

# Asymptotic Analysis

# Main idea:

Focus on how the runtime **scales** with  $n$  (the input size).

Some examples...

(Heuristically: only pay attention to the largest function of  $n$  that appears.)

| Number of operations                     |                         | Asymptotic Running Time                  |                         |
|--|-------------------------|--|-------------------------|
| Number of operations                     | Asymptotic Running Time | Number of operations                     | Asymptotic Running Time |
| $\frac{1}{10} \cdot n^2 + 100$           | $O(n^2)$                | $\frac{1}{10} \cdot n^2 + 100$           | $O(n^2)$                |
| $0.063 \cdot n^2 - .5n + 12.7$           | $O(n^2)$                | $0.063 \cdot n^2 - .5n + 12.7$           | $O(n^2)$                |
| $100 \cdot n^{1.5} - 10^{10000}\sqrt{n}$ | $O(n^{1.5})$            | $100 \cdot n^{1.5} - 10^{10000}\sqrt{n}$ | $O(n^{1.5})$            |
| $11 \cdot n \log(n) + 1$                 | $O(n \log(n))$          | $11 \cdot n \log(n) + 1$                 | $O(n \log(n))$          |
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# Why is this a good idea?

- Suppose the running time of an algorithm is:

$$T(n) = 10n^2 + 3n + 7 \text{ ms}$$

This constant factor of 10  
depends a lot on my  
computing platform...

These lower-order  
terms don't really  
matter as  $n$  gets large.

We're just left with the  $n^2$  term!  
That's what's meaningful.

# Pros and Cons of Asymptotic Analysis

## Pros:

- Abstracts away from hardware- and language-specific issues.
- Makes algorithm analysis much more tractable.
- Allows us to meaningfully compare how algorithms will perform on large inputs.

## Cons:

- Only makes sense if  $n$  is large (compared to the constant factors).

$1000000000n$   
is “better” than  $n^2$  ?!?!

pronounced “big-oh of ...” or sometimes “oh of ...”

# Informal definition for $O(g(n))$

- A function grows *no faster* than a certain rate

# Formal definition for $O(g(n))$

- For a given function of  $n$ ,  $g(n)$
- $O(g(n))$  is the *set of functions* such that,

$$O(g(n))$$

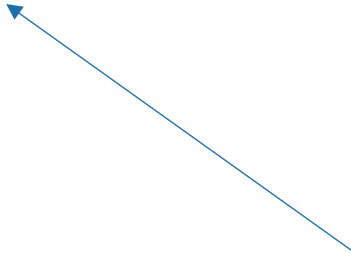
$$= \{$$

$f(n)$ : there exist positive constants  $c$  and  $n_0$  such that  
 $0 \leq f(n) \leq cg(n)$  for all  $n \geq n_0$

$$\}$$

# Formal definition for $O(g(n))$

- Let  $T(n)$ ,  $g(n)$  be functions of positive integers.
  - Think of  $T(n)$  as a runtime: positive and increasing in  $n$ .
- We say “ $T(n)$  is  $O(g(n))$ ” if:
  - for all large enough  $n$ ,
  - $T(n)$  is at most some constant multiple of  $g(n)$ .

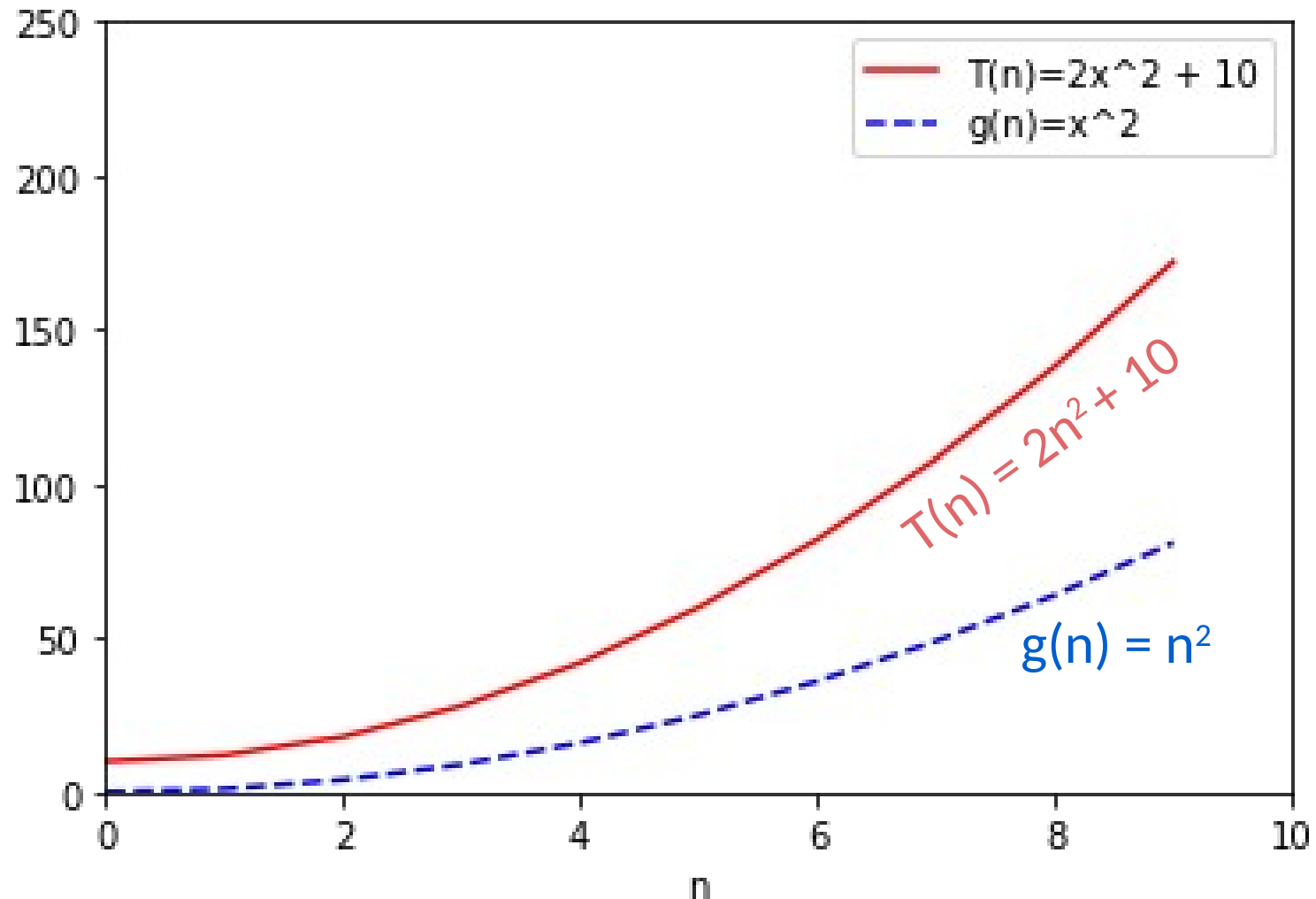


Here, “constant” means “some number that doesn’t depend on  $n$ .”

