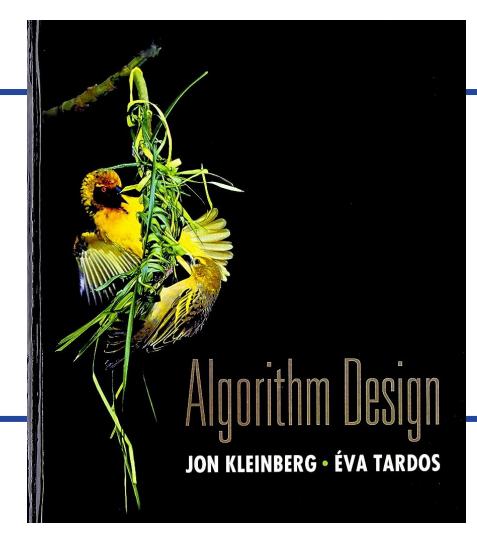


CS 310: Algorithms

# Lecture 4

**Instructor:** Naveed Anwar Bhatti





# Chapter 2: Basics of Algorithm Analysis



## Reasons to analyze algorithms

- Predict performance
- Compare algorithms
- Provide guarantees
- Improve performance
- Understand theoretical basis

Primary practical reason: avoid performance bugs.



client gets poor performance because programmer did not understand performance characteristics



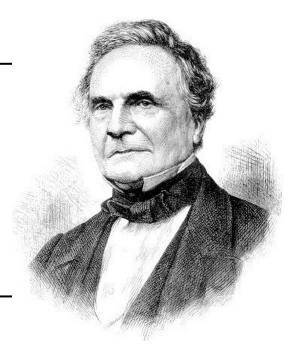


## A strikingly modern thought

"As soon as an <u>Analytic Engine</u> exists, it will necessarily guide the future course of the science. Whenever any result is sought by its aid, the question will arise—By what course of calculation can these results be arrived at by the machine in the shortest time?" — Charles Babbage (1864)



how many times do you have to turn the crank?



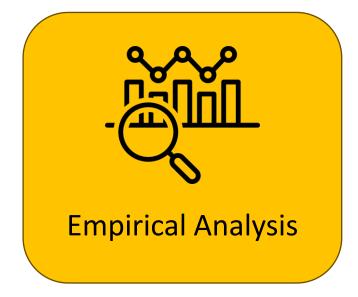


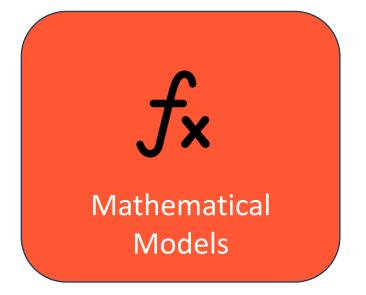
## How can we define efficiency?

"An algorithm is efficient if, when implemented, it runs quickly on real input instances."



## How can we measure efficiency? (Running Time)







Asymptotic Order of Growth



## **Empirical Analysis**

Problem: Find a triplet in an array whose sum is Zero

### **Experimental Setup:**





#### 1K array





#### 2K array

tick tick

#### 4K array



tick tick



## **Empirical Analysis**

### System independent effects.

- Algorithm.
- Input data.

We are interested in measuring this

### System dependent effects.

- Hardware: CPU, memory, cache, ...
- Software: compiler, operating system, garbage collector, ...

