



10-607  
Computational  
Foundations for  
Machine Learning

Computational  
Complexity

Instructor: Pat Virtue

# Plan

## Computational Complexity

- Counting operations
- Big-O
- Complexity classes
- Proving a Big-O relationship holds
- Proving a Big-O relationship does *not* hold

# How many statements are executed?

```
15  int search(int x, int[] A, int n)
16  {
17      for (int i = 0; i < n; i++)
18      {
19          if (A[i] == x) {
20              return i;
21          }
22      }
23      return -1;
24  }
```

If x is not in A...  
how times are these  
statements executed?

1

`i = 0`

$n+1$

`i < n`

$n$

`if (A[i] == x)`

$n$

`i++`

1

`return -1`

# How many **operations** are executed?

```
15  int search(int x, int[] A, int n)
16  {
17      for (int i = 0; i < n; i++)
18      {
19          if (A[i] == x) {
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```

If x is not in A...  
how times **operations** are  
executed?

1     i = 0  
n+1   i < n  
3n    if (A[i] == x) ←  
2n    i++     i = i + 1  
1     return -1

# How many **operations** are executed?

How many program operations are required to compute:

- L2 norm of vector
- Vector dot product
- Frobenius norm of matrix
- Matrix-vector multiplication
- Matrix-matrix multiplication

