

## Lec 2: Underlying Concepts and Review

Some slides borrowed from CS 240.

Thanks to Anna Lubiw and other previous CS 341 instructors.

- Problem Description
- Models of Computation
- Asymptotic Notation

## Problems (terminology)

Recall from CS 240 the terminology used so we can precisely characterize what we mean by efficiency.

**Problem:** Given a problem instance, carry out a particular computational task. A specification of an infinite set of inputs and corresponding outputs.

**Problem Instance:** An *Input* for the specified problem.

**Problem Solution:** The *Output* (correct answer) for the specified problem instance.

**Size of a problem instance:**  $Size(I)$  is a positive integer which is a measure of the size of the instance  $I$ .

**Analysis:** Measure the *Time* and *Space* used by the algorithm as a function of the input size.

# Algorithms and Programs

**Algorithm:** An algorithm is a *step-by-step process* (e.g., described in pseudo-code) for carrying out a series of computations, given an arbitrary problem instance  $I$ .

**Solving a problem:** An Algorithm  $A$  *solves* a problem  $\Pi$  if, for every instance  $I$  of  $\Pi$ ,  $A$  finds (computes) a valid solution for the instance  $I$  in finite time.

**Program:** A program is an *implementation* of an algorithm using a specified computer language.

In this course, our emphasis is on algorithms (as opposed to programs or programming).

# Models of Computation

We often use **Pseudocode**: similar to a typical programming language but **intended for a human to read** - uses common programming structures and conventions but omits machine and language specific details.

In contrast, a program is a method of communicating an algorithm to a computer.

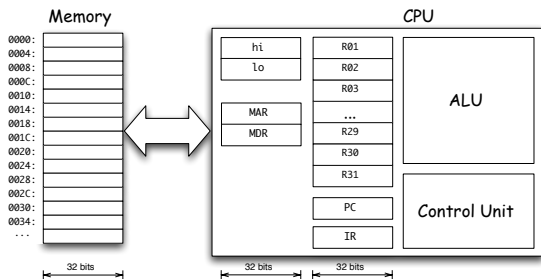
## Pseudocode

- omits obvious details, e.g. variable declarations,
- has limited if any error detection,
- sometimes uses English descriptions,
- sometimes uses mathematical notation.

Size of an integer? Cost of elementary operations?

# Random Access Machine

- Abstracts Assembly language
- “Random access” can access memory location  $i$  in one step



Size of a memory location? A good compromise:

- Word RAM - each memory location holds one word. Assume number of bits in word is  $\Theta(\log n)$  where  $n$  is the input size.
- E.g. Given array  $A[1..n]$ , an index  $i \in [1..n]$  fits in a word.

# Other Models of Computation

## Circuit Family

- Abstracts hardware circuitry

## Turing Machine

- Abstract human computer working with pencil and paper.
- Has a read/write head and infinite tape of cells (move left/right 1 cell at a time) - specific cell access may not be 1 step; time to access memory location  $i$  is proportional to  $i$ .

## Special purpose or Structured models of computing

- Comparison-based model for sorting:  $\Omega(n \log n)$  lower bound

# Runtime of an Algorithm

Runtime depends on input  $\Rightarrow$  express runtime as a function of input size.

- Model of Computation specifies how to count input size.
- We expect the runtime to increase as input size increases.

Let  $T_{\mathcal{A}}(I)$  denote the running time of an algorithm  $\mathcal{A}$  on instance  $I$ .

For a given size  $n$ , there are various inputs.

How do we combine runtimes to a single number?

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How do we combine runtimes to a single number?

**Worst-case complexity** of an algorithm: take the worst  $I$

- $T_{\mathcal{A}}(n) = \max\{T_{\mathcal{A}}(I) \text{ where } I \text{ is an input of size } n\}$
- Often simply say  $T(n)$  (or  $T(I)$ ) if  $\mathcal{A}$  is understood.

**Average-case complexity** of an algorithm: average over  $I$

- Often difficult to analyze, depends on input distribution, etc



# Asymptotic Analysis

- Want simple functions: e.g.  $n \log n$ ,  $n^2$ , etc
- Machine independent so ignore coefficients (multiplicative factors) and lower order terms.

Often use Big-Oh, an upper bound.

Want the tightest bound.

- $f(n) = 7n^2 + 13n + 27 \in O(?)$
- $10^{100}n \in O(?)$
- $2^{n+1} \in O(?)$
- $(n + 1)! \in O(?)$

# Order Notation Summary

**$O$ -notation:**  $f(n) \in O(g(n))$  if there exist constants  $c > 0$  and  $n_0 \geq 0$  such that  $|f(n)| \leq c |g(n)|$  for all  $n \geq n_0$ .

**$\Omega$ -notation:**  $f(n) \in \Omega(g(n))$  if there exist constants  $c > 0$  and  $n_0 \geq 0$  such that  $c |g(n)| \leq |f(n)|$  for all  $n \geq n_0$ .

**$\Theta$ -notation:**  $f(n) \in \Theta(g(n))$  if there exist constants  $c_1, c_2 > 0$  and  $n_0 \geq 0$  such that  $c_1 |g(n)| \leq |f(n)| \leq c_2 |g(n)|$  for all  $n \geq n_0$ .

**$o$ -notation:**  $f(n) \in o(g(n))$  if for all constants  $c > 0$ , there exists a constant  $n_0 \geq 0$  such that  $|f(n)| \leq c |g(n)|$  for all  $n \geq n_0$ .

**$\omega$ -notation:**  $f(n) \in \omega(g(n))$  if for all constants  $c > 0$ , there exists a constant  $n_0 \geq 0$  such that  $c |g(n)| \leq |f(n)|$  for all  $n \geq n_0$ .

