayMax(A, n):

*Input*: An array A storing  $n \ge 1$  integers

*Output*: The maximum element in *A* 

#### Method calls and return statements

- ► Method call:
  - ► [object.]method(args)
  - ightharpoonup Example: arrayMax(A, 13)

- ► Return statements:
  - **return** value

# Array indexing and field selection

- ightharpoonup A[i] represents the *i*th cell in the array A
  - ▶ An array of size n is indexed from A[0] to A[n-1]

- ▶ Dot (.) notation is used to access fields within an structure/object
  - ightharpoonup r. key represents the key field in the structure named r

#### **Decision structures**

```
    General form:
        if condition then
        true-actions
        [else
            false-actions]
        end
```

```
    Specific example:
    if currentMax < A[k] then</li>
    currentMax ← A[k]
    end
```

# Repetition structures: while loops

General form:while condition doactionsend

Specific example:

while k < n do  $count \leftarrow 2 * count$   $k \leftarrow k + 1$ 

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end

#### Repetition structures: repeat loops

► General form:

```
repeat
actions
until condition
```

➤ Specific example:

```
repeat count \leftarrow 2 * count k \leftarrow k + 1 until k \ge n
```

# Repetition structures: for loops

► Specific example:

```
for k \leftarrow 1 to n-1
count \leftarrow 2 * count
end
```

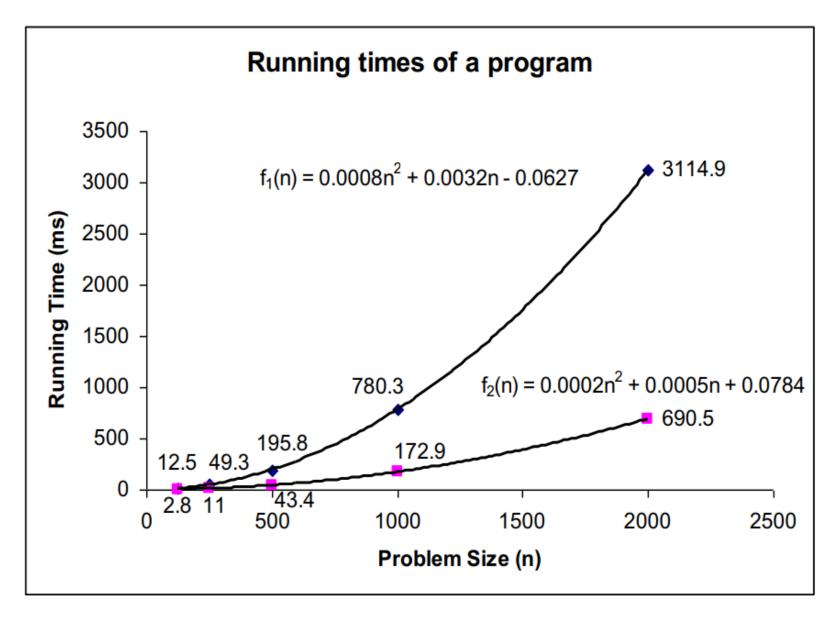
#### A metric for measuring algorithm runtime

- ► Two approaches
  - ► Counting primitive operations and computing upper and lower bounds
    - ► Covered in CSC 115/116, CSC 225, CSC 326, CSC 425, CSC 426, etc.
  - ▶ Instrument the code to measure computer clock cycles
    - ► Covered in in SENG 265, (and sometimes in projects of higher level Systems courses)

# What is the running time?

```
Algorithm arrayMax(A, n)
   Input: An array A storing n \ge 1 integers
   Output: The maximum element in A
   currentMax \leftarrow A[0]
   for k \leftarrow 1 to n-1 do
      if currentMax < A[k] then
         currentMax \leftarrow A[k]
      end
   end
   return currentMax
```

#### Instrumentation to measure clock cycles (SENG 265)



# Instrumentation to measure clock cycles (SENG 265)

- ► Limitations of this approach:
  - ▶ We must first implement the algorithm (in Java, C, C++, etc)
  - ▶ We need to determine which inputs to test, and how many
  - ▶ We must run all experiments in the same, controlled environment

- ► Earlier in this unit, our claim was that the methodology we will use in the course will:
  - ► take all possible inputs into account
  - compare the efficiency of two different algorithms independent from the underlying hardware and software configurations
  - be able to perform the analysis before needed to implement a solution

# Using our methodology

```
Algorithm arrayMax(A, n)
   Input: An array A storing n \ge 1 integers
   Output: The maximum element in A
   currentMax \leftarrow A[0]
   for k \leftarrow 1 to n-1 do
      if currentMax < A[k] then
         currentMax \leftarrow A[k]
      end
   end
   return currentMax
```

Could count line numbers

Why are some arrows colored differently?