```
def norm(a):  
    ss=0  
    for i in range(len(a)):  
        ss = ss + a[i]*a[i]  
    norm = np.sqrt(ss)  
    return norm
```



Operations:

- Arithmetic operations (e.g. + or \*\*)
- Logical operations (e.g., and)
- Comparison operations (e.g., <=)
- Structure accessing operations (e.g. array indexing like A[i])
- Simple assignment such as copying a value into a variable
- Calls to library functions that don't depend on size of input (e.g., print)
- Control Statements (e.g. if X>5)

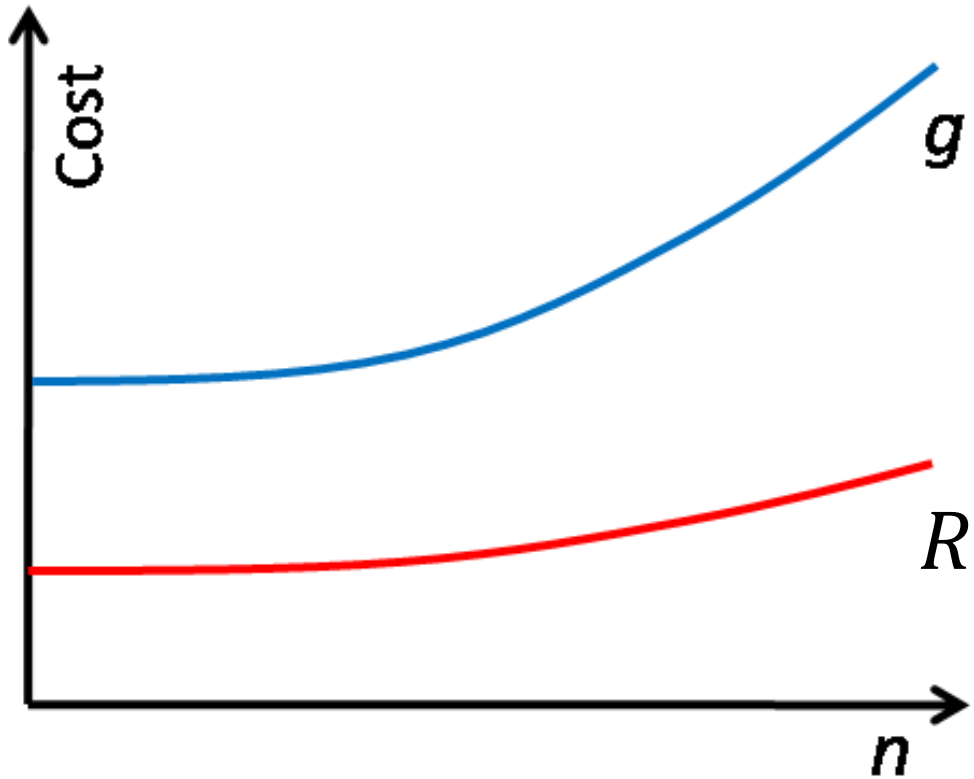
**Be careful with function calls that scale with the size of the input**

# Exercise

Counting operations handout

# Comparing functions of $n$

Which is better?



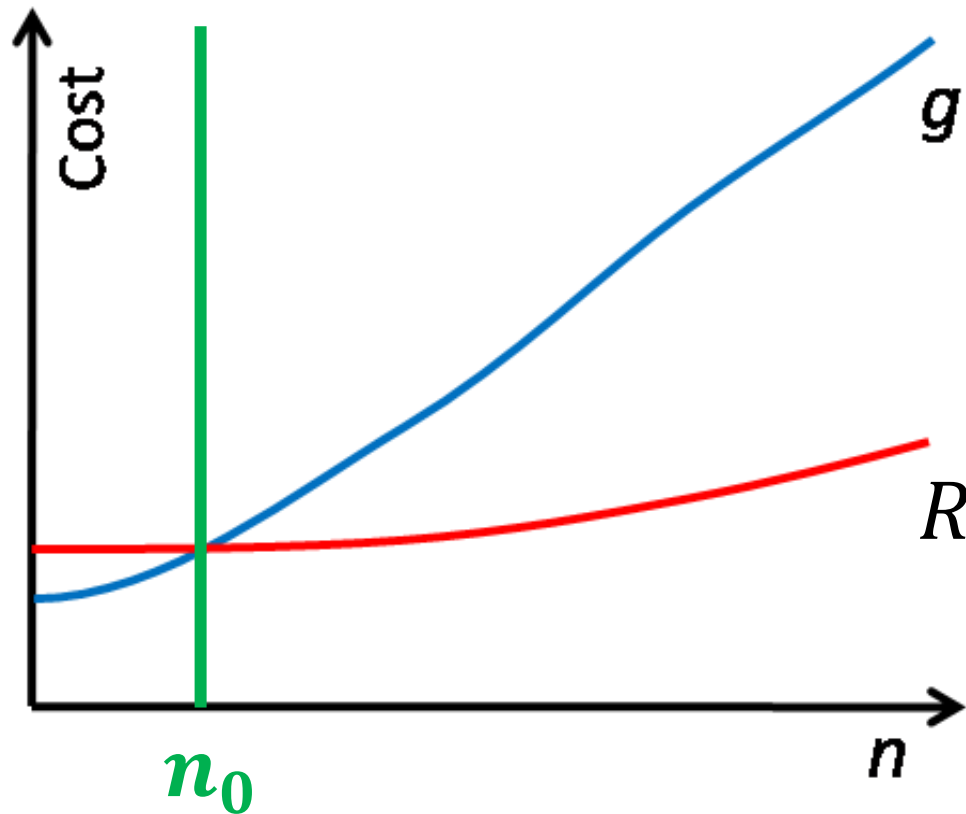
$R$  is better than  $g$   
if

$$R(n) \leq g(n)$$

for all  $n$

# Comparing functions of $n$

Which is better?



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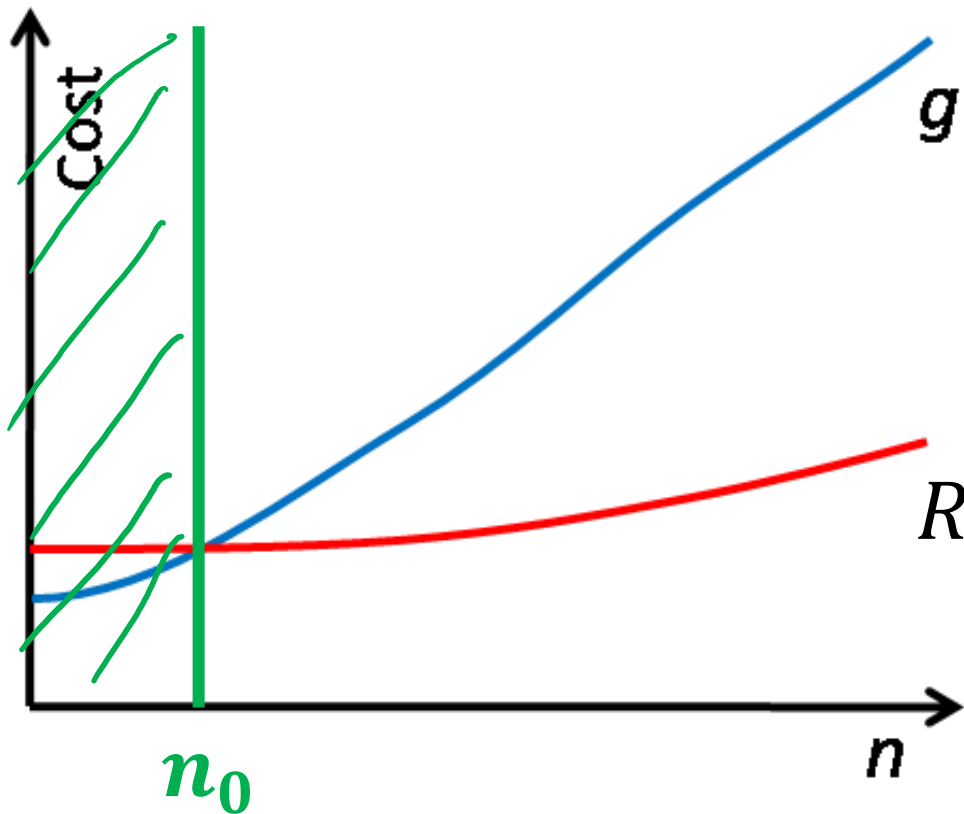
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for all  $n$



# Comparing functions of $n$

Which is better?



R is better than g

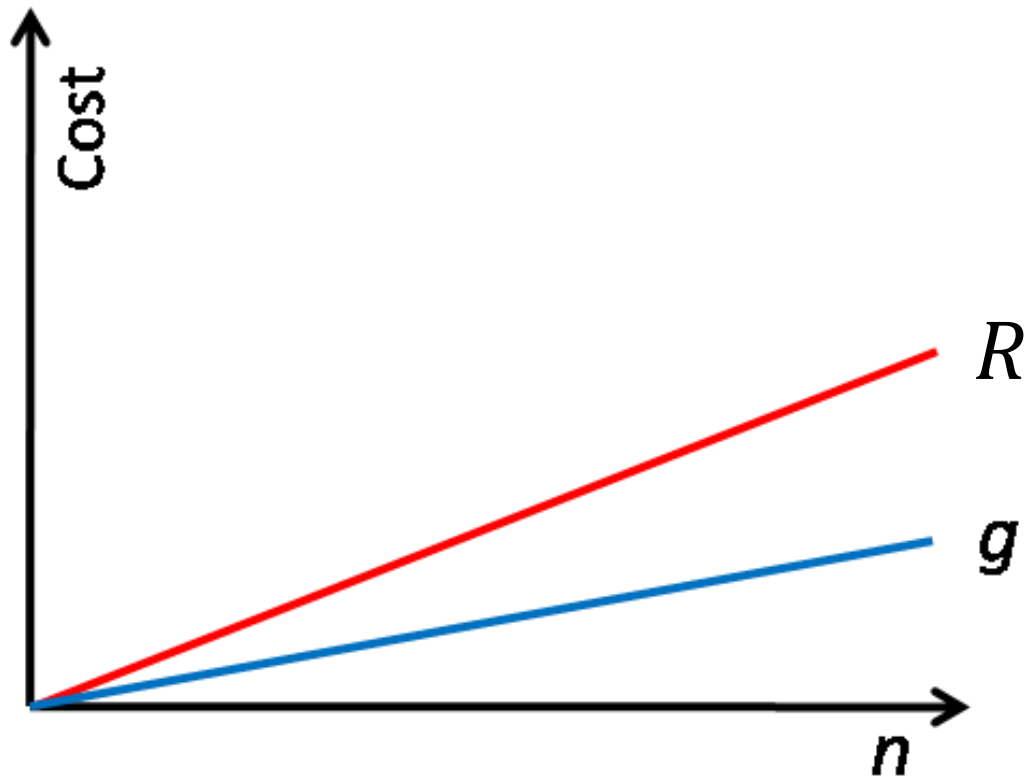
if **there exists  $n_0$ , such that**

$$R(n) \leq g(n)$$

for all  $n \geq n_0$

# Comparing functions of $n$

Which is better?



$R$  is better than  $g$

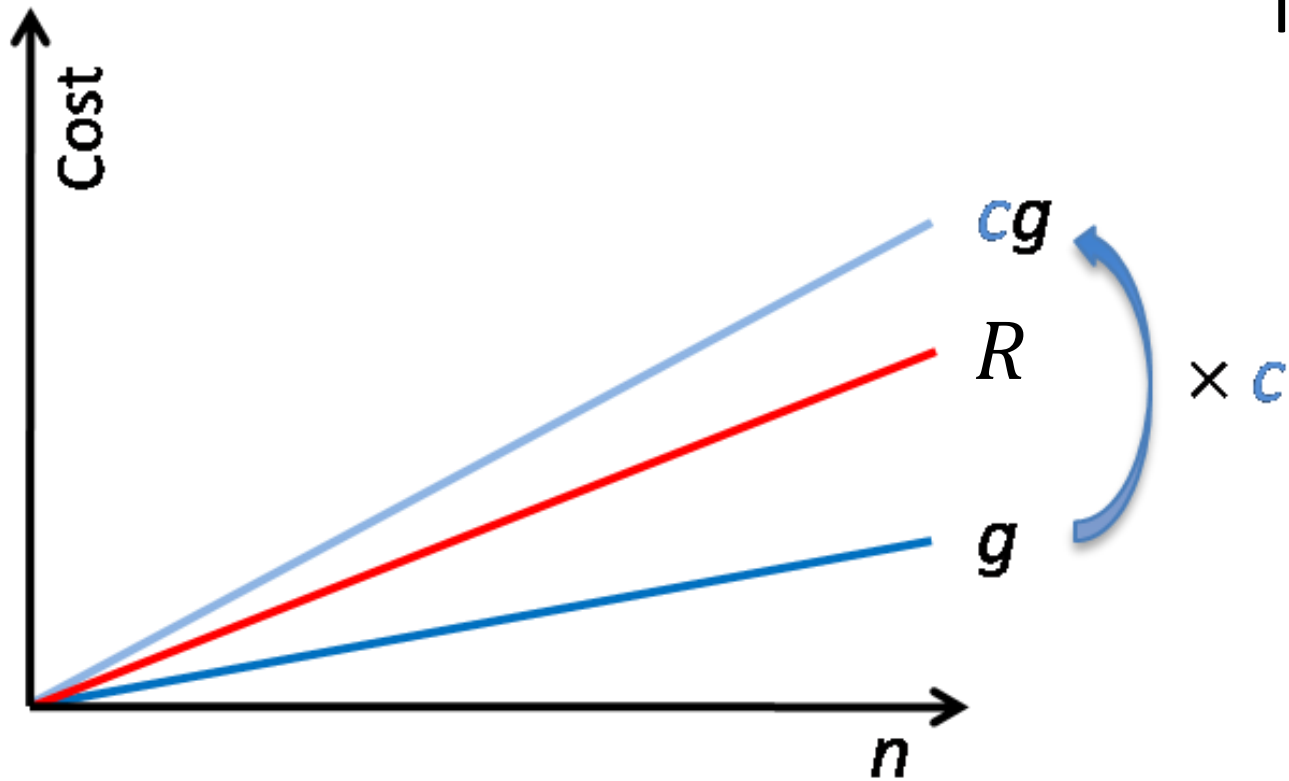
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Which is better?



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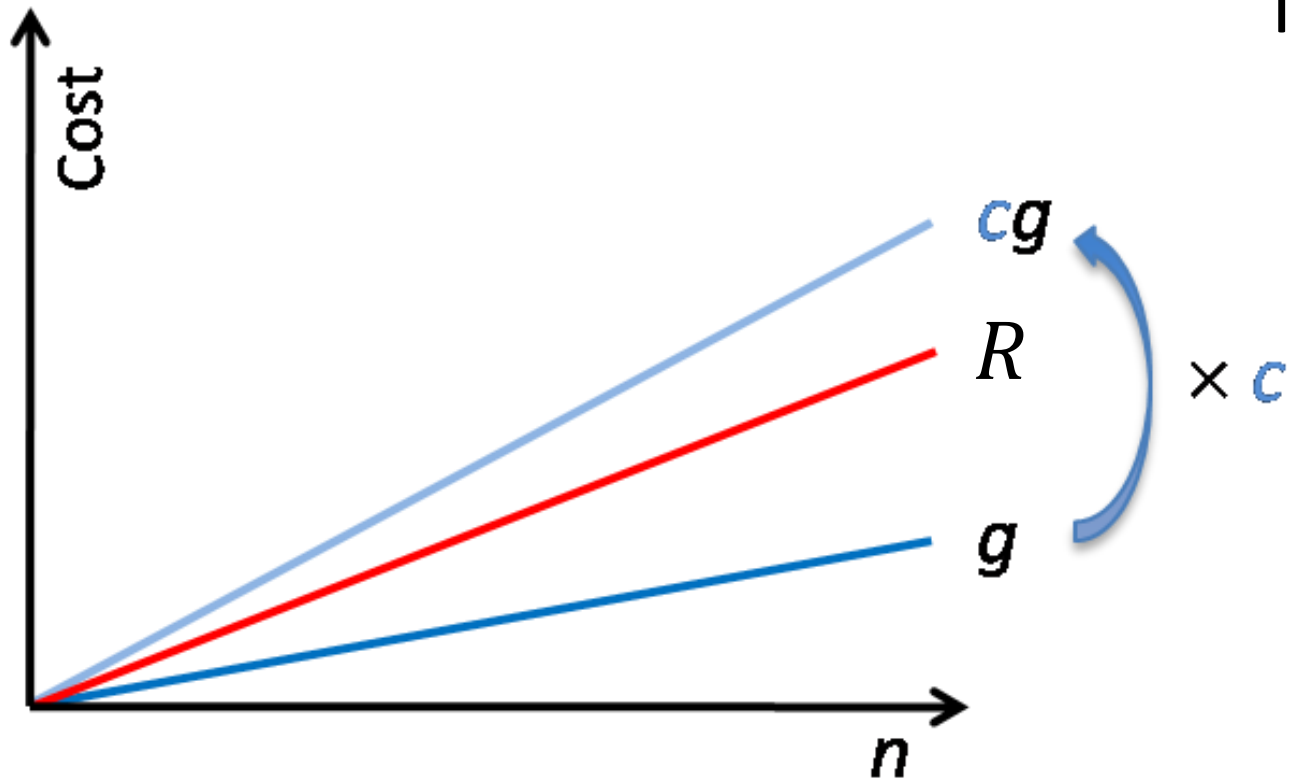