# Example

$$2n^2 + 10 = O(n^2)$$

for large enough  $n$ ,  
 $T(n)$  is at most some constant  
multiple of  $g(n)$ .

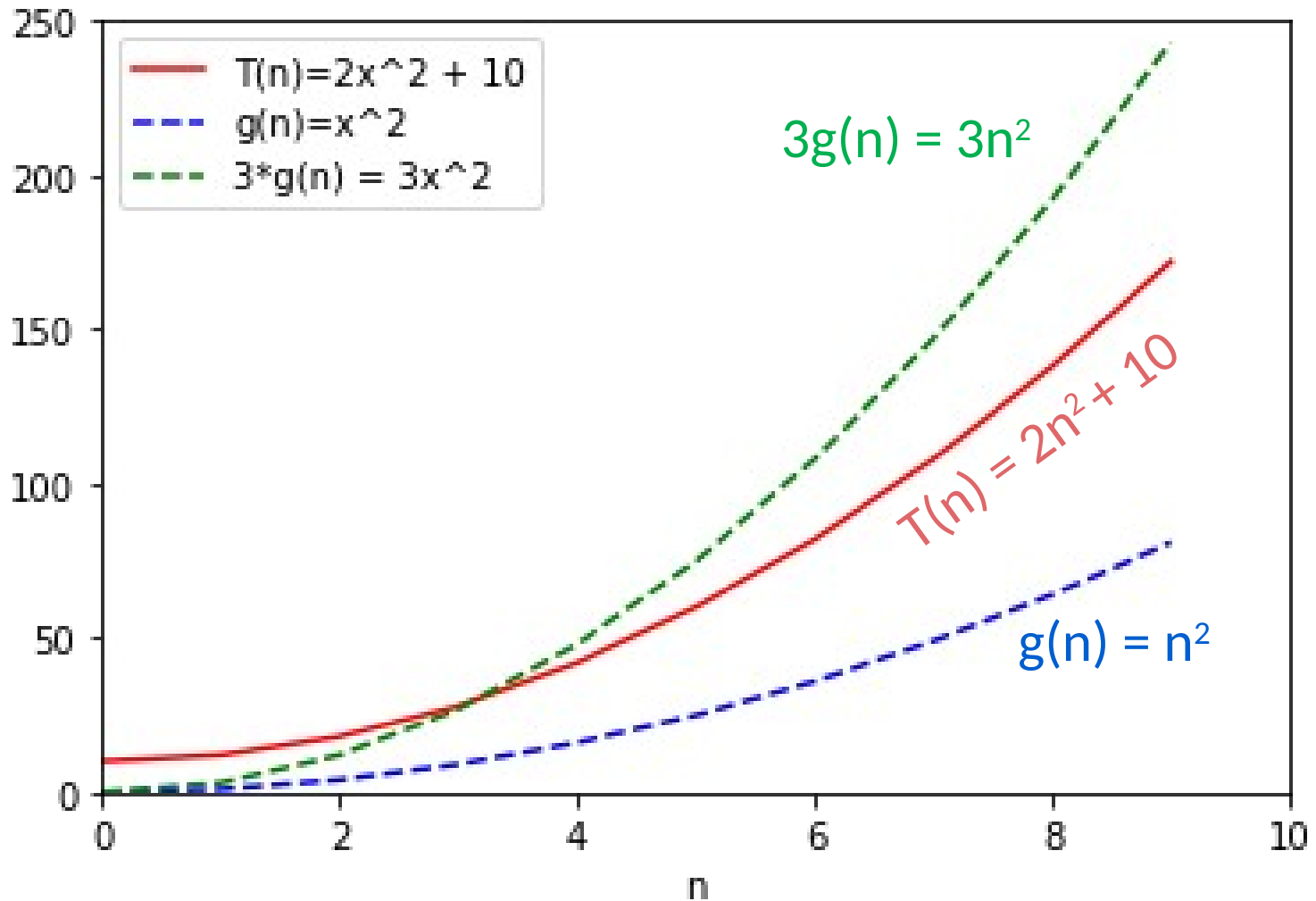




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$$2n^2 + 10 = O(n^2)$$

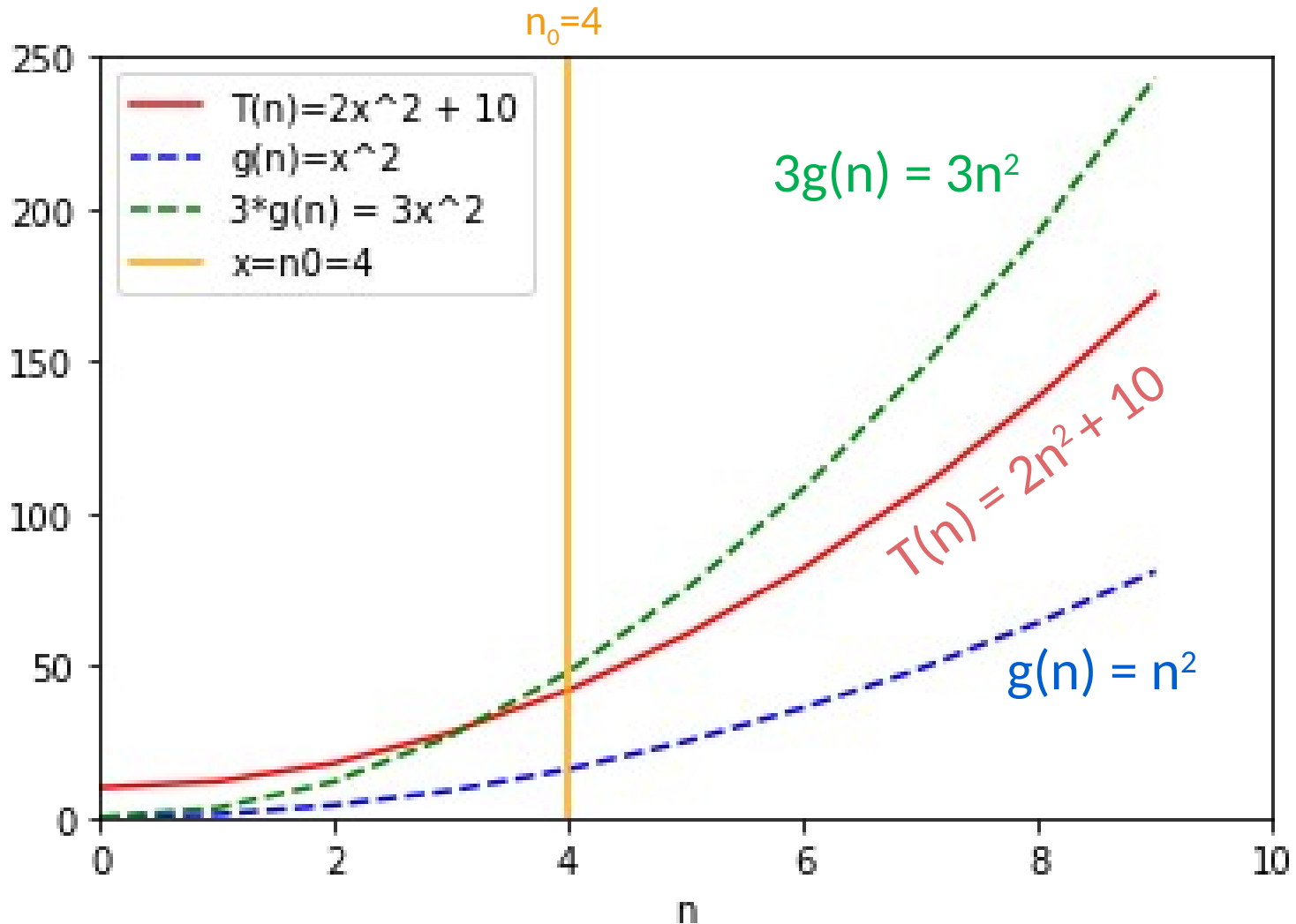
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# Example

