



Middlebury

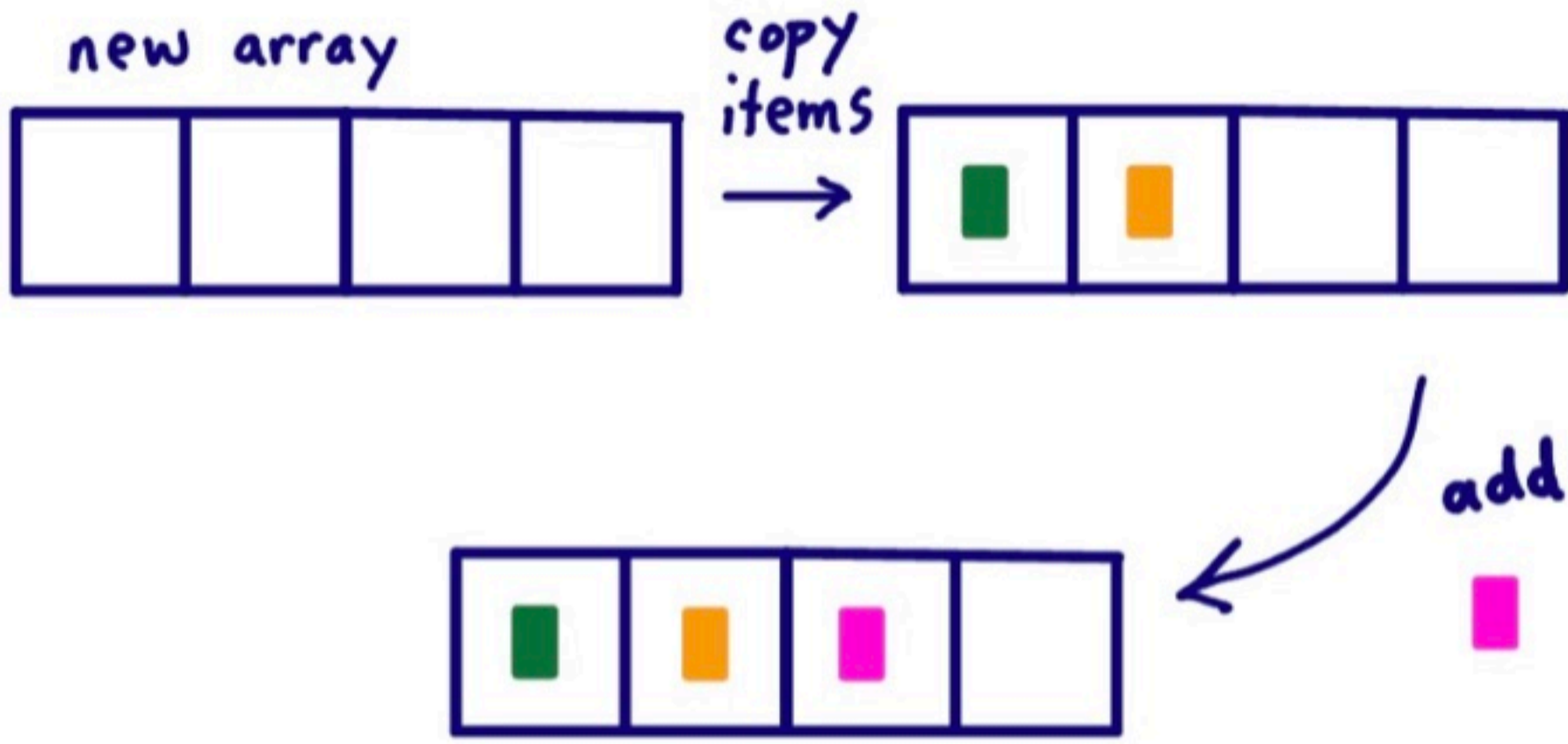
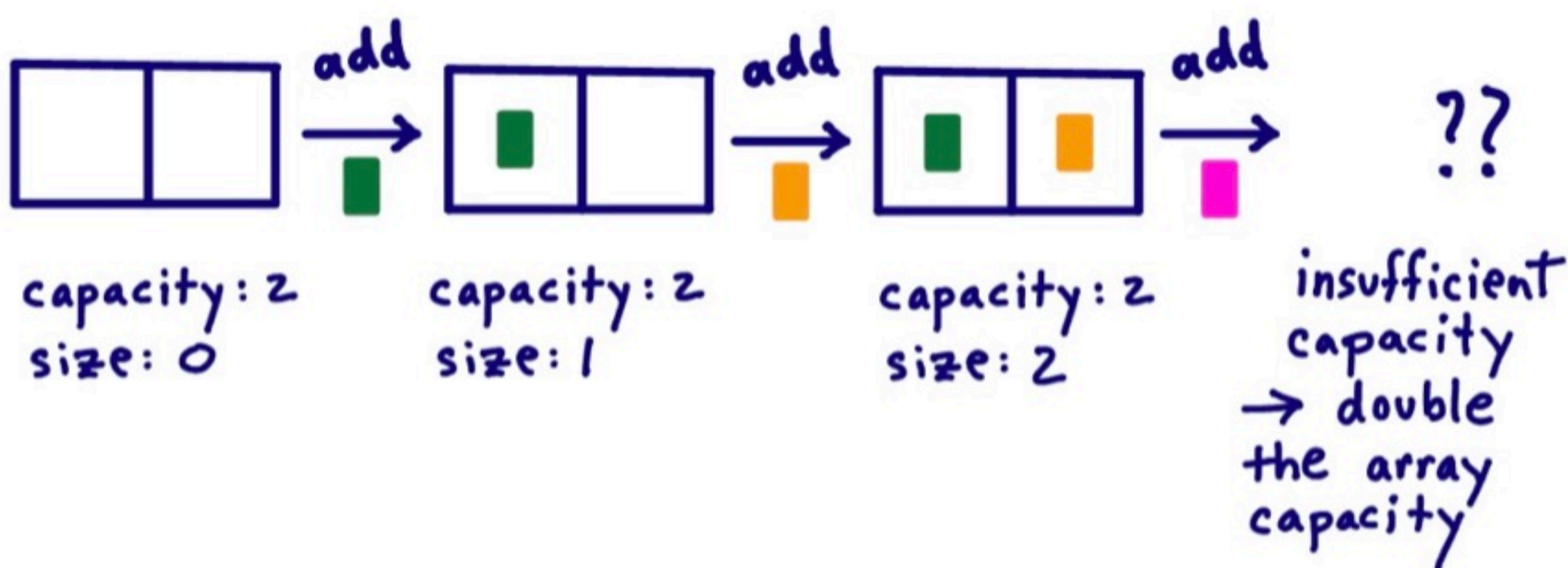
# CSCI 201: Data Structures