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# Comparing functions of $n$

Which is better?



$R$  is better than  $g$

if there exists  $n_0$  and  $c$ , s.t.

$$R(n) \leq cg(n)$$

for all  $n \geq n_0$

# A set of better functions

R is better than g

if there exists  $n_0$  and  $c$ , s.t.

$$R(n) \leq cg(n)$$

for all  $n \geq n_0$

A set of better functions

**The set of all functions  $R$  where**  
there exists  $n_0$  and  $c$ , s.t.

$$R(n) \leq cg(n)$$

for all  $n \geq n_0$

Definition of Big O of  $g$

A set of better functions

## Definition of Big O of g

**The set of all functions R where**  
there exists  $n_0$  and  $c$ , s.t.

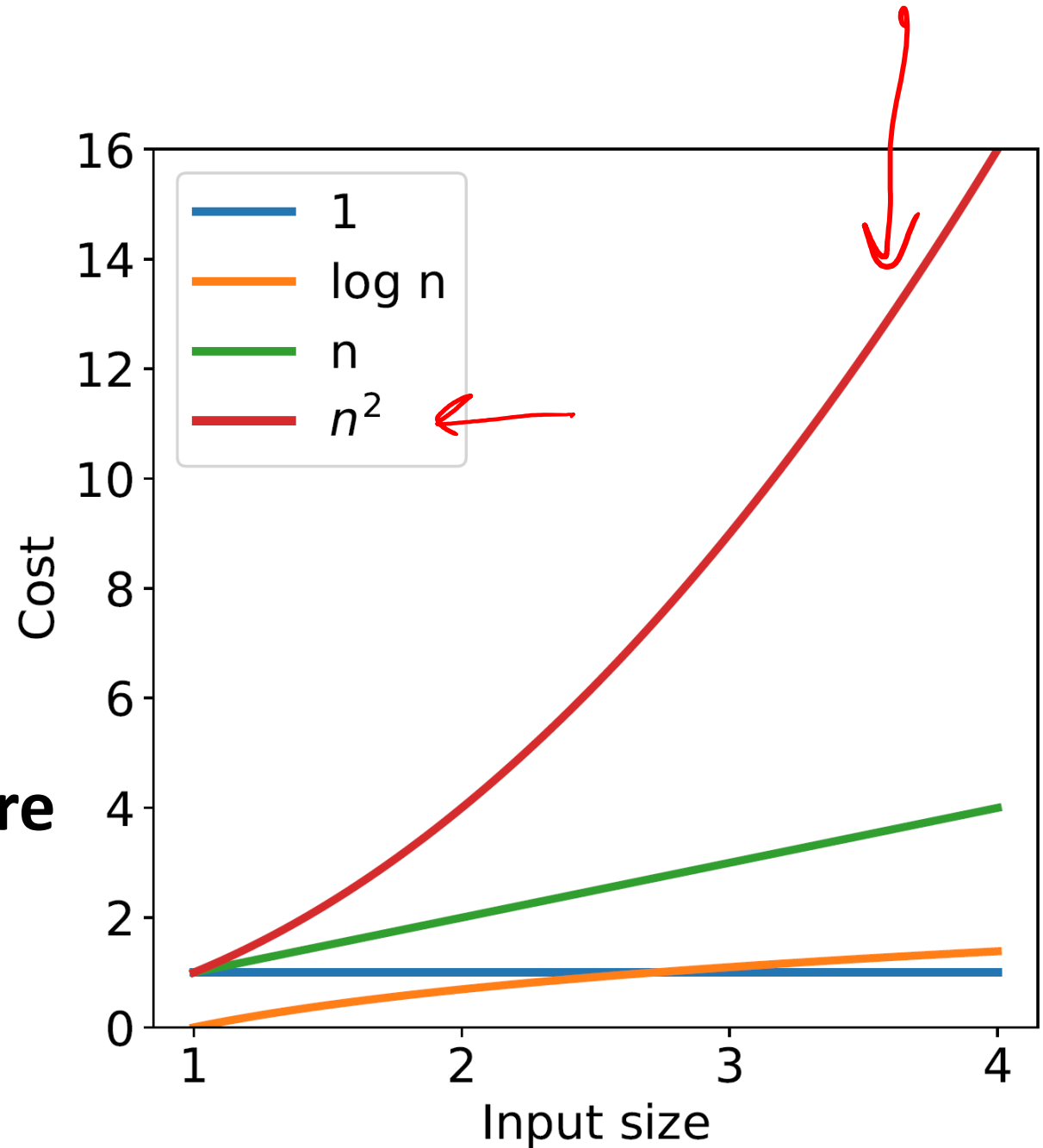
$$R(n) \leq cg(n)$$

for all  $n \geq n_0$

**e.g. The set of all functions R where**  
there exists  $n_0$  and  $c$ , s.t.

$$R(n) \leq cn^2$$

for all  $n \geq n_0$



# A set of better functions

## Definition of Big O of g

The set of all functions R where there exists  $n_0$  and  $c$ , s.t.

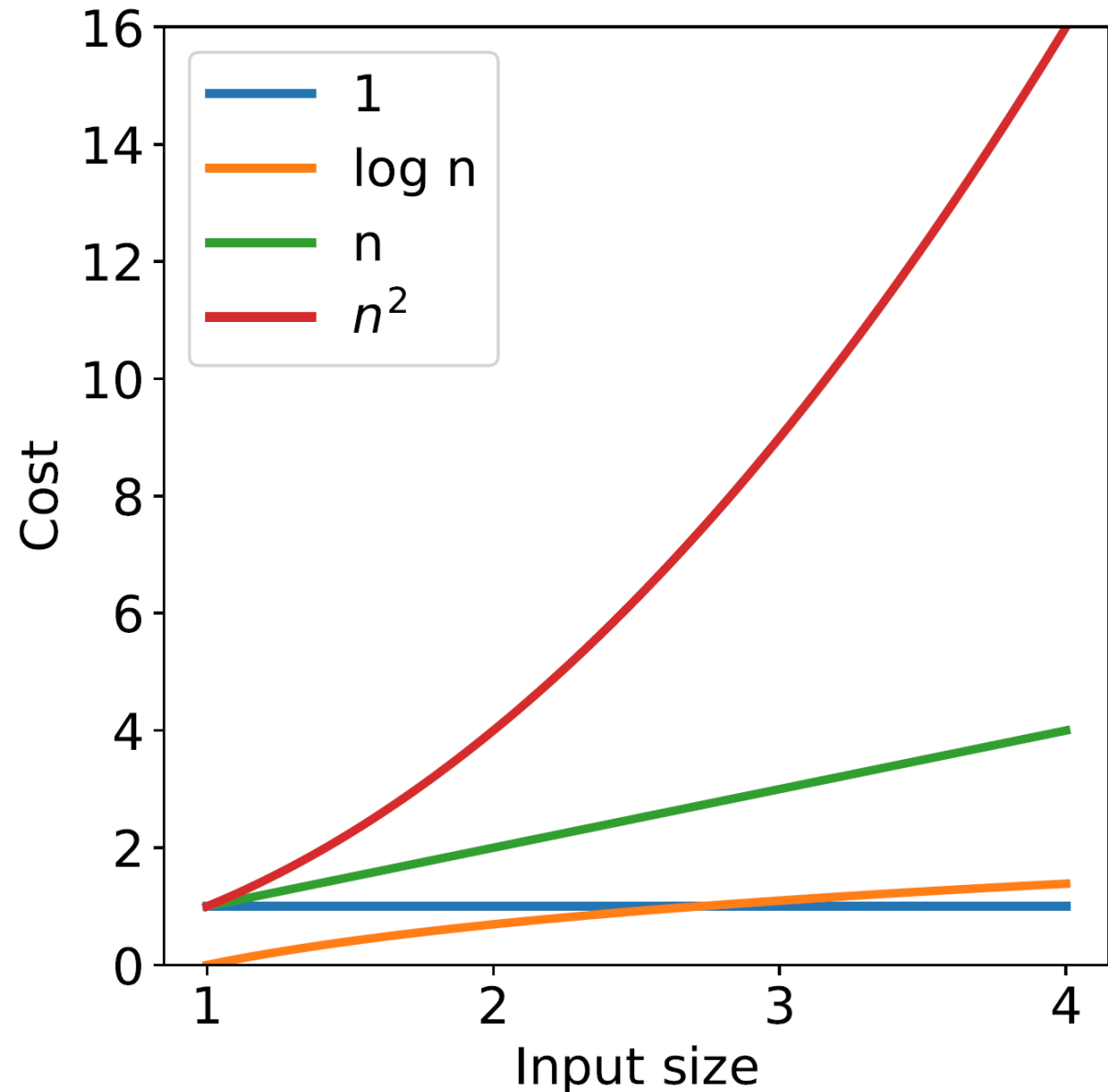
$$R(n) \leq cg(n)$$

for all  $n \geq n_0$

e.g.  $O(n^2)$  is a set that includes:

$$R_1(n) = 1 \quad R_3(n) = n$$

$$R_2(n) = \log n \quad R_4(n) = n^2$$





# A set of better functions

## Definition of Big O of g

The set of all functions R where there exists  $n_0$  and  $c$ , s.t.

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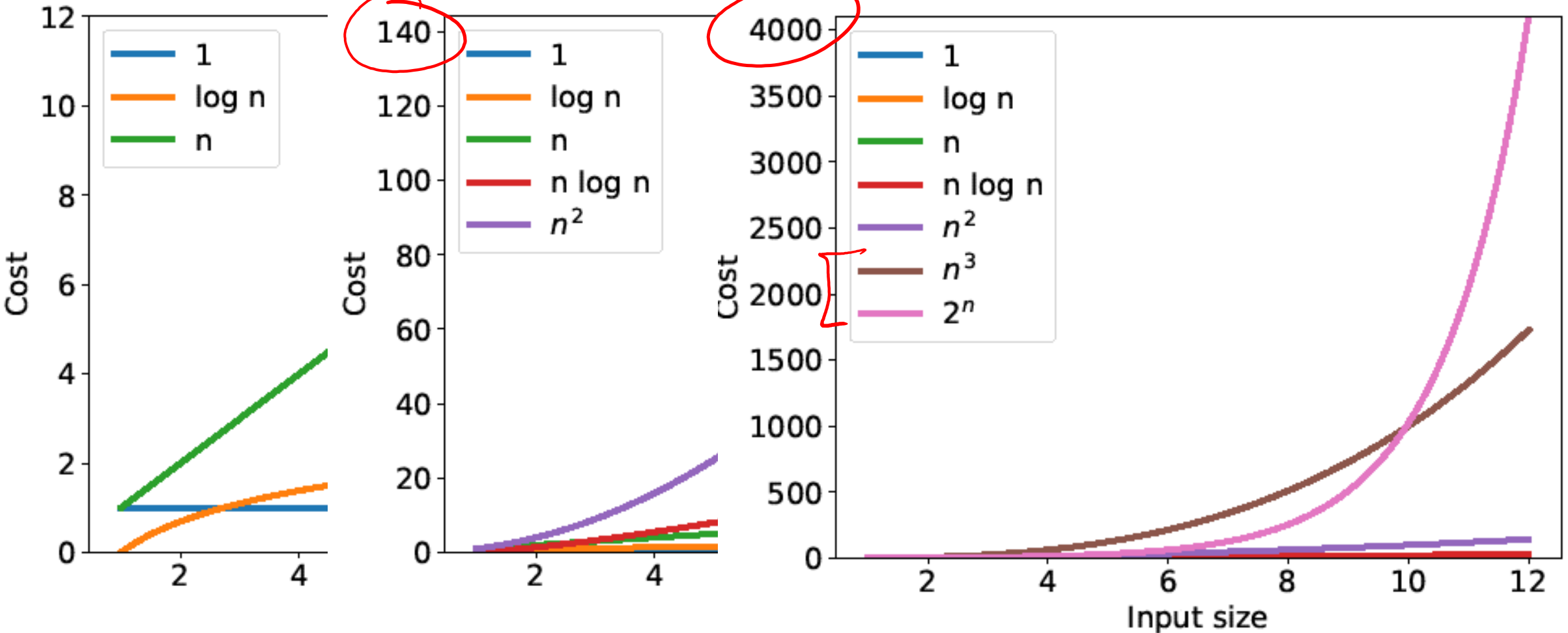
