

# CS 395 – Analysis of Algorithms

## Chapter 2 – Fundamentals of the Analysis of Algorithm Efficiency

### Read 2.1

- Algorithm Analysis Framework.
- Weekly assignment 3

When you have a good proof, it's because it's proven.

— Jean Chretien

# Analysis of algorithms

## ⌘ Issues:

- correctness
- time efficiency
- space efficiency
- optimality

## ⌘ Approaches:

- theoretical analysis
- empirical analysis

**Empirical:** based on, concerned with, or verifiable by observation or experience rather than theory

## Measuring an input's size

- **Input size is influenced by**
  - the **data representation**, e.g. matrix
  - the **operations of the algorithm**, e.g. spell-checker
  - the **properties of the objects** in the problem, e.g. checking if a given integer is a prime number

The choice of an appropriate size metric can be influenced by operations of the algorithm in question.

For example, how should we measure an input's size for a **spell-checking** algorithm? If the algorithm

We should make a special note about measuring input size for algorithms solving problems such as

## Units for measuring running time

- Using standard time units is not appropriate
- Counting all operations in an algorithm is excessively difficult and not needed
- **The Approach: Identify and count the basic operation(s) in the algorithm**
- **Basic Operations:**
  - Applied to all input items in order to carry out the algorithms
  - Contribute most towards the running time of the algorithm

**Time efficiency : the number of repetitions of the basic operation as a function of input size**

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Using standard time units is not appropriate

- If we measure in ms, what happens if we run it on a faster computer?
- What happens on a time sharing system when the program is waiting?
- What if the input size changes?
- What about start-up costs that only make a difference with small inputs?

## Input size and basic operation examples

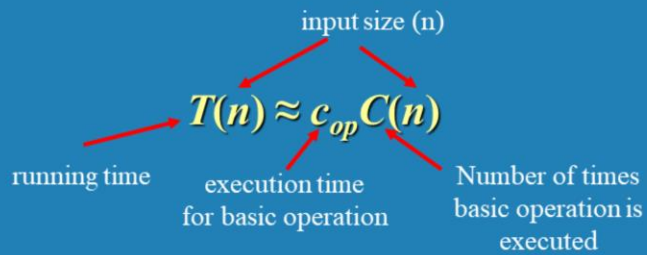
<i>Problem</i>	<i>Input size measure</i>	<i>Basic operation</i>
Searching for key in a list of $n$ items	Number of list's items, i.e. $n$	Key comparison
Multiplication of two matrices	Matrix dimensions or total number of elements	Multiplication of two numbers
Checking primality of a given integer $n$	$n$ 's size = number of digits (in binary representation)	Division
Typical graph problem	#vertices and/or edges	Visiting a vertex or traversing an edge

# Theoretical analysis of time efficiency



Time efficiency is analyzed by determining the number of repetitions of the basic operation as a function of input size

Q Basic operation: the operation that contributes most towards the running time of the algorithm



## Types of formulas for basic operation's count

⌚ **Exact formula**

e.g.,  $C(n) = n(n-1)/2$

⌚ **Formula indicating order of growth with specific multiplicative constant**

e.g.,  $C(n) \approx 0.5 n^2$

⌚ **Formula indicating order of growth with unknown multiplicative constant**

e.g.,  $C(n) \approx cn^2$

## Order of growth

⌘ Most important: Order of growth within a constant multiple as  $n \rightarrow \infty$

⌘ Example:

- How much faster will algorithm run on computer that is twice as fast?
- How much longer does it take to solve problem of double input size?

Example:  $cn^2$

$\Rightarrow$  how much faster on twice as fast computer? (2)

$\Rightarrow$  how much longer for  $2n$ ? (4)

## Asymptotic order of growth



A way of comparing functions that ignores constant factors and small input sizes

- ⌚  $O(g(n))$ : class of functions  $f(n)$  that grow no faster than  $g(n)$
- ⌚  $\Theta(g(n))$ : class of functions  $f(n)$  that grow at same rate as  $g(n)$
- ⌚  $\Omega(g(n))$ : class of functions  $f(n)$  that grow at least as fast as  $g(n)$



