### Algorithms: A Brief Introduction

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### Algorithms Formal Definition

#### Definition

An algorithm is a sequences of unambiguous instructions for solving a problem. Algorithms must be

- Finite must eventually terminate.
- ► Complete *always* gives a solution when there is one.
- ► Correct (sound) *always* gives a "correct" solution.

For an algorithm to be an acceptable solution to a problem, it must also be effective. That is, it must give a solution in a "reasonable" amount of time.

There can be many algorithms for the same problem.

### Pseudo-code

Algorithms are usually presented using some form of *pseudo-code*. Good pseudo-code is a balance between clarity and detail.

Bad pseudo-code gives too many details or is too implementation specific (i.e. actual C++ or Java code or giving every step of a sub-process).

Good pseudo-code abstracts the algorithm, makes good use of mathematical notation and is easy to read.

### Algorithms

**Brief Introduction** 

Real	World
Objec	cts
Relat	ions
Actio	ns

Computing World Data Structures, ADTs, Classes Relations and functions Operations

Problems are specified by (1) a formulation and (2) a query.

Formulation is a set of objects and a set of relations between them

Query is the information one is trying to extract from the formulation, the question to answer.

 $\ensuremath{\textbf{Algorithms}}^1$  are methods or procedures that solve instances of a problem

<sup>1</sup>"Algorithm" is a distortion of *al-Khwarizmi*, a Persian mathematician

# Algorithms

General Techniques

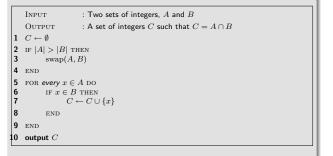
There are many broad categories of Algorithms: Randomized algorithms, Monte-Carlo algorithms, Approximation algorithms, Parallel algorithms, et al.

Usually, algorithms are studied corresponding to relevant data structures. Some general styles of algorithms include

- 1. Brute Force (enumerative techniques, exhaustive search)
- 2. Divide & Conquer
- 3. Transform & Conquer (reformulation)
- 4. Greedy Techniques

### Good Pseudo-code Example

INTERSECTION



Latex notation: \leftarrow.

### Designing An Algorithm

A general approach to designing algorithms is as follows.

- 1. Understand the problem, assess its difficulty
- 2. Choose an approach (e.g., exact/approximate, deterministic/probabilistic)
- 3. (Choose appropriate data structures)
- 4. Choose a strategy
- 5. Prove termination
- 6. Prove correctness
- 7. Prove completeness
- 8. Evaluate complexity
- 9. Implement and test it.
- 10. Compare to other known approaches and algorithms.

### MAX

Pseudo-code

MAX

### IN Οt

	INPUT	: A set $A = \{a_1, a_2, \dots, a_n\}$ of integers.
	Output	: An index $i$ such that $a_i = \max\{a_1, a_2, \dots, a_n\}$
1	$\mathrm{index} \gets 1$	
2	For $i = 2$ ,	$\dots, n$ do
3	IF $a_i$	$> a_{index}$ THEN
4		$index \leftarrow i$
5	END	
6	END	
7	$\mathbf{output}\ i$	

### Other examples

Check Bubble Sort and Insertion Sort in your textbooks, which you have seen ad nauseum, in CSE155, CSE156, and will see again in CSE310.

I will be glad to discuss them with any of you if you have not seen them yet.

### MAX

When designing an algorithm, we usually give a formal statement about the problem we wish to solve.

Problem

**Given** a set  $A = \{a_1, a_2, \ldots, a_n\}$  integers. **Output** the index i of the maximum integer  $a_i$ .

A straightforward idea is to simply store an initial maximum, say  $a_1$  then compare it to every other integer, and update the stored maximum if a new maximum is ever found.

### MAX

Analysis

This is a simple enough algorithm that you should be able to:

- Prove it correct
- Verify that it has the properties of an algorithm.
- Have some intuition as to its cost.

