# Data Structures Chapter 1

- 1. Recursion
- 2. Performance Analysis
- 3. Asymptotic Analysis
  - Revisit Step Count
  - Asymptotic Analysis
  - Asymptotic Notations

- Why step count?
- It is to compare the time complexities of two programs that compute the same function and also to predict the growth rate in run time.
- Example: Let's compute the step count for three programs and compare their time complexities.
- 1. T<sub>add</sub>(n) adding two numbers
- 2. T<sub>sum</sub>(n) adding list of numbers
- 3.  $T_{mtx}(n)$  adding two matrix

Program add	step count
<pre>float add(int a, int b) {   return a + b; }</pre>	1

Program <b>sum of list</b>	step count	
<pre>float sum(float list[], int n) {   float total = 0;   int i;</pre>	1	
for (i=0; i <n; +="list[i];&lt;/td" i++)="" total=""><td>n + 1 n</td></n;>	n + 1 n	
return total;	1	
}		

Program <b>sum of matrix</b>	step count
<pre>void add(int a[][MAX_SIZE], int b[][MAX_SIZE],</pre>	rows + 1 rows * (cols+1) rows * cols

- $T_{add}(n) = 2$
- $T_{sum(n)} = 1 + 2(n+1) + 2n + 1 = 4n + 4$ = c \* n + c'
- $T_{mtx(n)} = 2 rows * cols + 2 rows + 1$ =  $a * n^2 + b * n + c$

■ 
$$T_{sum(n)} = 1 + 2(n+1) + 2n + 1 = 4n + 4$$
  $\rightarrow O(n)$   
=  $c * n + c'$ 

$$T_{mtx(n)} = 2 rows * cols + 2 rows + 1$$

$$= a * n^2 + b * n + c$$

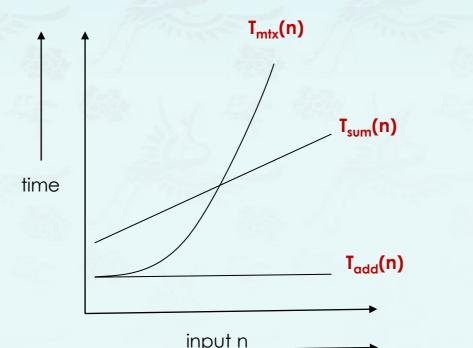
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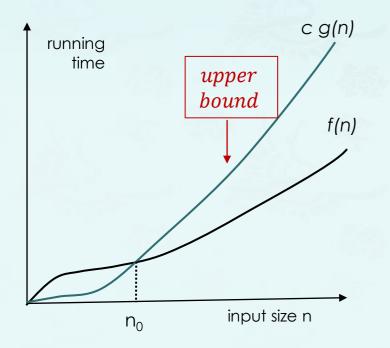
$$= a * n^2 + b * n + c$$

$$\rightarrow O(n^2)$$



- The "Big-Oh" Notation:
- Let f(n) and g(n) be functions mapping nonnegative integers to real numbers. We say that f(n) is O(g(n)) iff there are positive constants c and  $n_0$  such that
  - $f(n) \leq c g(n)$ , for  $n \geq n_0$ .
  - Then it is pronounced as "f(n) is big 0h of g(n) or f(n) = O(g(n))"

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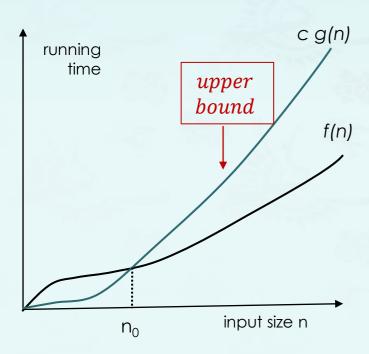


**Example**: Justify that the function 8n - 2 is O(n). Given f(n) = 8n - 2, g(n) = n, we need to find c and  $n_0$  such that  $8n - 2 \le c n$  for every integer  $n \ge n_0$ .

An easy choice among many is c = 8 and  $n_0 = 1$ . Therefore, f(n) is O(n).