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e.g.  $O(n^2)$  is a set that includes:

$$\begin{array}{ll} R_1(n) = 1 & R_3(n) = n \\ R_2(n) = \log n & R_4(n) = n^2 \end{array}$$

$$\begin{array}{l} R_1 \in O(n^2) \\ R_2 \in O(n^2) \\ R_3 \in O(n^2) \\ R_4 \in O(n^2) \end{array}$$

# Complexity Classes



# Complexity Classes

<b>Complexity class</b>	<b>Conventional name</b>
$O(1)$	Constant
$O(\log n)$	Logarithmic
$O(n)$	Linear
$O(n \log n)$	" $n \log n$ "
$O(n^2)$	Quadratic
$O(n^3)$	Cubic
$O(2^n)$	Exponential

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

# Complexity Classes

Can determine whether or not a problem can be solved at all!

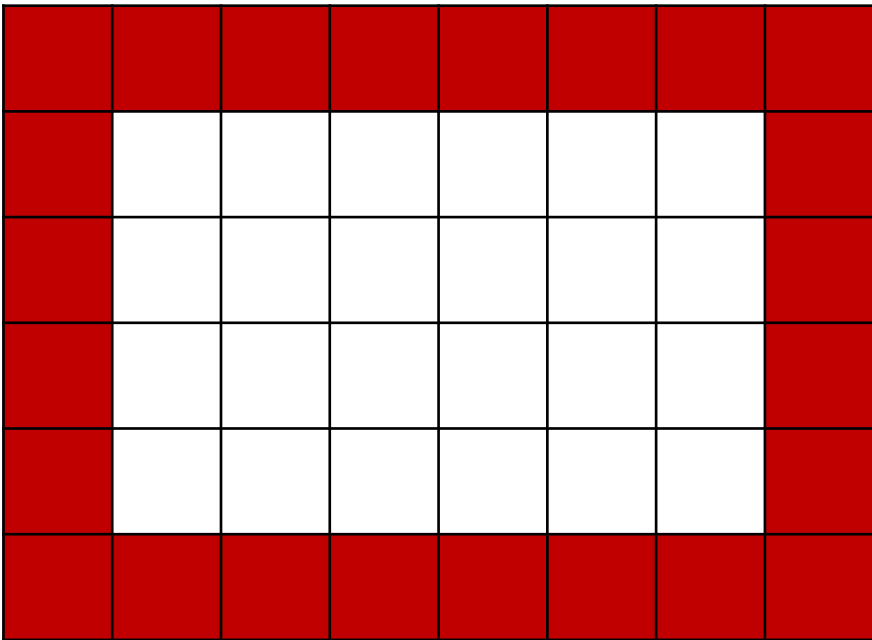
	$n$	$n \log_2 n$	$n^2$	$n^3$	$1.5^n$	$2^n$	$n!$
$n = 10$	< 1 sec	< 1 sec	< 1 sec	< 1 sec	< 1 sec	< 1 sec	4 sec
$n = 30$	< 1 sec	< 1 sec	< 1 sec	< 1 sec	< 1 sec	18 min	$10^{25}$ years
$n = 50$	< 1 sec	< 1 sec	< 1 sec	< 1 sec	11 min	36 years	very long
$n = 100$	< 1 sec	< 1 sec	< 1 sec	1 sec	12,892 years	$10^{17}$ years	very long
$n = 1,000$	< 1 sec	< 1 sec	1 sec	18 min	very long	very long	very long
$n = 10,000$	< 1 sec	< 1 sec	2 min	12 days	very long	very long	very long
$n = 100,000$	< 1 sec	2 sec	3 hours	32 years	very long	very long	very long
$n = 1,000,000$	1 sec	20 sec	12 days	31,710 years	very long	very long	very long

**Table 2.1** The running times (rounded up) of different algorithms on inputs of increasing size, for a processor performing a million high-level instructions per second. In cases where the running time exceeds  $10^{25}$  years, we simply record the algorithm as taking a very long time.

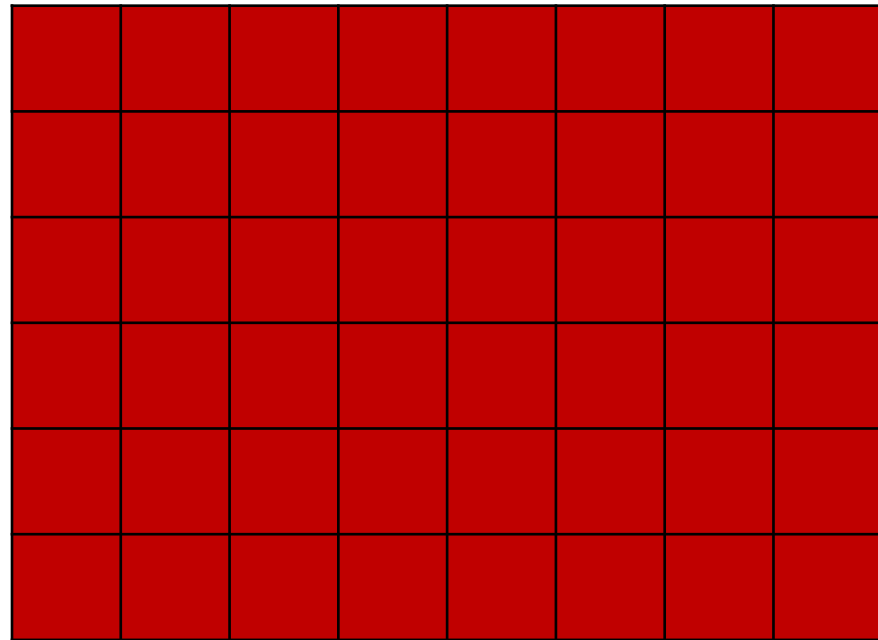
