

# Complexity

**Cost**

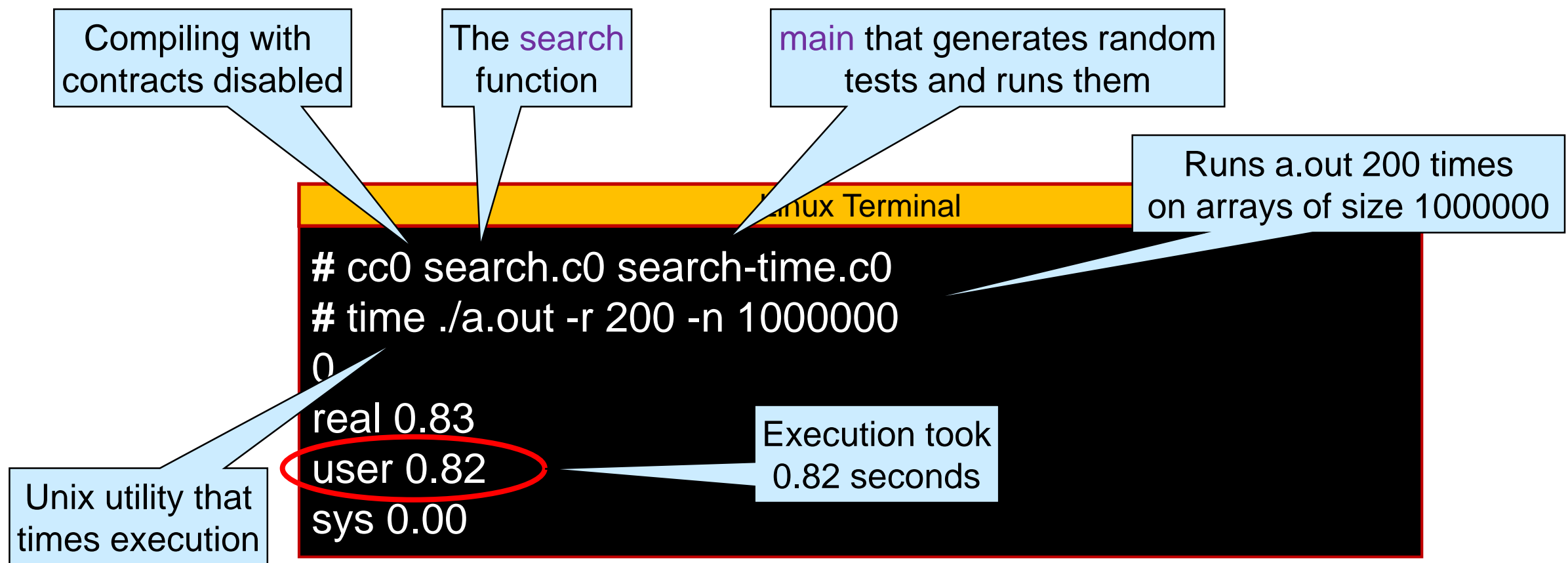
# Final Code for search

```
int search(int x, int[] A, int n)
//@requires n == \length(A);
/*@ensures (\result == -1 && !is_in(x, A, 0, n))
           || (0 <= \result && \result < n && A[\result] == x);
@*/
{
    for (int i = 0; i < n; i++)
        //@loop_invariant 0 <= i && i <= n;
        //@loop_invariant !is_in(x, A, 0, i);
        {
            if (A[i] == x) return i;
        }
    return -1;
}
```

- **How long** does it take to run?
  - *with contract-checking off*

# What do we mean by “How long”?

- First idea: wall-clock time
  - Time the code takes to run on a benchmark



# What do we mean by “How long”?

- Wall-clock time

- Gives different results depending on
  - what else is running on computer
  - what specific computer it is running on
- Is this a useful notion of “how long”?

```
Linux Terminal
# cc0 search.c0 search-time.c0
# time ./a.out -r 200 -n 1000000
0
real 0.83
user 0.82
sys 0.00
# time ./a.out -r 200 -n 1000000
0
real 0.91
user 0.90
sys 0.00
# time ./a.out -r 200 -n 1000000
0
real 0.81
user 0.80
sys 0.00
```

Different runs of the **same** code take **different** times!

# What do we mean by “How long”?

- *Wall-clock time*

- *Is this a useful notion of “how long”?*

- Time is *about double* when we double the length of the array

- not exactly double though

```
Linux Terminal
# cc0 search.c0 search-time.c0
# time ./a.out -r 200 -n 1000000
0
real 0.83
user 0.82
sys 0.00
# time ./a.out -r 200 -n 2000000
0
real 1.62
user 1.61
sys 0.00
# time ./a.out -r 200 -n 4000000
0
real 3.17
user 3.15
sys 0.01
```

# What do we mean by “How long”?

Can we do better than wall-clock time?

What are we looking for? A measure that is

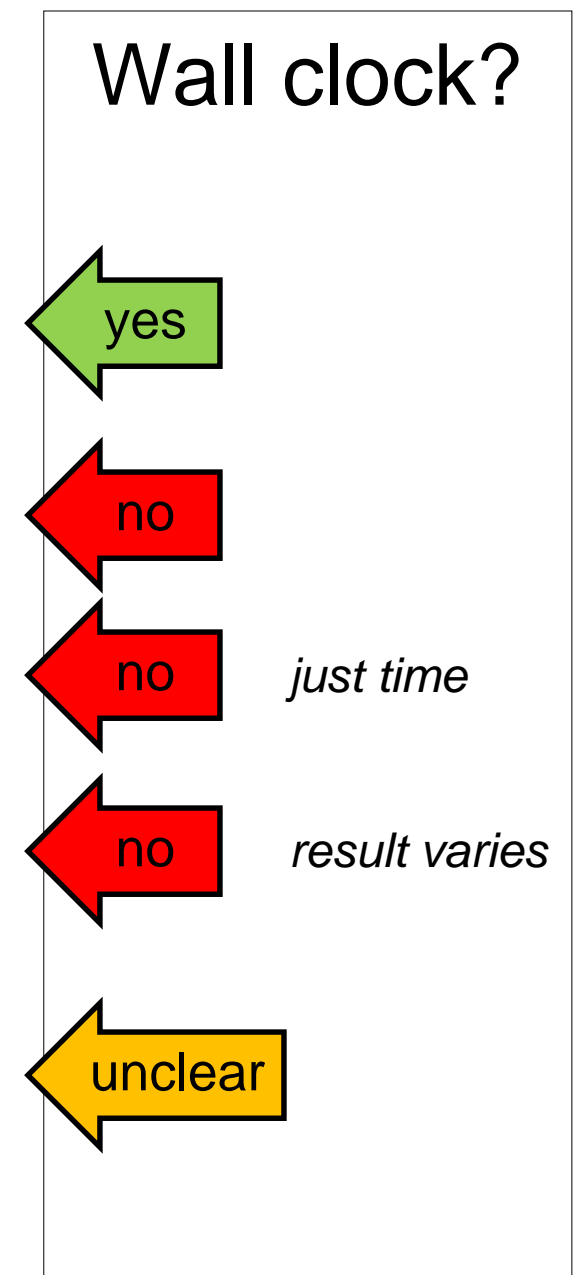
- **general**

- applicable to a large class of programs (and algorithms)
- independent of particular hardware
- applicable to many types of resources
  - time, space, energy, ...

- **mathematically rigorous**

- **useful**

- help us select among various algorithms for the same problem
  - e.g., POW vs. mystery function for exponentiation



# What do we mean by “How long”?

- Second idea: count the number of execution steps
  - How many operations do we do to search an  $n$ -element array?

```
int search(int x, int[] A, int n) {  
    for (int i = 0; i < n; i++) {  
        if (A[i] == x) return i;  
    }  
    return -1;  
}
```

Omitting  
contracts

- $i = 0$  1 step
- loop  $n$  times
  - $i < n$  1 step
  - $\text{if } (A[i] == x)$  1 step
  - $i++$  1 step
- $\text{return } -1$  1 step

$3n + 2$  steps



# What do we mean by “How long”?

## Step count

- $3n + 2$  steps to search an  $n$ -element array
- Always?
  - only if element is not found
- This is a **worst-case analysis**
  - Gives an **upper bound** on the number of steps

```
int search(int x, int[] A, int n) {  
    for (int i = 0; i < n; i++) {  
        if (A[i] == x) return i;  
    }  
    return -1;  
}
```

# What do we mean by “How long”?

## Step count

- $3n + 2$  steps to search an  $n$ -element array
- Depends only on  $n$ 
  - value of  $x$  doesn't matter
  - contents of  $A$  doesn't matter
    - other than its length
- $n$  is a **measure of the input** of the function
- Let's call the (upper bound on the) number of steps  $T(n)$

```
int search(int x, int[] A, int n) {  
    for (int i = 0; i < n; i++) {  
        if (A[i] == x) return i;  
    }  
    return -1;  
}
```

$$T(n) = 3n + 2$$

# What do we mean by “How long”?

## Step count

- *What is a step?*
  - is `i++` one step? 2 steps? 3 steps?
  - what about `if (A[i] == x)` ?
  - ... this gets complicated
- Each instruction takes a *constant* number of steps
  - exact number is tricky to tell, but it's constant
- In the worst case, `search` makes
  - a constant *b* number of steps outside the loop
  - a constant *a* number of steps in each iteration of the loop
- So,

```
int search(int x, int[] A, int n) {  
    for (int i = 0; i < n; i++) {  
        if (A[i] == x) return i;  
    }  
    return -1;  
}
```

• `i = 0`  
• `return -1`  
• `i < n`

**OOPS!!!**  
loop guard  
runs  $n+1$   
times!

• `i < n`  
• `if (A[i] == x)`  
• `i++`

$$T(n) = an + b$$

Note that *a* and *b*  
make it hard to  
plot exactly

# What do we mean by “How long”?

- Step count tells us “how long” a function takes to run
  - **about** “how long”
    - we can’t easily learn the constants  $a$ ,  $b$ , ...
  - **at most** “how long”
    - in the worst case
- We need to develop math that deals with
  - “about”
  - “at most”to reason about “how long” a function takes to run

That’s big-O

# What do we mean by “How long”?

- Step count tells us “how long” a function takes to run
  - Time (here, number of steps) is a type of **resource**
- Other resources of interest
  - Space: how much memory does the function use?
  - Energy: how much energy does running it consume?
  - Connectivity: how many network connection does it make?
  - ...

In this course, we will be mainly interested in execution time

- The amount of resources a function uses is called its **cost**
  - $T(n) = an + b$  is a **cost function**

Argument is  
a **measure** of  
the input

Computes the **cost**  
of executing it on  
an input of size  $n$

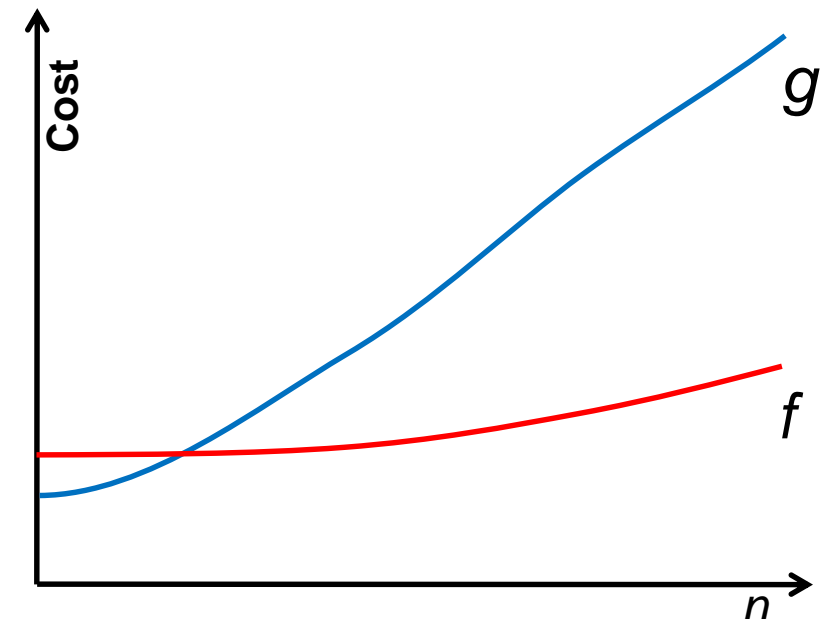
We will often keep the constants implicit:  
•  $3n+2$  when we really mean  
•  $an+b$   
Our new math will need to deal with this