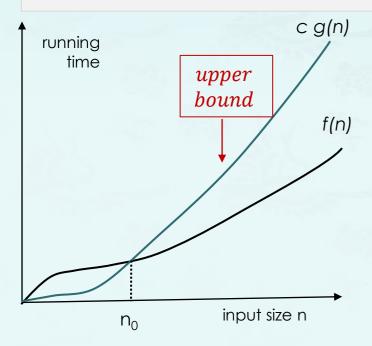
**Example:** Find c and  $n_0$  to justify that the function 7n + 5 is O(n).

7n + 5 is O(n), we have to find c and  $n_0$  such that  $7n + 5 \le c$  n  $for <math>n \ge n_0$ 



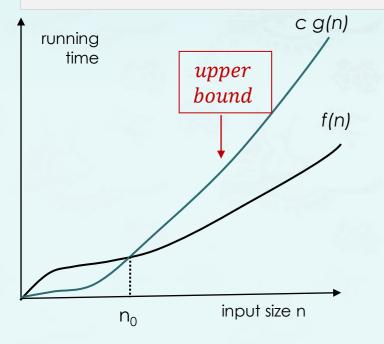
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```
7\mathbf{n}+5 is O(\mathbf{n}), we must find c and n_0 such that 7\mathbf{n}+5 \le c n for n \ge n_0 7\mathbf{n}+5 \le 7 n + n 7\mathbf{n}+5 \le 8 n, for n \ge n_0=5 Therefore, 7\mathbf{n}+5 \le c n for c=8 and n_0=5
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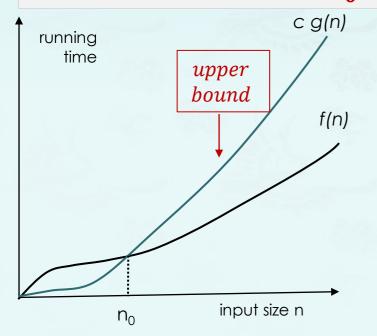
```
7n + 5 \le c n for n \ge n_0

7n + 5 \le 12 n for n \ge n_0 = 1

Therefore, 7n + 5 \le c n for c = 12 and n_0 = 1
```

• Example: Find c and  $n_0$  to justify that the function  $27n^2 + 16n$  is  $O(n^2)$ .

```
27n^2 + 16n is O(n^2), we must find c and n_0 such that For 16n \le n^2 27n^2 + 16n \le 27n^2 + n^2 27n^2 + 16n \le 28n^2 for n \ge n_0 = 16 Hence, c = 28 and n_0 = 16, Therefore, f(n) = O(n^2).
```



```
27n^2+16n is \boldsymbol{O}(n^2), we have to find \boldsymbol{c} and \boldsymbol{n_0} such that 27n^2+16n\leq 43n^2 27n^2+16n\leq 43n^2 for n\geq \boldsymbol{n_0}=1 Hence, c=43 and \boldsymbol{n_0}=1, Therefore, f(n)=\boldsymbol{O}(n^2).
```

#### More Examples:

1) 
$$3n + 2 =$$

2) 
$$3n + 3 =$$

3) 
$$100n + 6 =$$

4) 
$$10n^2 + 4n + 2 =$$

5) 
$$6 * 2^n + n^2 =$$

6) 
$$3n + 3 =$$

7) 
$$10n^2 + 4n + 2 =$$

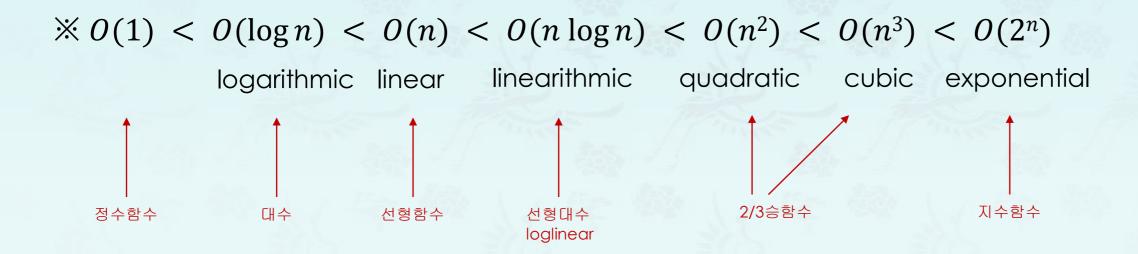
$$(3)$$
  $3n + 2 \neq 0$   $(1)$  as  $3n + 2$  is **not**  $\leq c$  for any  $c$  and all  $n, n \geq n_0$ .