But we are also measuring this

**Bad News:** Sometimes difficult to get precise measurements.

Good News: We can generate accurate mathematical models



## **Mathematical Model: Examples**

#### **Problem**

Find a index in an array whose value is **Zero** 

Find a twins in an array whose value is **Zero** 

Find a triplet in an array whose sum is **Zero** 

#### **Algorithm**

```
int count = 0;
 for (int i = 0; i < N; i++)
     if (a[i] == 0)
        count++;
int count = 0;
for (int i = 0; i < N; i++)
    for (int j = i+1; j < N; j++)
        if (a[i] + a[j] == 0)
          count++;
for (int i = 0; i < N; i++)
 for (int j = i+1; j < N; j++)
   for (int k = j+1; k < N; k++)
     if (a[i] + a[i] + a[k] == 0)
         count++;
```

#### **Mathematical Model**

 $\boldsymbol{n}$ 

$$n(n-1) = n^2 - n$$

$$n(n-1)(n-2)$$
  
=  $n^3 - 3n^2 + 2n$ 

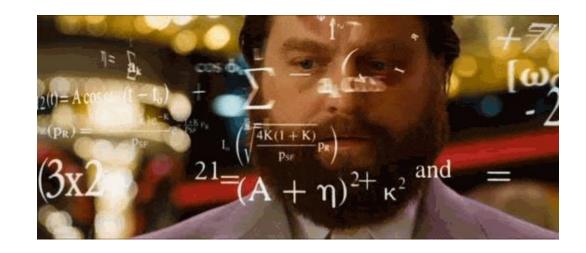


## Mathematical Model: We need Simplification

In principle, accurate mathematical models are available.

#### In practice,

- Formulas can be complicated.
- Advanced mathematics might be required.
- Exact models best left for experts.



Last option left: Use **Asymptotic** models



# Section 2.2: **Asymptotic Order of Growth**



## **Asymptotic Model:**

Estimate running time (or memory) as a function of input size N.

- Ignore lower order terms.
- Ignore constants
- when **N** is large, terms are negligible
- when **N** is small, we don't care



## **Asymptotic Model: Examples**

#### **Problem**

Find a index in an array whose value is **Zero** 

Find a twins in an array whose value is **Zero** 

Find a triplet in an array whose sum is **Zero** 

#### **Algorithm**

## **Mathematical Model**

 $\boldsymbol{n}$ 

$$n(n-1)$$

$$= n^2 - n$$

$$n(n-1)(n-2)$$
  
=  $n^3 - 3n^2 + 2n$ 

#### **Asymptotic Model**

 $\sim n$ 

 $\sim n^2$ 

 $\sim n^3$ 

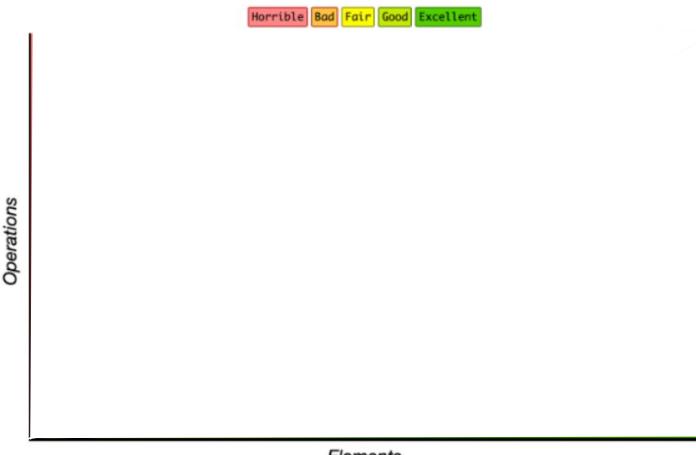


## Common Asymptotic order-of-growth classifications

**Good News:** The set of functions

1,  $\log N$ , N,  $N \log N$ ,  $N^2$ ,  $N^3$ ,  $2^N$  and N!

suffice to describe the order of growth of most common algorithms



Elements



## **Common Asymptotic order-of-growth classifications**

order of growth	name	typical code framework	description	example
1	constant	a = b + c;	statement	add two numbers
$\log N$	logarithmic	while (N > 1) { N = N/2; }	divide in half	binary search
N	linear	for (int i = 0; i < N; i++) { }	single loop	find the maximum
$N\log N$	linearithmic		divide and conquer	mergesort
N <sup>2</sup>	quadratic	<pre>for (int i = 0; i &lt; N; i++)   for (int j = 0; j &lt; N; j++)       { }</pre>	double loop	check all pairs
N 3	cubic	<pre>for (int i = 0; i &lt; N; i++)   for (int j = 0; j &lt; N; j++)     for (int k = 0; k &lt; N; k++)       { }</pre>	triple loop	check all triples
2 <sup>N</sup>	exponential		exhaustive search	fibonacci number



