

# CSE 250

## Data Structures

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212 Capen Hall

**Day 04**  
**Runtime Analysis**  
Textbook Ch. 7.3-7.4

# Announcements

- Dr. Kennedy will be giving lecture on Friday and Monday
- PA 0 is due Friday
- Start PA 1 early!

# From Lecture 01...

## Option 1

- Very fast Prepend, Get First
- Very slow Get Nth

## Option 2

- Very fast Get Nth, Get First
- Very slow Prepend

## Option 3

- Very fast Get Nth, Get First
- Occasionally slow Prepend

# From Lecture 01...

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**What is fast? slow?**

# Attempt #1: Wall-clock time?

- What is fast?
  - 10s? 100ms? 10ns?
  - ...it depends on the task
- Algorithm vs Implementation
  - Compare Grace Hopper's implementation to yours
- What machine are you running on?
  - Your old laptop? A lab machine? The newest, shiniest processor?
- What bottlenecks exist? CPU vs IO vs Memory vs Network...

# Attempt #1: Wall-clock time?

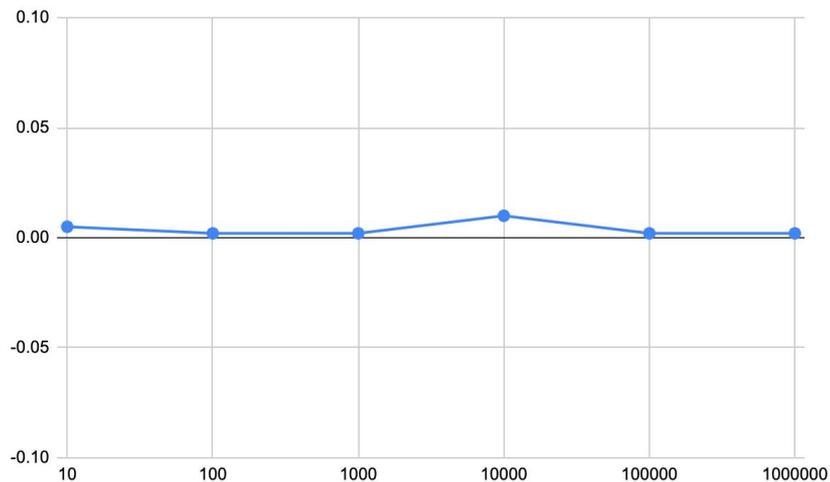
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**Wall-clock time is not terribly useful...**

