#### CS 171: Introduction to Computer Science II

#### **Algorithm Analysis**

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# Today

- Hw1 discussion
- Recap: linear search and binary search
- Algorithm Analysis
- Big-O Notation
- Loop Analysis

# Hw1 Discussion

- Read the instructions carefully
- Think before you code
- Useful classes/methods
  - -ArrayList
  - -Random Number generation

# ArrayList

- Use generics parameterized types
  - Type parameters have to be instantiated as reference types
- Autoboxing
  - Autoboxing: Automatically casting a primitive type to a wrapper type
  - –Auto-unboxing: automatically casting a wrapper type to a primitive type

```
ArrayList<Integer> numbers = new ArrayList<Integer>();
numbers.add(1001);
int mynumber = numbers.remove(0);
```

# ArrayList

- Useful methods
  - -add(E e): Appends the specified element to the end of this list
  - -size(): returns the number of elements in this list
  - –remove(int index): Removes the element at the specified position in this list. Shifts any subsequent elements to the left (subtracts one from their indices)
  - –get(int index): Returns the element at the specified position in this list.

```
ArrayList<Integer> numbers = new ArrayList<Integer>();
numbers.add(1001);
int n = numbers.size();
int mynumber2 = numbers.get(0);
int mynumber = numbers.remove(0);
```

# Random number generation

 If you want to generate random test numbers Math.random() method:

double x = Math.random();

This generates a double between [0.0, 1.0].

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# Search in an Array

- Unordered array: ~N
- Order array: ~lgN



#### Hunts Needle in a Haystack

How LONG does it take to find a needle in a haystack? Jim Moran, Washington, D. C., publicity man, recently dropped a needle into a convenient pile of hay, hopped in after it, and began an intensive search for (a) some publicity and (b) the needle. Having found the former, Moran abandoned the needle hunt.

## Review question 1

- The maximum number of elements to examine to complete binary search of 30 elements is:
  - -A: 1
  - -B: 30
  - -C: 7
  - -D: 5

## **Review Question 2**

- True or false: It is generally faster to find an existing item in an ordered array than a missing one (item not there).
- Trust or false: It is generally faster to search an item in an ordered array than in an unordered array of the same size

# **Algorithm Analysis**

 An algorithm is a method for solving a problem expressed as a sequence of steps that is suitable for execution by a computer (machine)

-E.g. Search, insertion, deletion in an array

- We are interested in designing good algorithms
  - -Linear search vs. binary search
- Good algorithms
  - -Running time
  - -Space usage (amount of memory required)

# Running time of an algorithm

- Running time typically increases with the input size (problem size)
- Also affected by hardware and software environment
- We would like to focus on the relationship between the running time and the input size



#### How to measure running time

- Experimental studies
- Theoretical analysis

# **Experimental Studies**

- Write a program implementing the algorithm
- Run the program with inputs of varying size and composition
- Use a method like System.currentTimeMillis() to get an accurate measure of the actual running time
- Plot the results



## Limitations of Experiments

- It is necessary to implement the algorithm, which may be difficult
- Results may not be indicative of the running time on other inputs not included in the experiment.
- In order to compare two algorithms, the same hardware and software environments must be used

#### MY HOBBY: EXTRAPOLATING



# Mathematical Analysis - insight

- Total running time of a program is determined by two primary factors:
  - -Cost of executing each statement (property of computer, Java compiler, OS)
  - –Frequency of execution of each statement (property of program and input)

# **Algorithm Analysis**

- Algorithm analysis:
  - -Determine frequency of execution of statements
  - –Characterizes running time as a function of the input size
- Benefit:
  - -Takes into account all possible inputs
  - –Allows us to evaluate the speed of an algorithm independent of the hardware/software environment

# Analysis Method

- Count the number of primitive operations executed as a function of input size
- A primitive operation corresponds to a low-level (basic) computation with a constant execution time
  - -Evaluating an expression
  - -Assigning a value to a variable
  - -Indexing into an array
- The number of primitive operations is a good estimate that is proportional to the running time of an algorithm

#### Average-case vs. worst-case

- An algorithm may run faster on some inputs than it does on others (with the same input size)
- Average case: taking the average over all possible inputs of the same size
  - -Depends on input distribution
- Best case
- Worst case
  - -Easier analysis
  - -Typically leads to better algorithms



# Loop Analysis

- Programs typically use loops to enumerate through input data items
- Count number of operations or steps in loops
- Each statement within the loop is counted as a step

## Example 1

```
double sum = 0.0;
for (int i = 0; i < n; i ++) {
    sum += array[i];
}</pre>
```

#### How many steps?

Only count the loop statements (update to the loop variable i is ignored).

