

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

The diagram illustrates a terminal window with the following content:

```
# cc0 search.c0 search-time.c0
# time ./a.out -r 200 -n 1000000
0
real 0.83
user 0.82
sys 0.00
```

Callouts and annotations:

- Compiling with contracts disabled**: Points to the `cc0` compiler command.
- The `search` function**: Points to the `search.c0` file.
- `main` that generates random tests and runs them**: Points to the `search-time.c0` file.
- Runs a.out 200 times on arrays of size 1000000**: Points to the `./a.out -r 200 -n 1000000` command.
- Unix utility that times execution**: Points to the `time` command.
- Execution took 0.82 seconds**: Points to the `user 0.82` output line, which is circled in red.

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

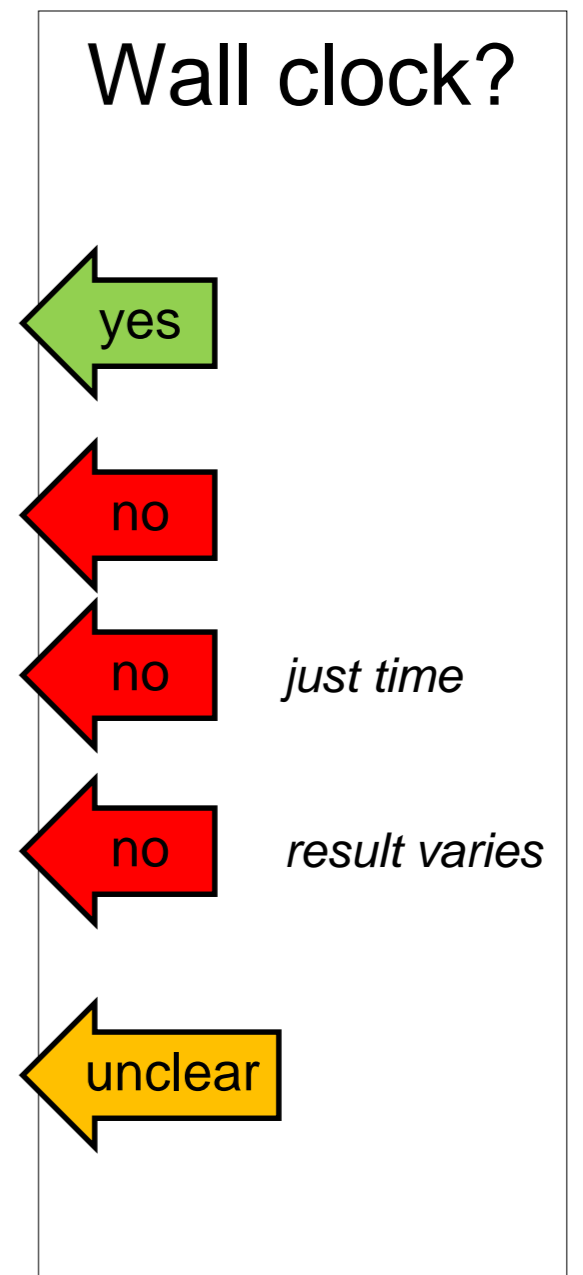
- **general**

- 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

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

- n is a **measure of the input** of the function
- Let's call the (upper bound on the) number of steps $T(n)$

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

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

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

- 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,

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

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

• `i = 0`
• `return -1`
• `i < n`

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

• `i < n`
• `if (A[i] == x)`
• `i++`

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” as 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
- $an+b$

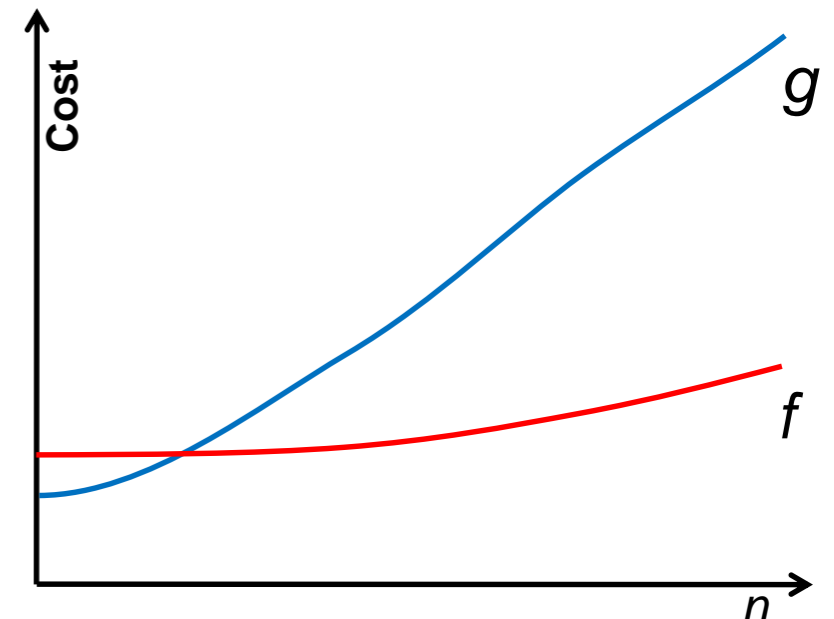
Our new math will need to deal with this