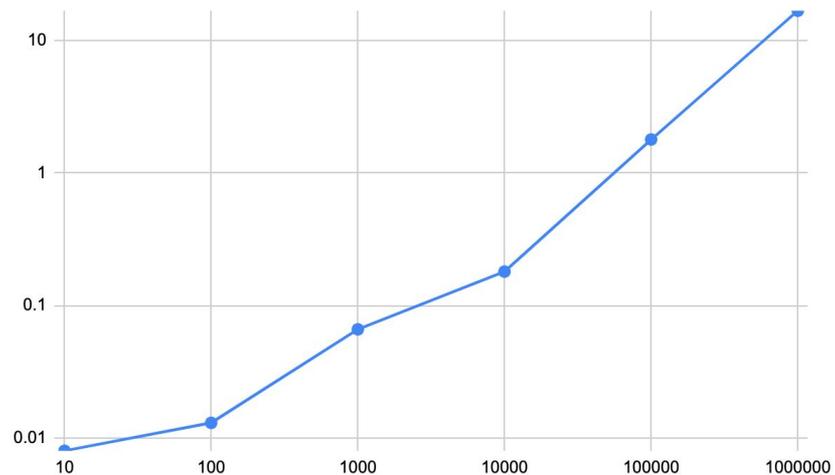
**Let's do a quick demo...**

# Comparing Random Access for Array vs List

## Array

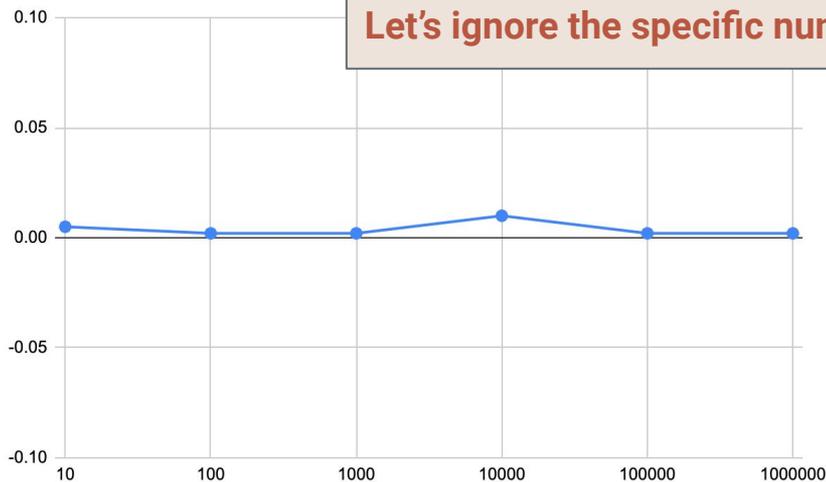


## List

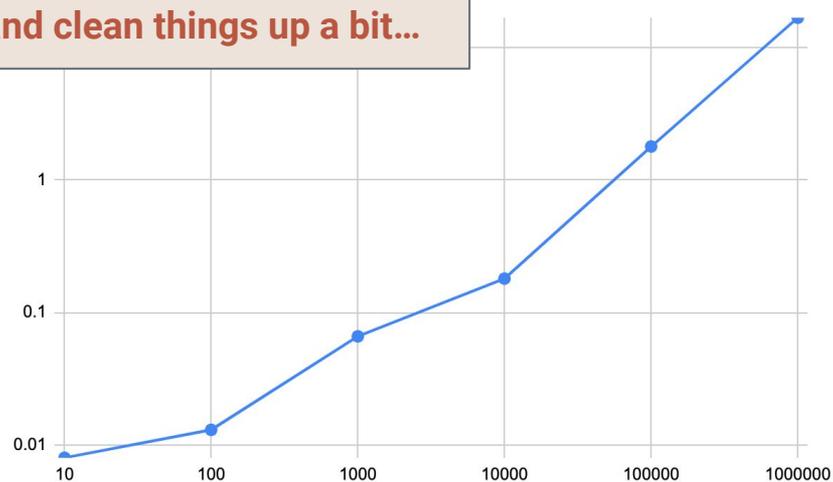


# Comparing Random Access for Array vs List

## Array



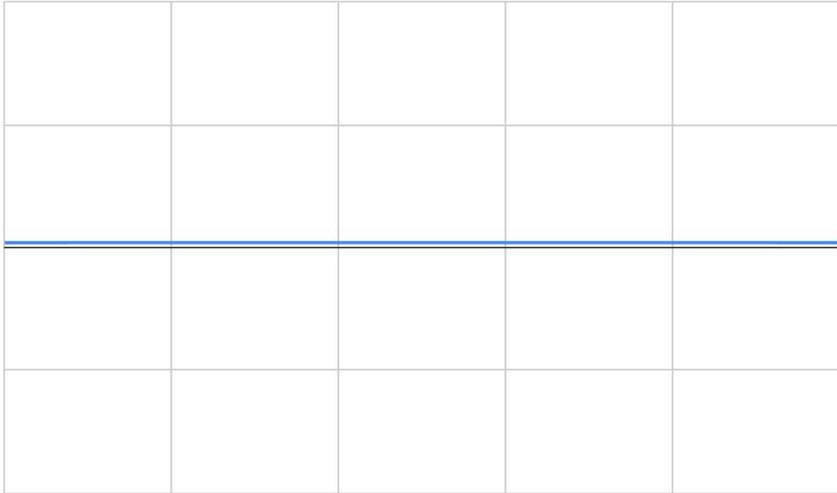
## List



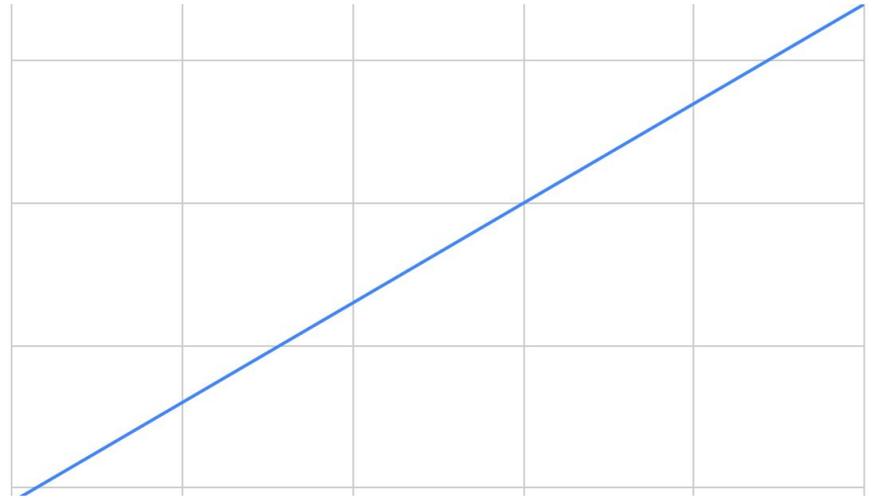
Let's ignore the specific numbers and clean things up a bit...

# Comparing Random Access for Array vs List

**Array**



**List**

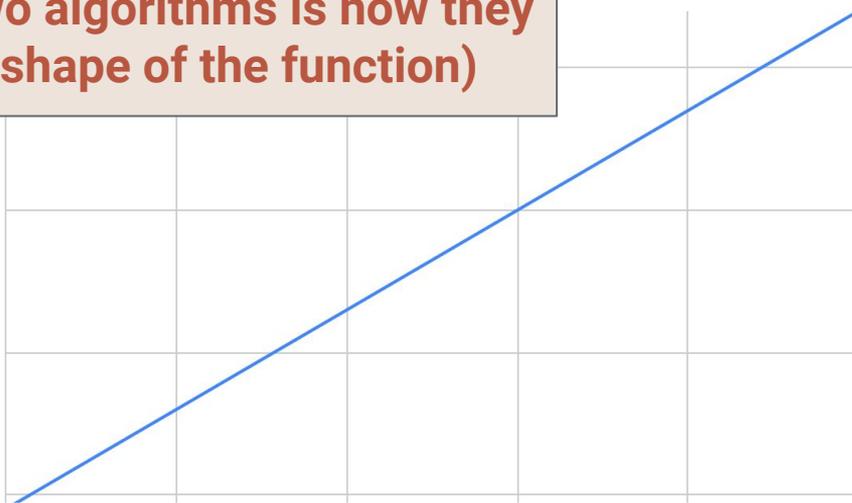
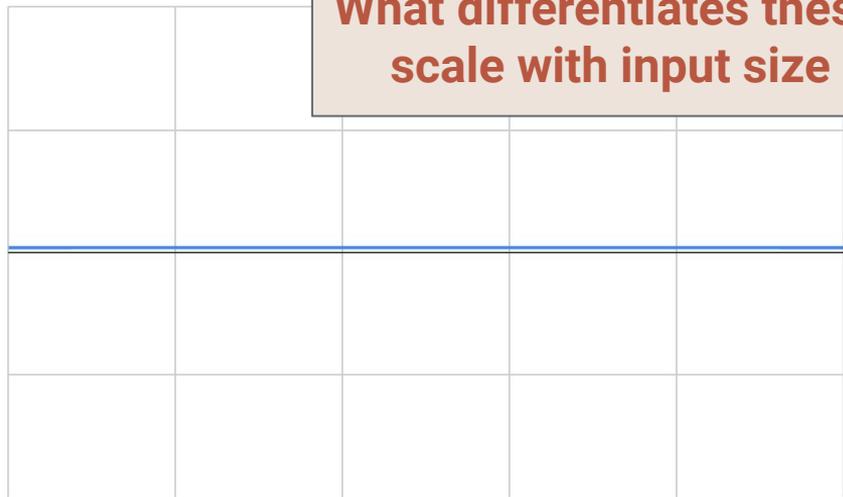