# Comparing Cost

# Comparing Cost

- Given two C0 functions that solve the same problem
  - F has cost  $f(n)$
  - G has cost  $g(n)$we want to answer the question “is F better than G?”
- We will do so by answering the question  
“is  $f$  better than  $g$ ?”
  - **How do we define this?**
- $f$  and  $g$  are functions that
  - take a natural number as input
  - return a natural number as output

a non-negative integer

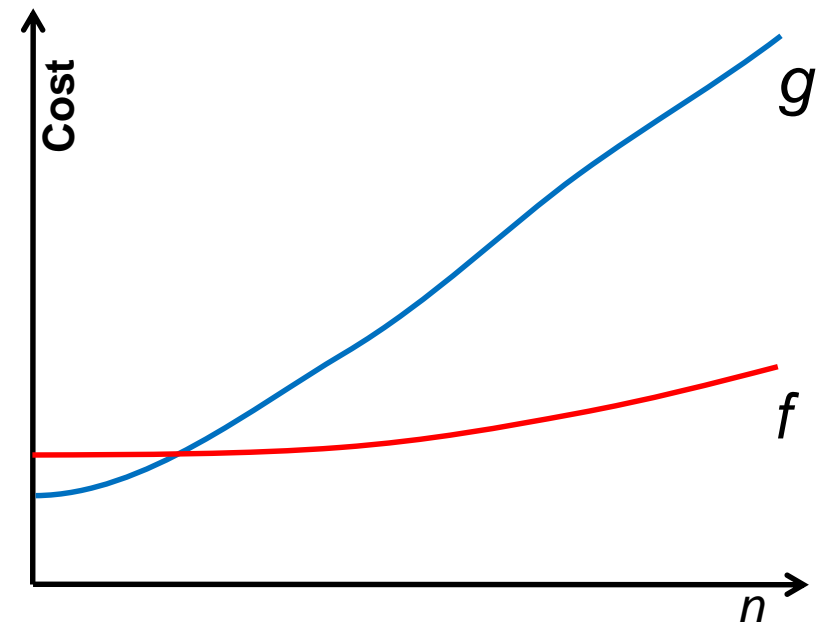
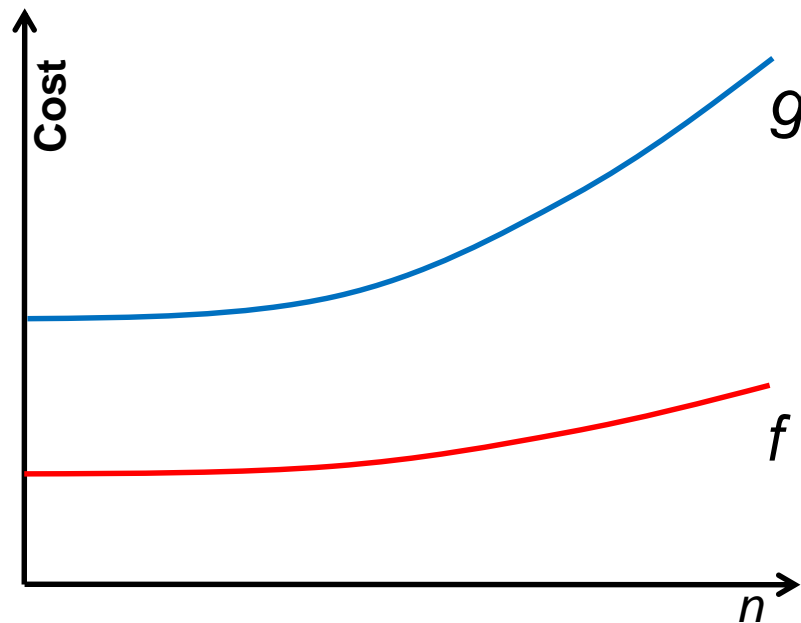


# “Is $f$ better than $g$ ?”

- Attempt #1:

“ $f$  is better than  $g$ ” if  
for all  $n$ ,  $f(n) \leq g(n)$

It's Ok if  $f(n) = g(n)$  for some (or all)  $n$

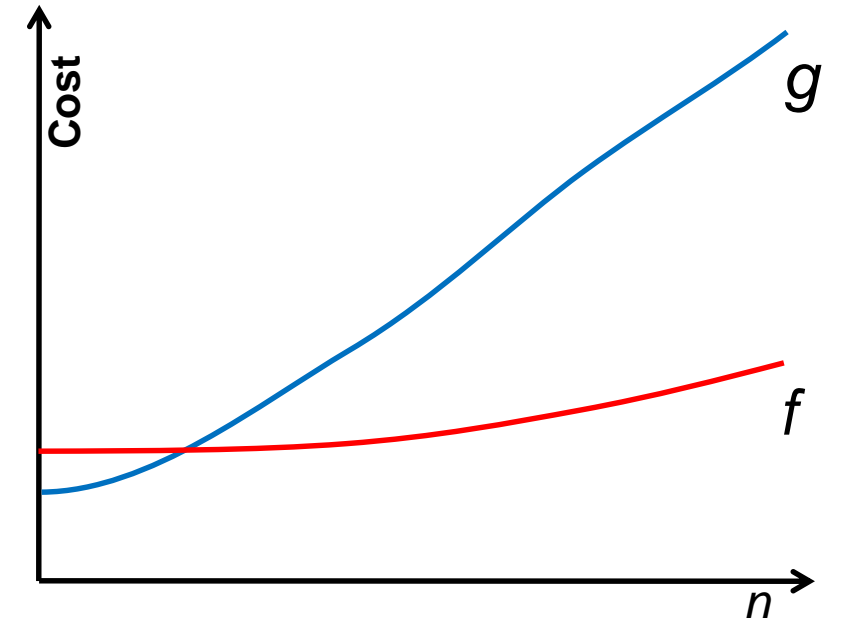




# “Is $f$ better than $g$ ?”

- Attempt #1:

“ $f$  is better than  $g$ ” if  
for all  $n$ ,  $f(n) \leq g(n)$



- But is this useful?

- This  $f$  is initially worse than  $g$
- but  $f$  is better beyond a certain point
- For small inputs, both costs are low
  - 0.12 ms vs. 0.23 ms doesn't matter for most applications
- For large inputs, we want lower cost
  - 1.35 ms vs. 200 years matters for all applications



- **Solution:** ignore small inputs

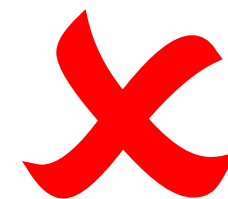
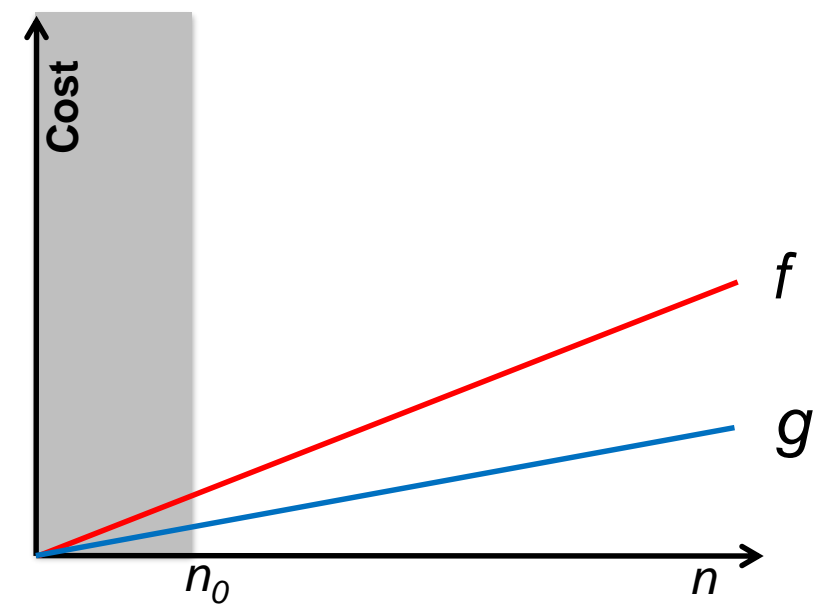
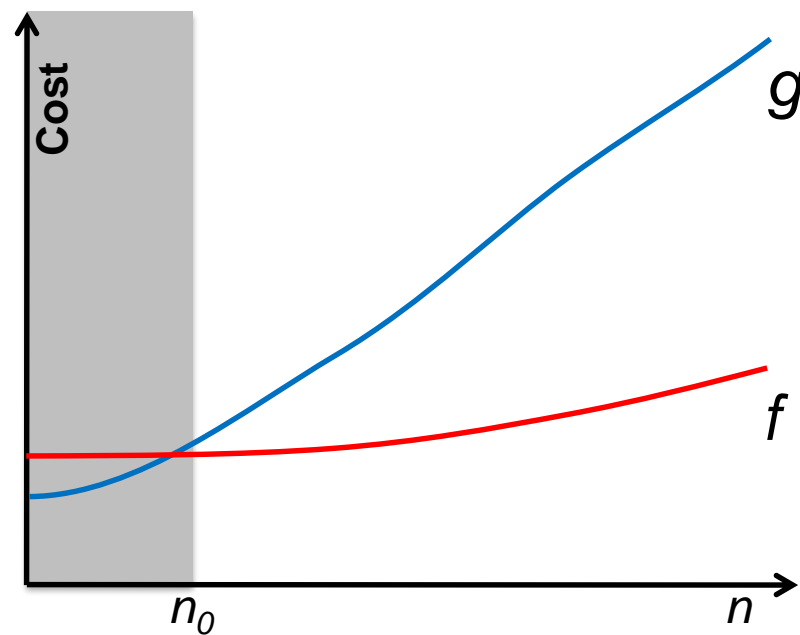
- **Asymptotic** notion of cost

# “Is $f$ better than $g$ ?”

- Attempt #2:

“ $f$  is better than  $g$ ” if

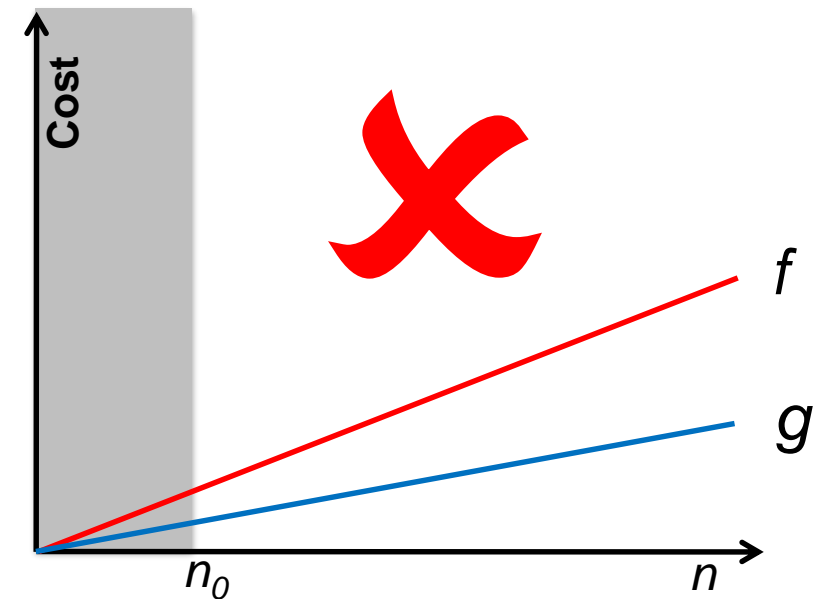
there exists a natural number  $n_0$  such that  
for all  $n \geq n_0$ ,  $f(n) \leq g(n)$



# “Is $f$ better than $g$ ?”

- Attempt #2:

“ $f$  is better than  $g$ ” if  
there exists a natural number  $n_0$  s.t.  
for all  $n \geq n_0$ ,  $f(n) \leq g(n)$



- But is this useful?