$$(3)$$
 9)  $10n^2 + 4n + 2 \neq O(n)$ 

- Preferred Big-Oh usage:
- Pick the tightest bound. If f(N) = 5N, then:
  - $f(N) = O(N^5)$
  - $f(N) = O(N^3)$
  - $f(N) = O(N \log N)$
  - $f(N) = O(N) \leftarrow \text{preferred or right!}$

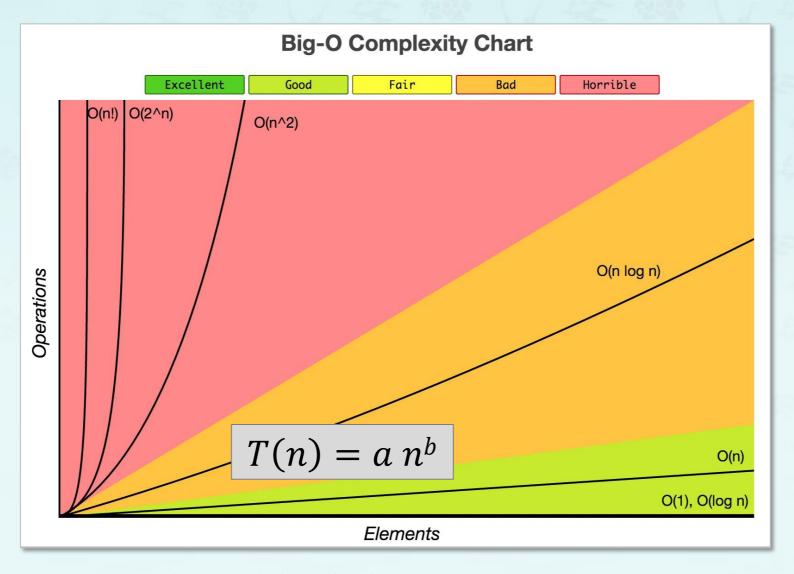
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  - $f(N) = O(N \log N)$
  - f(N) = O(N) ← preferred or right!
- Ignore constant factors and low order terms:
  - f(N) = O(N), not f(N) = O(5N)
  - $f(N) = O(N^3)$ , not  $f(N) = O(N^3 + N^2 + 15)$
  - Wrong:  $f(N) \leq O(g(N))$
  - Wrong:  $f(N) \ge O(g(N))$
  - Right: f(N) = O(g(N))

- Suppose two algorithms, A and B, solving the same problem have the running time of O(n) and  $O(n^2)$ , respectively.
- Then algorithm A is asymptotically better than algorithm B.



$$T(n) = a n^b$$

$$O(1) < O(\log n) < O(n) < O(n \log n) < O(n^2) < O(n^3) < O(2^n)$$



[Omega]  $f(n) = \Omega$  (g(n)) iff there exist positive constants c and  $n_0$  such that  $f(n) \ge c g(n)$ , for  $n \ge n_0$ .

[Omega]  $f(n) = \Omega$  (g(n)) iff there exist positive constants c and  $n_0$  such that  $f(n) \ge c g(n)$ , for  $n \ge n_0$ .

Example: Let's suppose we have

$$f(n) = 5n^2 + 2n + 1$$
$$g(n) = n^2$$

For all  $n \ge 0$ , this (2n + 1) will be  $\ge$  to 1, **if** we have c = 5 and  $n_0 = 0$ .

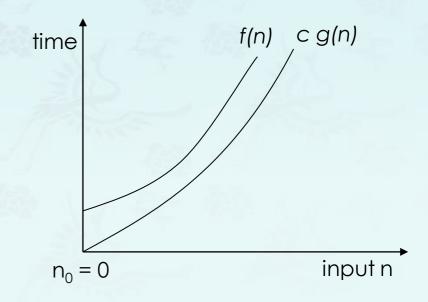
Then,  $5 n^2 \le f(n)$ , for all  $n \ge 0$ 

**Therefore**, we can say that the time complexity of f(n) is  $\Omega(n^2)$ ;

[Omega]  $f(n) = \Omega$  (g(n)) iff there exist positive constants c and  $n_0$  such that  $f(n) \ge c g(n)$ , for  $n \ge n_0$ .