# Comparing Random Access for Array vs List

**Array**

**List**

What differentiates these two algorithms is how they scale with input size (the shape of the function)



# When is an algorithm “fast”?

- To give a useful solution, we should take “scale” into account
  - How does the runtime change as we change the size of the input (number of users, records, pixels, elements, etc)
- Don’t think in terms of wall-time, think in terms of “number of steps”

# Scaling Examples

- “Five steps plus Ten steps per user”
- “Ten steps per network connection. Each node has connections to 1% of the other nodes in the system”
- “Seven steps for every possible pair of elements
- “For each user, Ten steps, plus Three steps per post”

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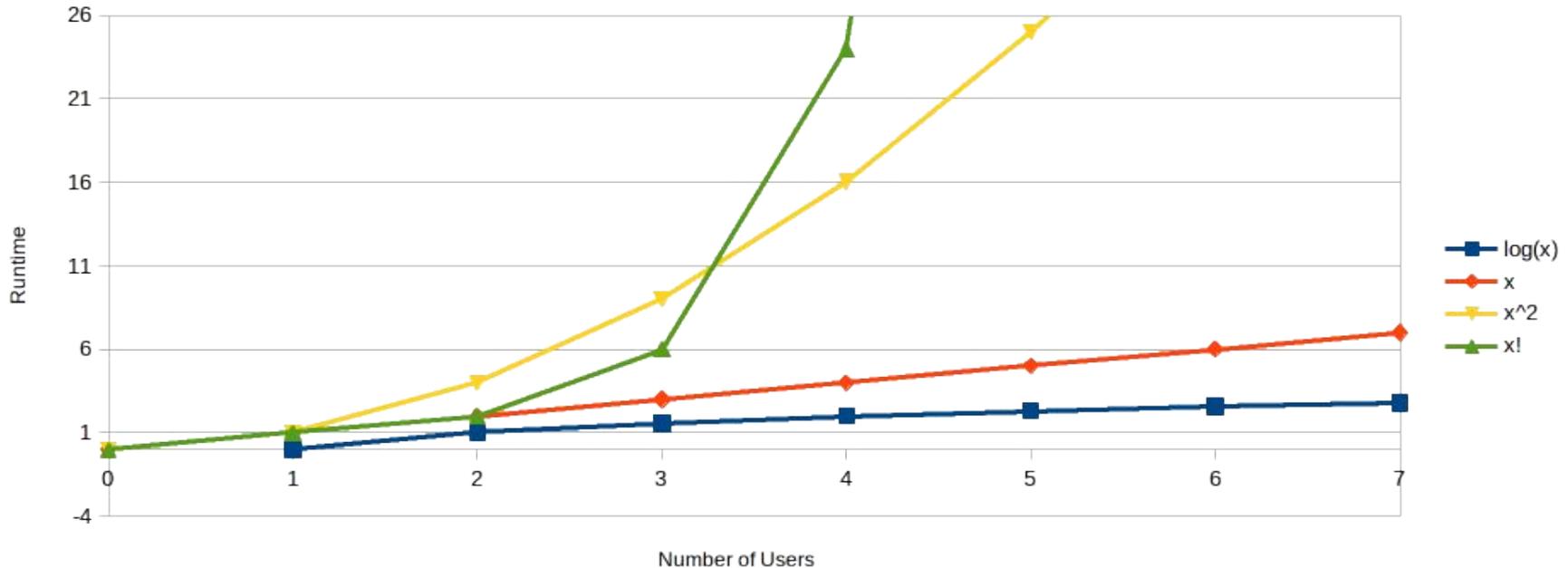
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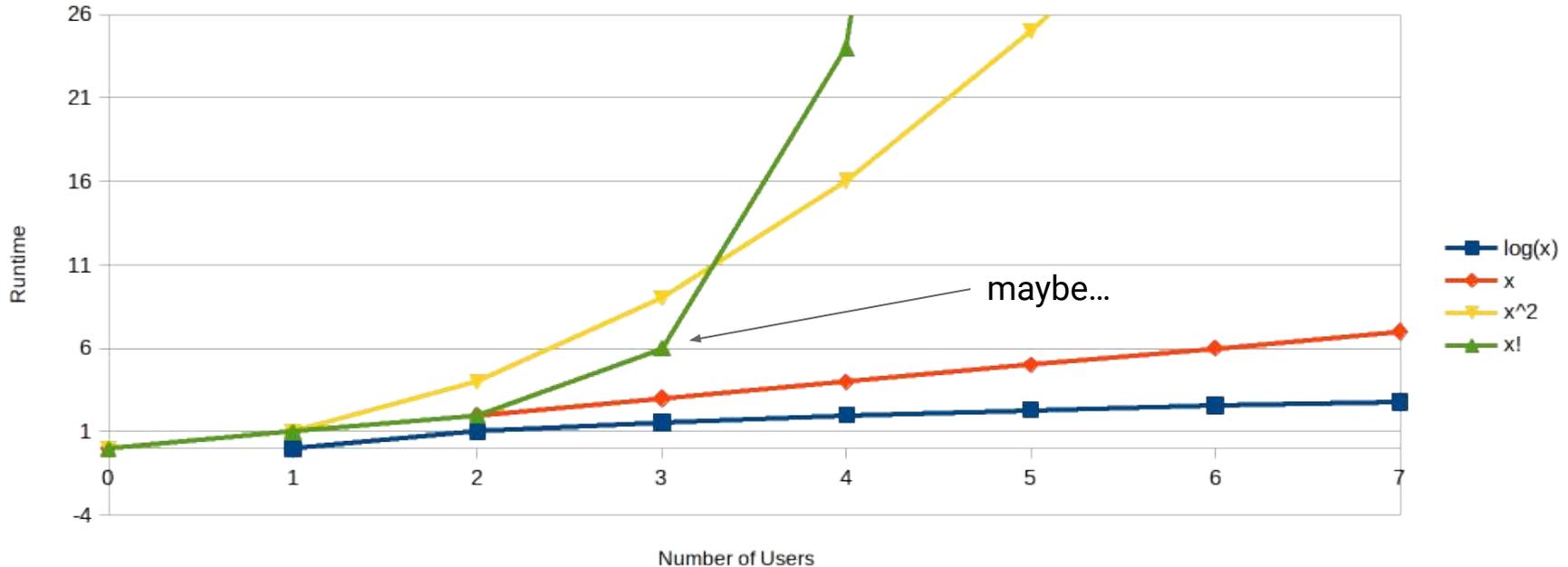
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  - $|\text{Users}| \times (10 + 3 \times |\text{Posts}|)$

