CS24 Week 6 Lecture 2

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Overview

- More complexity analysis
- Recursion

More Complexity Analysis

Measuring Efficiency

- How might we determine the efficiency of a program?
 - Benchmarks tend to be too specific (new hardware? How big of inputs do we test?)
 - Better approach: define a formula in terms of the input size

```
int sum2(int* arr, int length) {
  int s = 0, x, y;
  for (x = 0; x < length; x++) {
    for (y = 0; y < length; y++) {
      s += arr[x] + arr[y];
    }
  return s;
}
```

```
int sum2(int* arr, int length) {
  int s = 0, x, y;
  for (x = 0; x < length; x++)  {
    for (y = 0; y < length; y++) {
      s += arr[x] + arr[y];
    }
  return s;
}
```

Constant time, done once. Call this c_1 .

int sum2(int* arr, int length) { int s = 0, x, y;for (x = 0; x < length; x++) { for (y = 0; y < length; y++){ s += arr[x] + arr[y];} return s; }

Constant time, done once. Call this c_2 .

```
int sum2(int* arr, int length) {
  int s = 0, x, y;
  for (x = 0; x < length; x++) {
    for (y = 0; y < length; y++) {
      s += arr[x] + arr[y];
    }
  return s;
}
```

Constant time, done length times. Call this c3.

```
int sum2(int* arr, int length) {
  int s = 0, x, y;
  for (x = 0; x < length; x++) {
    for (y = 0; y < length; y++) {
      s += arr[x] + arr[y];
    }
  return s;
}
```

Constant time, done length **times. Call this** c₄.

```
int sum2(int* arr, int length) {
  int s = 0, x, y;
  for (x = 0; x < length; x++)  {
    for (y = 0; y < length; y++) {
      s += arr[x] + arr[y];
    }
  return s;
}
```

Constant time, done length times. Call this c5.

int sum2(int* arr, int length) { int s = 0, x, y;for (x = 0; x < length; x++) { for (y = 0; y < length; y++) { s += arr[x] + arr[y]; } return s; }

Constant time, done length * length times. Call this c₆.

```
int sum2(int* arr, int length) {
  int s = 0, x, y;
  for (x = 0; x < length; x++)  {
    for (y = 0; y < length; y++)  {
      s += arr[x] + arr[y];
    }
  return s;
}
```

Constant time, done length * length times. Call this c7.

```
int sum2(int* arr, int length) {
  int s = 0, x, y;
  for (x = 0; x < length; x++)  {
    for (y = 0; y < length; y++) {
      s += arr[x] + arr[y];
    }
  return s;
}
```

Constant time, done length * length times. Call this c₈.

```
int sum2(int* arr, int length) {
  int s = 0, x, y;
  for (x = 0; x < length; x++)  {
    for (y = 0; y < length; y++) {
      s += arr[x] + arr[y];
    }
  return s;
}
```

Constant time, done once. Call this C9.

• We are left with the following formula:

 $c_1 + c_2 + (length * c_3) + (length * c_4) + (length * c_5) + (length * length * c_6) + (length * length * c_7) + (length * length * c_8) + c_9$

- The specific values of constants are unimportant as long as they are positive
- We can replace all these with the value 1 as far as Big O notation is concerned

 $c_1 + c_2 + (length * c_3) + (length * c_4) + (length * c_5) + (length * length * c_6) + (length * length * c_7) + (length * length * c_8) + c_9$

- The specific values of constants are unimportant as long as they are positive
- We can replace all these with the value 1 as far as Big O notation is concerned

1 + 1 + (length * 1) + (length * 1) + (length * 1) + (length * length * 1) + (length * length * 1) + (length * length * 1) + 1

- The specific values of constants are unimportant as long as they are positive
- We can replace all these with the value 1 as far as Big O notation is concerned

3 + length + length + length + (length * length) + (length * length) + (length * length)

- The specific values of constants are unimportant as long as they are positive
- We can replace all these with the value 1 as far as Big O notation is concerned

3 + 3(length) + 3(length * length)

- The specific values of constants are unimportant as long as they are positive
- We can replace all these with the value 1 as far as Big O notation is concerned

1+ length + (length * length)

- The specific values of constants are unimportant as long as they are positive
- We can replace all these with the value 1 as far as Big O notation is concerned

$1 + length + length^2$

- With sums, we always choose the larger sum
- A variable is always larger than a constant

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$$length^2$$

- Observe that length is really N, the input size
- For this example, we are done