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

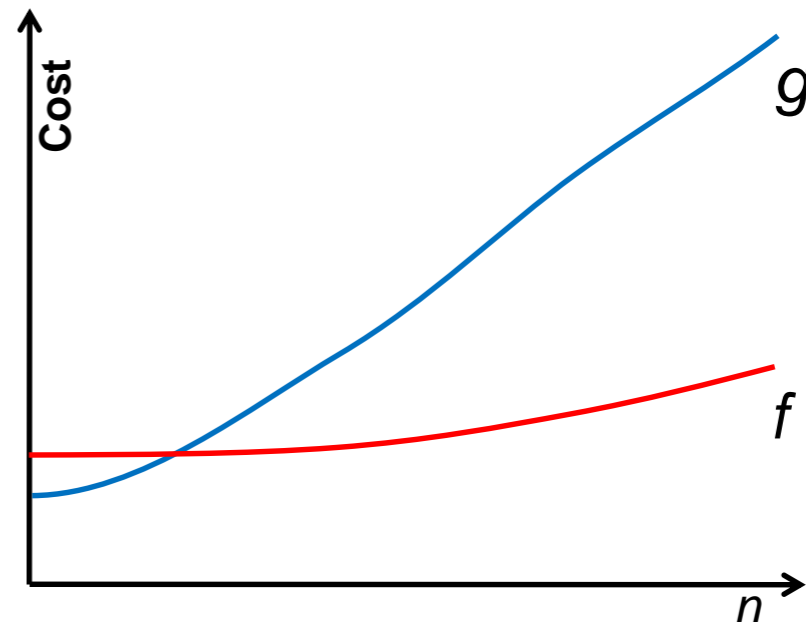
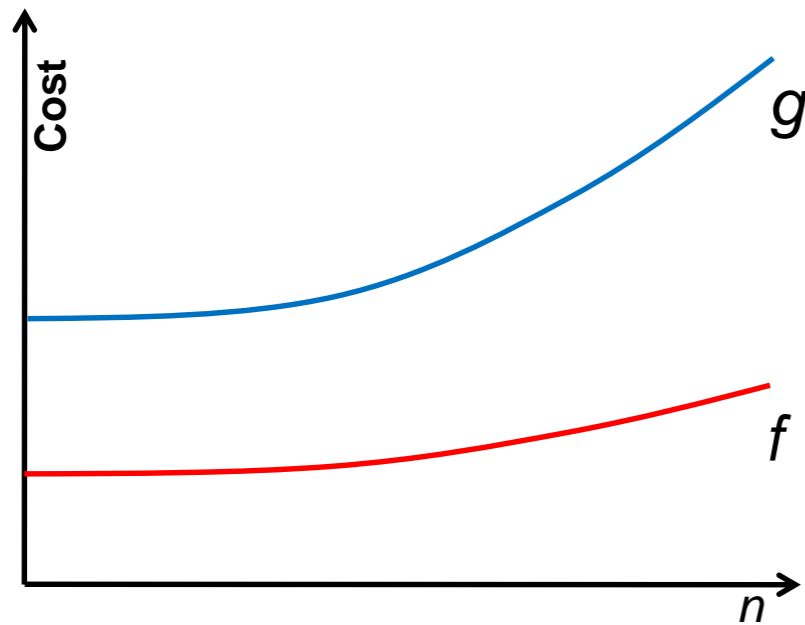


“Is f better than g ?”

- 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?

- f is initially worse than g

- but f is better beyond a certain point

- For small inputs, both costs are low

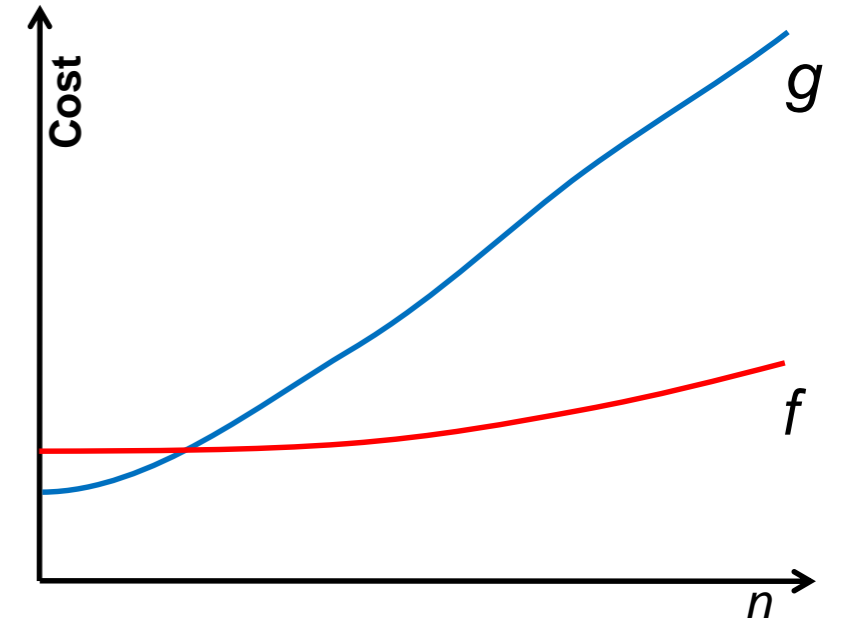
 - 0.12 ms vs. 0.23 ms doesn't matter for most applications