## **Algorithm Analysis: Live Poll 1**

Suppose you have algorithms with the running time of  $\mathbf{n^2}$  (Assume these are the exact number of operations performed as a function of the input size  $\mathbf{n}$ ) and you have a computer that can perform  $\mathbf{10^{10}}$  operations per second. You need to compute a result in at most an hour of computation.

What is the largest input size **n** for which you would be able to get the result within an hour?

- A.  $6 \times 10^6$
- B.  $36 \times 10^{12}$
- $C. 6 \times 10^{10}$
- $D. 100 \times 10^{10}$



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#### Algorithm Analysis: Poll 1

96 responses

Only people in my organization can respond, Record name

1. Suppose you have algorithms with the running time of n<sup>2</sup> and you have a computer that can...

6 × 10^6	51%
36 × 10^12	46%
6 x 10^10	2%
100 x 10^10	1%

< 1/1 >



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## **Asymptotic Order of Growth: Live Poll 1**

First, let's find out the total number of operations the computer can perform in an hour:

1 hour = 60 minutes = 3600 seconds

Operations per second=  $10^{10}$ 

Total operations in an hour =  $3600 \times 10^{10}$ 

We need  $n^2 \leq 3600 \times 10^{10}$ 

Taking the square root of both sides,

$$n \leq 60 \times 10^5$$



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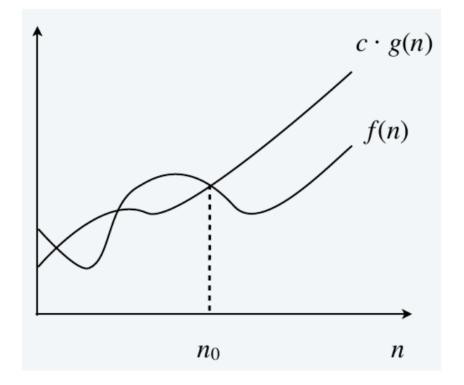


# O, $\Omega$ , and $\Theta$



## **Asymptotic Order of Growth – Big O notation**

• *Upper bounds:* f(n) is O(g(n)) if there exist constants c > 0 and  $n_0 \ge 0$  such that for all  $n \ge n_0$  we have  $f(n) \le c \cdot g(n)$ .



- Ex:  $f(n) = 32n^2 + 17n + 1$ 
  - f(n) is  $O(n^2)$
  - Can we say f(n) is  $O(n^3)$ ?



## **Asymptotic Order of Growth: Live Poll 1**

We know  $f(n) = 32n^2 + 17n + 1$  is  $O(n^2)$ . What is the value of C and n0?

A. 
$$C=1$$
 and  $n0=1$ 

- D. C=50 and n0=1
- E. None of above

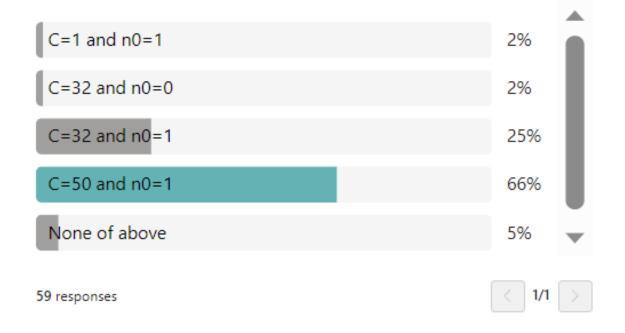


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#### Asymptotic Order of Growth: Live Poll 1