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# Formal definition of $O(g(n))$

- Let  $T(n)$ ,  $g(n)$  be functions of positive integers.
  - Think of  $T(n)$  as a runtime: positive and increasing in  $n$ .
- Formally,

$$T(n) = O(g(n))$$

“If and only if”



“For all”

$$\exists c, n_0 > 0, \text{ s. t. } \forall n \geq n_0,$$

“There exists”

$$T(n) \leq c \cdot g(n)$$

“such that”

# Example

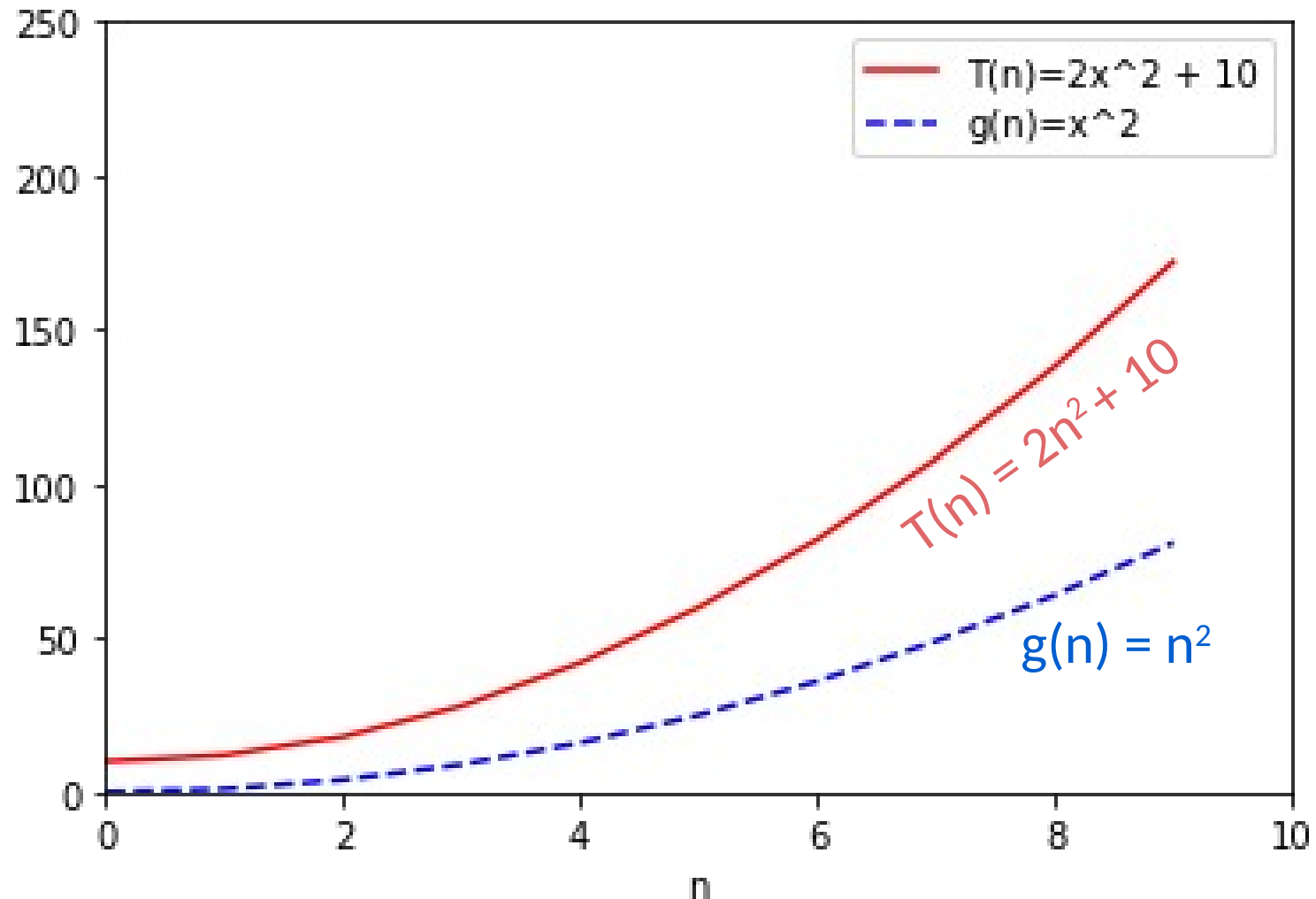
$$2n^2 + 10 = O(n^2)$$

$$T(n) = O(g(n))$$

$$\Leftrightarrow$$

$$\exists c, n_0 > 0 \text{ s.t. } \forall n \geq n_0,$$

$$T(n) \leq c \cdot g(n)$$



# Example

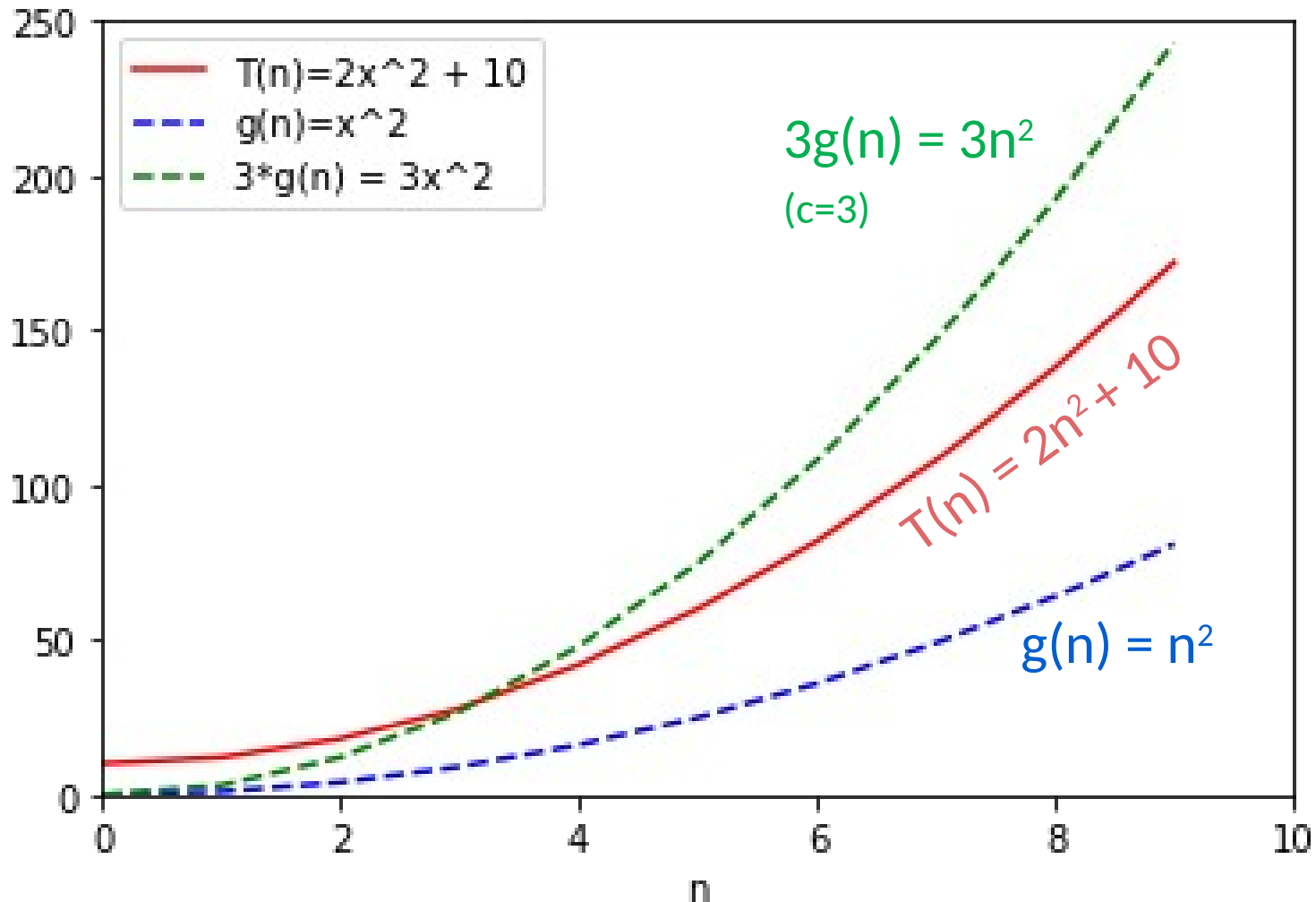
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# Example

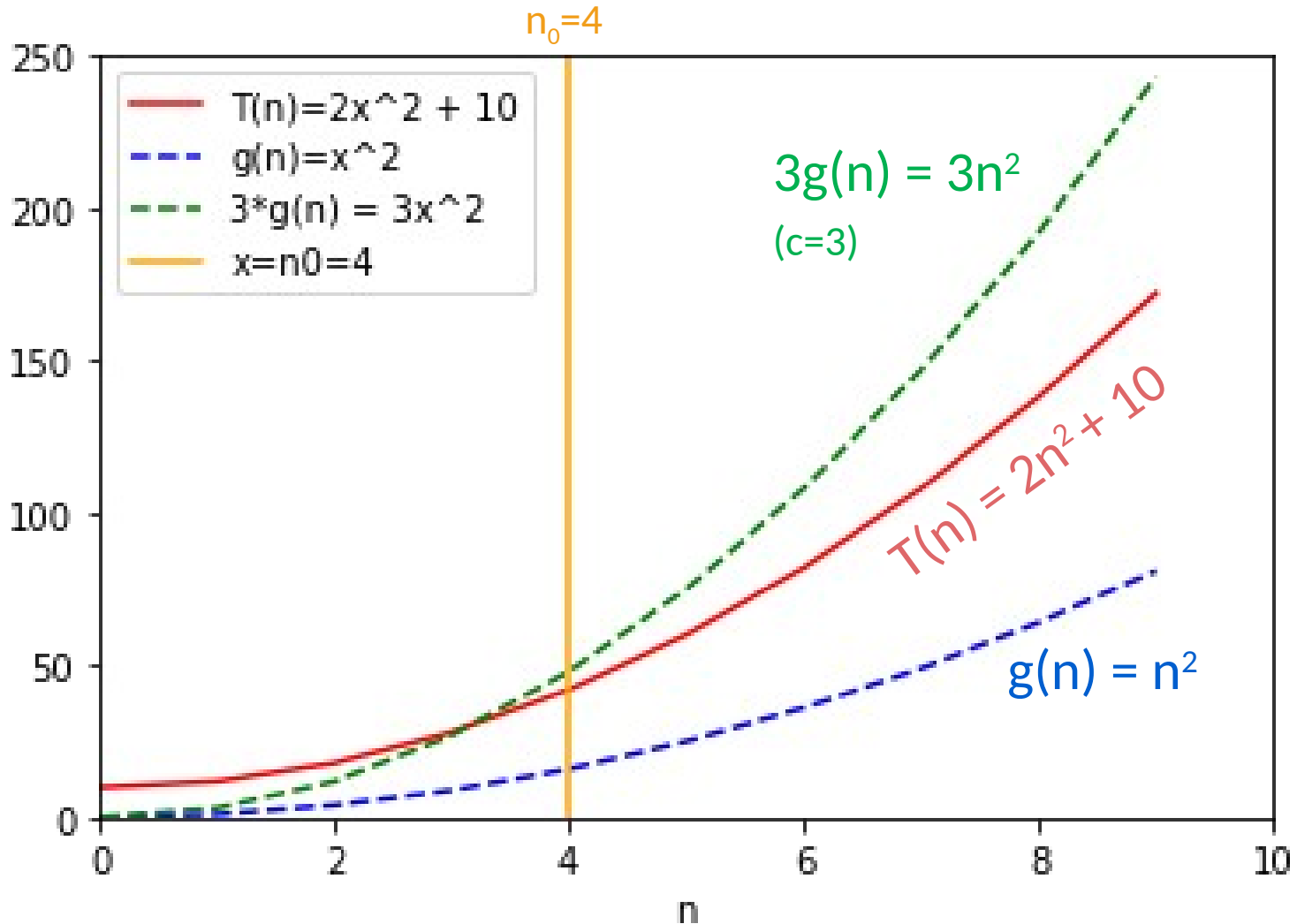
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# Example