- 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

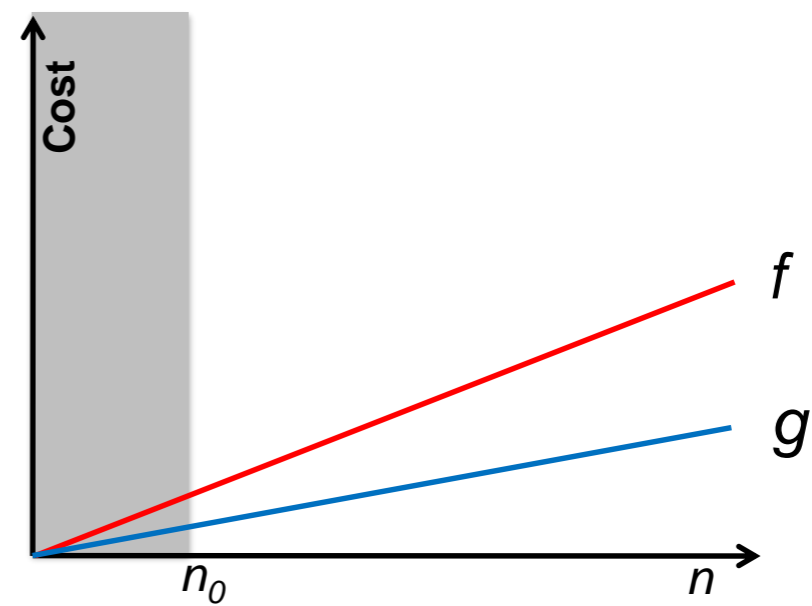
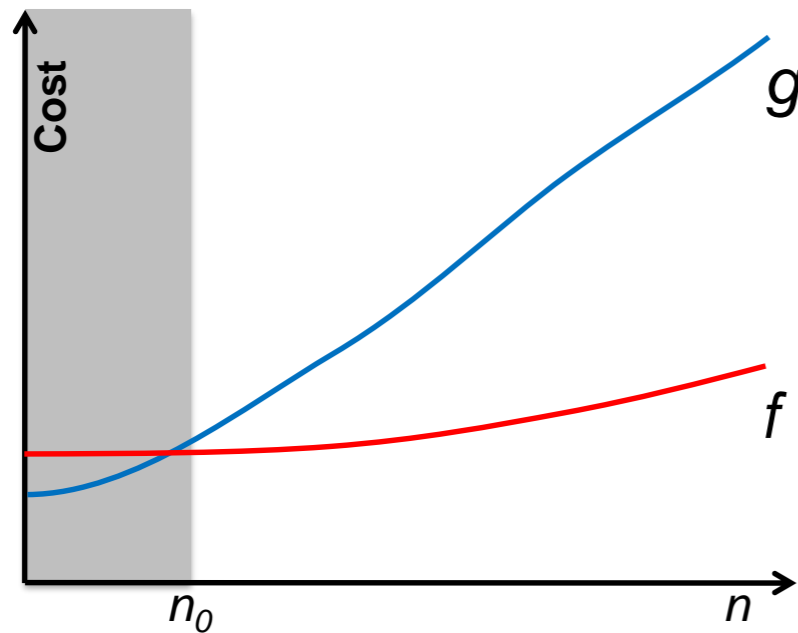


“Is f better than g ?”

- Attempt #2:

“ f is better than g ” if

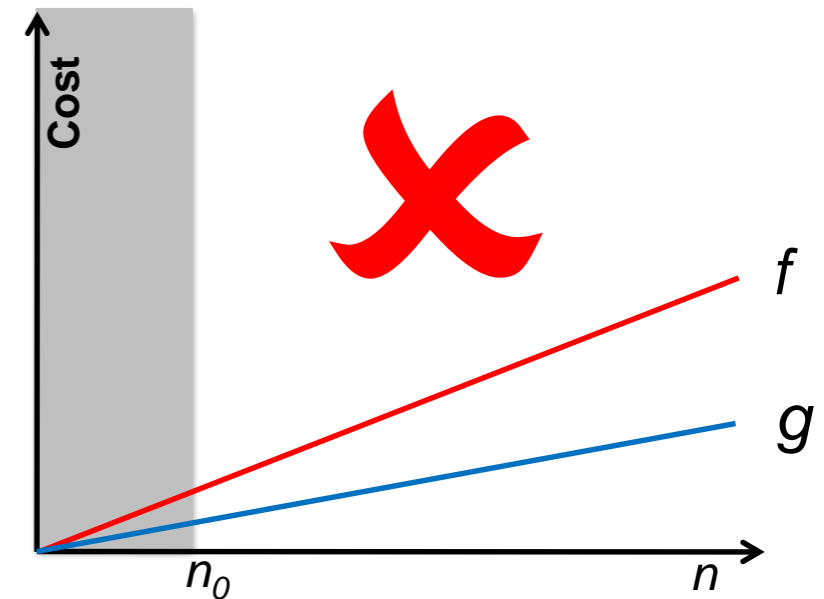
there exists a natural number n_0 such that
for all $n \geq n_0$, $f(n) \leq g(n)$



“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?