# Runtime as a Function



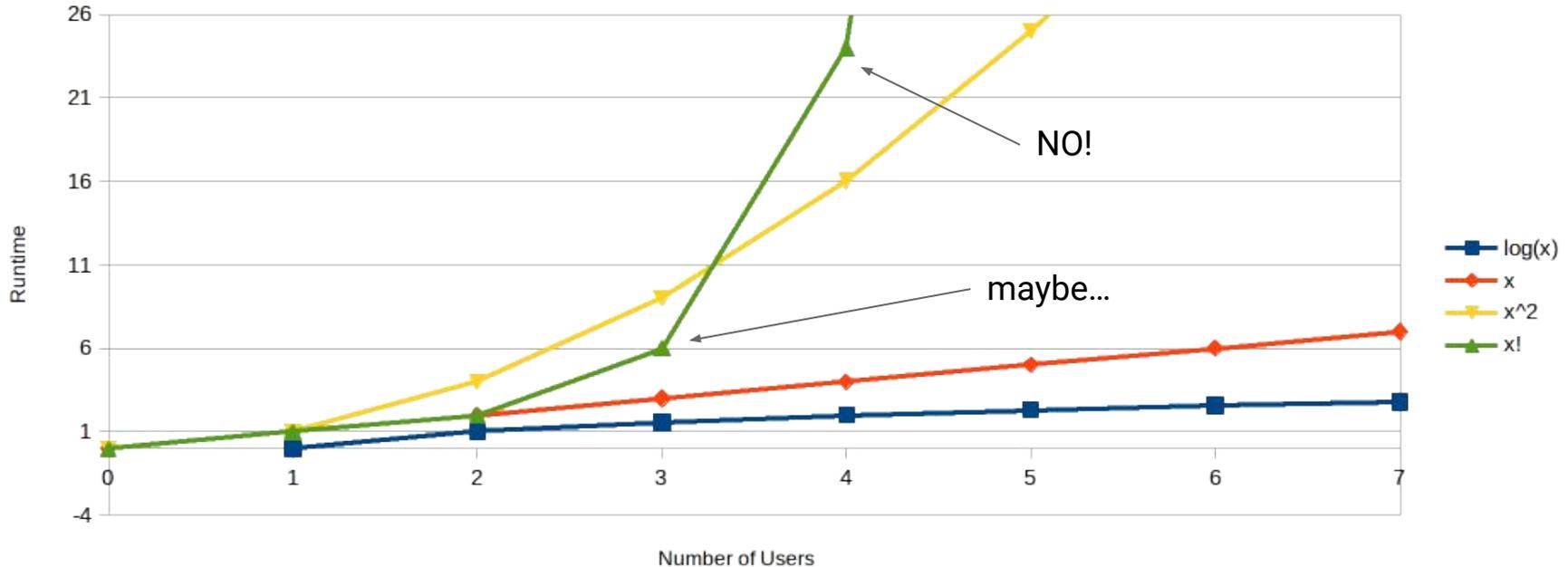
**Would you consider an algorithm that takes  $|\text{Users}|!$  number of steps?**