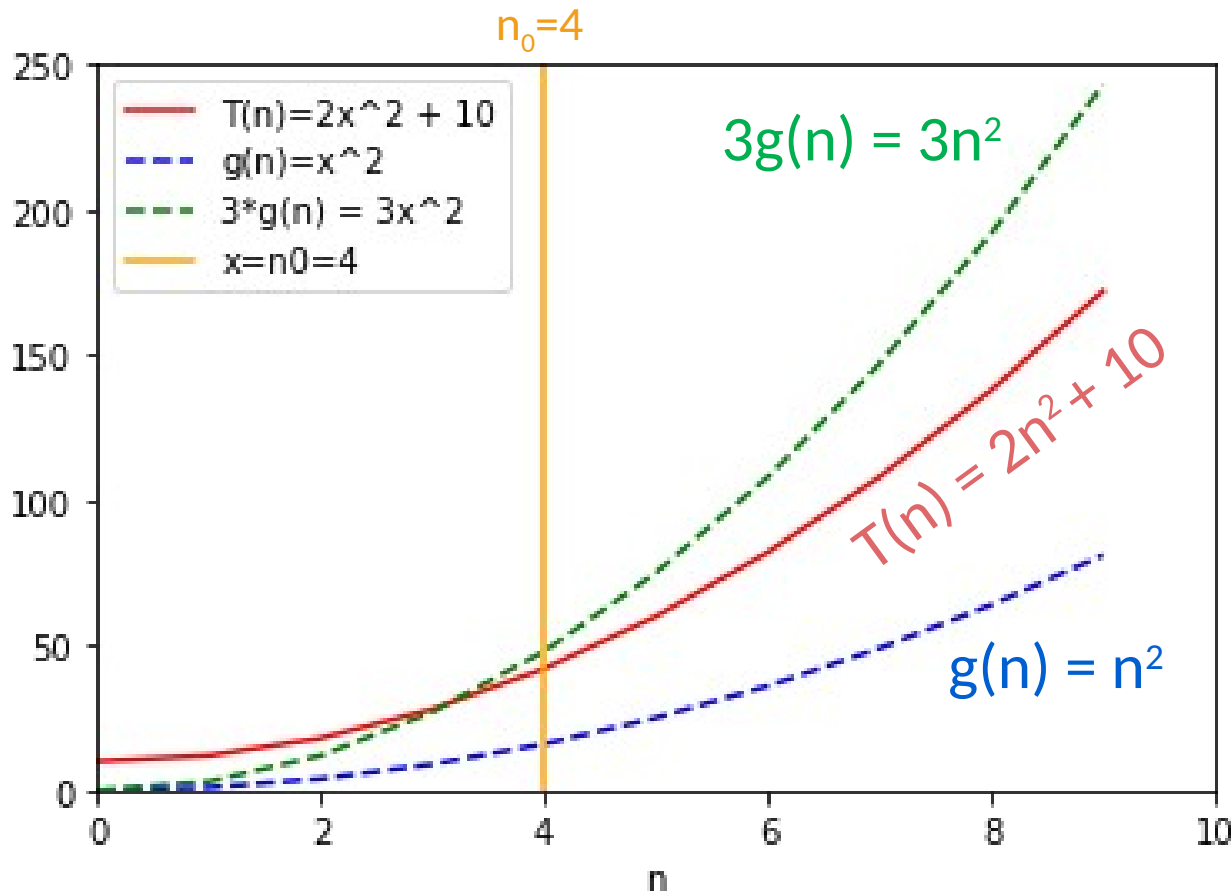
$$2n^2 + 10 = O(n^2)$$

$$T(n) = O(g(n))$$

$$\Leftrightarrow$$

$$\exists c, n_0 > 0 \text{ s.t. } \forall n \geq n_0,$$

$$T(n) \leq c \cdot g(n)$$



Formally:

- Choose  $c = 3$
- Choose  $n_0 = 4$
- Then:

$$\forall n \geq 4,$$

$$2n^2 + 10 \leq 3 \cdot n^2$$

# Same example

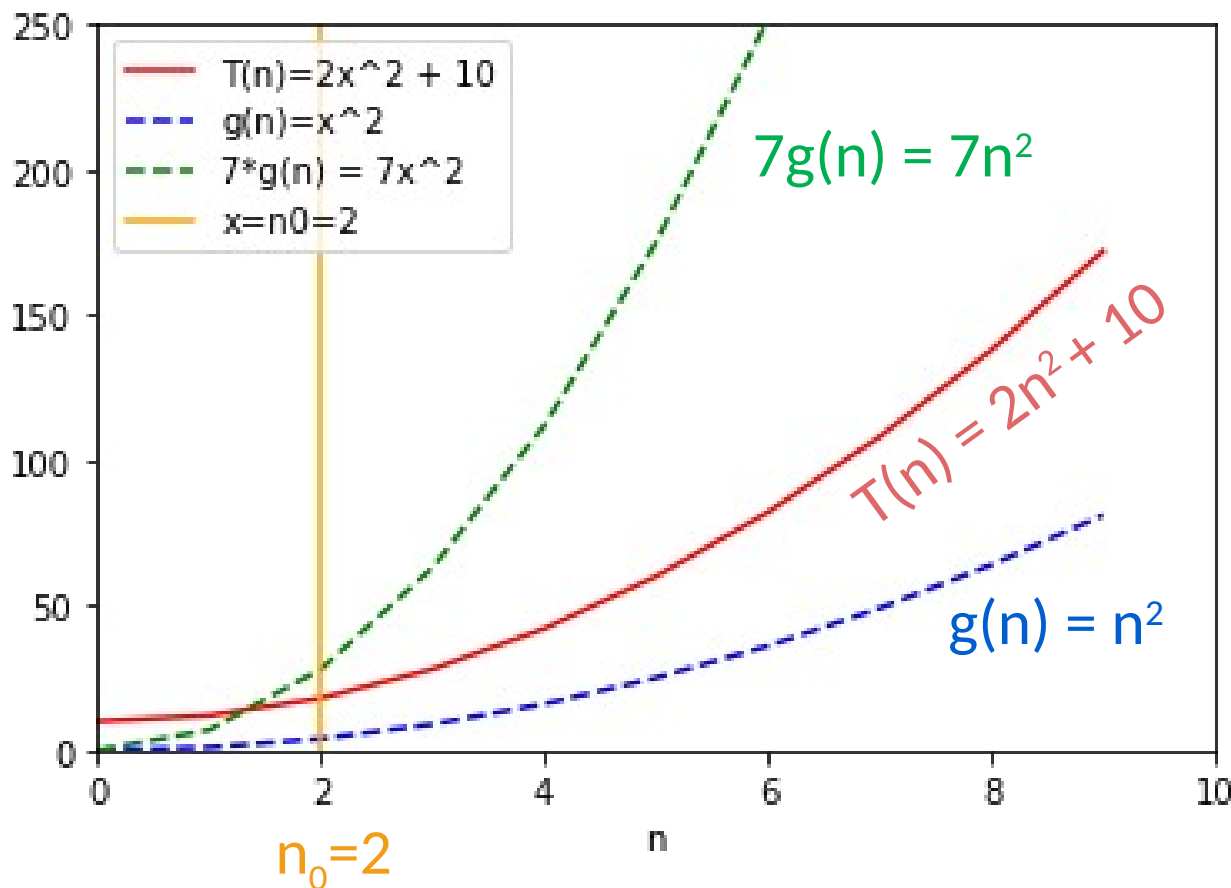
$2n^2 + 10 = O(n^2)$

$$T(n) = O(g(n))$$

$\Leftrightarrow$

$$\exists c, n_0 > 0 \text{ s.t. } \forall n \geq n_0,$$

$$T(n) \leq c \cdot g(n)$$



Formally:

- Choose  $c = 7$
- Choose  $n_0 = 2$
- Then:

$$\forall n \geq 2,$$

$$2n^2 + 10 \leq 7 \cdot n^2$$



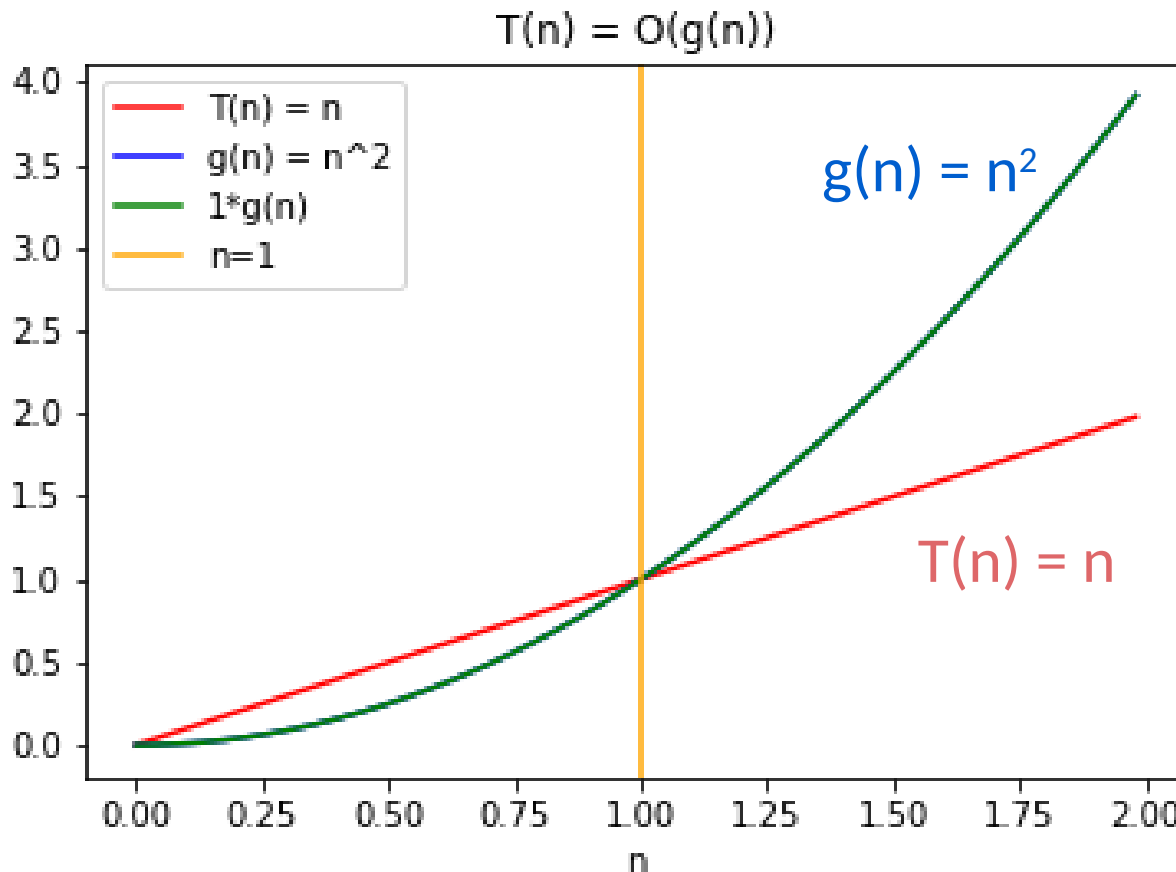
$O(g(n))$  is an upper bound  
 $n = O(n^2)$

$$T(n) = O(g(n))$$

$$\Leftrightarrow$$

$$\exists c, n_0 > 0 \text{ s.t. } \forall n \geq n_0,$$

$$T(n) \leq c \cdot g(n)$$



- Choose  $c = 1$
- Choose  $n_0 = 1$
- Then

$$\forall n \geq 1,$$

$$n \leq n^2$$

# Informal definition for $\Omega(g(n))$

- A function grows *at least as fast as* a certain rate

# Formal definition for $\Omega(g(n))$

- For a given function of  $n$ ,  $g(n)$
- $\Omega(g(n))$  is the *set of functions* such that,

$$\begin{aligned} &\Omega(g(n)) \\ = &\{ \\ &\quad f(n) : \text{there exist positive} \\ &\quad \text{constants } c \text{ and } n_0 \text{ such that} \\ &\quad 0 \leq cg(n) \leq f(n) \text{ for all } n \geq n_0 \\ &\} \end{aligned}$$

# $\Omega(g(n))$ means a lower bound

- We say “ $T(n)$  is  $\Omega(g(n))$ ” if, for large enough  $n$ ,  $T(n)$  is at least as big as a constant multiple of  $g(n)$ .