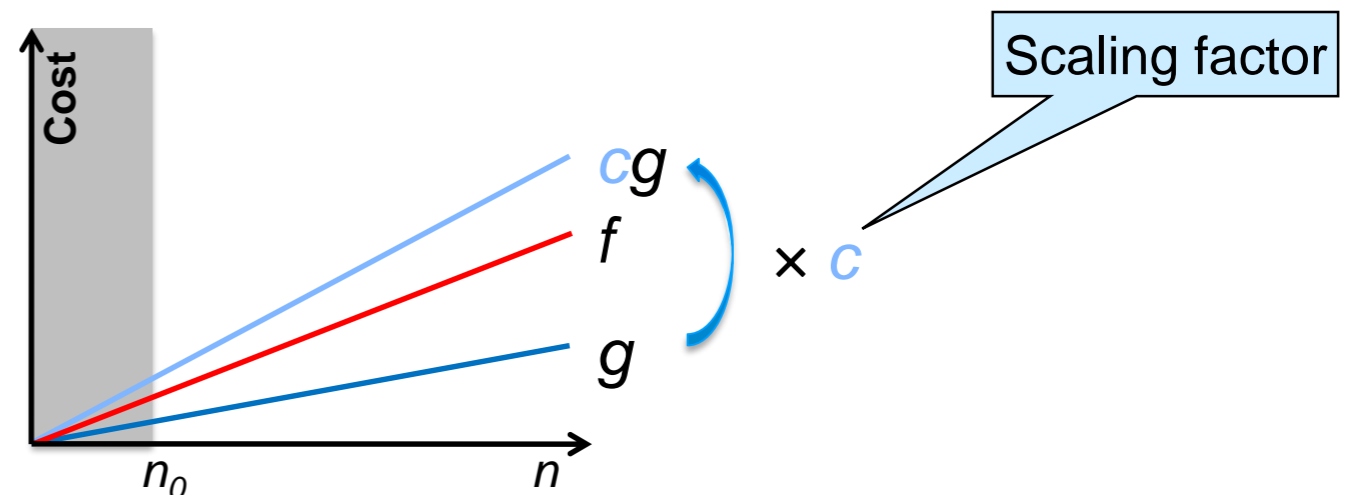
- f and g are both linear functions

- $f(n) = a_1n$ 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

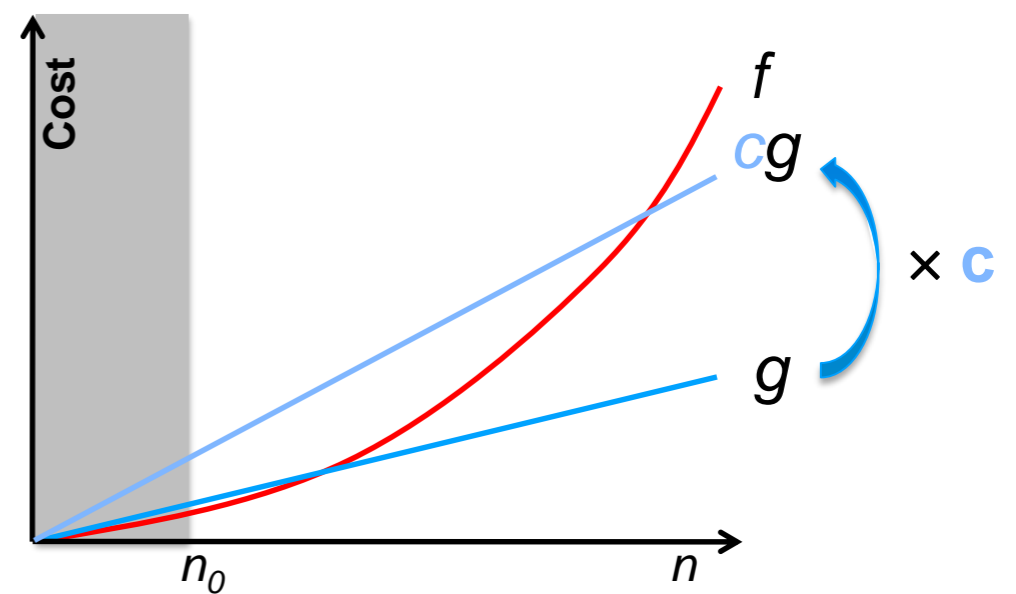
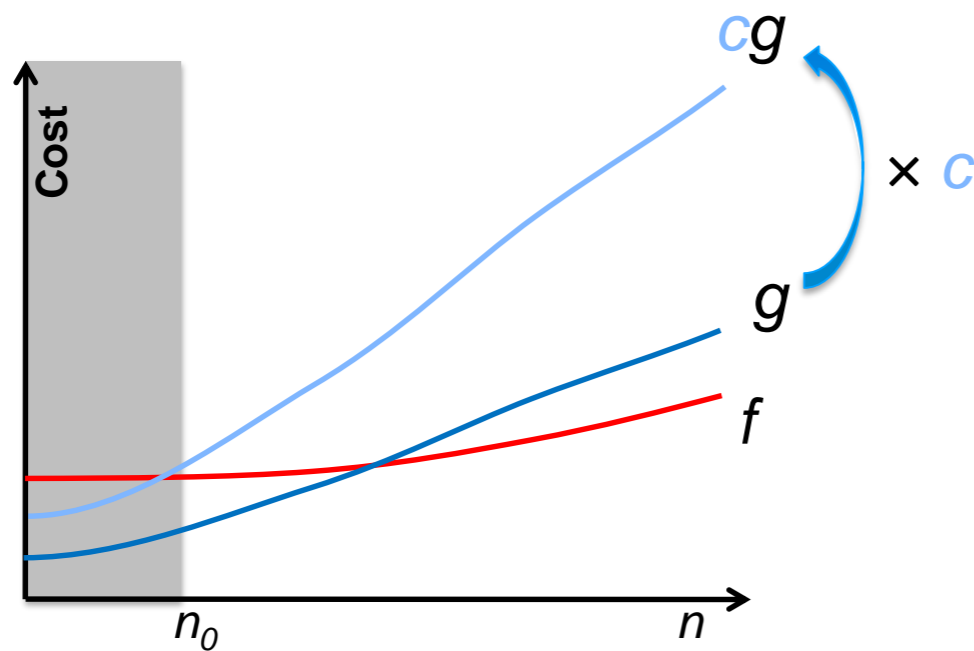


“Is f better than g ?”