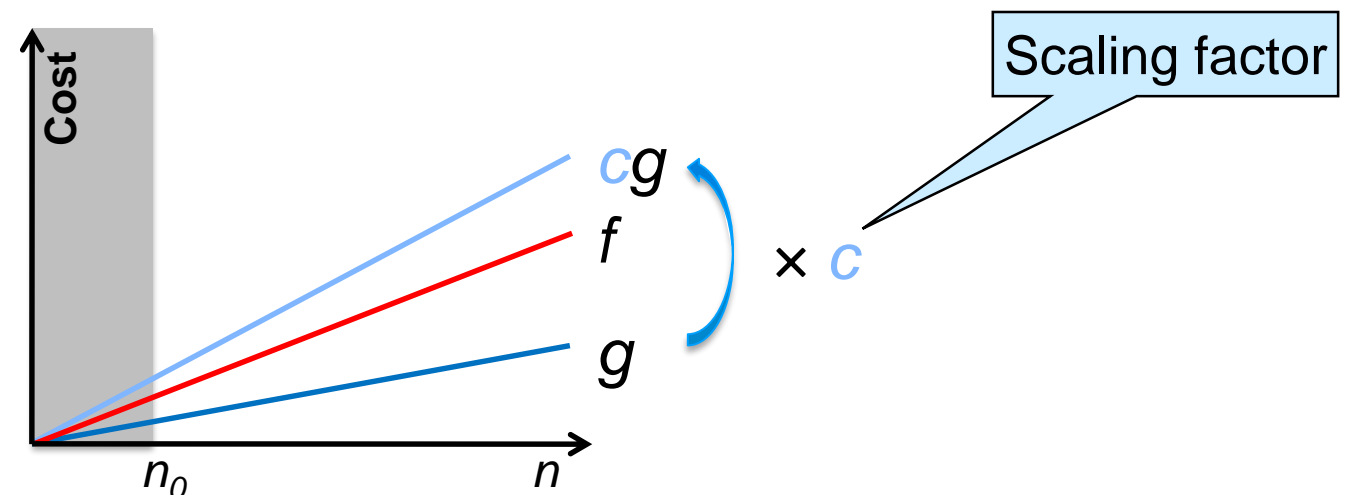
- These  $f$  and  $g$  are both linear functions

- $f(n) = a_1n$  and  $g(n) = a_2n$

- $a_1$  and  $a_2$  summarize unknown instruction-level step constants

- we can't easily know if  $a_1 > a_2$  or  $a_1 < a_2$  or even  $a_1 = a_2$

- **Solution:** scale  $g$

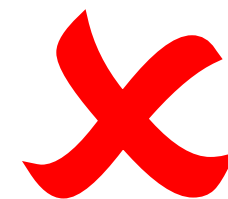
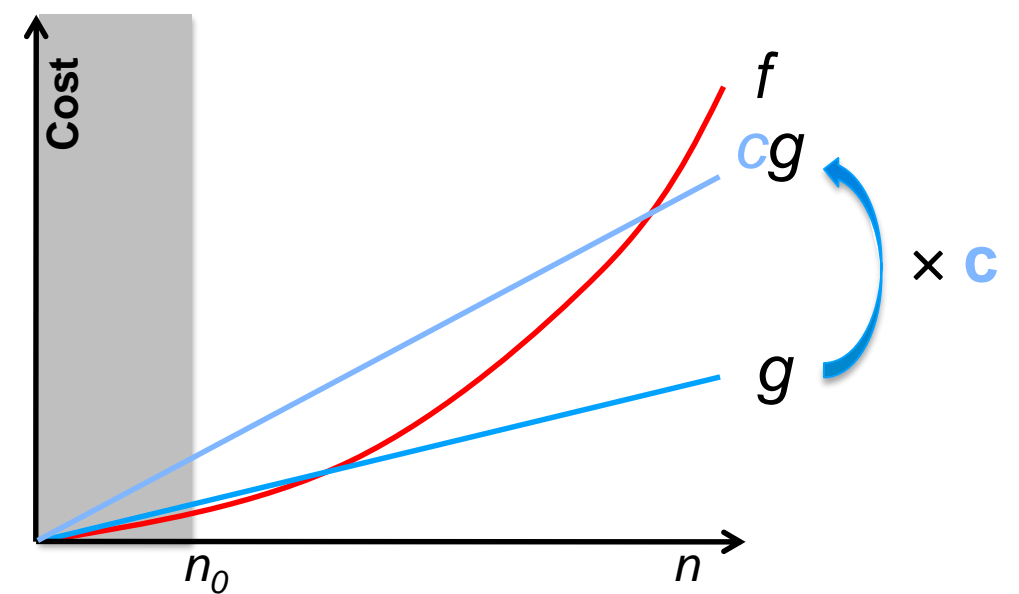
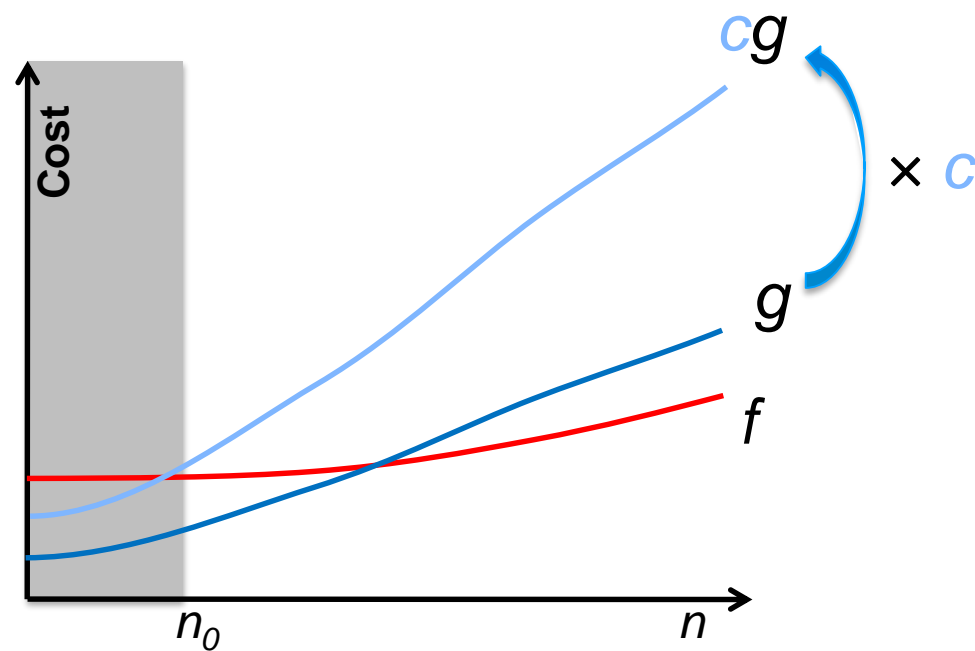


# “Is $f$ better than $g$ ?”

- Final attempt:

“ $f$  is better than  $g$ ” if

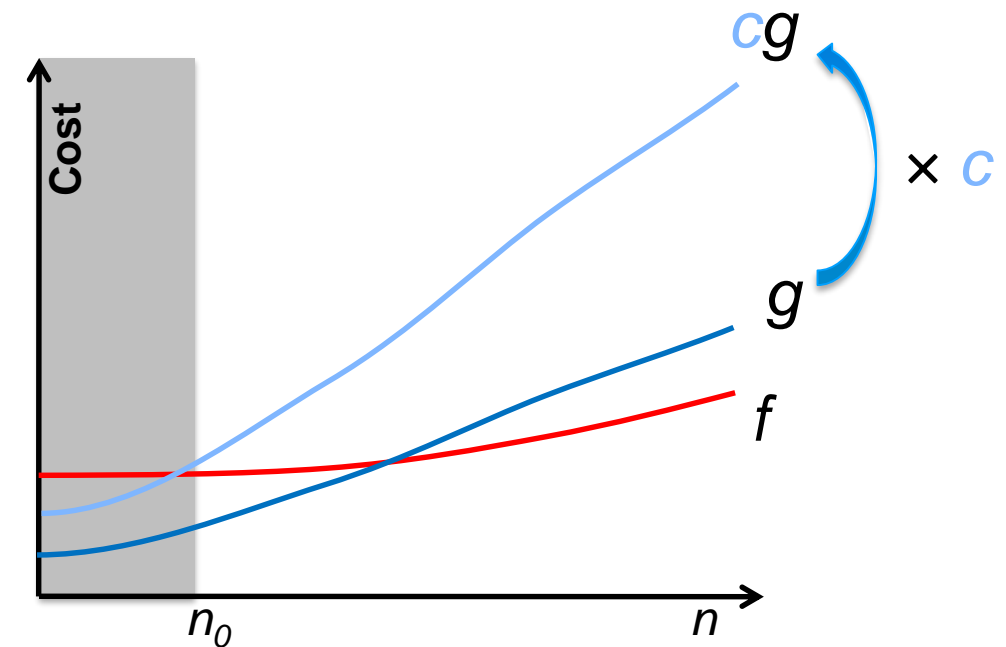
there exists a natural number  $n_0$  *and a real  $c > 0$*  s.t.  
for all  $n \geq n_0$ ,  $f(n) \leq c g(n)$



# Big O

# Big-O

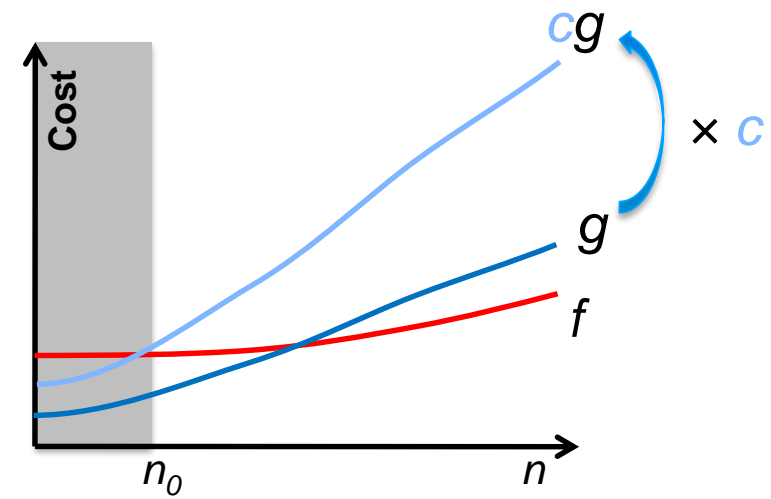
- Rather than “ $f$  is better than  $g$ ”, we say  $f \in O(g)$ 
  - “ $f$  is in big-O of  $g$ ”



$f \in O(g)$  if  
there exists a natural number  $n_0$  and a real  $c > 0$  s.t.  
for all  $n \geq n_0$ ,  $f(n) \leq c g(n)$

- $O(g)$  is a **set**:  
 $O(g) = \{ f \text{ s.t. there exists a natural number } n_0$   
and a real  $c > 0$  s.t.  
for all  $n \geq n_0$ ,  $f(n) \leq c g(n) \}$

# Big-O



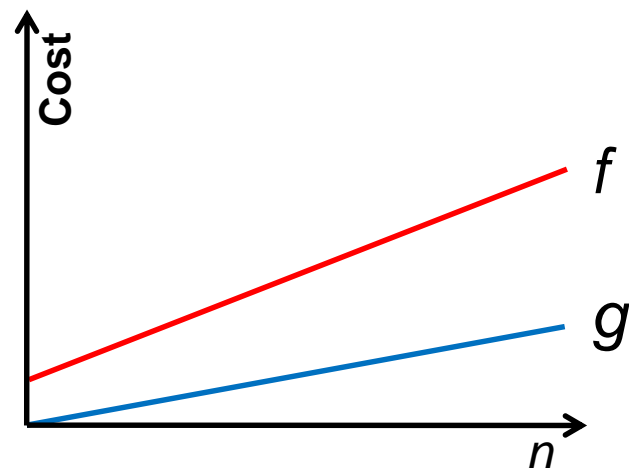
- Given two concrete functions  $f$  and  $g$ , how to tell if  $f \in O(g)$ ?
  - do the math
    - find  $n_0$  and  $c$  and show that the definition holds
    - or show that the definition cannot hold for any  $n_0$  or  $c$
  - recall what you learned in your calculus classes
    - enough for most of this course

# Big-O

- Is  $3n + 2 \in O(n)$ ?

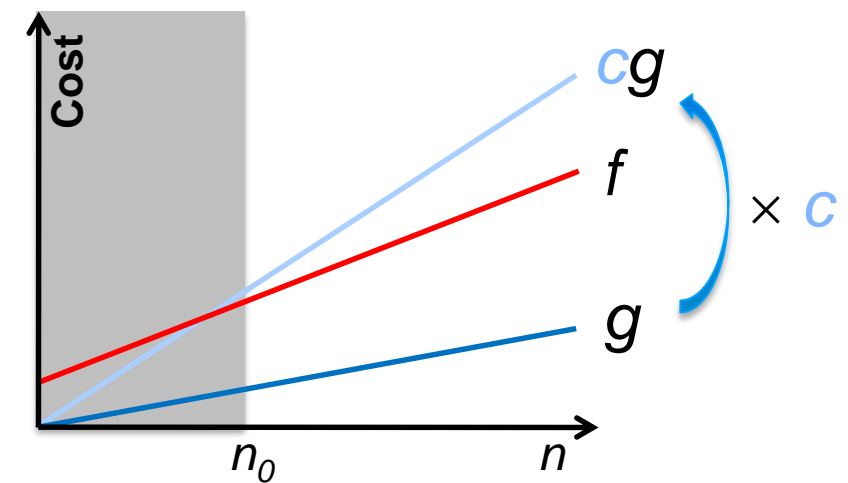
$$f(n) = 3n + 2$$

$$g(n) = n$$



Take  $c = 3.5$   
and  $n_0 = 4$

Any bigger values  
work too!