- Formally,

$$T(n) = \Omega(g(n))$$

$$\Leftrightarrow$$

$$\exists c, n_0 > 0 \text{ s.t. } \forall n \geq n_0,$$

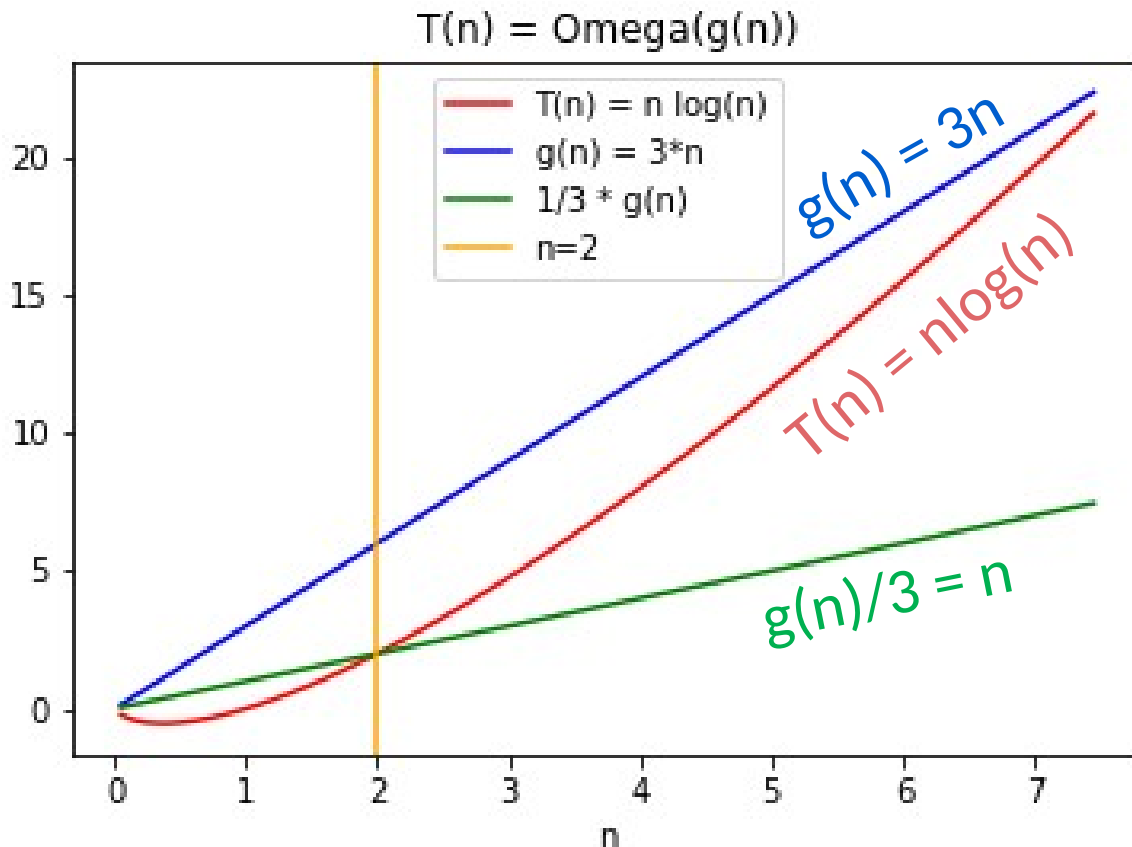
$$c \cdot g(n) \leq T(n)$$

Switched these!!

# Example

## $n \log_2(n) = \Omega(3n)$

$$T(n) = \Omega(g(n)) \Leftrightarrow \exists c > 0, n_0 \text{ s.t. } \forall n \geq n_0, c \cdot g(n) \leq T(n)$$



- Choose  $c = 1/3$
- Choose  $n_0 = 2$
- Then

$$\forall n \geq 2,$$

$$\frac{3n}{3} \leq n \log_2(n)$$

# Informal definition for $\Theta(g(n))$

- A function grows *precisely at* a certain rate

$\Theta(g(n))$  means both!

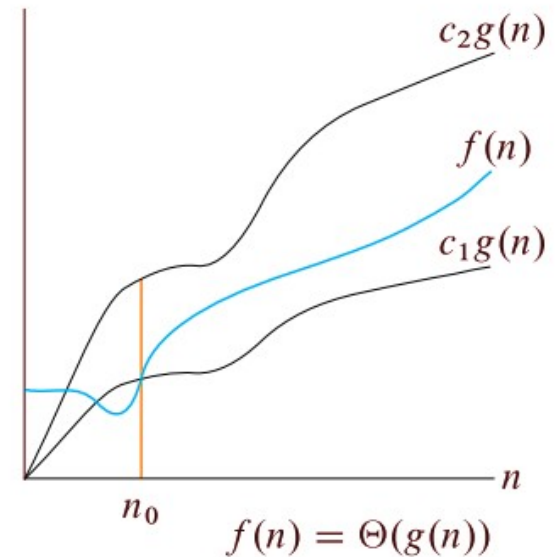
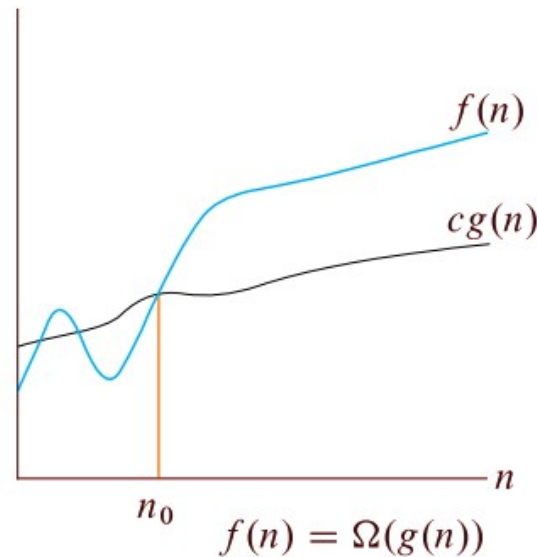
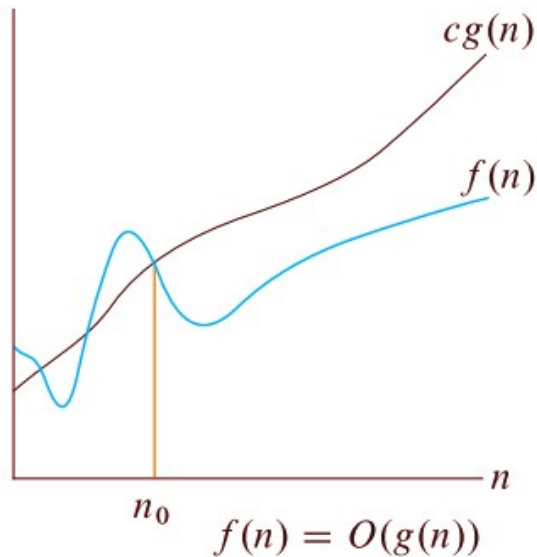
- We say “ $T(n)$  is  $\Theta(g(n))$ ” iff both:

$$T(n) = O(g(n))$$

and

$$T(n) = \Omega(g(n))$$

# Summary of Asymptotic Notations





# Non-Example:

$n^2$  is not  $O(n)$

$$\begin{aligned} T(n) = O(g(n)) \\ \Leftrightarrow \\ \exists c > 0, n_0 \text{ s.t. } \forall n \geq n_0, \\ T(n) \leq c \cdot g(n) \end{aligned}$$