That is, how many "steps" would it take for this algorithm to complete its run? What constitutes a step? How do we measure the complexity of the step?

These questions will be answered in the next few lectures, for now let us just take a look at a couple more examples.

### Greedy algorithm Optimization

In many problems, we wish to not only find a solution, but to find the best or optimal solution.

A simple technique that works for some optimization problems is called the greedy technique.

As the name suggests, we solve a problem by being greedy-that is, choosing the best, most immediate solution (i.e. a local solution).

However, for some problems, this technique is not guaranteed to produce the best *globally optimal* solution.

## Example

Change-Making Problem

For anyone who's had to work a service job, this is a familiar problem: we want to give change to a customer, but we want to minimize the number of total coins we give them.

Problem

Given An integer n and a set of coin denominations  $(c_1,c_2,\ldots,c_r)$  with  $c_1>c_2>\cdots>c_r$ 

**Output** A set of coins  $d_1, d_2, \cdots, d_k$  such that  $\sum_{i=1}^k d_i = n$  and k is minimized.

# Change-Making Algorithm

Analysis

Will this algorithm always produce an optimal answer?

Consider a coinage system:

- where  $c_1 = 20, c_2 = 15, c_3 = 7, c_4 = 1$
- $\blacktriangleright$  and we want to give 22 "cents" in change.

What will this algorithm produce?

Is it optimal?

It is *not* optimal since it would give us one  $c_4$  and two  $c_1$ , for three coins, while the optimal is one  $c_2$  and one  $c_3$  for two coins.

### Change-Making Algorithm

Proof.

Provin

- ▶ Let  $C = \{d_1, d_2, ..., d_k\}$  be the solution given by the greedy algorithm for some integer n. By way of contradiction, assume there is *another* solution  $C' = \{d'_1, d'_2, ..., d'_l\}$  with l < k.
- ► Consider the case of quarters. Say in C there are q quarters and in C' there are q'. If q' > q, the greedy algorith would have used q'.
- Since the greedy algorithm uses as many quarters as possible, n = q(25) + r. where r < 25, thus if q' < q, then in n = q'(25) + r',  $r' \ge 25$  and so C' does not provide an optimal solution.
- ▶ Finally, if *q* = *q*′, then we continue this argument on dimes and nickels. Eventually we reach a contradiction.
- Thus, C = C' is our optimal solution.

### Example

### Change-Making Algorithm

### CHANGE

6 END	$+ \sum_{k=1}^{k} d =$	with c1 2
2 FOR $i = 1, \dots, r$ DO 3 WHILE $n \ge c_i$ DO 4 $C \leftarrow C \cup \{c_i\}$ 5 $n \leftarrow n - c_i$ 6 END	$\sum_{i=1}^{n} a_i = n$ and $k$ is	
$ \begin{array}{llllllllllllllllllllllllllllllllllll$		$C \gets \emptyset$
$ \begin{array}{ll} 4 & C \leftarrow C \cup \{c_i\} \\ 5 & n \leftarrow n - c_i \\ 6 & \text{END} \end{array} $		For $i = 1, \ldots, r$ do
6 END		
•		$n \leftarrow n - c_i$
7 END		END
		END
8 output C		output C

# Change-Making Algorithm Optimal?

What about the US currency system—is the algorithm correct in this case?

Yes, in fact, we can prove it by contradiction.

For simplicity, let  $c_1 = 25, c_2 = 10, c_3 = 5, c_4 = 1$ .

### Change-Making Algorithm Proving optimality

Why (and where) does this proof fail in our previous counter example to the general case?

We need the following lemma:

If n is a positive integer then n cents in change using quarters, dimes, nickels, and pennies using the fewet coins possible

- 1. has at most two dimes, at most one nickel, at most most four pennies, and
- 2. cannot have two dimes and a nickel.

The amount of change in dimes, nickels, and pennies cannot exceed 24 cents.