# Input

Nothing special about variable  $n$ , we could have input variables  $w$  and  $h$

$$O(w + h)$$



$$O(wh)$$



# Plan

## Computational Complexity

- ✓ Counting operations
- ✓ Big-O
- ✓ Complexity classes
  - Proving a Big-O relationship holds
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# Definition of Big O

## Definition of $R(n) \in O(f(n))$

Let  $R(n)$  be a function, be the running time of some program as a function of the input size  $n$ . We assume that:

1.  $n$  is an integer  $\geq 0$
  2.  $R(n) \geq 0$  for all  $n$
- $f(n)$  is a function defined on  $n$ . We say that " $R(n)$  is in  $O(f(n))$ " if there exists a constant  $c > 0$  and an integer  $n_0$  such that, for all integers  $n \geq n_0$ , we have  $R(n) \leq cf(n)$ .

## Witnesses

$n_0$  and  $c$  are called witnesses that  $R(n)$  is in  $O(f(n))$ . Finding such witnesses is a form of proof of  $R(n)$  being in  $O(f(n))$ .

# Proving a Big-O Relationship Holds

Template to prove  $R(n) \in O(f(n))$

1. State the witnesses  $n_0$  and  $c$  as specific constants, e.g.,  $n_0 = 32$  and  $c = 5$ .
2. By appropriate algebraic manipulation, show that if  $n \geq n_0$  then  $R(n) \leq cf(n)$ .



# Proving a Big-O Relationship Holds

Example: Prove  $(n + 1)^2 \in O(n^2)$

Suppose  $R(0) = 1$ ,  $R(1) = 4$ ,  $R(2) = 9$ , and in general  $R(n) = (n + 1)^2$ .