- Final attempt:

“ f is better than g ” if

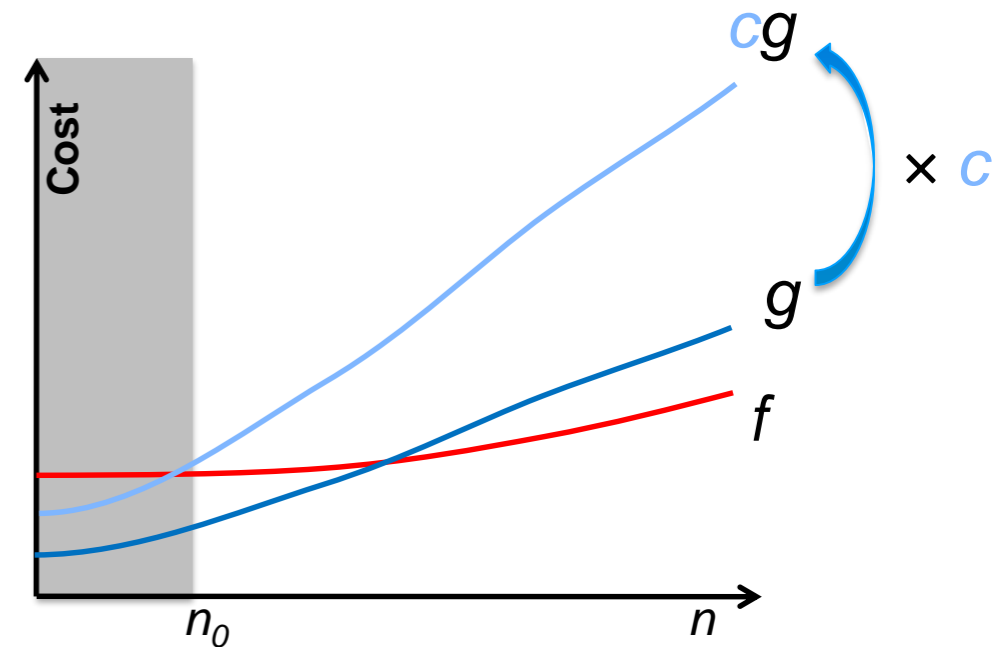
there exists a natural number n_0 and a real $c > 0$ s.t.
for all $n \geq n_0$, $f(n) \leq c g(n)$



Big O

Big-O

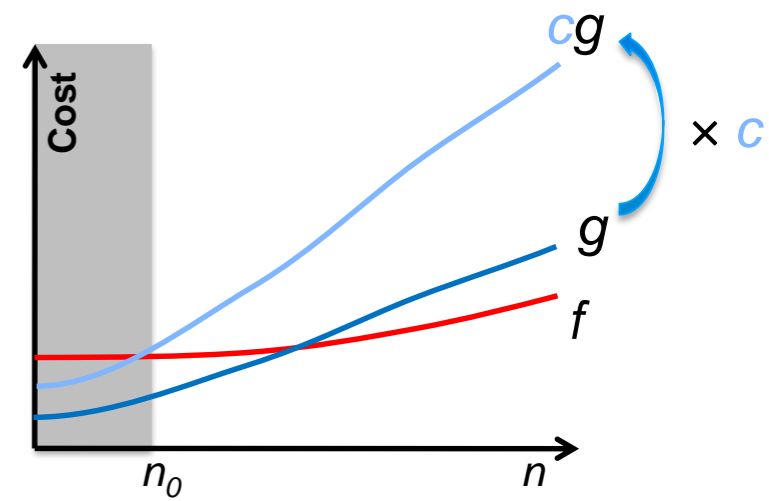
- Rather than “ f is better than g ”, we say $f \in O(g)$
 - “ f is in big-O of g ”



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

Big-O



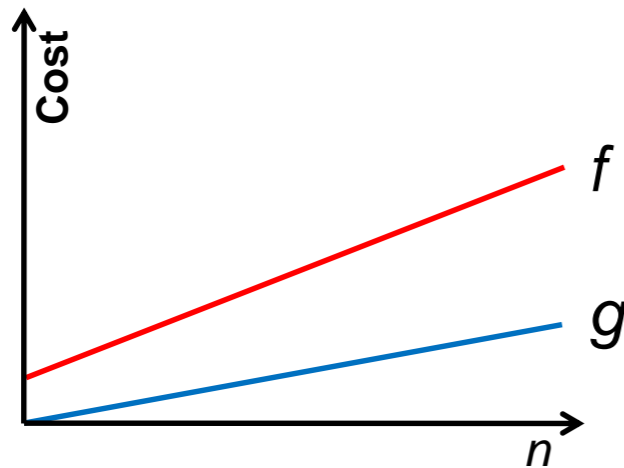
- Given two concrete functions f and g , how to tell if $f \in O(g)$?
 - do the math
 - find n_0 and c and show that the definition holds
 - or show that the definition cannot hold for any n_0 or c
 - recall what you learned in your calculus classes
 - enough for most of this course