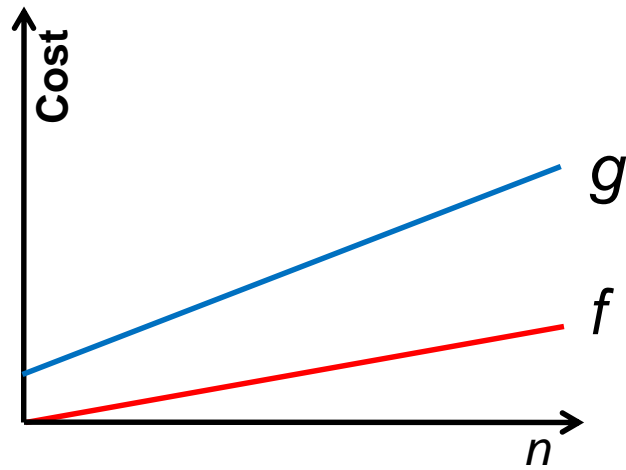


# Big-O

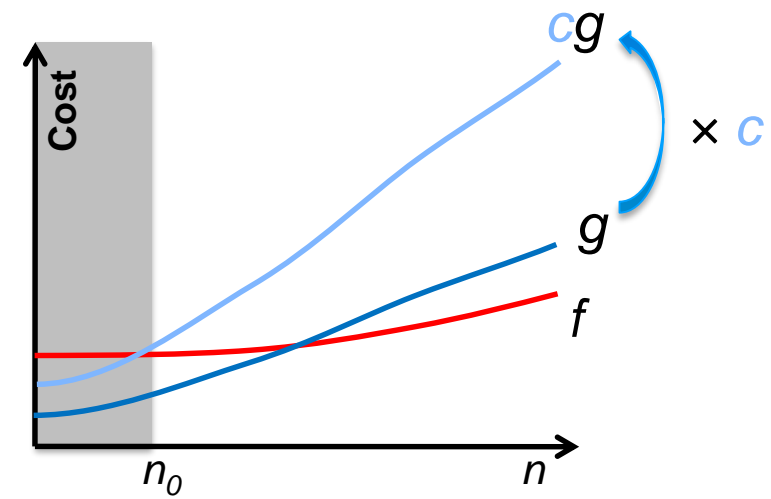
- Is  $n \in O(3n + 2)$ ?

$$f(n) = n$$

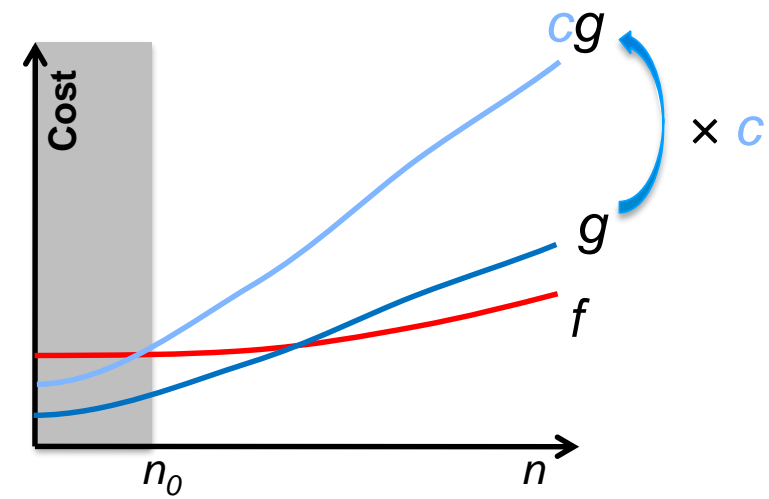
$$g(n) = 3n + 2$$



Take  $c = 1.0$   
and  $n_0 = 0$



# Big-O



- Every **linear cost function** is in  $O(n)$
- Every linear cost function is also in  $O(3n + 2)$
- As sets,  $O(n) = O(3n + 2)$ 
  - $O(n)$  is simpler however
  - $g(n) = n$  is the **simplest** linear function
- We describe a cost function that is linear by saying that it is in  $O(n)$

# Big-O

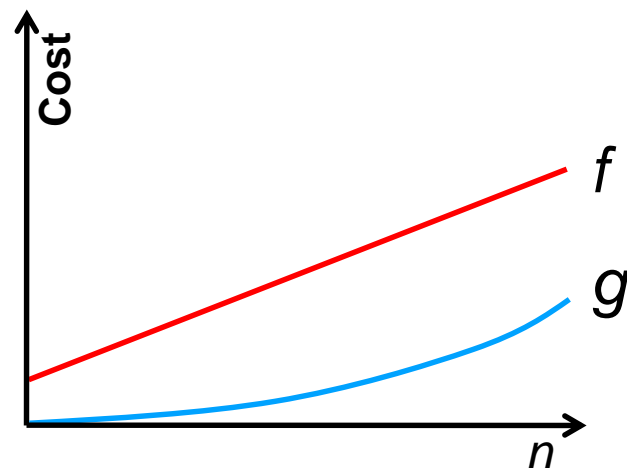
- Similarly, every **quadratic cost function** is in  $O(n^2)$ 
  - $g(n) = n^2$  is the **simplest** quadratic function
- We describe a cost function that is quadratic by saying that it is in  $O(n^2)$
- In general, every **polynomial cost function of degree  $p$**  is in  $O(n^p)$ 
  - $g(n) = n^p$  is the **simplest** polynomial of degree  $p$ 
    - We can ignore the terms with a smaller exponent
- We describe a cost function that is polynomial of degree  $p$  by saying that it is in  $O(n^p)$

# Big-O

- Is  $3n + 2 \in O(n^2)$ ?

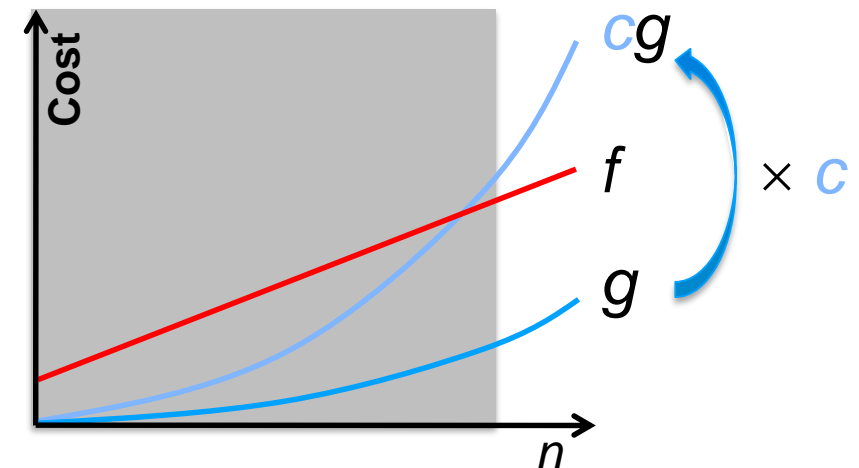
$$f(n) = 3n + 2$$

$$g(n) = n^2$$



Take  $c = 1.5$   
and  $n_0 = 3$

Any bigger values  
work too!



# Big-O

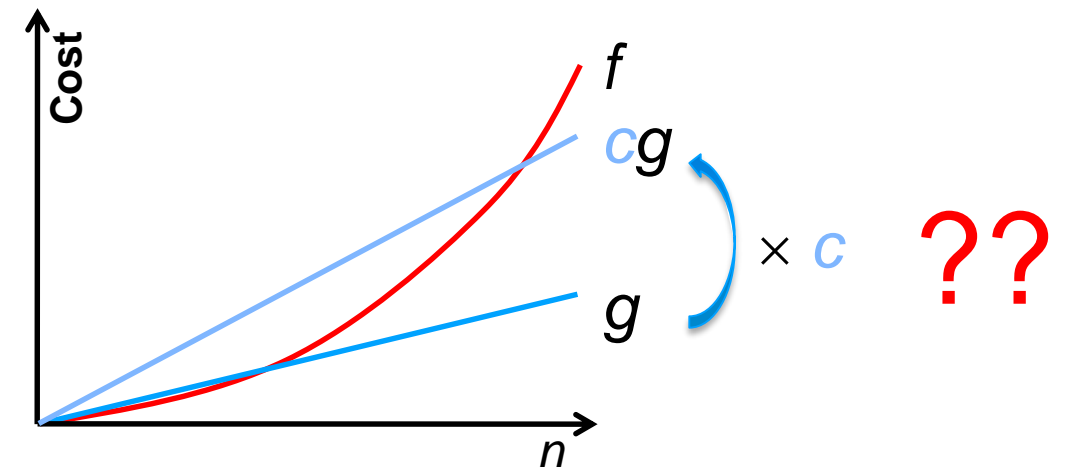
- $3n + 2 \in O(n)$  and  $3n + 2 \in O(n^2)$ 
  - $O(n)$  is **tighter** however
- Every linear function is in  $O(n^2)$ 
  - $O(n) \subseteq O(n^2)$
- In general, if  $p \leq q$ 
  - $O(n^p) \subseteq O(n^q)$

# Big-O

- Is  $n^2 \in O(3n + 2)$ ?

$$f(n) = n^2$$

$$g(n) = 3n + 2$$



- $n^2$  eventually dominates  $c(3n+2)$  no matter the scaling factor  $c$
- $n^2 \notin O(3n + 2)$

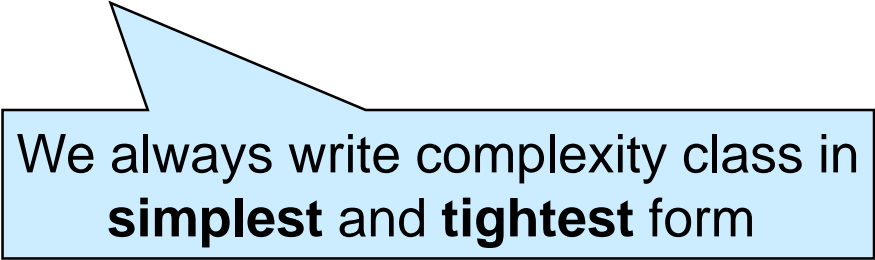


- Quadratic functions are **not** in  $O(n)$ 
  - $O(n^2) \not\subset O(n)$
  - $O(n) \subset O(n^2)$

# **Complexity Classes**

# Complexity Classes

- We learned that  $O(n) \subset O(n^2)$ 
  - $O(n) \subseteq O(n^2)$
  - but  $O(n^2) \not\subseteq O(n)$
- $O(n)$  and  $O(n^2)$  are called **complexity classes**
  - Simplest and tightest expressions for sets of cost functions



We always write complexity class in **simplest** and **tightest** form



# Complexity Classes

Main complexity classes we will use in this course

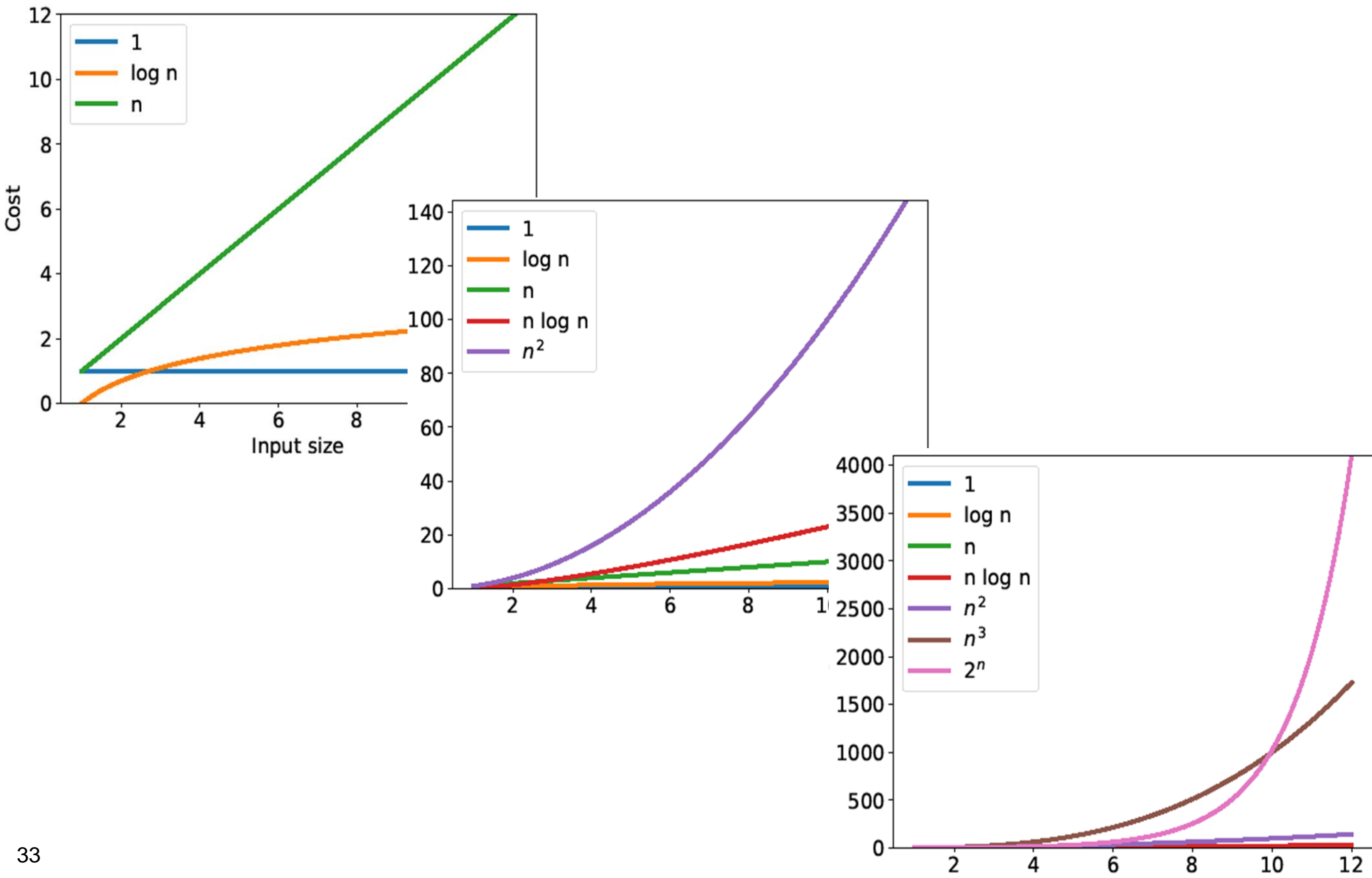
Complexity class	Common name
$O(1)$	Constant
$O(\log n)$	Logarithmic
$O(n)$	Linear
$O(n \log n)$	(just “ $n \log n$ ”)
$O(n^2)$	Quadratic
$O(2^n)$	Exponential