Spring 2025

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## Lecture 4M: Complexity

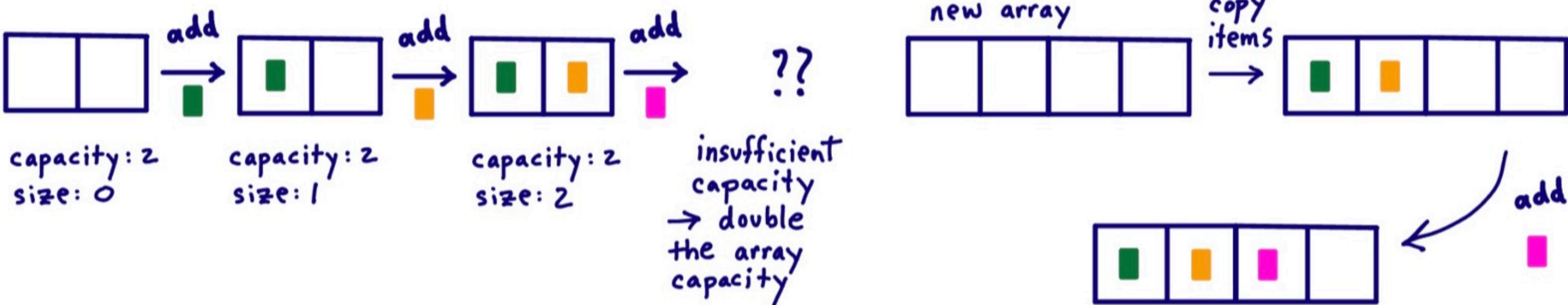
# Remember our decision to *double* the capacity of a **DIYList** when we ran out of space during a call to **add**?





# Goals for today:

- Analyze the runtime cost of our **add** method for a **DIYList** as we call it many times.
- Characterize how functions grow as the inputs get very large.
- Use big-O notation to describe the running time of algorithms.





We also want our analysis to be computer-independent.