Big-O

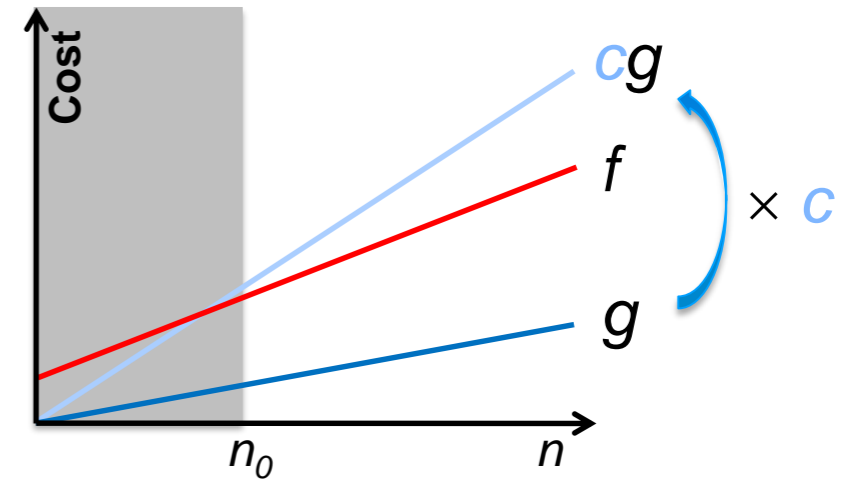
- Is $3n + 2 \in O(n)$?

$f(n) = 3n + 2$

$g(n) = n$



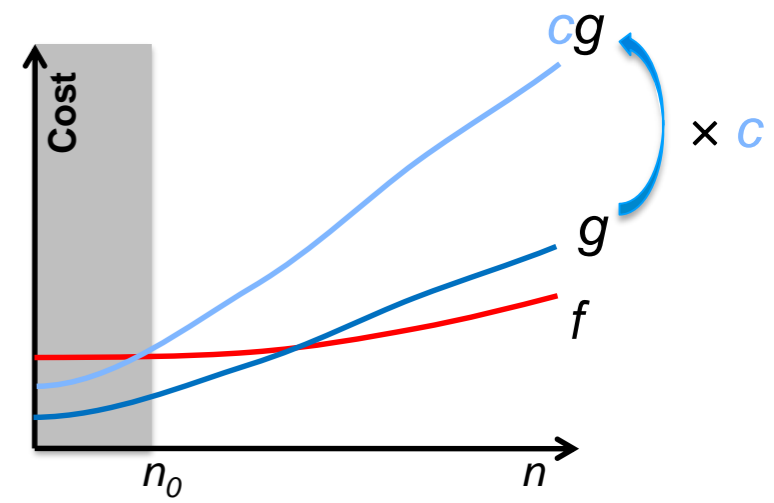
Take $c = 3.5$
and $n_0 = 4$



Any bigger values
work too!



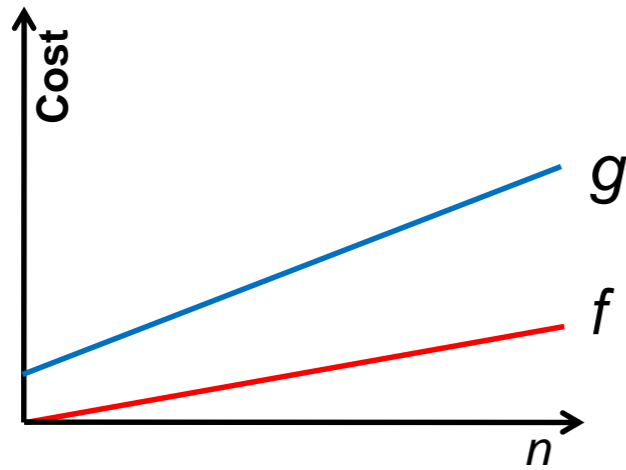
Big-O



● Is $n \in O(3n + 2)$?