Example: Let's suppose we have

$$f(n) = 5n^2 + 2n + 1$$
$$g(n) = n^2$$



Omega notation gives us the lower bound of the growth rate of a function.

[Omega]  $f(n) = \Omega$  (g(n)) iff there exist positive constants c and  $n_0$  such that  $f(n) \ge c g(n)$ , for  $n \ge n_0$ 

#### More Example:

1)
$$3n + 2 = \Omega(n)$$
 since  $3n + 2 \ge 3n$  for  $n \ge 1$ 

2)
$$3n + 3 = \Omega(n)$$
 since  $3n + 3 \ge 3n$  for  $n \ge 1$ 

3)
$$100n + 6 = \Omega(n)$$
 since  $100n + 6 \ge 100n$  for  $n \ge 1$ 

4)
$$100n^2 + 4n + 2 = \Omega(n^2)$$
 since  $100n^2 + 4n + 2 \ge n^2$  for  $n \ge 1$ 

5)6 \* 
$$2^n + n^2 = \Omega(2^n)$$
 since  $6 * 2^n + n^2 \ge 2^n$  for  $n \ge 1$ 

Omega notation gives us the lower bound of the growth rate of a function.

[Theta]  $f(n) = \Theta(g(n))$  iff there exist positive constants c and  $n_0$  such that

• 
$$c_1g(n) \le f(n) \le c_2g(n)$$
, for  $n \ge n_0$ .

Example: Let's suppose we have

$$f(n) = 5n^2 + 2n + 1$$
$$g(n) = n^2$$

• Then, we can choose  $c_1 = 5$ ,  $c_2 = 8$ , and  $n_0 = 1$ ; and our inequality will hold. Therefore, we can say that the time complexity of

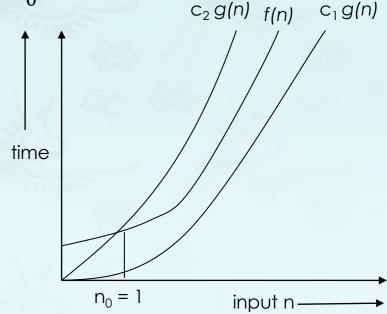
$$f(n) = 5n^2 + 2n + 1 = \Theta(n^2)$$

[Theta]  $f(n) = \Theta(g(n))$  iff there exist positive constants c and  $n_0$  such that

• 
$$c_1g(n) \le f(n) \le c_2g(n)$$
, for  $n \ge n_0$ .

Example: Let's suppose we have

$$f(n) = 5n^2 + 2n + 1$$
$$g(n) = n^2$$



•  $\Theta$  **notation** best describes or give the best idea about the growth rate of the function because it gives us a **tight bound** unlike O and  $\Omega$  which give us **upper bound** and **lower bound**, respectively.

[Theta]  $f(n) = \Theta(g(n))$  iff there exist positive constants c and  $n_0$  such that

• 
$$c_1g(n) \le f(n) \le c_2g(n)$$
, for  $n \ge n_0$ .

#### More Examples:

1) 
$$3n + 2 = \Theta(n)$$
  
since  $3n \le 3n + 2 \le 4n$  for all  $n \ge 2$ ,  $c_1 = 3$ ,  $c_2 = 4$ , and  $n_0 = 2$ 

$$2)3n + 3 = \Theta(n)$$

3) 
$$10n^2 + 4n + 2 = \Theta(n^2)$$

4) 
$$6 * 2^n + n^2 = \Theta(2^n)$$

$$5) 10 * \log n + 4 = \Theta(\log n)$$

#### Asymptotic Analysis - Quiz

- Example: Running time estimates empirical analysis
  - Personal computer executes 10° compares/second
  - Supercomputer executes 10<sup>13</sup> compares/second

	Selection sort ( N <sup>2</sup> )			Selection sort ( $N^2$ ) Merge sort ( $N \log_2 N$ )		
N	Million	10 million	Billion	Million	10 million	Billion
PC	16.7 min			instant	0.2 sec	
Super Com	0.1 sec	77		Instant	Instant	Instant

 $log_{10}2 \cong 0.3$ 86,400sec/day instant < 0.1 sec Use a reasonable or understandable time units. Do not say, for example, "3660 days" nor "1220 seconds", but 10.0 years or 20.3 min, respectively.

**X Bottom line**: Good algorithms are better than supercomputers.

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