We do not write  
the base of the  
logarithm!

$$O(1) \subset O(\log n) \subset O(n) \subset O(n \log n) \subset O(n^2) \subset O(2^n)$$

- There are many more, though

# Complexity Classes



# **Complexity of Linear Search**

# What do we mean by “How long”?

- $T(n) = an + b$

- So  $T(n) \in O(n)$

Simplest and tightest  
class

```
int search(int x, int[] A, int n) {  
    for (int i = 0; i < n; i++) {  
        if (A[i] == x) return i;  
    }  
    return -1;  
}
```

- Linear search has **worst case complexity**  $O(n)$ 
  - It has **linear complexity** in the size  $n$  of its input
  - That's why it is called *linear* search

# What do we mean by “How long”?

- We can often determine the big-O class of a function without writing down its cost function

- For each operation, note
  - its cost
  - how many times it is executed

Add up

	Cost	Tally	
<code>int search(int x, int[] A, int n) {</code>			
<code>// int i = 0 happens before the loop</code>	$O(1)$	$O(1)$	
<code>for (int i = 0; i &lt; n; i++) { // at most</code>	$n$ times		
<code>// i &lt; n happens before each iteration</code>	$O(1)$	$O(n)$	
<code>if (A[i] == x)</code>	$O(1)$	$O(n)$	
<code>return i;</code>	$O(1)$	$O(n)$	
<code>// i++ happens last in the body</code>	$O(1)$	$O(n)$	
<code>}</code>			
<code>return -1;</code>	$O(1)$	$O(n)$	Complexity of search(x, A, n)
<code>}</code>			

# Big-O

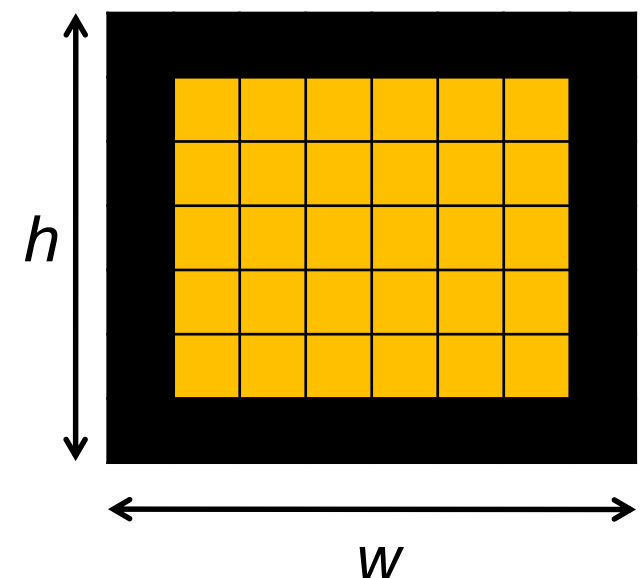
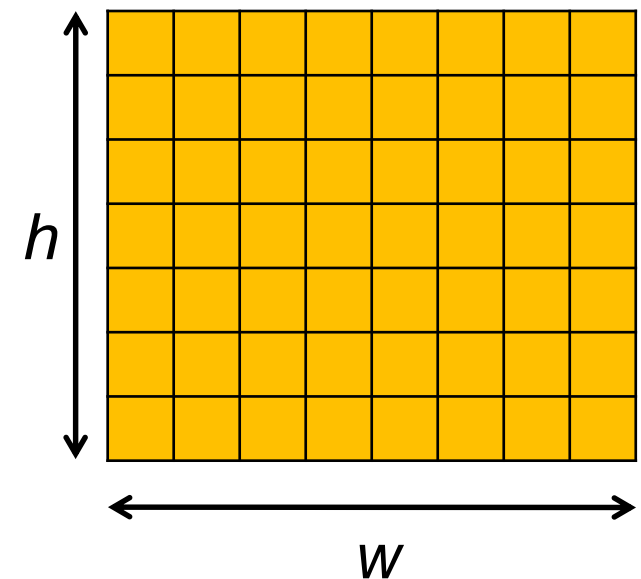
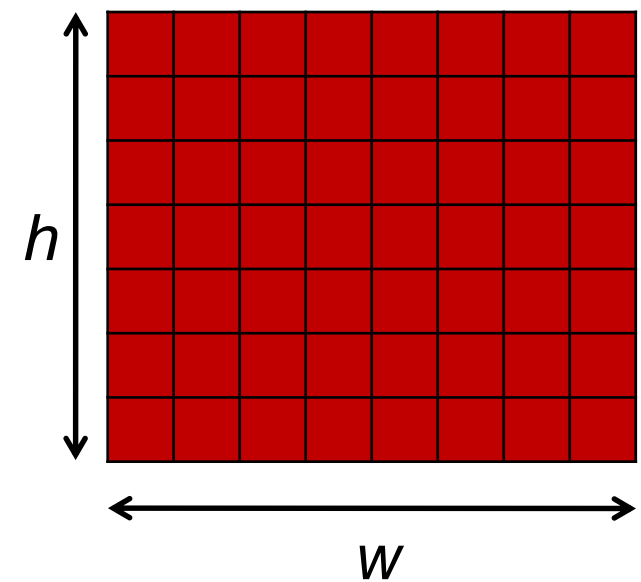
- There is nothing special about “ $n$ ”
  - If we call the size of the input  $k$ ,
  - then this function has cost  $O(k)$ 
    - still linear

```
int search(int x, int[] A, int k) {  
    for (int i = 0; i < k; ++i) {  
        if (A[i] == x) return i;  
    }  
    return -1;  
}
```

# Big-O

An input may have multiple size parameters

- For example, an image has a width  $w$  and a height  $h$
- Brightening an image:  $O(wh)$ 
  - We change every one of its  $w \times h$  pixels
- Putting a border:  $O(w+h)$ 
  - we touch about  $2(w + h)$  pixels



# Big-O

- Here's the definition of big-O with 2 variables

$f(x,y) \in O(g(x,y))$  if

there exist natural numbers  $x_0, y_0$  and a real  $c > 0$  s.t.

for all  $x \geq x_0$  and  $y \geq y_0$ ,  $f(x,y) \leq c g(x,y)$

➤ There are other definitions, but this is the one we will use

- It is asymptotic in both  $x$  and  $y$

➤ no need for special cases for small values of  $x$  or  $y$

□ e.g., if  $x=0$  and  $y$  grows very big

- This generalizes to any number of variables



# **Towards a Better Search**

# Algorithms vs. Problems

- Linear search has cost  $O(n)$
- But this is only one of the many algorithms to search an array

```
int search(int x, int[] A, int n) {  
    for (int i = 0; i < n; i++) {  
        if (A[i] == x) return i;  
    }  
    return -1;  
}
```

- Can a different algorithm find an element faster?
- **No**: the **problem** of searching a *generic* n-element array has complexity  $O(n)$ 
  - some algorithms have worse complexity
  - but no algorithm has better complexity
    - unless we radically change what we mean by “step”
- Can we do better for arrays with common characteristics?
  - Say, the array is **sorted**


# Searching Sorted Arrays

A is sorted

- the segment  $A[0, n)$  is sorted
  - (useful for later)

x: 4

	0	1	2	3	4
A:	3	5	7	8	12



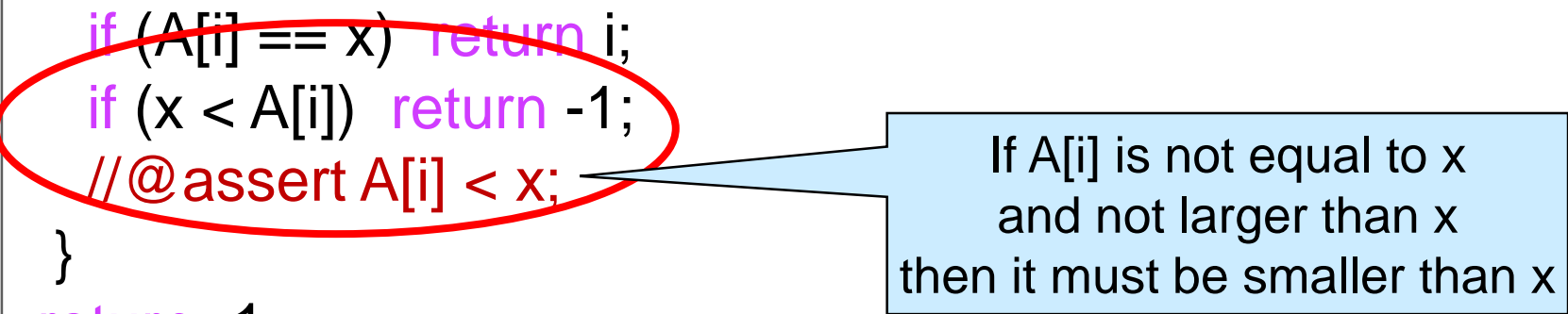
## ● Idea:

- If we find an element larger than x, we can stop searching
  - All elements after it will also be larger than x
  - That's because A is sorted

```
int search(int x, int[] A, int n)
//@requires n == \length(A);
//@requires is_sorted(A, 0, n);
/*@ensures (\result == -1 && !is_in(x, A, 0, n))
           || (0 <= \result && \result < n && A[\result] == x);
@*/
{
    for (int i = 0; i < n; i++)
        //@loop_invariant 0 <= i && i <= n;
        //@loop_invariant !is_in(x, A, 0, i);
        {
            if (A[i] == x) return i;
        }
    return -1;
}
```

# Searching Sorted Arrays

```
int search(int x, int[] A, int n)
//@requires n == \length(A);
//@requires is_sorted(A, 0, n);
/* @ensures (\result == -1 && !is_in(x, A, 0, n))
           || (0 <= \result && \result < n && A[\result] == x);
@*/
{
    for (int i = 0; i < n; i++)
        //@loop_invariant 0 <= i && i <= n;
        //@loop_invariant !is_in(x, A, 0, i);
        {
            if (A[i] == x) return i;
            if (x < A[i]) return -1;
            //@assert A[i] < x;
        }
    return -1;
}
```



# Searching Sorted Arrays

- What is the cost of this **search**?
- Still  $O(n)$ 
  - **worst case** is when searching an element bigger than anything in A
- But that's just *one* algorithm for searching in a sorted array
  - Can we do better?
  - ... next lecture ...

```
int search(int x, int[] A, int n)
//@requires is_sorted(A, 0, n);
{
    for (int i = 0; i < n; i++) {
        if (A[i] == x) return i;
        if (x < A[i]) return -1;
        //@assert A[i] < x;
    }
    return -1;
}
```

# Searching Sorted Arrays

- Is this code safe?

- Yes, no new array accesses

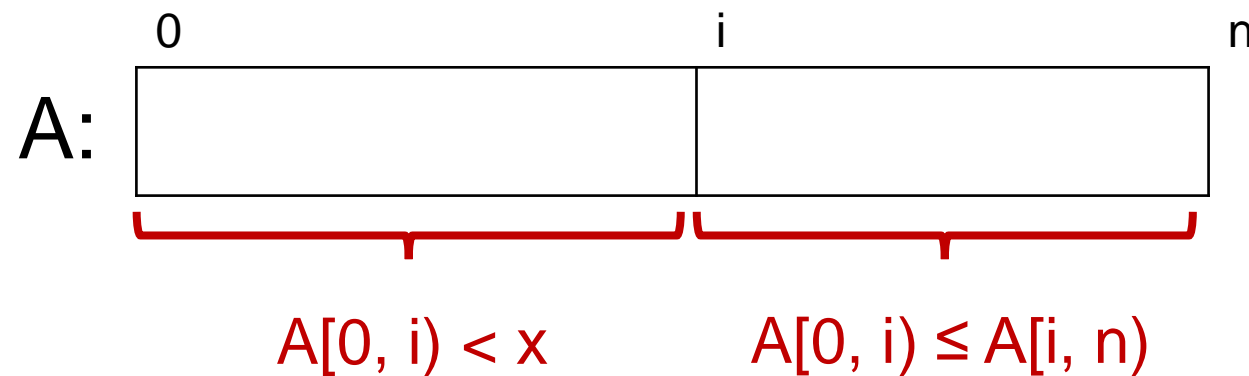
- Is it correct?