$$\begin{aligned} n_0 = 1 \quad (n+1)^2 &= n^2 + 2n + 1 \leq n^2 + \underline{2n^2} + 1 \\ n \leq n^2 &\leq n^2 + 2n^2 + n^2 \\ 1 \leq n^2 &\leq 4n^2 \\ &\leq cn^2 \quad c=4 \\ &\quad n_0=1 \end{aligned}$$

# Proving a Big-O Relationship Holds

Example: Prove  $(n + 1)^2 \in O(n^2)$

Suppose  $R(0) = 1$ ,  $R(1) = 4$ ,  $R(2) = 9$ , and in general  $R(n) = (n + 1)^2$ .

We can say that  $R(n) \in O(n^2)$ , by choosing witnesses  $n_0 = 1$  and  $c = 4$ :

- Expand  $(n + 1)^2 = n^2 + 2n + 1$
- if  $n \geq 1$ , we know that  $n \leq n^2$  and  $1 \leq n^2$
- Thus  $n^2 + 2n + 1 \leq n^2 + 2n^2 + n^2 = 4n^2$ .

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## Choosing witnesses

We could have also picked  $n_0 = 3$  and  $c = 2$ .

However, we *can't* pick  $n_0 = 0$  with any  $c$  (why?). But that doesn't matter, because we only need to find one pair of witnesses  $n_0$  and  $c$ .

# Proving a Big-O Relationship Holds

Example: Prove  $(n + 1)^2 \in O(n^2)$

But  $(n + 1)^2$  is bigger than  $n^2$ !!!

It may seem odd that  $(n + 1)^2 \in O(n^2)$  even though  $(n + 1)^2 > n^2$ .

But being in  $O(f(n))$  does not mean “less than”  $f(n)$ .

In fact,  $(n + 1)^2$  is also in big-O of any fraction of  $n^2$ , for example:

$$(n + 1)^2 \in O(n^2/100) \text{ with witnesses } n_0 = 1 \text{ and } c = 400$$

# Proving a Big-O Relationship Holds

## Quick tips and important points

### Constant factors don't matter

For any positive constant  $d$  and any function that is  $O(f(n))$  is also  $O(df(n))$ . (Choose  $n_0 = 0$  and  $c = 1/d$ .)