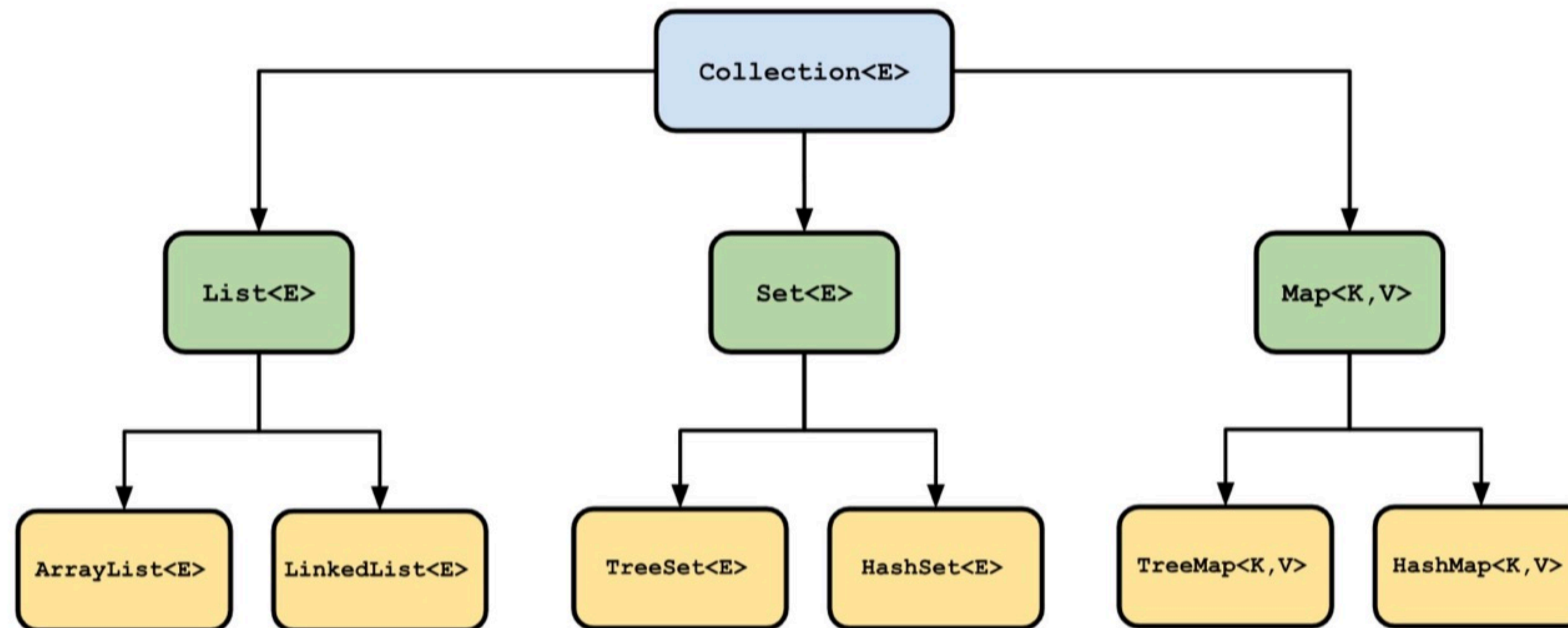
**Two types of resources to consider:**

- **Processor cycles:** number of operations per second a machine can perform.
- **Memory:** space for storing data while program is running (RAM, cache).

← 2 GHz  
← 16 GB



The **Collections** framework describes the efficiency an implemented method should provide.



The **size**, **isEmpty**, **get**, **set**, **iterator**, and **listIterator** operations run in constant time. The **add** operation runs in amortized constant time, that is, adding  $n$  elements requires  $O(n)$  time. All of the other operations run in linear time (roughly speaking).



## What kinds of things in our programs might affect the runtime?

for-loops, while-loops

conditionals

function calls

assignment =

comparison <, <=, >, >=, ==, !=

logical &&, ||

arithmetic +, -, \*, /, ++, --



Analyzing how many = we're doing in the **add** method when *doubling* the capacity (as needed). Assume we start with a capacity of **1**.