# Example 1: Solution

```
double sum = 0.0;
for (int i = 0; i < n; i ++) {
    sum += array[i];
}
```

#### How many steps?

Loop will be executed n times; and there is 1 loop statement. So overall:

N

#### Example 2

```
double sum = 0.0;
for (int i = 0; i < n; i += 2) {
    sum += array[i];
}
```

#### How many steps?

#### **Example 2: Solution**

```
double sum = 0.0;
for (int i = 0; i < n; i += 2) {
    sum += array[i];
}
```

#### How many steps?

Loop will be executed n/2 times. So overall:

*n*/2

#### Example 3 – Multiple Loops

for (int i = 0; i < n; i ++) {
 for (int j = 0; j < n; j ++) {
 int x = i\*j;
 sum += x;
 }
}</pre>

How many steps?

#### Example 3 – Solution

# - for (int i = 0; i < n; i ++) { for (int j = 0; j < n; j ++) { int x = i\*j; sum += x; } \_</pre>

How many steps?

2 loops, each loop *n* times, so overall:  $2 n^2$ 

# Increase of Cost w.r.t. *n*

- Example 1 takes twice as many steps (n) as Example 2 (n/2), but both of them are <u>linear</u> to the input size n
  - If n is 3 times larger, both costs are 3 times larger
- Example 3 (2n<sup>2</sup>)is different:
  - If n is 3 times larger, it becomes 9 times more expensive.
  - Therefore the cost is <u>quadratic</u> w.r.t. to problem size.

# Increase of Cost with Growth of *n*

- In practice we care a lot about how the cost increases w.r.t. the problem size, rather than the absolute cost.
- Therefore we can ignore the constant scale factor in the cost function, and concentrate on the part relevant to *n*
- We need formal mathematical definitions and tools for comparing the cost

#### **Tilde Notation**

- Tilde notation: ignore insignificant terms
- Definition: we write f(n) ~ g(n) if f(n)/g(n) approaches 1 as n grows

- $2n + 10 \sim 2n$
- $3n^3 + 20n^2 + 5 \sim 3n^3$

# **Big-Oh Notation**

Given functions f(n) and 1
 g(n), we say that f(n) is
 O(g(n)) if there are
 positive constants
 c and n<sub>0</sub> such that

 $f(n) \leq cg(n)$  for  $n \geq n_0$ 

• Example: 2n + 10 is O(n)- pick c = 3 and  $n_0 = 10$ 



# **Big-Oh Example**

• Example: the function  $n^2$  is <u>not</u> O(n)

$$-n^2 \leq cn$$

- $-n \leq c$
- The above
  inequality cannot
  be satisfied since c
  must be a constant



# **Big-Oh and Growth Rate**

- The big-Oh notation gives an <u>upper bound</u> on the growth rate of a function
- The statement "f(n) is O(g(n))" means that the growth rate of f(n) is no more than the growth rate of g(n)
- We can use the big-Oh notation to rank functions according to their growth rate

#### Important Functions in Big-Oh Analysis

- –Constant: 1
- –Logarithmic: log n<sub>4096</sub>
- –Linear: *n*
- $-N-Log-N: n \log n$
- –Quadratic: *n*<sup>2</sup>
- **–**Cubic: *n*<sup>3</sup>
- –Polynomial: *n*<sup>d</sup>
- -Exponential:  $2^n$
- -Factorial: *n*!

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# **Growth Rate**

From calculus we know that: in terms of the order:
 *exponentials* > polynomials > logarithms > constant.

 $O(3^{n})$ Exponential  $O(n^3)$ Polynomial  $O(n^2)$ Log-linear  $O(n\log n)$ Linear O(n)Log  $\log n$ ) **Increasing order** Constant

#### **Big-Oh Analysis**

• Write down cost function *f*(*n*)

1. Look for highest-order term

- 2. Drop constant factors
- Examples
  - $-3n^3 + 20n^2 + 5$ -n log n + 10

#### Example 4

for (int i = 0; i < n; i ++) {
 for (int j = i; j < n; j ++) {
 sum += i\*j;
 }
}</pre>

#### **Example 4: Solution**

for (int i = 0; i < n; i ++) { for (int j = i; j < n; j ++) { sum += i\*j; }  $n+(n-1)+(n-2)+...+1+0 = \frac{n(n+1)}{2}$  is

$$0.5(n^2 + n) \rightarrow O(n^2)$$

#### Example 5

```
double product = 1.0;
for (int i = 1; i <= n; i *= 2) {
    product *= i;
}</pre>
```

# Example 5: Solution

• This has a logarithmic cost:

 $O(\log_2 n)$ 

or  $O(\log n)$  as the change of base is merely a matter of a constant factor.

# Search in Ordered vs. Unordered Array

- What's the big O function for linear search?
- Binary search?

# Search in Ordered vs. Unordered Array

- What's the big O function for linear search?
   O(N)
- Binary search? O(IgN)
- Binary search has much better running time, particularly for large-scale problems

#### **Review Question**

• What is the Order of growth (big-oh) of the following code?

```
for (int i=1; i<=N; ++i){
    for (int j=1; j<N; j*=2){
        count++;
    }
}</pre>
```

#### Summary

- Today
  - -Algorithm Analysis
  - -Loop Analysis

- Next lecture
  - -Simple sorting algorithms and analysis