Only people in my organization can respond, Record name

1. We know  $f(n) = 32n^2 + 17n + 1$  is  $O(n^2)$ . What is the value of C and n0?



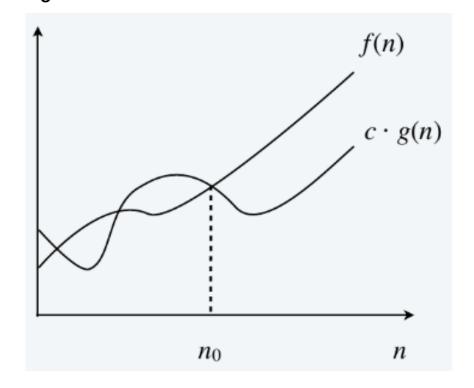


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## **Asymptotic Order of Growth – Big Omega notation**

• Lower bounds. f(n) is  $\Omega(g(n))$  if there exist constants c > 0 and  $n_0 \ge 0$  such that for all  $n \ge n_0$  we have  $f(n) \ge c \cdot g(n)$ .



- Ex:  $f(n) = 32n^2 + 17n + 1$ 
  - f(n) is  $\Omega(n^2)$ ,  $\Omega(n)$

#### Asymptotic Order of Growth: Live Poll 2

Only people in my organization can respond, Record name

#### 1. Which is an equivalent definition of big Omega notation?

f(n) is  $\Omega(g(n))$  if g(n) is O(f(n))14% f(n) is  $\Omega(g(n))$  if there exist constants c > 0 such that  $f(n) \ge$ 34%  $c \cdot g(n) \ge 0$  for infinitely many n

Neither A nor B. 2%

56 responses

Both A and B.





50%

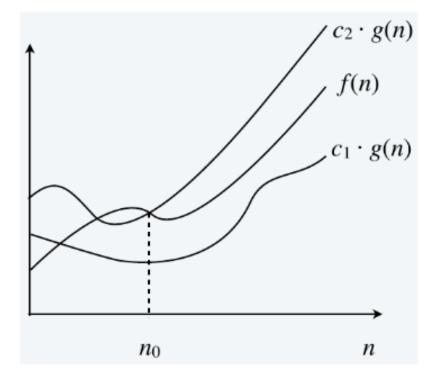


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## **Asymptotic Order of Growth – Big Theta notation**

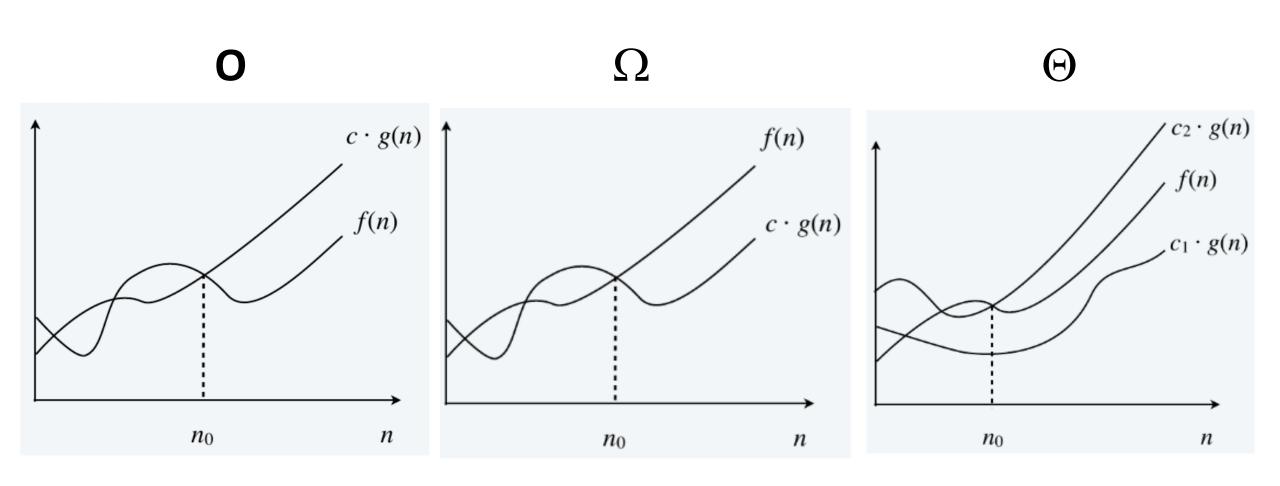
• Tight bound. f(n) is  $\Theta(g(n))$  if there exist constants  $c_1 > 0$ ,  $c_2 > 0$  and  $n_0 \ge 0$  such that for all  $n \ge n_0$  we have  $c_1 \cdot g(n) \le f(n) \le c_2 \cdot g(n)$ .