- Original argument still holds
- but we have a new place where the function returns
  - Is it correct there?
  - no good argument
  - we need new insight

```
int search(int x, int[] A, int n)
//@requires n == \length(A);
//@requires is_sorted(A, 0, n);
/* @ensures (\result == -1 && !is_in(x, A, 0, n))
           || (0 <= \result && \result < n && A[\result] == x);
@*/
{
    for (int i = 0; i < n; i++)
        //@loop_invariant 0 <= i && i <= n;
        //@loop_invariant !is_in(x, A, 0, i);
        {
            if (A[i] == x) return i;
            if (x < A[i]) return -1;
            //@assert A[i] < x;
        }
    return -1;
}
```

# Searching Sorted Arrays

- What do we know at iteration  $i$ ?
  - Let's draw pictures!



```
int search(int x, int[] A, int n)
//@requires n == \length(A);
//@requires is_sorted(A, 0, n);
/*@ensures (\result == -1 && !is_in(x, A, 0, n))
           || (0 <= \result && \result < n &&
              A[\result] == x);
           @*/
{
    for (int i = 0; i < n; i++)
        //@loop_invariant 0 <= i && i <= n;
        //@loop_invariant !is_in(x, A, 0, i);
        {
            if (A[i] == x) return i;
            if (x < A[i]) return -1;
            //@assert A[i] < x;
        }
    return -1;
}
```

- $A[0, i) < x$ : every element in segment  $A[0, i)$  is less than  $x$ 
  - because we would have returned otherwise
- $A[0, i) \leq A[i, n)$ : everything in  $A[0, i)$  is  $\leq$  everything in  $A[i, n)$ 
  - because  $A$  is sorted

# Reasoning about Array Segments

- $A[0, i) < x$ , etc are useful to reason about array segments
- Implement them into *specification* functions: **arrayutil**

➤  $A[lo, hi)$  is sorted

➤  $x \in A[lo, hi)$

➤  $x < A[lo, hi)$

➤  $x \leq A[lo, hi)$

➤  $x > A[lo, hi)$

➤  $x \geq A[lo, hi)$

➤  $A[lo_1, hi_1) < B[lo_2, hi_2)$

➤  $A[lo_1, hi_1) \leq B[lo_2, hi_2)$

➤  $A[lo_1, hi_1) > B[lo_2, hi_2)$

➤  $A[lo_1, hi_1) \geq B[lo_2, hi_2)$

`is_sorted(A, lo, hi)`

`is_in(x, A, lo, hi)`

`lt_seg(x, A, lo, hi)`

`le_seg(x, A, lo, hi)`

`gt_seg(x, A, lo, hi)`

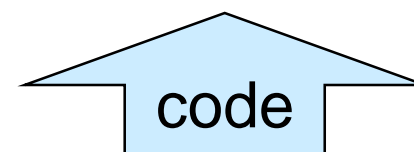
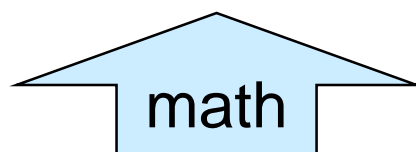
`ge_seg(x, A, lo, hi)`

`lt_segs(A, lo1, hi1, B, lo2, hi2)`

`le_segs(A, lo1, hi1, B, lo2, hi2)`

`gt_segs(A, lo1, hi1, B, lo2, hi2)`

`ge_segs(A, lo1, hi1, B, lo2, hi2)`



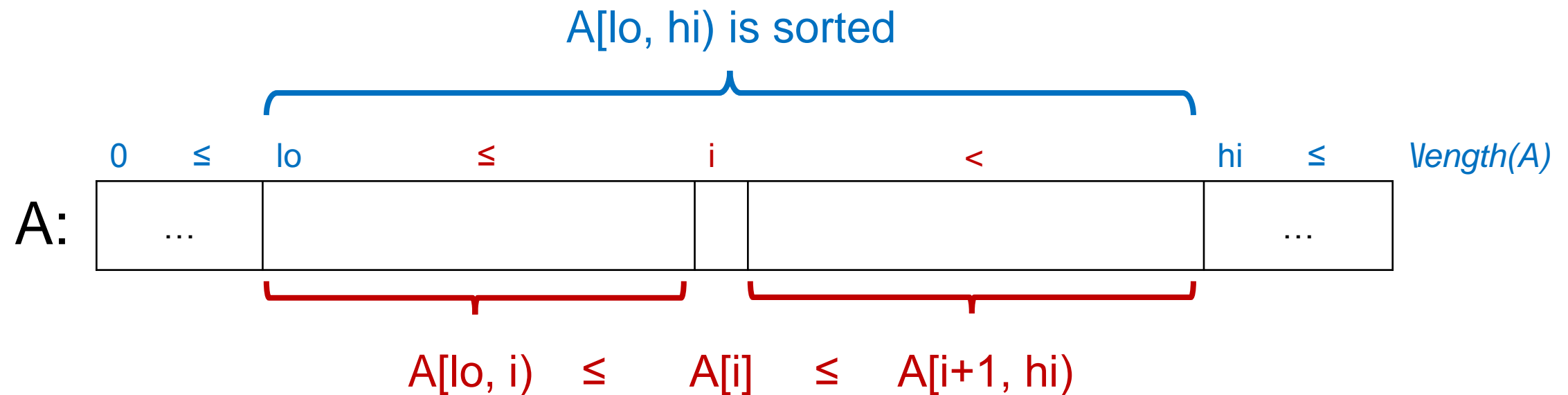
See **arrayutil.c0** file

Includes  
contracts



# Reasoning about Sorted Arrays

- If an array (segment) is sorted, what do we know?



- for every element  $A[i]$

- $lo \leq i < hi$

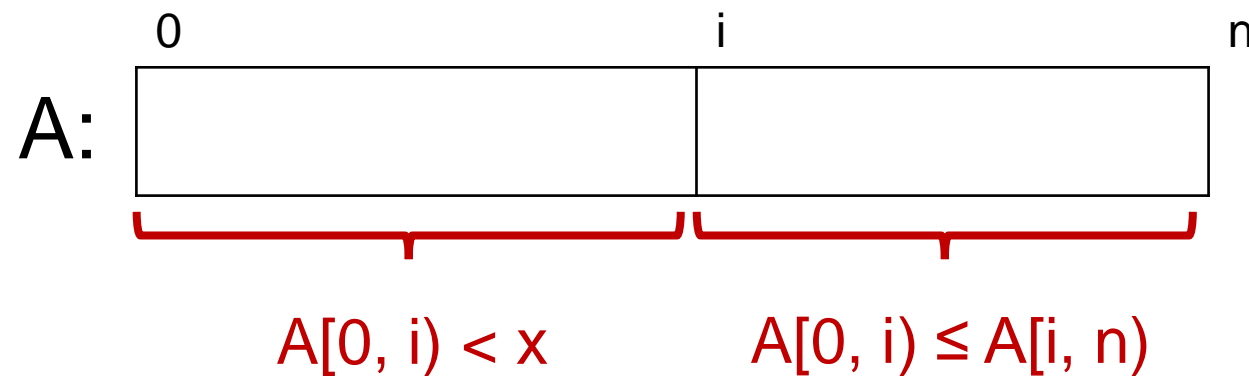
$A[lo, hi)$  can't be empty

- $A[lo, i) \leq A[i] \leq A[i+1, hi)$

- $A[lo, i)$  and  $A[i+1, hi)$  are sorted

# Searching Sorted Arrays

- What do we know at iteration  $i$ ?
  - Let's draw pictures!



```
int search(int x, int[] A, int n)
//@requires n == \length(A);
//@requires is_sorted(A, 0, n);
/*@ensures (\result == -1 && !is_in(x, A, 0, n))
           || (0 <= \result && \result < n &&
              A[\result] == x);          @*/
{
    for (int i = 0; i < n; i++)
        //@loop_invariant 0 <= i && i <= n;
        //@loop_invariant gt_seg(x, A, 0, i);
        //@loop_invariant le_segs(A, 0, i, A, i, n):
        {
            if (A[i] == x) return i;
            if (x < A[i]) return -1;
            //@assert A[i] < x;
        }
    return -1;
}
```

- Candidate loop invariants
  - $gt\_seg(x, A, 0, i)$ : that's  $A[0, i) < x$ 
    - This implies  $!is\_in(x, A, 0, i)$ , that's  $x \notin A[0, i)$
  - $le\_segs(A, 0, i, A, i, n)$ : that's  $A[0, i) \leq A[i, n)$

# Searching Sorted Arrays

Is this code correct?

➤ **To show:**  $\text{!is\_in}(x, A, 0, n)$   
(assuming invariants are valid)

- |                       |  |
|-----------------------|--|
| A. $A[0, i) < x$      | by line 10 (LI 2)                      |
| B. $x \notin A[0, i)$ | by math on A                           |
| C. $x < A[i]$         | by line 14 (conditional)               |
| D. $x < A[i, n)$      | by math on C and line 3 (precondition) |
| E. $x \notin A[0, n)$ | by math on B, D                        |

```
1. int search(int x, int[] A, int n)
2. // @requires n == \length(A);
3. // @requires is_sorted(A, 0, n);
4. /* @ensures (\result == -1 && !is_in(x, A, 0, n))
5.           || (0 <= \result && \result < n &&
6.              A[\result] == x);          @*/
7. {
8.     for (int i = 0; i < n; i++)
9.         // @loop_invariant 0 <= i && i <= n;
10.        // @loop_invariant gt_seg(x, A, 0, i);
11.        // @loop_invariant le_segs(A, 0, i, A, i, n);
12.        {
13.            if (A[i] == x) return i;
14.            if (x < A[i]) return -1;
15.            // @assert A[i] < x;
16.        }
17.     return -1;
18. }
```

# Searching Sorted Arrays

$x > A[0, i)$  is a **valid** loop invariant

## INIT

➤ **To show:**  $x > A[0, i)$  initially

- A.  $i = 0$  by line 7
- B.  $x > A[0, 0)$  by definition of **gt\_seg**
  - $A[0, 0)$  is the empty array segment
- Nothing is in it

## PRES

➤ **To show:** if  $x > A[0, i)$ , then  $x > A[0, i')$

- A.  $i' = i + 1$  by line 8
- B.  $x > A[0, i)$  by assumption
- C.  $x > A[i]$  by line 15 (math on lines 13 and 14)
- D.  $x > A[0, i + 1)$  by math on B and C
- E.  $x > A[0, i')$  by math on A and D

```
1. int search(int x, int[] A, int n)
2. //@requires n == \length(A);
3. //@requires is_sorted(A, 0, n);
4. /*@ensures (\result == -1 && !is_in(x, A, 0, n))
5.           || (0 <= \result && \result < n &&
6.              A[\result] == x);          @*/
7. {
8.   for (int i = 0; i < n; i++)
9.     //@loop_invariant 0 <= i && i <= n;
10.    //@loop_invariant gt_seg(x, A, 0, i);
11.    //@loop_invariant le_segs(A, 0, i, A, i, n);
12.    {
13.      if (A[i] == x) return i;
14.      if (x < A[i]) return -1;
15.      //@assert A[i] < x;
16.    }
17.   return -1;
18. }
```

# Searching Sorted Arrays

$A[0,i) \leq A[i,n)$  is a **valid** loop invariant

## INIT

➤ **To show:**  $A[0, i) \leq A[i, n)$  initially

A.  $i = 0$  by line 7

B.  $A[0, 0) \leq A[i, n)$  by definition of **le\_segs**

○  $A[0, 0)$  is the empty array segment

➤ All the (**zero**) things in  $A[0, 0)$  are  $\leq$  everything in  $A[i, n)$

```
1. int search(int x, int[] A, int n)
2. // @requires n == \length(A);
3. // @requires is_sorted(A, 0, n);
4. /* @ensures (\result == -1 && !is_in(x, A, 0, n))
5.           || (0 <= \result && \result < n &&
6.              A[\result] == x);          @*/
7. {
8.   for (int i = 0; i < n; i++)
9.     // @loop_invariant 0 <= i && i <= n;
10.    // @loop_invariant gt_seg(x, A, 0, i):
11.    // @loop_invariant le_segs(A, 0, i, A, i, n);
12.    {
13.      if (A[i] == x) return i;
14.      if (x < A[i]) return -1;
15.      // @assert A[i] < x;
16.    }
17.   return -1;
18. }
```