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- For this example, we are done

$$O(N^2)$$

Big O Heuristics

- A non-loop is often O(1)
- A single loop is often O(N)
- A singly nested loop is often $O(N^2)$
- Not always true though we will see exceptions later in this class
 - Determining time complexity can be quite difficult in general

Recursion

Motivation

- A lot of problems are defined in terms of themselves (recursive)
 - You're already familiar with a lot!
- These demand solutions which are themselves recursive

Recursion

- Defining a problem in terms of:
 - Some simple trivial case
 - A more complex case which ultimately leads to the trivial case
- A way to define a problem in terms of itself

Trivial Case

- Often called the "base" case
- It represents a simple form of the problem

Recursive Case

- Defines problem in terms of itself
- Recursive cases should ultimately lead to base cases

My Two Cents on Recursion

- Phrased as a problem strictly with numbers, this seems magical and unintuitive
- Phrased as a problem over data structures, this makes more sense
 - Data structures themselves can have recursive structure
 - You're been familiar with recursive data structures, for **many, many years**



Example: Arithmetic Expressions

n is an Integer e is an Expression op is an Operator

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Example: Arithmetic Expression Evaluation

- A number evaluates to itself
- To evaluate an operation ($e_1 \text{ op } e_2$):
 - Evaluate e_1 to a number n_1
 - Evaluate e_2 to a number n_2
 - Evaluate n_1 op n_2

Example: Natural Language

- It is possible to take the majority of most natural languages and express them in a way that is similar to our arithmetic expression representation
- A clause containing another clause...

Example: Programming Languages

- Most programming languages work this way, too
- ifs can be nested in ifs...
 - At some point, we have to stop nesting the ifs, or else we won't have a program

Example: Linked Lists

- A linked list is either:
 - An empty list
 - A node holding an item (int below) and a pointer to another list

List = Empty | int List

Relationship to Operations

- The recursive structure of applicable data structures often mirrors the recursive structure of operations on those data structures
- Which cases might be interesting for a linked list?

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- The recursive structure of applicable data structures often mirrors the recursive structure of operations on those data structures
- Which cases might be interesting for a linked list?
 - Empty list (e.g., NULL)
 - Non-empty list (a node)

Example Problem

- Say we want to calculate the length of a linked list recursively
- A list is represented as a Node*
 - Base case?
 - Length of list besides first element?
 - Recursive case?

int length(Node* list);

Example Problem

Revised Problem

- Say we want to determine the length of a list, but with a tweak: we also take the length of the list so far
 - Base case?
 - Length of list besides first element?
 - Recursive case?
- What does the initial call look like?

int firstCall(Node* list); int length2(Node* list, int soFar);

int length2(Node* list, int soFar) {
 if (list == NULL) {

return soFar; // base case

} else {

// get the length of the rest of
// the list, and say that the
// length so far is + 1

int firstCall(Node* list) { return length2(list, 0);

}

Relationship to Loops

- length2 is more similar to an iterative implementation than it may seem at first
 - while dynamically inserts ifs as many times as needed
 - Recursion dynamically inserts the body of a function as many times as needed
- After doing these expansions, they basically look the same!

Recursion With Arrays

Recursion With Arrays

- If we look at arrays in a similar way as linked lists, operations become more clear
- The index acts like a pointer to a particular node
 - What is the base case?
 - Recursive case?

Recursion With Arrays

- If we look at arrays in a similar way as linked lists, operations become more clear
- The index acts like a pointer to a particular node
 - What is the base case?
 - Index out of array
 - Recursive case?
 - Index in array

Example

- Determine the sum of an array of integers, starting from a particular index. An array containing no elements has a sum of 0.
 - Base case?
 - Recursive case?

Example

- Determine the sum of an array of integers, starting from a particular index. An array containing no elements has a sum of 0.
 - Base case? index out of bounds (0)
 - Recursive case? index in bounds (current element + sum of rest)

int sumFromIndex(int* array, int length, int index) { if (index >= length) return 0; else { int restSum = sumFromIndex(array, length, index + 1);return restSum + array[index];

Recursion Pros

- If your recursive case is always guaranteed to reach a base case, infinite recursion is impossible (appeals to induction)
 - No more infinite loops!
- Vital for more complex recursive data structures (e.g., trees)
- Easier to understand :)

Recursion Cons

- If you're not careful, you can run out of stack space (a stack overflow)
 - Not written in a tail-recursive way
 - Compiler is too stupid to notice it's tailrecursive
 - Very large input