$f(n) = n$

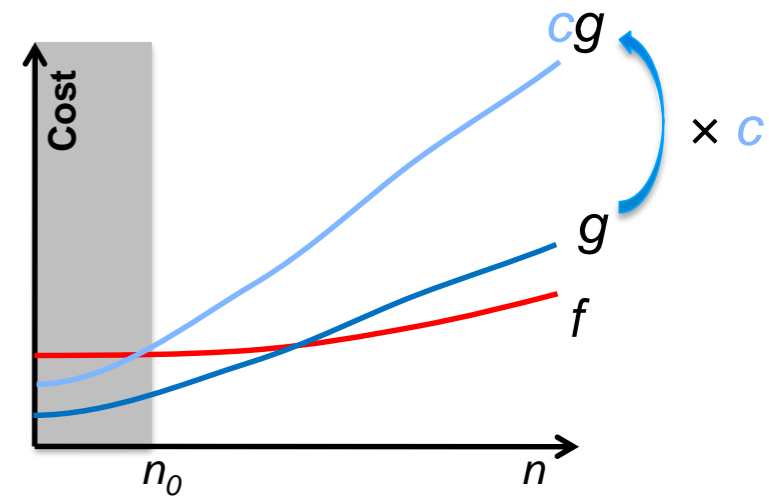
$g(n) = 3n + 2$



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

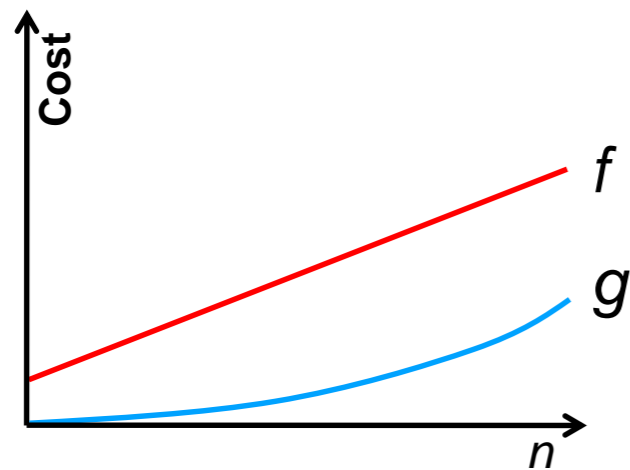
- Similarly, every **quadratic cost function** is in $O(n^2)$
 - $g(n) = n^2$ is the **simplest** quadratic function
- We describe a cost function that is quadratic by saying that it is in $O(n^2)$
- In general, every **polynomial cost function of degree p** is in $O(n^p)$
 - $g(n) = n^p$ is the **simplest** 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 $3n + 2 \in O(n^2)$?

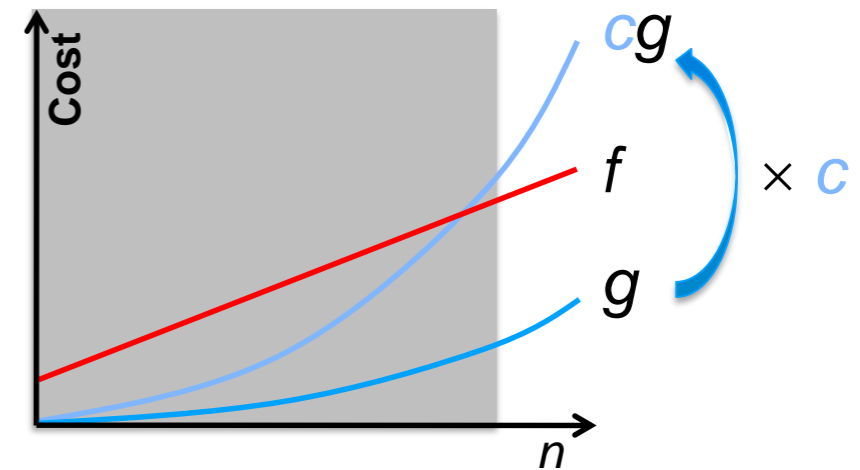
$$f(n) = 3n + 2$$

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



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 $p \leq q$
 - $O(n^p) \subseteq O(n^q)$

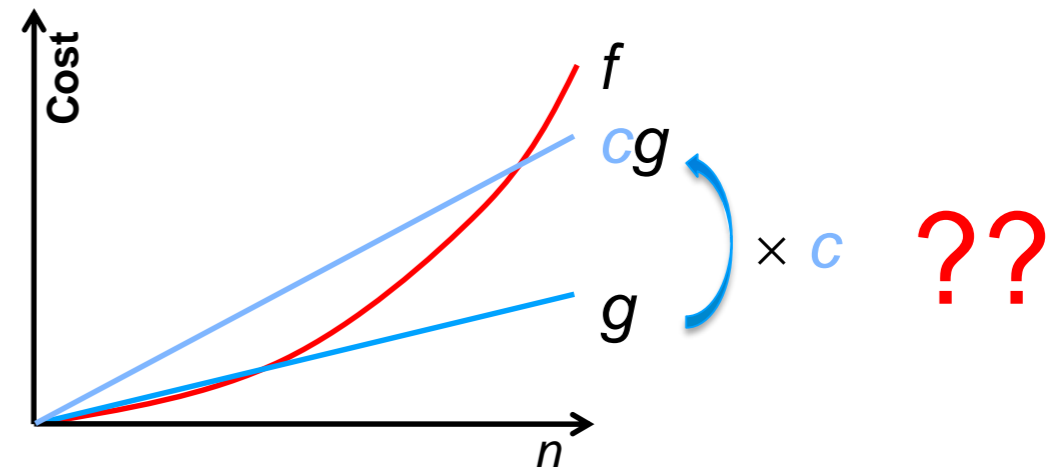
Big-O

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

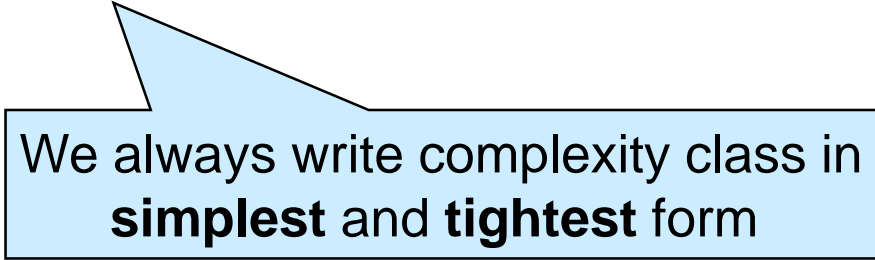


- Quadratic functions are **not** in $O(n)$
 - $O(n^2) \not\subset O(n)$
 - $O(n) \subset O(n^2)$