## PRES

➤ **To show:** if  $A[0, i) \leq A[i, n)$ , then  $A[0, i') \leq A[i', n)$

A.  $i' = i + 1$  by line 8

B.  $A[0, i) \leq A[i, n)$  by assumption

C.  $A[0, n)$  sorted by line 3 (precondition)

D.  $A[0, i + 1) \leq A[i + 1, n)$  by math on C

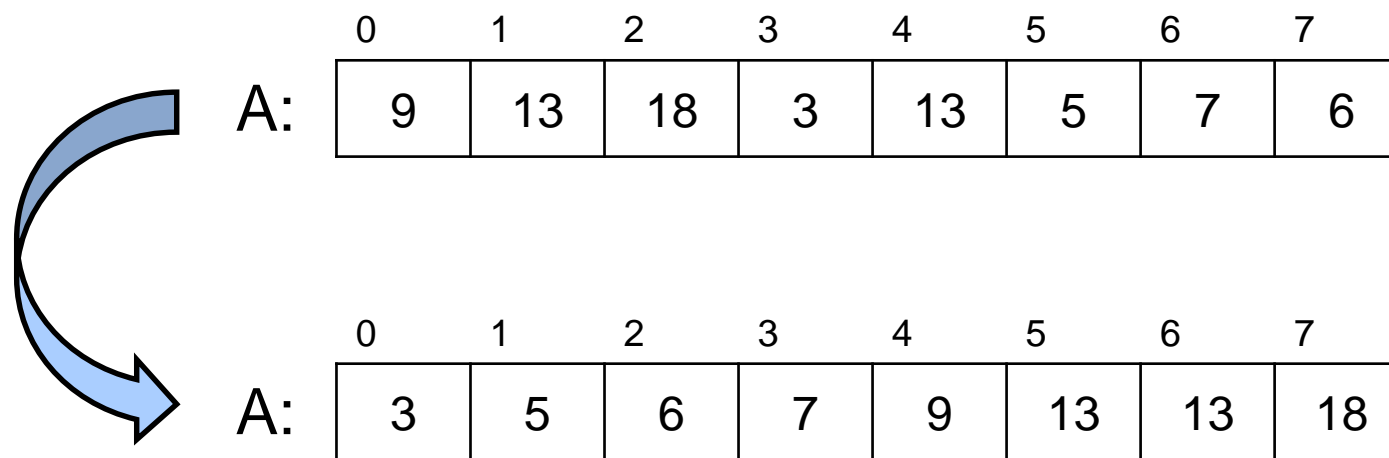
E.  $A[0, i') \leq A[i', n)$  by math on A and D

We actually don't need this loop invariant to prove correctness

# Selection Sort

# Sorting an Array

- Reorder the elements to put them in increasing order
  - Duplicate elements are allowed



- There are many algorithms to sort arrays

# Selection Sort


- Find element that shall go in  $A[0]$ 
  - Smallest element in  $A[0, n)$
- Swap it with  $A[0]$
- Find element that shall go in  $A[1]$ 
  - Smallest element in  $A[1, n)$
- Swap it with  $A[1]$
- ... carry on ...
- Stop when  $A$  is entirely sorted

A:

0	1	2	3	4	5	6	7
9	13	18	3	13	5	7	6

A:

0	1	2	3	4	5	6	7
3	13	18	9	13	5	7	6




A:

0	1	2	3	4	5	6	7
3	13	18	9	13	5	7	6

A:

0	1	2	3	4	5	6	7
3	5	18	9	13	13	7	6



A:

0	1	2	3	4	5	6	7
3	5	6	7	9	13	13	18



# Selection Sort

We need two operations

- find the minimum of an array segment  $A[lo, hi)$ 
  - and return its index

```
int find_min(int[] A, int lo, int hi)
//@requires 0 <= lo && lo < hi && hi <= \length(A);
/* @ensures lo <= \result && \result < hi
    && le_seg(A[\result], A, lo, hi); @*/ ;
```

$A[lo, hi)$  can't be empty

That's  $A[\text{result}] \leq A[lo, hi)$

- swap two elements of an array (given their indices)

```
// swaps A[i] and A[j]; all other elements are unchanged
void swap(int[] A, int i, int j)
//@requires 0 <= i && i < \length(A);
//@requires 0 <= j && j < \length(A);
```

returns no value  
swap modifies  
input array

We can't say this as  
a postcondition.  
We use a comment instead

*Implementation left as exercise*

# Selection Sort

- Let's capture our intuition about how it works in code
  - Generalization: sort array segment  $A[lo, hi)$

sort does not return anything either

but it modifies the input array

$A[lo, hi)$  can be empty

```
void sort(int[] A, int lo, int hi)
//@requires 0 <= lo && lo <= hi && hi <= \length(A);
//@ensures is_sorted(A, lo, hi);
{
    for (int i = lo; i < hi; i++)
    {
        int min = find_min(A, i, hi);
        swap(A, i, min);
    }
}
```

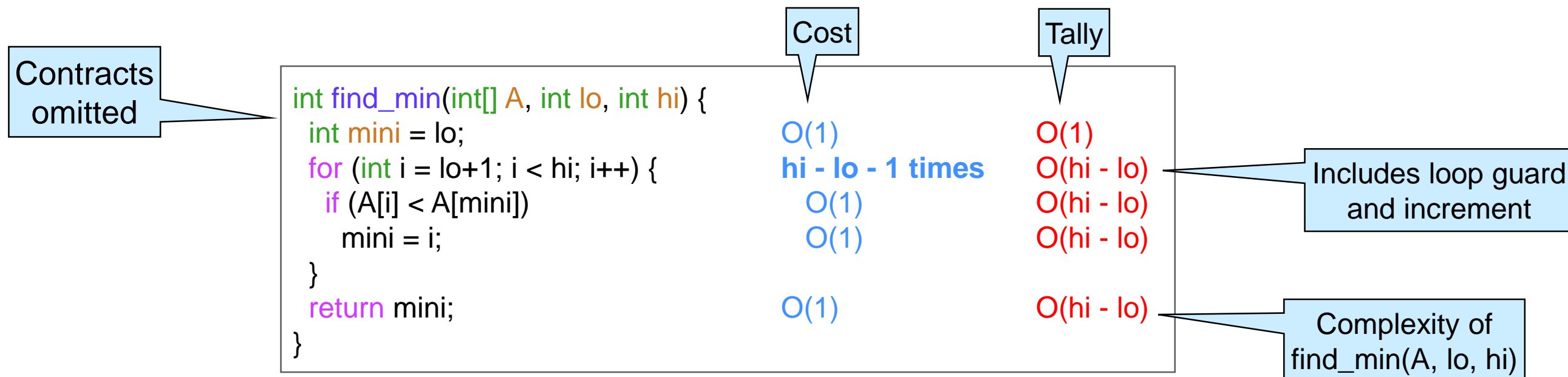
for every index  $i$  from  $lo$  to  $hi$

find the minimum of  $A[i, hi)$

swap it with  $A[i]$

# Cost of Selection Sort

- `find_min(A, lo, hi)`
  - finds the minimum of an array segment `A[lo, hi)`
  - and returns its index
  - it scans the entire segment once

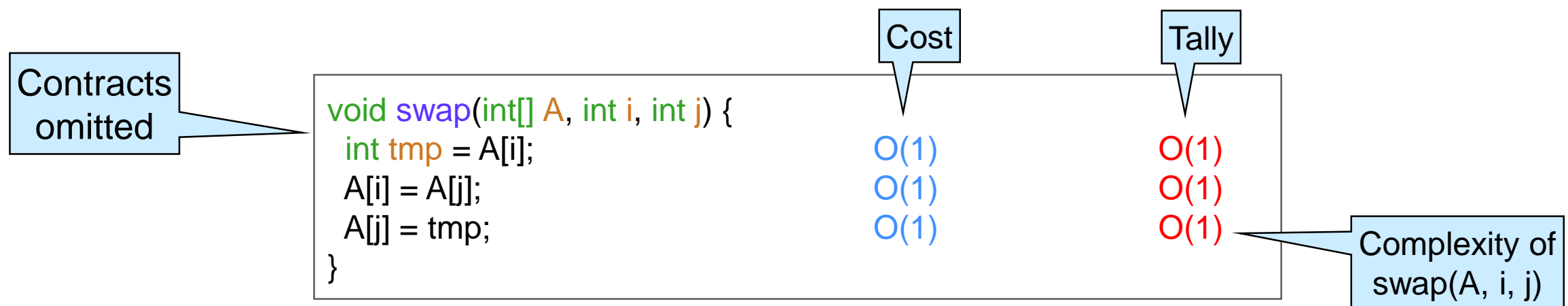


Cost:  $O(hi - lo)$

- note that it makes  $hi - lo - 1$  comparisons
  - the number of comparisons is a convenient proxy for our unit of cost

# Cost of Selection Sort

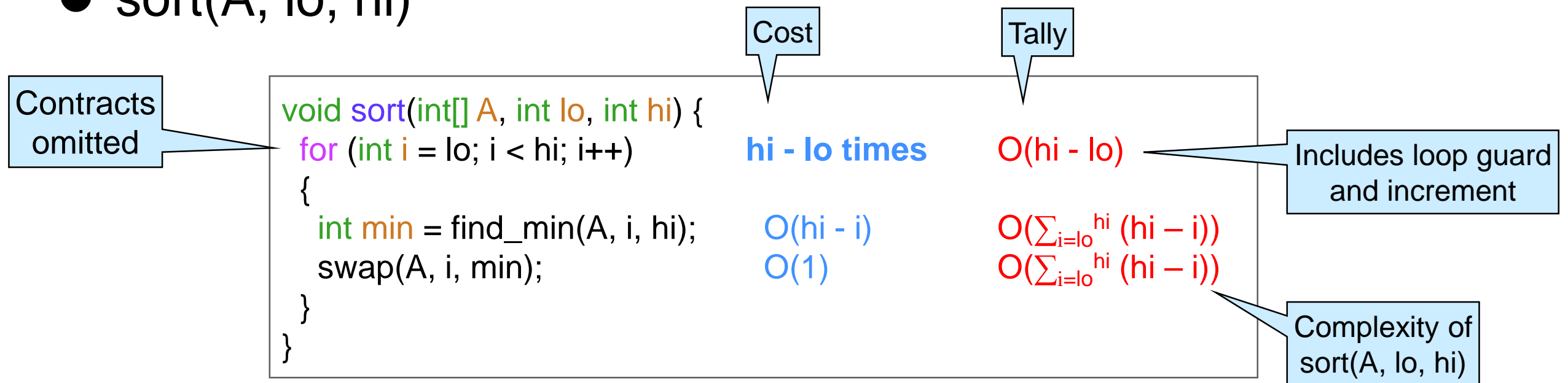
- `swap(A, i, j)`
  - simply swaps values at two indices:



Cost:  **$O(1)$**

# Selection Sort

## ● sort(A, lo, hi)



○ Let  $n = hi - lo$

➤ the length of the segment A[lo,hi)

then  $\sum_{i=lo}^{hi} (hi - i) = \sum_{j=0}^n j$

Carl Friedrich Gauss came up with this formula when he was 9 years old



$$0 + \dots + n = \sum_{j=0}^n j = n(n+1)/2$$

But not in simplest and tightest form!