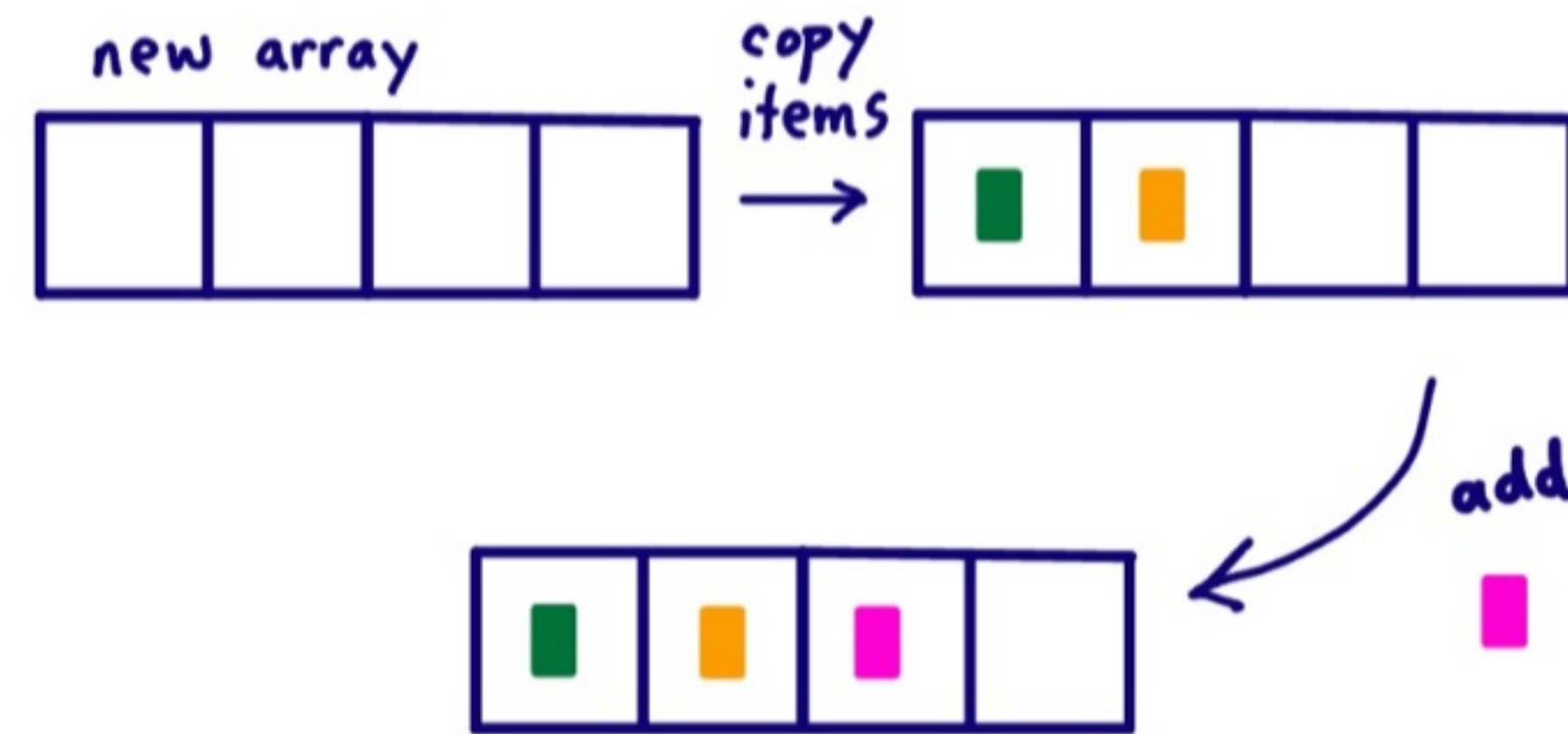
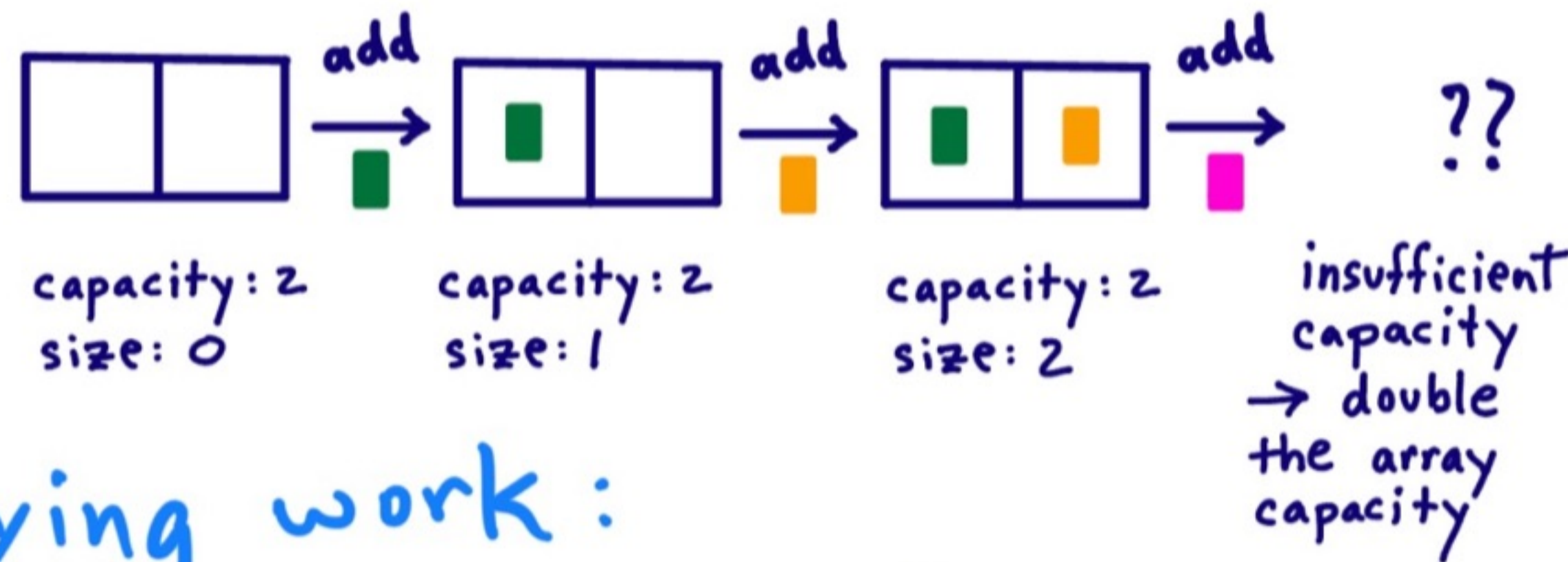
Define  
 $n = \# \text{ values}$   
 $m = \# \text{ resizings}$

```
public class DIYList {
    int size; // current number of items actually stored
    String[] items; // capacity is items.length

    public void add(String item) {
        // Is there enough space (capacity, i.e. items.length)?
        // If not, make more space and copy the old items.

        // Place item in items[size] and increment size.
    }
}
```

Doubling means  
 $n = 2^m + 1$  so  
 $2^m = n - 1$



Copying work:

$$1 + 2^1 + 2^2 + 2^3 + \dots + 2^m$$

$$= \frac{2^{m+1} - 1}{2 - 1} = 2^{m+1} - 1 = 2 \cdot 2^m - 1$$

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

Adding  $n$  new values is additional  $n$  steps  
 $\Rightarrow$  total  $3n - 3$  < >





# Two types of series we'll encounter:

## 1. Geometric:

$$1 + r + r^2 + r^3 + \dots + r^m = \frac{r^{m+1} - 1}{r - 1}$$

Here,  $r=2$

## 2. Arithmetic:

$$1 + 2 + 3 + \dots + n = \frac{n(n+1)}{2}$$

Derive in CS 200!