# Algebra of Order Notations

**Identity rule:**  $f(n) \in \Theta(f(n))$

**Transitivity:**

- If  $f(n) \in O(g(n))$  and  $g(n) \in O(h(n))$  then  $f(n) \in O(h(n))$ .
- If  $f(n) \in \Omega(g(n))$  and  $g(n) \in \Omega(h(n))$  then  $f(n) \in \Omega(h(n))$ .

**Maximum rules:** Suppose that  $f(n) > 0$  and  $g(n) > 0$  for all  $n \geq n_0$ .  
Then:

- $f(n) + g(n) \in O(\max\{f(n), g(n)\})$
- $f(n) + g(n) \in \Omega(\max\{f(n), g(n)\})$

Proof:  $\max\{f(n), g(n)\} \leq f(n) + g(n) \leq 2 \max\{f(n), g(n)\}$

# Runtime Examples

- $O(1)$  - counting a finite number of things
- $O(\log n)$  - binary search
- $O(n)$  - find max
- $O(n \log n)$  - sorting
- $O(n^2)$  - insertion sort
- $O(n^3)$  - multiplying two  $n \times n$  matrices
- $O(2^n)$  - try all subsets
- $O(n!)$  - try all orderings of a set; e.g. Travelling Salesman
- Also,  $\sqrt{n}$ ,  $\log \log n$ ,  $(\log n)^2$ , etc

# Multiple Variables

**Output sensitive:** Jarvis March  $O(n * h)$  where  $h$  is the number of edges in convex hull; i.e.  $h$  is a measure of the output.

**Multiple variables:** Graph  $G = (V, E)$  where  $|V| = n$  and  $|E| = m$ .  $m \in O(n^2)$  but in some instances  $(n + m)$  may have a tighter  $\Theta$ -bound.

**$O$ -notation:**  $f(n, m) \in O(g(n, m))$  if there exist constants  $c > 0$  and  $n_0, m_0 \geq 0$  such that  $|f(n, m)| \leq c |g(n, m)|$  for all  $n \geq n_0, m \geq m_0$ .

# Reductions

- Simply put: Using a known algorithm to solve a new problem.

## Problem

### 2-SUM

**Instance:** Array  $A[1 \dots n]$  of numbers and target number  $m$

**Find:**  $i, j$  s.t.  $A[i] + A[j] = m$  (if they exist)

Note: Its sometimes simpler to index arrays from 1 to  $n$ .

*Algorithm1*( $A, n, m$ )

1. **for**  $i \leftarrow 1$  **to**  $n$  **do**
2.     **for**  $j \leftarrow 1$  **to**  $n$  **do**
3.         **if**  $A[i] + A[j] = m$  **then**
4.             **return** FOUND
5. **return** FAIL

## Reductions continued

Algorithm 2: Use algorithms Sort and BinarySearch.

```
Algorithm2( $A, n, m$ )  
1.   Sort  $A$   
2.   for  $i \leftarrow 1$  to  $n$  do  
3.        $j \leftarrow \text{BinarySearch}(A, m - A[i])$   
4.       if  $A[i] + A[j] = m$  then  
5.           return FOUND  
6.   return FAIL
```

Runtime:  $O(n \log n) + O(n \log n) \in O(n \log n)$

## Reductions continued

Algorithm 3: Improve the 2nd phase.

```
Algorithm3( $A, n, m$ )
1.   Sort  $A$ 
2.    $i, j \leftarrow 1, n$ 
3.   while  $i \leq j$  do
4.        $sum \leftarrow A[i] + A[j]$ 
5.       if  $sum > m$  then
6.            $j \leftarrow j - 1$ 
7.       elseif  $sum < m$  then
8.            $i \leftarrow i + 1$ 
9.       else
10.          return FOUND
11.  return FAIL
```

Runtime:  $O(n \log n) + O(n) \in O(n \log n)$  but  $O(n)$  after sorting.

Correctness Invariant: if a solution exists:  $i^* \leq j^*$  then  $i^* \geq i, j^* \leq j$



## Reductions continued

### Problem

#### **3-SUM**

**Instance:** Array  $A[1 \dots n]$  of numbers and target number  $m$

**Find:**  $i, j, k$  s.t.  $A[i] + A[j] + A[k] = m$  (if they exist)

**Reduce** 3-SUM to 2-SUM.

# Reductions continued

## Problem

### 3-SUM

**Instance:** Array  $A[1 \dots n]$  of numbers and target number  $m$

**Find:**  $i, j, k$  s.t.  $A[i] + A[j] + A[k] = m$  (if they exist)

**Reduce** 3-SUM to 2-SUM.

Note:  $A[i] + A[j] + A[k] = m$  so  $A[i] + A[j] = m - A[k]$

Algorithm

- Run 2-SUM with target  $m - A[k]$  for  $k = 1, \dots, n$ .

Runtime:  $O(n \cdot n \log n) \in O(n^2 \log n)$ ?

## Reductions continued

### Problem

#### 3-SUM

**Instance:** Array  $A[1 \dots n]$  of numbers and target number  $m$

**Find:**  $i, j, k$  s.t.  $A[i] + A[j] + A[k] = m$  (if they exist)

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Algorithm

- Run 2-SUM with target  $m - A[k]$  for  $k = 1, \dots, n$ .

Runtime:  $O(n \cdot n \log n) \in O(n^2 \log n)$ ?

Don't need to sort over and over - use Alg 3 but only sort once.

Runtime:  $O(n \log n) + O(n^2) \in O(n^2)$

Faster Algorithms?