- Ex:  $f(n) = 32n^2 + 17n + 1$ 
  - f(n) is  $\Theta(n^2)$



# **Asymptotic Order of Growth**



# Notation Notation

- Slight abuse of notation. T(n) = O(f(n)).
  - Asymmetric:
    - $f(n) = 5n^3$ ;  $g(n) = 3n^2$
    - $f(n) = O(n^3) = g(n)$
    - but  $f(n) \neq g(n)$ .
  - Better notation:  $T(n) \in O(f(n))$ .

- Meaningless statement. Any comparison-based sorting algorithm requires at least O(n log n) comparisons.
  - Statement doesn't "type-check."
  - Use  $\Omega$  for lower bounds.

## **Properties**

## Transitivity.

- If f = O(g) and g = O(h) then f = O(h).
- If  $f = \Omega(g)$  and  $g = \Omega(h)$  then  $f = \Omega(h)$ .
- If  $f = \Theta(g)$  and  $g = \Theta(h)$  then  $f = \Theta(h)$ .

## Additivity.

- If f = O(h) and g = O(h) then f + g = O(h).
- If  $f = \Omega(h)$  and  $g = \Omega(h)$  then  $f + g = \Omega(h)$ .
- If  $f = \Theta(h)$  and  $g = \Theta(h)$  then  $f + g = \Theta(h)$ .



## **Asymptotic Bounds for Some Common Functions**

- Polynomials.  $a_0 + a_1 n + ... + a_d n^d$  is  $\Theta(n^d)$  if  $a_d > 0$ .
- Polynomial time. Running time is O(nd) for some constant d independent of the input size n.

Logarithms. O(log<sub>a</sub> n) = O(log<sub>b</sub> n) for any constants a, b > 0.

can avoid specifying the base

• Logarithms. For every x > 0,  $\log n = O(n^x)$ .

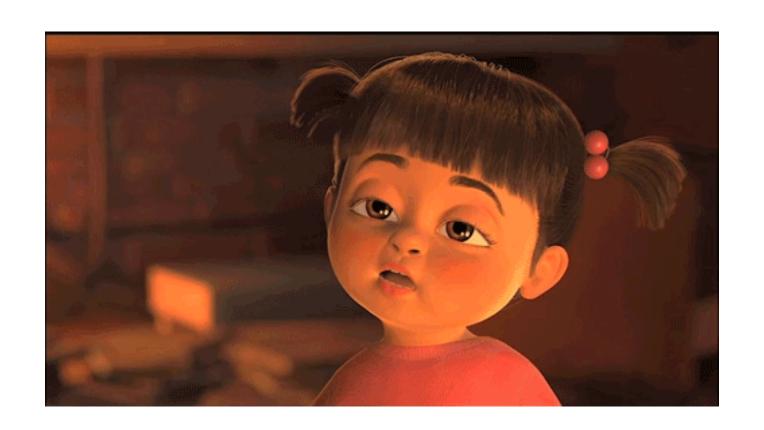
log grows slower than every polynomial

• Exponentials. For every r > 1 and every d > 0,  $n^d = O(r^n)$ .

every exponential grows faster than every polynomial



# Thanks a lot



If you are taking a Nap, wake up.....Lecture Over