So the total number of = when calling **add**  $n$  times is:

$$3n - 3$$

```
public class DIYList {
    int size; // current number of items actually stored
    String[] items; // capacity is items.length

    public void add(String item) {
        // Is there enough space (capacity, i.e. items.length)?
        // If not, make more space and copy the old items.

        // Place item in items[size] and increment size.
    }
}
```

Averaged over  $n$  calls to **add** (with  $n$  getting very large):

$$\frac{3n-3}{n} = 3 - \frac{3}{n} \quad \text{as } n \rightarrow \infty, \quad \frac{3}{n} \rightarrow 0$$

$$\boxed{= 3}$$

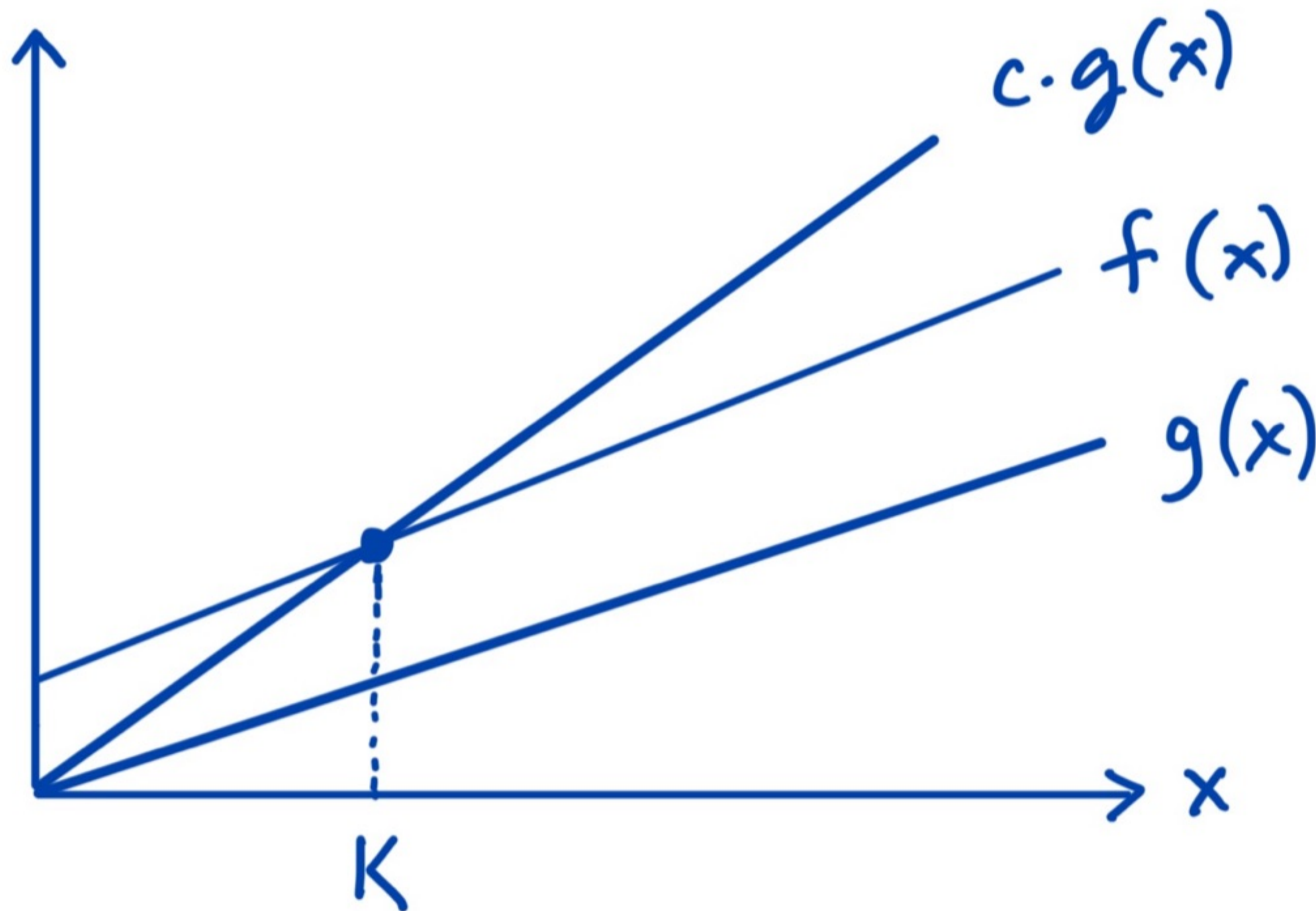
← constant time  
amortized over  
 $n$  adds



We need a better way to analyze running time of algorithms.