# Cost of Selection Sort

```
void sort(int[] A, int lo, int hi) {  
    for (int i = lo; i < hi; i++) {  
        int min = find_min(A, i, hi);  
        swap(A, i, min);  
    }  
}
```

- Assume the array segment  $A[lo, hi)$  has length  $n$
- Number of comparisons to sort an  $n$ -element array segment

$$n(n-1)/2$$

- $n(n-1)/2 \in O(n^2)$

Selection sort has cost in  $O(n^2)$

That's  $O((hi-lo)^2)$   
in terms of  $lo$  and  $hi$

# Selection Sort

```
void sort(int[] A, int lo, int hi)
//@requires 0 <= lo && lo <= hi && hi <= \length(A);
//@ensures is_sorted(A, lo, hi);
{
    for (int i = lo; i < hi; i++)
    {
        int min = find_min(A, i, hi);
        swap(A, i, min);
    }
}
```

# Is this Code Safe?

```
int find_min(int[] A, int lo, int hi)
//@requires 0 <= lo && lo < hi && hi <= \length(A);
/*@ensures ... @*/;
```

- find\_min(A, i, hi)

➤ **To show:**  $0 \leq i < hi \leq \text{length}(A)$

A.  $hi \leq \text{length}(A)$  by line 2

B.  $i < hi$  by line 5

C.  $0 \leq i$  *oops! we need the usual loop invariant*

➤ *//@loop\_invariant*  $lo \leq i$ ; then

a.  $0 \leq lo$  by line 2

b.  $lo \leq i$  by new LI

c.  $0 \leq i$  by math



```
1. void sort(int[] A, int lo, int hi)
2. //@requires 0 <= lo && lo <= hi && hi <= \length(A);
3. //@ensures is_sorted(A, lo, hi);
4. {
5.   for (int i = lo; i < hi; i++)
6.   {
7.     int min = find_min(A, i, hi);
8.     swap(A, i, min);
9.   }
10. }
```



# Is this Code Safe?

- `swap(A, i, min)`

➤ **To show:**  $0 \leq \text{min} < \text{length}(A)$

- |                                      |  |   |   |
|--------------------------------------|--|---|---|
| A. $0 \leq i$                        | by lines 2 and 6                           | } | $0 \leq \text{min}$ by math             |
| B. $i \leq \text{min}$               | by postconditions of <code>find_min</code> |   |   |
| C. $\text{min} < \text{hi}$          | by postconditions of <code>find_min</code> | } | $\text{min} < \text{length}(A)$ by math |
| D. $\text{hi} \leq \text{length}(A)$ | by line 2                                  |   |   |

➤ **To show:**  $0 \leq i$   
 $\&\& i < \text{length}(A)$   
(just proved for `find_min`)



```
int find_min(int[] A, int lo, int hi)
//@requires 0 <= lo && lo < hi && hi <= \length(A);
/*@ensures lo <= \result && \result < hi
    && le_seg(A[\result], A, lo, hi); @*/ ;

void swap(int[] A, int i, int j)
//@requires 0 <= i && i < \length(A);
//@requires 0 <= j && j < \length(A);
```

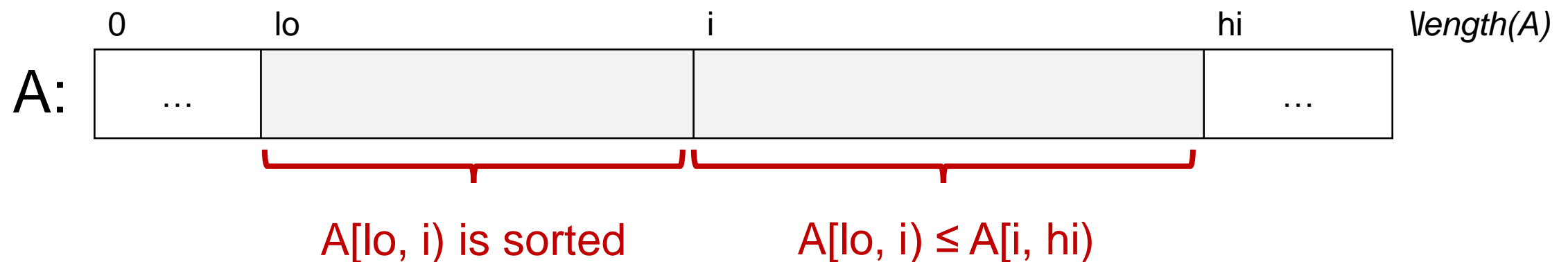
```
1. void sort(int[] A, int lo, int hi)
2. //@requires 0 <= lo && lo <= hi && hi <= \length(A);
3. //@ensures is_sorted(A, lo, hi);
4. {
5.     for (int i = lo; i < hi; i++)
6.         //@loop_invariant lo <= i;
7.         {
8.             int min = find_min(A, i, hi);
9.             swap(A, i, min);
10.        }
11. }
```

Added

# Is this Code Correct?

- To show:  $\text{is\_sorted}(A, \text{lo}, \text{hi})$
- What do we know at iteration  $i$ ?
  - Let's draw pictures!

```
void sort(int[] A, int lo, int hi)
//@requires 0 <= lo && lo <= hi && hi <= \length(A);
//@ensures is_sorted(A, lo, hi);
{
  for (int i = lo; i < hi; i++)
    //@loop_invariant lo <= i;
    {
      int min = find_min(A, i, hi);
      swap(A, i, min);
    }
}
```

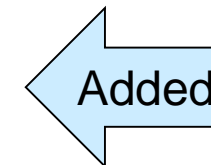


- Candidate loop invariants
  - $\text{lo} \leq i \ \&\& \ i \leq \text{hi}$
  - $\text{is\_sorted}(A, \text{lo}, i)$
  - $\text{le\_segs}(A, \text{lo}, i, A, i, \text{hi})$

# Selection Sort

- Resulting code

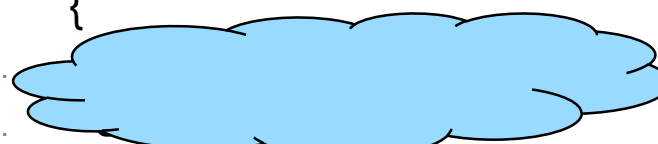
```
1. void sort(int[] A, int lo, int hi)
2. //@requires 0 <= lo && lo <= hi && hi <= \length(A);
3. //@ensures is_sorted(A, lo, hi);
4. {
5.     for (int i = lo; i < hi; i++)
6.         //@loop_invariant lo <= i && i <= hi;
7.         //@loop_invariant is_sorted(A, lo, i);
8.         //@loop_invariant le_segs(A, lo, i, A, i, hi);
9.         {
10.            int min = find_min(A, i, hi);
11.            swap(A, i, min);
12.        }
13. }
```



- We will need to prove that the added invariants are valid

# Correctness

➤ **To show:**  $\text{is\_sorted}(A, \text{lo}, \text{hi})$   
(assuming invariants are valid)

```
1. void sort(int[] A, int lo, int hi)
2. // @requires 0 <= lo && lo <= hi && hi <= \length(A);
3. // @ensures is_sorted(A, lo, hi);
4. {
5.     for (int i = lo; i < hi; i++)
6.         // @loop_invariant lo <= i && i <= hi;
7.         // @loop_invariant is_sorted(A, lo, i);
8.         // @loop_invariant le_segs(A, lo, i, A, i, hi);
9.     {
10.         
11.     }
12. }
13. }
```

- A.  $i \geq \text{hi}$  by line 5 (negation of loop guard)
- B.  $i \leq \text{hi}$  by line 6 (LI 1)
- C.  $i = \text{hi}$  by math on A, B
- D.  $\text{is\_sorted}(A, \text{lo}, \text{hi})$  by line 8 (LI 2) and C

○ This is a standard EXIT argument

➤ But are the loop invariants valid?

We didn't need LI 3  
 $A[\text{lo}, i] \leq A[i, \text{hi})$

# Selection Sort

Are the loop invariants valid?

## INIT

- To show:  $lo \leq lo$  by math
- To show:  $lo \leq hi$  by line 2 (preconditions)
- To show:  $A[lo, lo)$  sorted by math (empty interval)
- To show:  $A[lo, lo) \leq A[lo, hi)$  by math (empty interval)

## PRES

- To show: if  $lo \leq i \leq hi$ , then  $lo \leq i' \leq hi$

*Proof left as exercise*

```
1. void sort(int[] A, int lo, int hi)
2. //@requires 0 <= lo && lo <= hi && hi <= \length(A);
3. //@ensures is_sorted(A, lo, hi);
4. {
5.   for (int i = lo; i < hi; i++)
6.     //@loop_invariant lo <= i && i <= hi;
7.     //@loop_invariant is_sorted(A, lo, i);
8.     //@loop_invariant le_segs(A, lo, i, A, i, hi);
9.   {
10.    int min = find_min(A, i, hi);
11.    swap(A, i, min);
12.   }
13. }
```

# Selection Sort

Are the loop invariants valid?

**PRES**

➤ **To show:** if  $A[lo, i)$  is sorted,  
then  $A[lo, i')$  is sorted

```
1. void sort(int[] A, int lo, int hi)
2. //@requires 0 <= lo && lo <= hi && hi <= \length(A);
3. //@ensures is_sorted(A, lo, hi);
4. {
5.   for (int i = lo; i < hi; i++)
6.     //@loop_invariant lo <= i && i <= hi;
7.     //@loop_invariant is_sorted(A, lo, i);
8.     //@loop_invariant le_segs(A, lo, i, A, i, hi);
9.     {
10.      int min = find_min(A, i, hi);
11.      swap(A, i, min);
12.    }
13. }
```

A.  $i' = i + 1$

B.  $A[lo, i)$  is sorted

C.  $A[lo, i) \leq A[i, hi)$

D.  $A[lo, i) \leq A[i]$

E.  $A[lo, i')$  is sorted

by line 5 (step

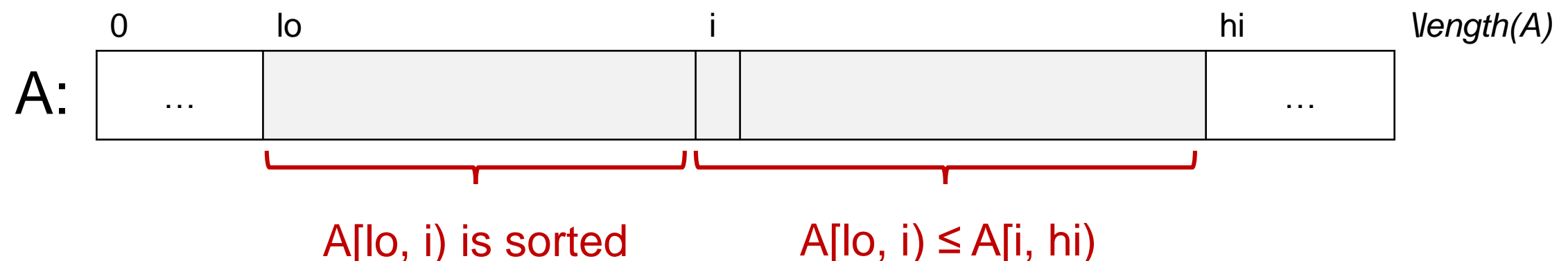
assumption

by line 8 (LI 3)

by math on C and line 5 (loop guard)

by math on A and D

This is where  
we need LI 3!



# Selection Sort

Are the loop invariants valid?

**PRES**

➤ **To show:** if  $A[lo, i) \leq A[i, hi)$ ,  
then  $A[lo, i') \leq A[i', hi)$

```

1. void sort(int[] A, int lo, int hi)
2. // @requires 0 <= lo && lo <= hi && hi <= \length(A);
3. // @ensures is_sorted(A, lo, hi);
4. {
5.   for (int i = lo; i < hi; i++)
6.     // @loop_invariant lo <= i && i <= hi;
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8.     // @loop_invariant le_segs(A, lo, i, A, i, hi);
9.     {
10.      int min = find_min(A, i, hi);
11.      swap(A, i, min);
12.    }
13. }

```

A.  $i' = i+1$

by line 5

B.  $A[lo, i) \leq A[i, hi)$

assumption

C.  $A[min] \leq A[i, hi)$

by postcondition of `find_min`

D.  $A[i] \leq A[i, hi)$

**after** `swap` by definition (in comment)

E.  $A[i] \leq A[i+1, hi)$

by math

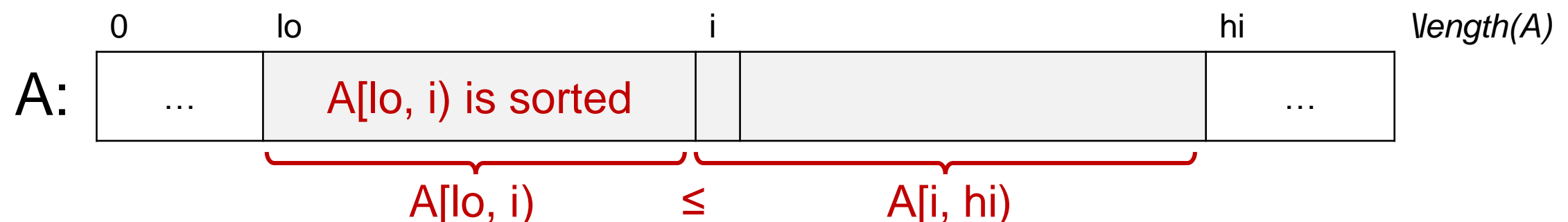
F.  $A[lo, i) \leq A[i]$

by math on B and definition of `swap`

G.  $A[lo, i') \leq A[i', hi)$

by math on A, F and E

after  
**swap**



# Selection Sort

```
1. void sort(int[] A, int lo, int hi)
2. //@requires 0 <= lo && lo <= hi && hi <= \length(A);
3. //@ensures is_sorted(A, lo, hi);
4. {
5.     for (int i = lo; i < hi; i++)
6.         //@loop_invariant lo <= i && i <= hi;
7.         //@loop_invariant is_sorted(A, lo, i);
8.         //@loop_invariant le_segs(A, lo, i, A, i, hi);
9.         {
10.            int min = find_min(A, i, hi);
11.            swap(A, i, min);
12.        }
13. }
```

- We have proved it correct