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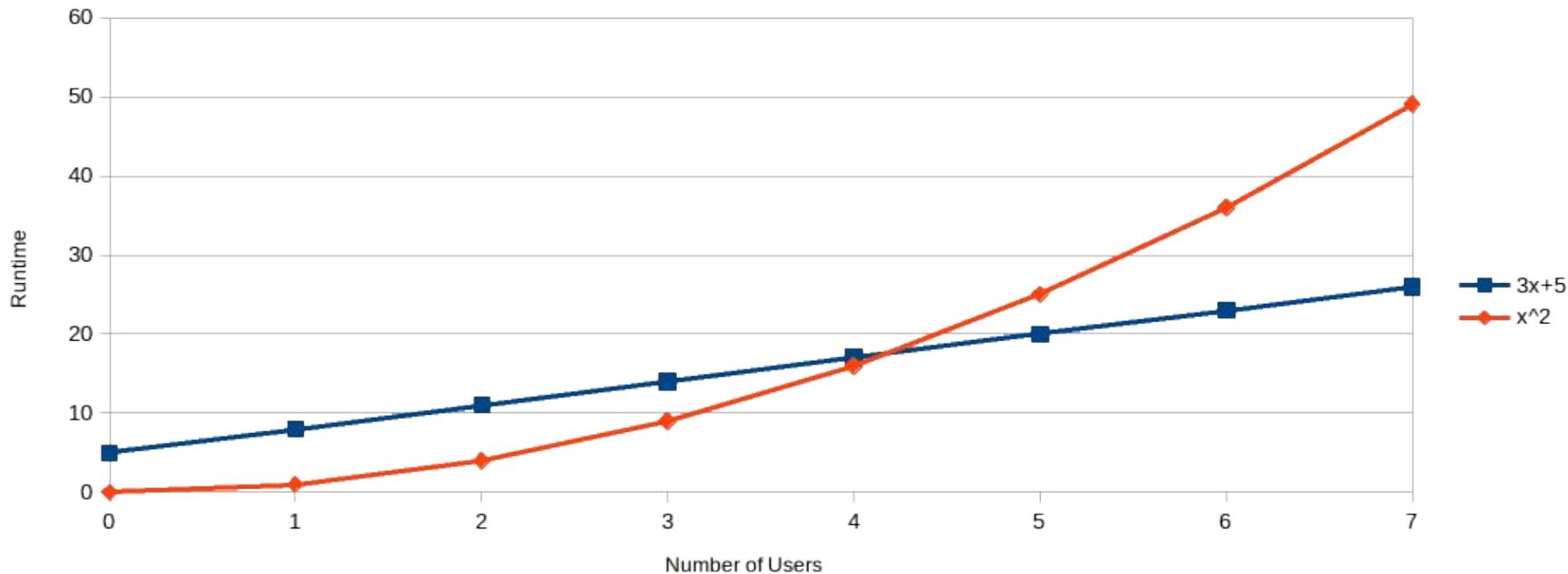
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Which is better?  $3x|\text{Users}|+5$  or  $|\text{Users}|^2$

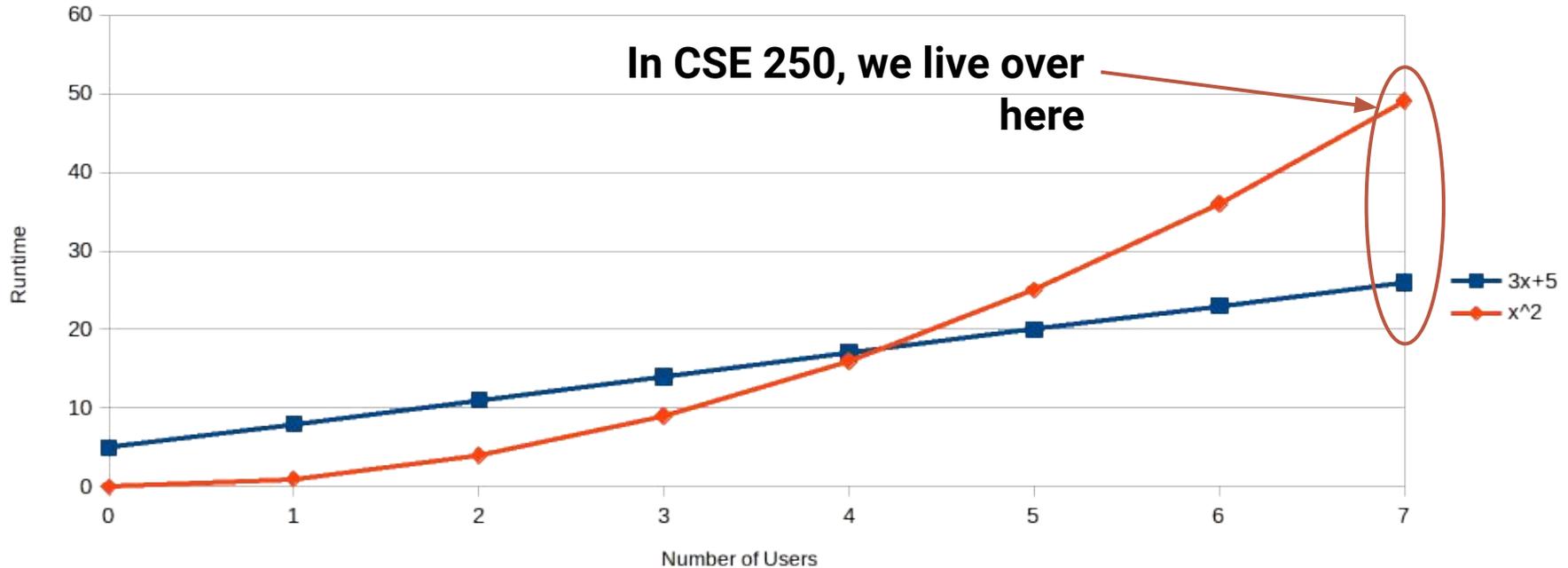
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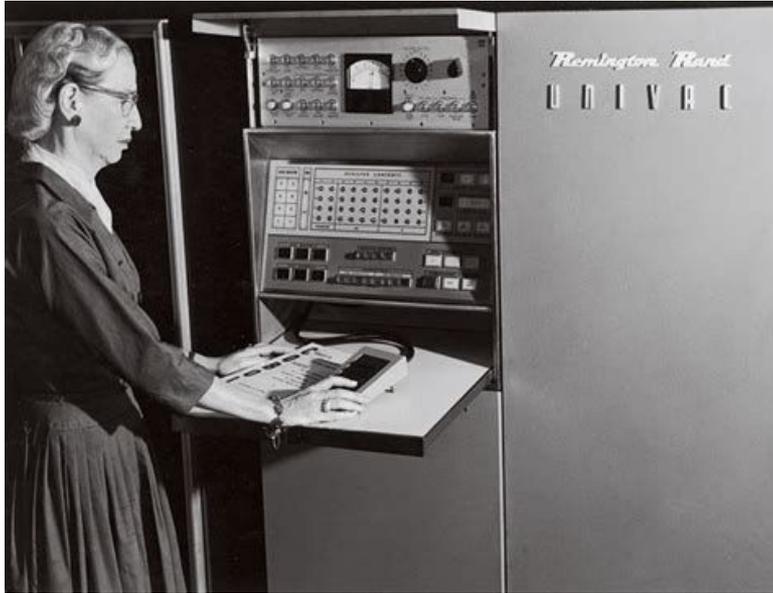
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  - Rank functions based on how they behave at large scales