**Big-O notation:** Given functions  $f$ ,  $g$ , we say that  $f(x)$  is  $\mathcal{O}(g(x))$  if-and-only-if there exist constants  $c > 0$  and  $k$  such that

$$|f(x)| \leq c \cdot |g(x)|, \quad \text{for all } x \geq k$$



$f(x)$  is eventually  
no larger than  
some constant  
multiple of  $g(x)$



Example: Show that  $x^2 + 2x + 1$  is  $\mathcal{O}(x^2)$ .

Show  $x^2 + 2x + 1 \leq c \cdot x^2$

Rewrite as  $(c-1)x^2 - 2x - 1 \geq 0$

Choose  $c=2 \Rightarrow$  Show  $x^2 - 2x - 1 \geq 0$

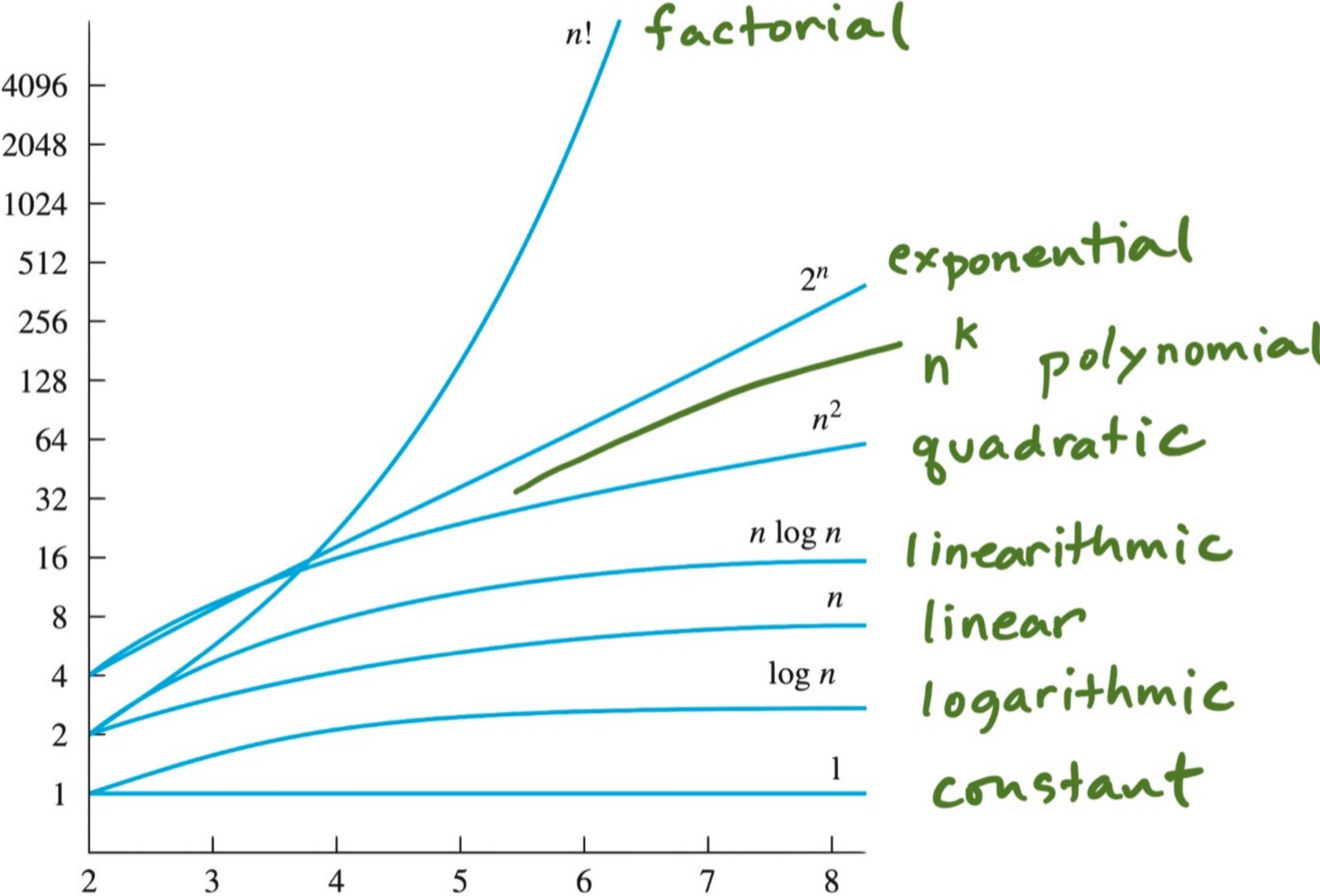
Find  $k$  : try  $x=2$  :  $(2)^2 - 2(2) - 1 = -1 \geq 0$ ? no  
try  $x=3$  :  $(3)^2 - 2(3) - 1 = 2 \geq 0$ ? yes

So constants  $c=2$  and  $k=3$

work to show  $x^2 + 2x + 1$  is  $\mathcal{O}(x^2)$



# Common functions used in big-O estimates.



(Discrete Mathematics and Its Applications 7th Ed., Rosen)





# We usually want to express our algorithm runtime using the *tightest bound*.

We'll often use  $T(n)$  to represent algorithm runtime in terms of input size  $n$ .