- Proof by contradiction:
- Suppose that  $n^2 = O(n)$ .
- Then there is some positive  $c$  and  $n_0$  so that:
$$\forall n \geq n_0, \quad n^2 \leq c \cdot n$$
- Divide both sides by  $n$ :
$$\forall n \geq n_0, \quad n \leq c$$
- That's not true!!! What about  $n = n_0 + c + 1$ ?
  - Then  $n \geq n_0$ , but  $n > c$ .
- Contradiction!

# Take-away from examples

- To prove  $T(n) = O(g(n))$ , you have to come up with  $c$  and  $n_0$  so that the definition is satisfied.
- To prove  $T(n)$  is **NOT**  $O(g(n))$ , one way is **proof by contradiction**:
  - Suppose (to get a contradiction) that someone gives you a  $c$  and an  $n_0$  so that the definition *is* satisfied.
  - Show that this someone must be lying to you by deriving a contradiction.

# Formal definition of $o(g(n))$

- For a given function of  $n$ ,  $g(n)$
- $o(g(n))$  is the *set of functions* such that,

$$o(g(n))$$

$$= \{$$

$f(n)$ : there exist positive constants  $c$  and  $n_0$  such that  
 $0 \leq f(n) < cg(n)$  for all  $n \geq n_0$

$$\}$$

# Formal definition of $\omega(g(n))$

- For a given function of  $n$ ,  $g(n)$
- $\omega(g(n))$  is the *set of functions* such that,

$$\begin{aligned} &\omega(g(n)) \\ = &\{ \\ &\quad f(n): \text{there exist positive} \\ &\quad \text{constants } c \text{ and } n_0 \text{ such that} \\ &\quad 0 \leq cg(n) < f(n) \text{ for all } n \geq n_0 \\ &\} \end{aligned}$$

# Asymptotic Analysis

BUBBLE-SORT (A)

1 `n = length[A]`

2 `for i = 1 to n - 1`

$n-1$  iterations on the outer loop



3 `for j = i + 1 to n`

$n-i-1$  iterations on the inner loop

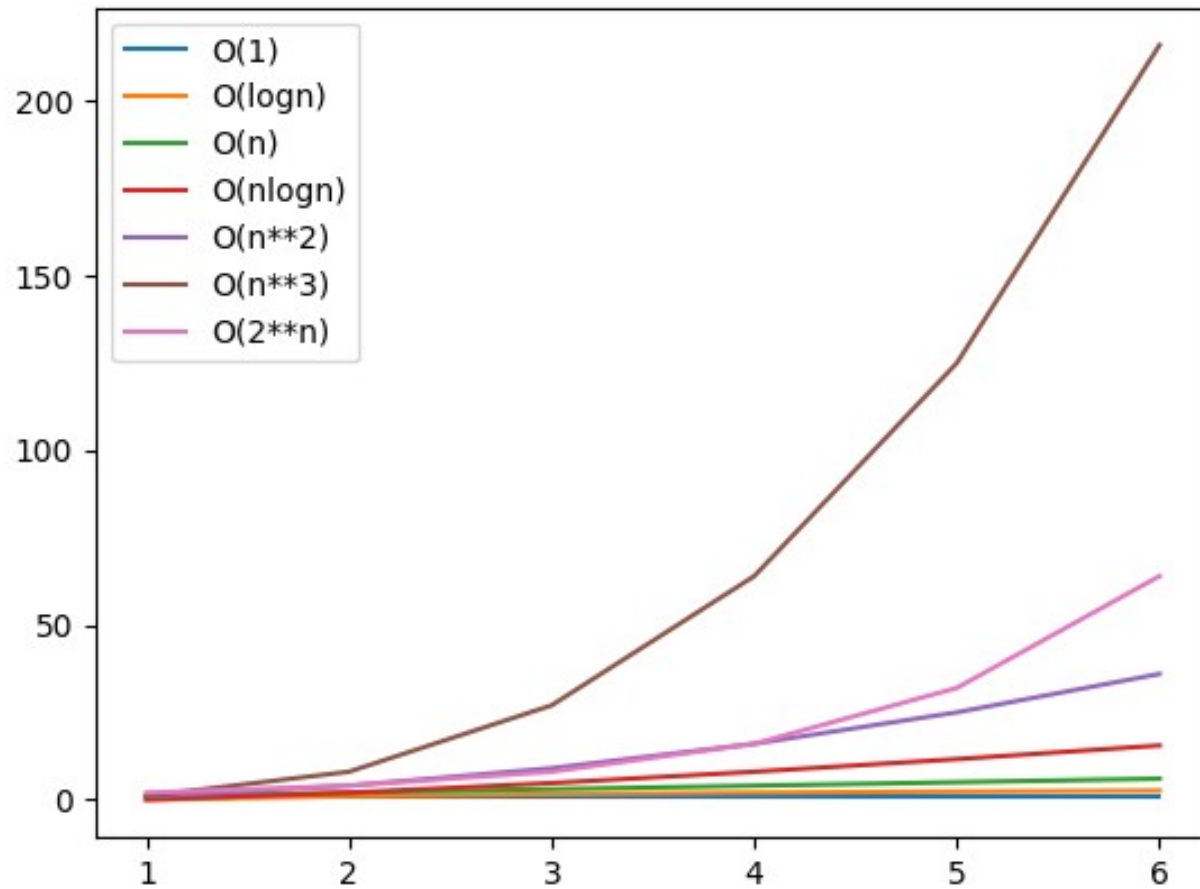


4 `if A[j] < A[j - 1]`

5 `exchange A[j] with A[j - 1]`

Runtime  $O(n^2)$

# Common Bounds



# Some Notations

ˆ Asymptotic notations are defined as sets.

- But we use,

$$f(n) = O(g(n)) \text{ instead of } f(n) \in O(g(n))$$

- We can also write,

$$T(n) = 2T\left(\frac{n}{2}\right) + \theta(n)$$

- Provide the simplest and most precise bounds possible

# Reference

- CLRS Chapter 3