## Definition of Big-O

**Definition 1:** A function  $t(n)$  is said to be in  $O(g(n))$ , denoted  $t(n) \in O(g(n))$ , if  $t(n)$  is **bounded above** by some positive constant multiple of  $g(n)$  for all large  $n$ , i.e., if there exist some positive constant  $c$  and some nonnegative integer  $n_0$  such that

$$t(n) \leq c g(n) \text{ for all } n \geq n_0.$$

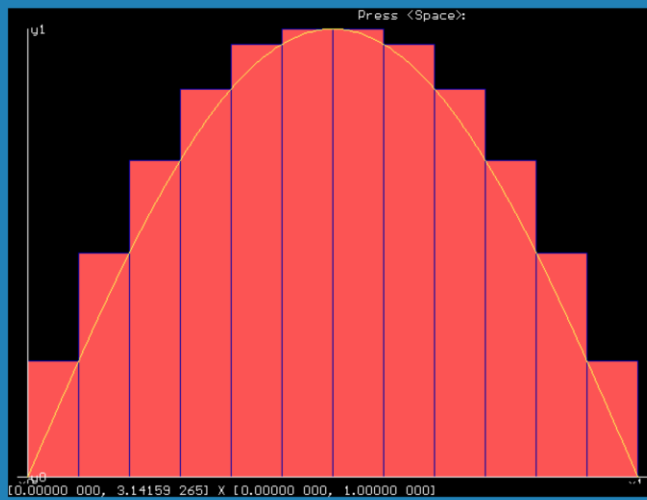
**Examples (find  $c$  and  $n_0$  so the above formula holds true):**

$$10n^2 + 4n \quad \text{is } O(n^2)$$

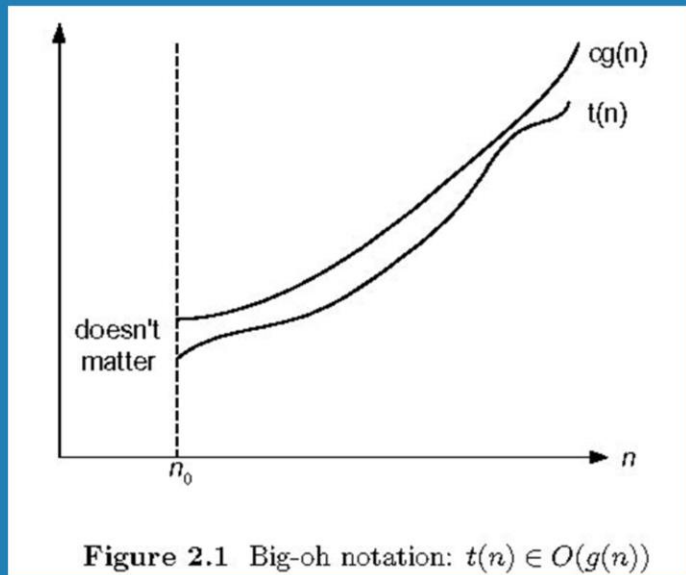
$$5n+20 \quad \text{is } O(n)$$

$c$  takes care of the faster computer  
 $n_0$  takes care of the startup costs.

The area in the red boxes is an **upper bound** area under the yellow curve.



# Big-oh



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It does not matter what happens before  $n_0$ , but after that  $t(n) < O(g(n))$

## Definition of Big-Ω

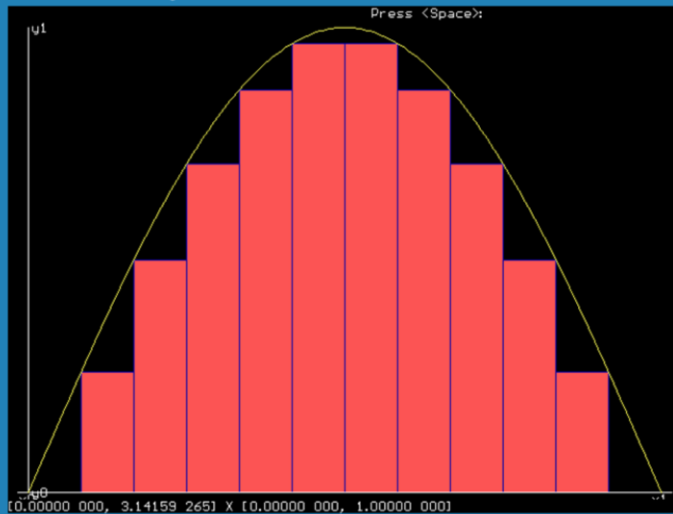
**Definition:** A function  $t(n)$  is said to be in  $\Omega(g(n))$ , denoted  $t(n) \in \Omega(g(n))$ , if  $t(n)$  is **bounded below** by some positive constant multiple of  $g(n)$  for all large  $n$ , i.e., if there exist some positive constant  $c$  and some nonnegative integer  $n_0$  such that

$$t(n) \geq c g(n) \text{ for all } n \geq n_0.$$

**Examples:**

- ❓ can we say  $n^3$  is  $\Omega(n^2)$ ?
- ❓ i.e.,  $n^3 \geq n^2$  with  $c=1, n_0=0$

The area in the red boxes is a **lower bound** area under the yellow curve.