### Low-order terms don't matter

Consider a polynomial  $R(n) = a_k n^k + a_{k-1} n^{k-1} + \dots + a_2 n^2 + a_1 n + a_0$  where the leading coefficient,  $a_k$ , is positive. We can throw away all terms except the term with the highest exponent,  $k$ , and we can ignore  $a_k$  (a constant), replacing it by 1.  $R(n) \in O(n^k)$ . (To prove, choose  $n_0 = 1$ , and  $c = \sum_{i \in \{1, \dots, k\} \mid a^i > 0} a^i$ .)

# Poll 1

Which of the following functions are in  $O(n^2)$ ?

Select all that apply.

1. 1

2.  $n$

3.  $n \log(n) \leq n \cdot n$

4.  $n^2$

5.  ~~$4n^2$~~

6.  ~~$4n^2 + n \log(n)$~~

7.  ~~$4n^2 + n \log(n) + n$~~

~~8.  $n^3$~~

~~9.  $n^3 + n^2$~~  ←

# Proving a Big-O Relationship Does Not Hold

Template to disprove  $R(n) \in O(f(n))$

1. Assume that witnesses  $n_0$  and  $c$  exist
2. Derive a contradiction

# Proving a Big-O Relationship Does Not Hold

Example: Prove that  $n^2$  is not in  $O(n)$

- Assume  $n^2 \in O(n)$   $cf(n)$
- Then there exist  $n_0$  and  $c$  such that  $n^2 \leq cn$  for all  $n \geq n_0$ .

Def

$$n_a > c$$

$$n_a > n_0$$

$$n \leq c$$

$$n_a \leq c \quad \text{Contradiction}$$



# Proving a Big-O Relationship Does Not Hold

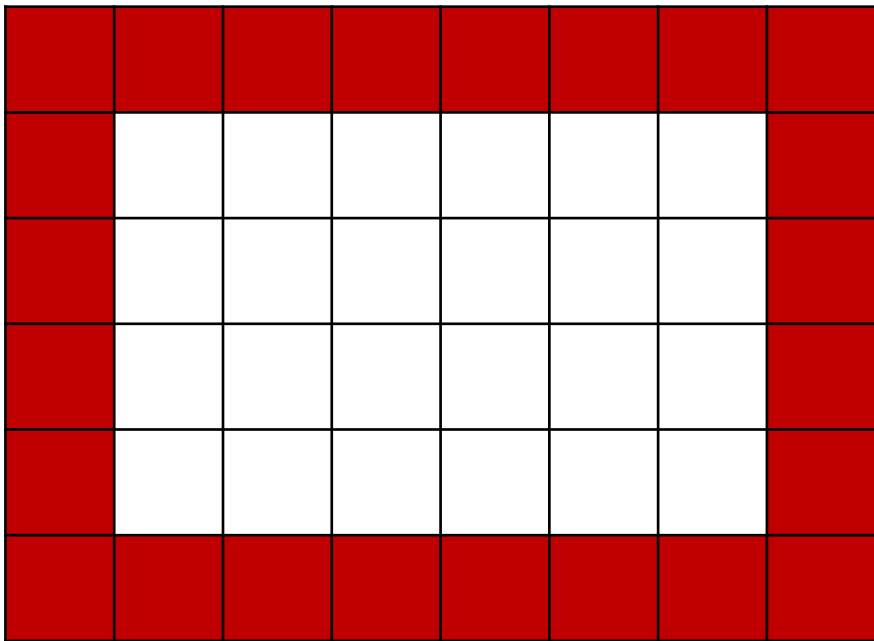
Example: Prove that  $n^2$  is not in  $O(n)$

- Assume  $n^2 \in O(n)$
- Then there exist  $n_0$  and  $c$  such that  $n^2 \leq cn$  for all  $n \geq n_0$ .
