

## Lecture 2: May 4th, 2017

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## 2.1 Algorithms and Programs

### 2.1.1 Terminology

**Definition 2.1** *Problem* : Given a problem instance, carry out a particular computational task

**Definition 2.2** *Problem Instance* : Input for the specified problem

**Definition 2.3** *Problem Solution* : Output for the specified problem

**Definition 2.4** *Size of a problem instance* :  $\text{Size}(I)$  is a positive integer which is measure of the size of the instance  $I$

**Definition 2.5** *Algorithm* : An algorithm is a step-by step process for carrying out a series of computations, given an arbitrary problem instance  $I$ .

**Definition 2.6** *Algorithm solving a problem* : An algorithm  $A$  solves a problem  $(x)$  if for every instance  $I$  of  $x$ ,  $A$  finds a valid solution for the instance  $I$  in finite time

**Definition 2.7** *Program* : A program is an implementation of an algorithm using a specified computer language

### 2.1.2 Order Notation

1. O-notation ( $\leq$ ) :  $f(n) \in O(g(n))$  if there exists constants  $c > 0$  and  $n_0 > 0$  such that  $0 \leq f(n) \leq cg(n)$  for all  $n \geq n_0$
2.  $\Omega$ -notation ( $\geq$ ) :  $f(n) \in \Omega(g(n))$  if there exists constants  $c > 0$  and  $n_0 > 0$  such that  $0 \leq cg(n) \leq f(n)$  for all  $n \geq n_0$
3.  $\theta$ -notation ( $=$ ) :  $f(n) \in \theta(g(n))$  if there exists constants  $c_1, c_2 > 0$  and  $n_0 > 0$  such that  $0 \leq c_1g(n) \leq f(n) \leq c_2g(n)$  for all  $n \geq n_0$
4. o-notation ( $<$ ) :  $f(n) \in o(g(n))$  if for all constants  $c > 0$  and  $n_0 > 0$  such that  $0 \leq f(n) \leq cg(n)$  for all  $n \geq n_0$
5.  $\omega$ -notation ( $>$ ) :  $f(n) \in \omega(g(n))$  if for all constants  $c > 0$  and  $n_0 > 0$  such that  $0 \leq cg(n) \leq f(n)$  for all  $n \geq n_0$

### 2.1.3 Complexity of algorithms

1. Average-case complexity of an algorithm :

$$T_A^{avg}(n) = \frac{1}{|\{I : Size(I) = n\}|} \sum_{\{I : Size(I) = n\}} T_A(I)$$

2. Worst-case complexity of an algorithm :

$$T_A(n) = \max\{T_A(I) : Size(I) = n\}$$

### 2.1.4 Growth Rates

- If  $f(n) \in \theta(g(n))$ , then growth rates are the same
- If  $f(n) \in o(g(n))$ , then growth rate of  $f(n)$  is less than  $g(n)$
- If  $f(n) \in \omega(g(n))$ , then the growth rate for  $f(n)$  is greater than the growth rate of  $g(n)$ .
- Typically  $f(n)$  may be "complicated" and  $g(n)$  is chosen to be a very simple function.

### 2.1.5 Common Growth Rates

- $\theta(1)$  (Constant Complexity)
- $\theta(\log n)$  (Logarithmic Complexity)
- $\theta(n)$  (Linear Complexity)
- $\theta(n \log n)$  (linearithmic)
- $\theta(n \log^k n)$ , for some constant  $k$  (quasi-linear)
- $\theta(n^2)$  (Quadratic Complexity)
- $\theta(n^3)$  (Cubic Complexity)
- $\theta(2^n)$  (Exponential Complexity)