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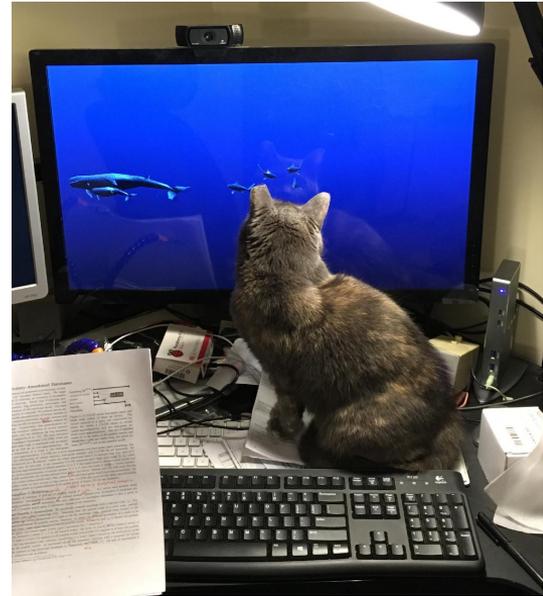
Which is better?  $3x|\text{Users}|+5$  or  $|\text{Users}|^2$

# Goal: Ignore implementation details



**Seasoned Pro Implementation**

**VS**



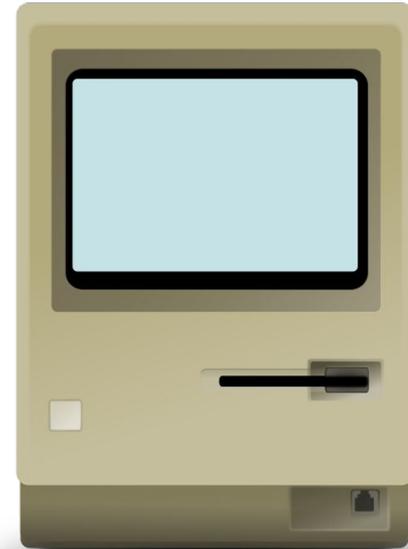
**Error 23: Cat on Keyboard**

# Goal: Ignore execution environment



**Intel i9**

**VS**



**Motorola 68000**

# Goal: Judge the Algorithm Itself

- How fast is a step? Don't care
  - Only count number of steps
- Can this be done in two steps instead of one?
  - “3 steps per user” vs “some number of steps per user”
  - Sometimes we don't care...sometimes we do

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  - Rank functions based on how they behave at large scales
- Decouple algorithm from infrastructure/implementation
  - Asymptotic notation...?

**And now a brief interlude...**

# Logarithms (refresher)

Let  $a, b, c, n > 0$

**Exponent Rule:**  $\log(n^a) = a \log(n)$

**Product Rule:**  $\log(an) = \log(a) + \log(n)$

**Division Rule:**  $\log(n/a) = \log(n) - \log(a)$

**Change of Base:**  $\log_b(n) = \log_c(n) / \log_b(c)$

**Log/Exponent are Inverses:**  $b^{\log_b(n)} = \log_b(b^n) = n$

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**In this class, always assume log base 2 unless specified otherwise**

**Now back to “fast”..**

# Attempt #2: Growth Functions

Not a function in code...but a mathematical function:

$$f(n)$$

**n: The “size” of the input**

ie: number of users, rows, pixels, etc

**f(n): The number of “steps” taken for input of size n**

ie: 20 steps per user, where  $n = |\text{Users}|$ , is  $20 \times n$

# Some Basic Assumptions:

Problem sizes are non-negative integers

$$n \in \mathbb{Z}^+ \cup \{0\} = \{0, 1, 2, 3, \dots\}$$

We can't reverse time...(obviously)

$$f(n) > 0$$

Smaller problems aren't harder than bigger problems

$$\forall n_1 < n_2, f(n_1) \leq f(n_2)$$

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$$f : \mathbb{Z}^+ \cup \{0\} \rightarrow \mathbb{R}^+$$

Smaller problems aren't harder than bigger problems

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# First Problem...