## Strategy:

1. Pick out fastest growing term in  $T(n)$ .
2. Drop coefficients.

Eg,  $1 + 100n^2$  :  $O(n)$ ? no  
*tightest*  $\rightarrow$   $O(n^2)$ ? yes  
 $O(n^3)$ ? yes

Determine a big-O bound for the following functions.

1.  $T(n) = 1 + 5n$ :  $O(n)$

2.  $T(n) = 1 + 5n^2$ :  $O(n^2)$

3.  $T(n) = 5 + 20n + 3n^2$ :  $O(n^2)$

4.  $T(n) = \frac{n^2(n^2+1)}{2}$ :  $(n^4 + n^2) / 2$  :  $O(n^4)$

5.  $T(n) = 5$ :  $O(1)$

6.  $T(n) = n(5 + \log n)$ :  $5n + n \log n$  :  $O(n \log n)$



# A few rules for inferring big-O bounds on algorithm runtime.

**consecutive statements:**  $T(n) = T(s_1) + T(s_2)$

```
statement1; // performing T(s1) amount of work  
statement2; // performing T(s2) amount of work
```

**for loop:**  $T(n) = n \times T(b)$ .

```
for (int i = 0; i < n; i++) {  
    // some block performing T(b) amount of work  
}
```

**nested for loop:**  $T(n, m) = n \times m \times T(b)$ .

```
for (int i = 0; i < n; i++) {  
    for (int j = 0; j < m; j++) {  
        // some block performing T(b) amount of work  
    }  
}
```

**if statements:**  $T(n) = T(c) + \max(T(b_i), T(b_e))$

```
if (condition) { // condition performs T(c) amount of work  
    body1; // performing T(bi) amount of work  
} else {  
    body2; // performing T(be) amount of work  
}
```



Determine  $T(n)$  (an expression for the number of operations performed by the following algorithms), then provide a big-O bound on  $T(n)$ . **Focus on counting ++**

Example 1:

```
int sum = 0;
for (int i = 0; i < n; i++) {
    for (int j = 0; j < m; j++) {
        sum++;
    }
}
```

*Handwritten annotations:  $n$  above the first loop,  $n \cdot m$  below the second loop, and  $n \cdot m$  at the bottom.*

$T(n,m) = 2nm + n$   
 $O(nm)$

Example 2:

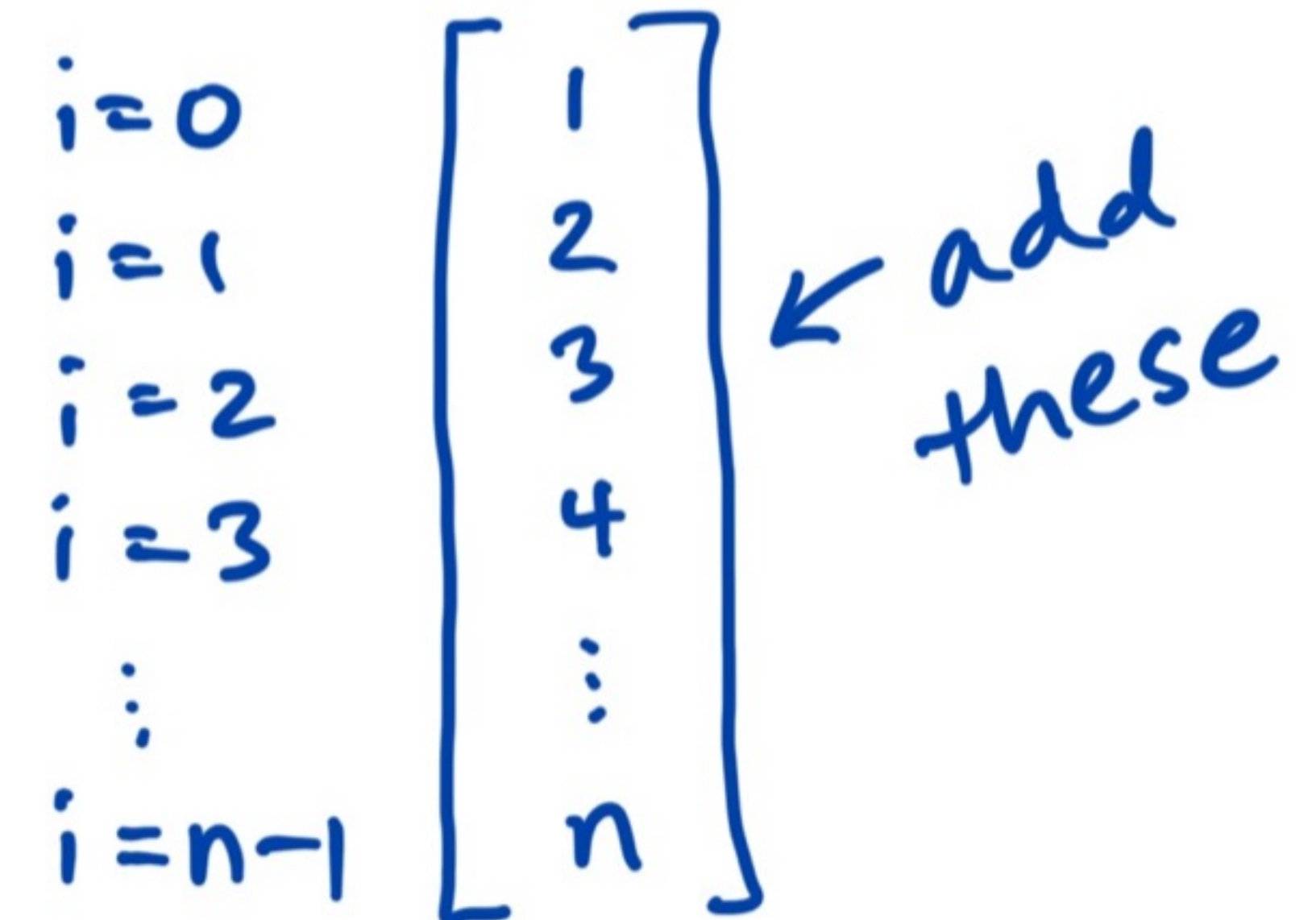
```
int sum = 0;
for (int i = 0; i < n; i++) {
    for (int j = 0; j < n; j++) {
        for (int k = 0; k < m; k++) {
            sum++;
        }
    }
}
```