- Let  $n_a$  be a value  $n_a > \max(n_0, c) + 1$
- Then  $(n_a)^2 \leq cn_a$
- Dividing both sides by  $n_a$ , we have  $n_a \leq c$ . **Contradiction!**
- Therefore  $n^2 \notin O(n)$

# Input

Nothing special about variable  $n$ , we could have input variables  $w$  and  $h$ .  
However, be careful to pay attention to which inputs we care about analyzing.

$$O(w + h)$$



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However, be careful to pay attention to which inputs we care about analyzing.

Nearest neighbor example

$N$ : data points

$$\vec{x}^{(i)} \in \mathbb{R}^M$$

$$\vec{x}_{\text{new}} \in \mathbb{R}^M$$

find  $i$  s.t.  $\min_i \|\vec{x}_{\text{new}} - \vec{x}^{(i)}\|_2$

$$O(MN) \quad O(N) \quad O(M)$$

# Exercise



What is the computation complexity of matrix multiplication of two  $N \times N$  matrices? Give the tightest complexity in its simplest form.

$$A \in \mathbb{R}^{N \times N}, B \in \mathbb{R}^{N \times N}, C \in \mathbb{R}^{N \times N}$$

$$C = AB$$

$$O(n^3)$$

$$O(n^3 + n^3 - n^2)$$

$$\forall i \in \{1..N\}, \forall j \in \{1..N\}$$

$$C_{i,j} = \sum_{k=1}^N A_{ik} B_{kj}$$

$$C = \text{zeros}(N, N)$$

$$O(n^4)$$

for  $i$  in  $1..N$

$$O(2^n)$$

for  $j$  in  $1..N$

$$O(n^n)$$

for  $k$  in  $1..N$

$$C[i,j] += A[i,k] * B[k,j]$$

## Exercise

Prove that  $n^3$  is in  $O(2^n)$

Prove that  $n^2 + 100$  is in  $O(n^4)$

Prove that  $\frac{1}{4}n^2 + n \log(n) + n$  is in  $O(n^2)$