We are still implementation dependent

$$f_1(n) = 20n$$

$$f_2(n) = 19n$$

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Does 1 extra step per  
element really matter...?

# First Problem...

We are still implementation dependent

$$f_1(n) = 20n$$

$$f_2(n) = 19n$$

$$f_3(n) = \frac{n^2}{2}$$

$f_1$  and  $f_2$  are much more “similar” to each other than they are to  $f_3$

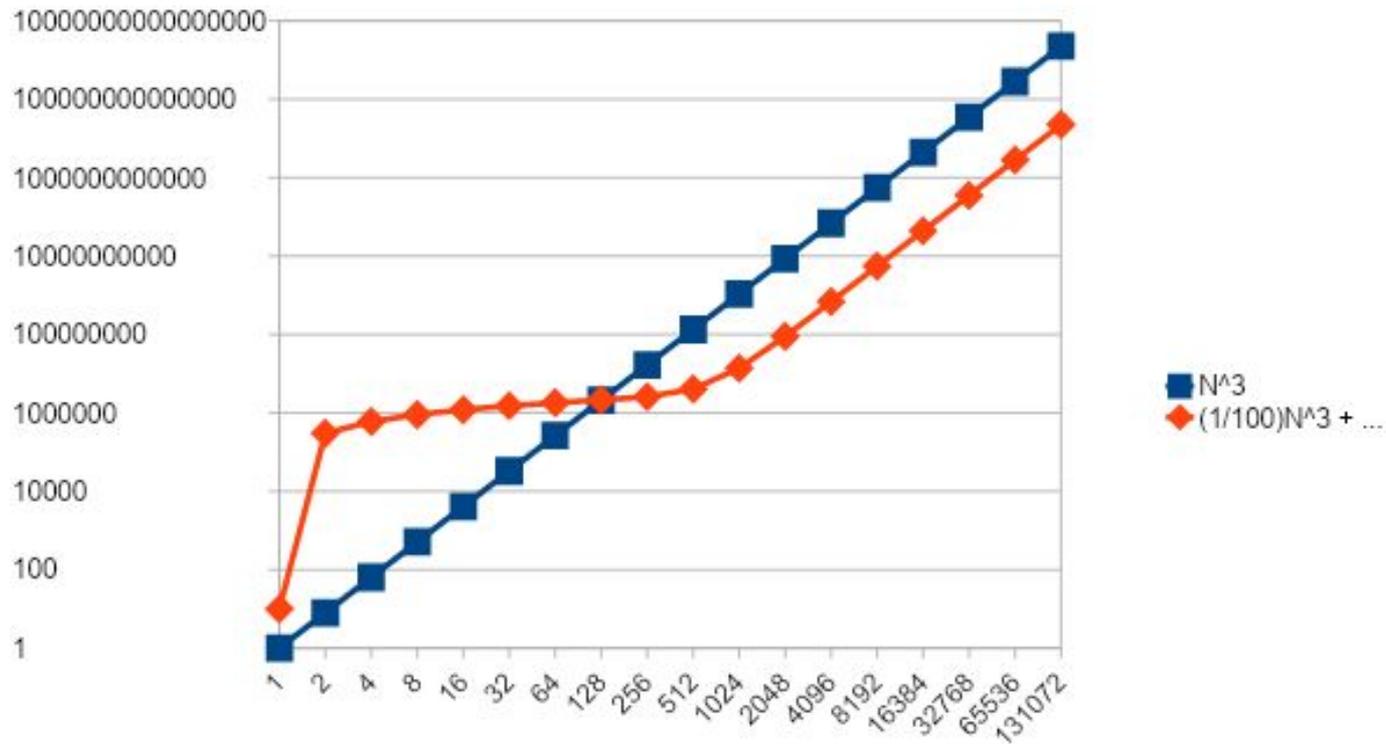
# How Do We Capture Behavior at Scale?

Consider the following two functions:

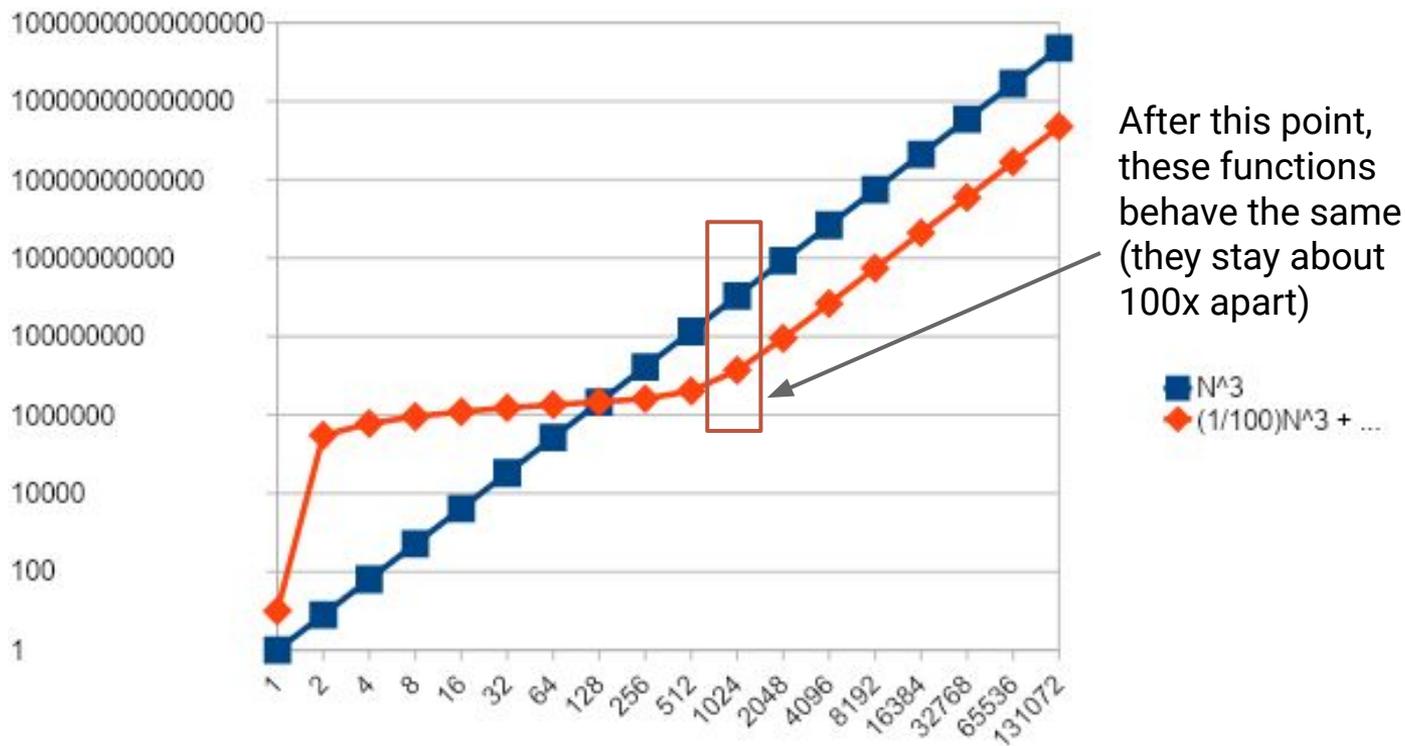
$$\frac{1}{100}n^3 + 10n + 1000000 \log(n)$$

$$n^3$$

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$$\lim_{n \rightarrow \infty} \frac{\frac{1}{100}n^3 + 10n + 1000000 \log(n)}{n^3}$$

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These terms go to 0

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# Attempt #3: Asymptotic Analysis

Consider two functions,  $f(n)$  and  $g(n)$

$$\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \infty$$

In this particular case,  $f$  grows w.r.t.  $n$  faster than  $g$

So...if  $f(n)$  and  $g(n)$  are the number of steps two different algorithms take on a problem of size  $n$ , which is better?

# Attempt #3: Asymptotic Analysis

**Case 1:**  $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \infty$  *(f grows faster; g is better)*

**Case 2:**  $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = 0$  *(g grows faster; f is better)*

**Case 3:**  $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \textit{some constant}$  *(f and g "behave" the same)*

# Goal of “Asymptotic Analysis”

We want to organize runtimes (growth functions)  
into different ***Complexity Classes***

Within the same complexity class, runtimes “behave  
the same”

# Goal of “Asymptotic Analysis”

**“Strategic Optimization” focuses on improving the complexity class of your code!**

# Back to Our Previous Example...

$$\frac{1}{100}n^3 + 10n + 1000000 \log(n)$$

The  $10n$  and  $1000000 \log(n)$  “don’t matter”

The  $1/100$  “does not matter”

# Back to Our Previous Example...

$$\frac{1}{100}n^3 + 10n + 1000000 \log(n)$$

The  $10n$  and  $1000000 \log(n)$  “don’t matter”

The  $1/100$  “does not matter”

**$n^3$  is the dominant term, and that determines the “behavior”**

# Why Focus on Dominating Terms?

$f(n)$	10	20	50	100	1000
$\log(\log(n))$	0.43 ns	0.52 ns	0.62 ns	0.68 ns	0.82 ns
$\log(n)$	0.83 ns	1.01 ns	1.41 ns	1.66 ns	2.49 ns
$n$	2.5 ns	5 ns	12.5 ns	25 ns	0.25 $\mu$ s
$n\log(n)$	8.3 ns	22 ns	71 ns	0.17 $\mu$ s	2.49 $\mu$ s
$n^2$	25 ns	0.1 $\mu$ s	0.63 $\mu$ s	2.5 $\mu$ s	0.25 ms
$n^5$	25 $\mu$ s	0.8 ms	78 ms	2.5 s	<b>2.9 days</b>
$2^n$	0.25 $\mu$ s	0.26 ms	<b>3.26 days</b>	<b><math>10^{13}</math> years</b>	<b><math>10^{284}</math> years</b>
$n!$	0.91 ms	<b>19 years</b>	<b><math>10^{47}</math> years</b>	<b><math>10^{141}</math> years</b>	

# Why Focus on Dominating Terms?

$$2^n \gg n^c \gg n \gg \log(n) \gg c$$