*Handwritten annotations:  $n^2$  above the second loop,  $n^2 m$  below the third loop, and  $n^2$  at the top right.*

~~$2n^2m + n^2 + n$~~   
 $O(n^2m)$

Example 3:

```
int sum = 0;
for (int i = 0; i < n; i++) {
    for (int j = 0; j <= i; j++) {
        sum++;
    }
}
```



$\frac{n(n+1)}{2} : O(n^2)$

Example 4:

```
int i = n;
while (i >= 1) {
    i = i / 2;
}
```

$O(\log n)$

Eg, take  $n = 32$   
 So  $i = 32, 16, 8, 4, 2, 1$   
 Loop runs 6 times





Describe the worst-case runtime complexity of the following methods in Big-O.

Use variables to describe complexity of data structures, eg,  $n = \text{nums.size()}$ ,  $n = \text{list1.size()}$ , and  $m = \text{list2.size()}$

Complexity of isEven:  $O(1)$

```
public static boolean isEven(int n) {  
    return (n % 2 == 0);  
}
```

Complexity of isPrime:  $O(n)$

```
public static boolean isPrime(int n) {  
    if (n < 2) {  
        return false;  
    }  
    for (int i = 2; i < n; i++) {  
        if (n % i == 0) return false;  
    }  
    return true;  
}
```

Complexity of factorialOf:  $O(n)$

```
public static long factorialOf(int n) {  
    switch(n) {  
        case 0:  
        case 1: return 1;  
        default: return n * factorialOf(n - 1);  
    }  
}
```

Complexity of sumProduct:  $O(n^2)$

```
public static Integer sumProduct(ArrayList nums) {  
    int sum = 0;  
    for (int i = 0; i < nums.size(); i++) {  
        for (int j = 0; j < nums.size(); j++) {  
            if (i != j) {  
                sum += nums.get(i) * nums.get(j);  
            }  
        }  
    }  
    return sum;  
}
```

Complexity of checking:  $O(nm)$

```
public static boolean checking(ArrayList list1, ArrayList list2) {  
    int value1;  
    int value2;  
    for (int i = 0; i < list1.size(); i++) {  
        value1 = list1.get(i);  
        for (int j = 0; j < list2.size(); j++) {  
            value2 = list2.get(j);  
            if (value1 == value2) return true;  
        }  
    }  
    return false;  
}
```

Complexity of lastElement:  $O(1)$

```
public static Integer lastElement(ArrayList nums) {  
    if (nums.size() == 0) return null;  
    else return nums.get(nums.size()-1);  
}
```

Complexity of createPairs:  $O(n^2)$

```
public static void createPairs(ArrayList nums) {  
    for (int i = 0; i < nums.size(); i++) {  
        for (int j = i+1; j < nums.size(); j++) {  
            System.out.println(nums.get(i) + ", " + nums.get(j));  
        }  
    }  
}
```





## See you on Wednesday!

- We'll use what we covered today to analyze some sorting algorithms.
- Work on [Homework 3](#)! Implement your own `DIYArrayListString`.
- Reminder that Noah ([go/noah](#)) and Smith ([go/smith](#)) have office hours throughout the week and the 201 Course Assistants have drop-in hours in the late afternoons/evenings ([go/cshelp](#)).