What are some other issues with counting line numbers?

#### **Primitive Operations**

- Our approach will be to count primitive operations:
  - Assignment statements (A)
  - ► Comparisons (C)
  - ► Boolean expressions (B)
  - ► Array indexing (I)
  - ► Selector/Object references (R)
  - ► Arithmetic operations:
    - ► Add/Subtract (S)
    - ► Multiplication/Division (M)
  - ► Trigonometric operations (sin, cos, tan) (T)
  - ► Method/function calls (M)

A for-loop declaration has multiple primitive operations (assignment, comparison, addition, possibly array indexing...)

#### Runtime analysis

- ► Categories of algorithm running times:
  - ▶ Best-case analysis:  $T_b(n)$
  - ightharpoonup Average-case analysis:  $T_a(n)$
  - ► Worst-case analysis: *T*(*n*)
- ▶ Which is the best choice?

```
Algorithm arrayMax(A, n)
    Input: An array A storing n \ge 1 integers
    Output: The maximum element in A
    currentMax \leftarrow A[0]
    for k \leftarrow 1 to n-1 do
        if currentMax < A[k] then
            currentMax \leftarrow A[k]
        end
    end
    return currentMax
```

#### Runtime analysis

- Average-case
  - > calculate the expected running times based on a given input data distribution
  - requires math and probability theory (we are getting there)
  - ▶ focused on more in CSC 226 (but mentioned in CSC 225 as well)
- ► Best-case
  - The minimum number of primitive operations (depending on n), given the most advantageous and best input configuration of size n
- ► Worst-case:
  - ▶ What is the maximum number of primitive operations (depending on *n*) executed by the algorithm, taken over all inputs of size *n*?
  - ▶ Worst-case analysis is most common and may aid in the design of the algorithm

```
Algorithm arrayMax(A, n)
   Input: An array A storing n \ge 1 integers
   Output: The maximum element in A
   currentMax \leftarrow A[0]
                                     1I + 1A
   for k \leftarrow 1 to n-1 do
                                     1A
      if currentMax < A[k] then
         currentMax \leftarrow A[k]
      end
   end
   return currentMax
```

```
Algorithm arrayMax(A, n)
```

*Input*: An array A storing  $n \ge 1$  integers

Output : The maximum element in A

How many times is our condition  $k \le n - 1$  executed?

The n-1 times the condition is true, when k is 1, 2, 3, ..., n-1.

Once when the loop is terminated, when k = n.

```
Algorithm arrayMax(A, n)
```

*Input*: An array A storing  $n \ge 1$  integers

1I + 1A

*Output*: The maximum element in *A* 

After each iteration of the loop, we will increment k by 1 (sum operation (S) followed by an assignment operation (A)).

This will happen at the end of all n-1 iterations of the loop.

```
currentMax \leftarrow A[0]
for k \leftarrow 1 to n-1 do
   if currentMax < A[k] then
    currentMax \leftarrow A[k]
end
end
return currentMax
```

1A + (n)\*(1C) + (n-1)\*(1S+1A)

**Algorithm** arrayMax(A, n)

return currentMax

*Input*: An array A storing  $n \ge 1$  integers

*Output*: The maximum element in *A* 

Similarly, for all n-1 iterations of the loop the if-statement is executed (an array index (I) and a comparison (C)).

And in the worst-case, we update currentMax each time.

```
\begin{array}{ll} \textit{currentMax} \leftarrow A[0] & \text{1I} + \text{1A} \\ & \text{for } k \leftarrow 1 \text{ to } n-1 \text{ do} & \text{1A} + (n)*(1C) + (n-1)*(1S+1A) \\ & \textit{if } \textit{currentMax} < A[k] \text{ then} & (n-1)*(1I + 1C) \\ & \textit{currentMax} \leftarrow A[k] & (n-1)*(1I + 1A) \\ & \text{end} \\ & \text{end} \end{array}
```

```
Algorithm arrayMax(A, n)
   Input: An array A storing n \ge 1 integers
   Output: The maximum element in A
   currentMax \leftarrow A[0]
                                     1I + 1A
   for k \leftarrow 1 to n-1 do
                                     1A + (n)*(1C) + (n-1)*(1S+1A)
      if currentMax < A[k] then (n-1)*(1I + 1C)
         currentMax \leftarrow A[k]
                                     (n-1)*(1I + 1A)
      end
   end
                                                 T(n) = 4 + (n - 1)(6) + n

T(n) = 7n - 2
   return currentMax
```