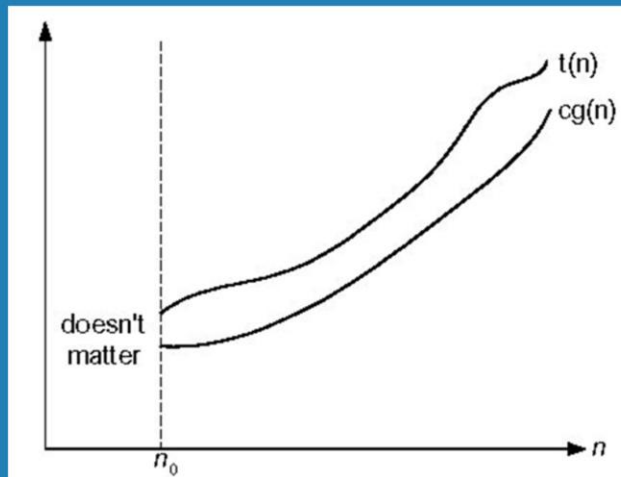


Picture from

[http://archives.math.utk.edu/visual.calculus/4/riemann\\_sums.3/microcalc.html](http://archives.math.utk.edu/visual.calculus/4/riemann_sums.3/microcalc.html)

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# Big-omega



**Fig. 2.2** Big-omega notation:  $t(n) \in \Omega(g(n))$

## Definition of Big- $\Theta$



**Definition:** A function  $t(n)$  is said to be in  $\Theta(g(n))$ , denoted  $t(n) \in \Theta(g(n))$ , if  $t(n)$  is **bounded both above and below** by some positive constant multiple of  $g(n)$  for all large  $n$ , i.e., if there exist some positive constant  $c_1$  and  $c_2$ , and some nonnegative integer  $n_0$  such that

$$c_2 g(n) \leq t(n) \leq c_1 g(n) \quad \text{for all } n \geq n_0.$$

**Example:**

$\frac{1}{2} n(n-1)$  is  $\Theta(n^2)$



# Big-theta

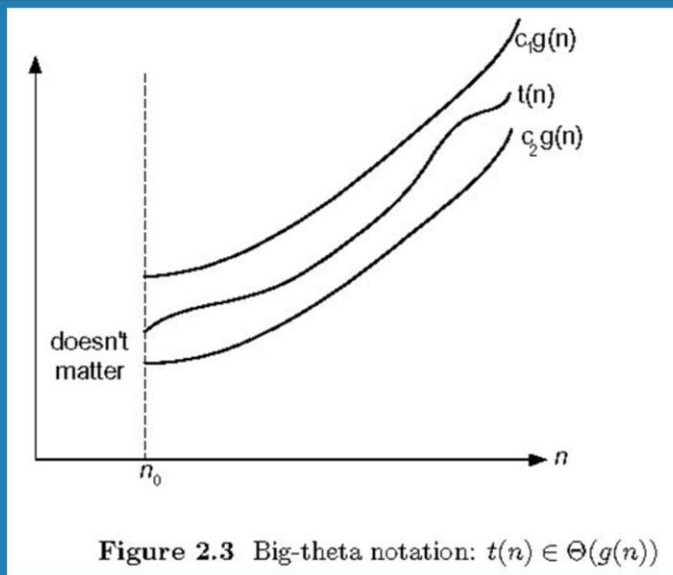


Figure 2.3 Big-theta notation:  $t(n) \in \Theta(g(n))$

Is  $3x^2 + 2x + 4 \in \theta(x^2)$ ?

Assume that  $T$ ,  $f$  and  $g$  are functions mapping the natural numbers  $\{0, 1, 2, 3, \dots\}$  into the positive reals.

**Definition: “Big Oh”** A function  $T(n)$  is in  $O(f(n))$  if there exist constants  $n_0 \geq 0$ , and  $c > 0$ , such that for all  $n \geq n_0$ ,  $T(n) \leq c * f(n)$ .

**Definition: “Omega”** A function  $T(n)$  is in  $\Omega(f(n))$  if there exist constants  $n_0 \geq 0$ , and  $c > 0$ , such that for all  $n \geq n_0$ ,  $T(n) \geq c * f(n)$ .

**Definition: “Theta”** The set  $\theta(g(n))$  of functions consists of  $\Omega(g(n)) \cap O(g(n))$ .

Do on the overhead.