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

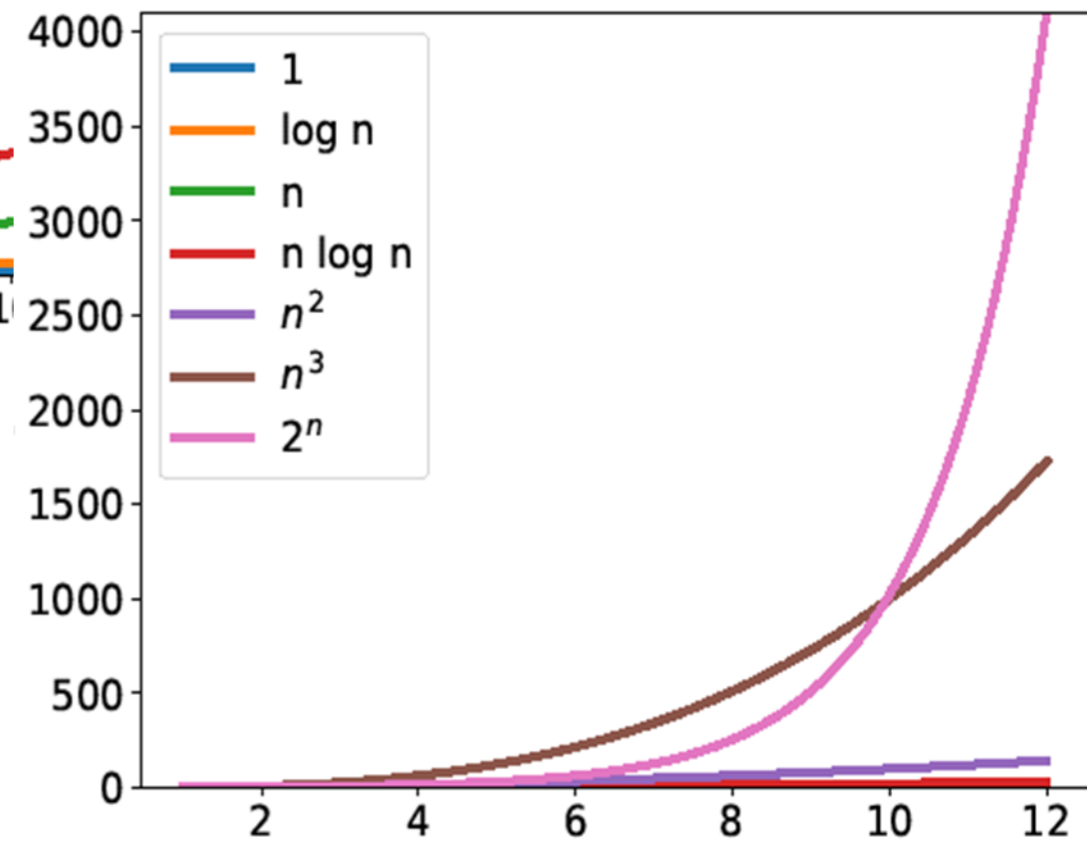
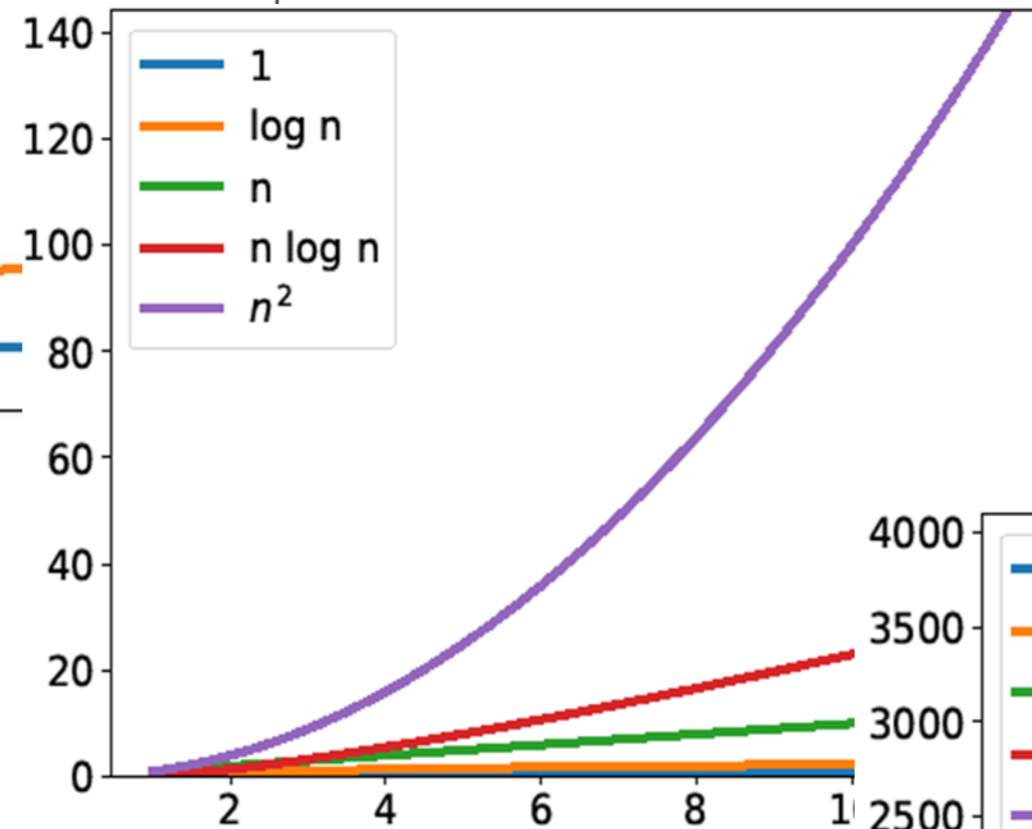
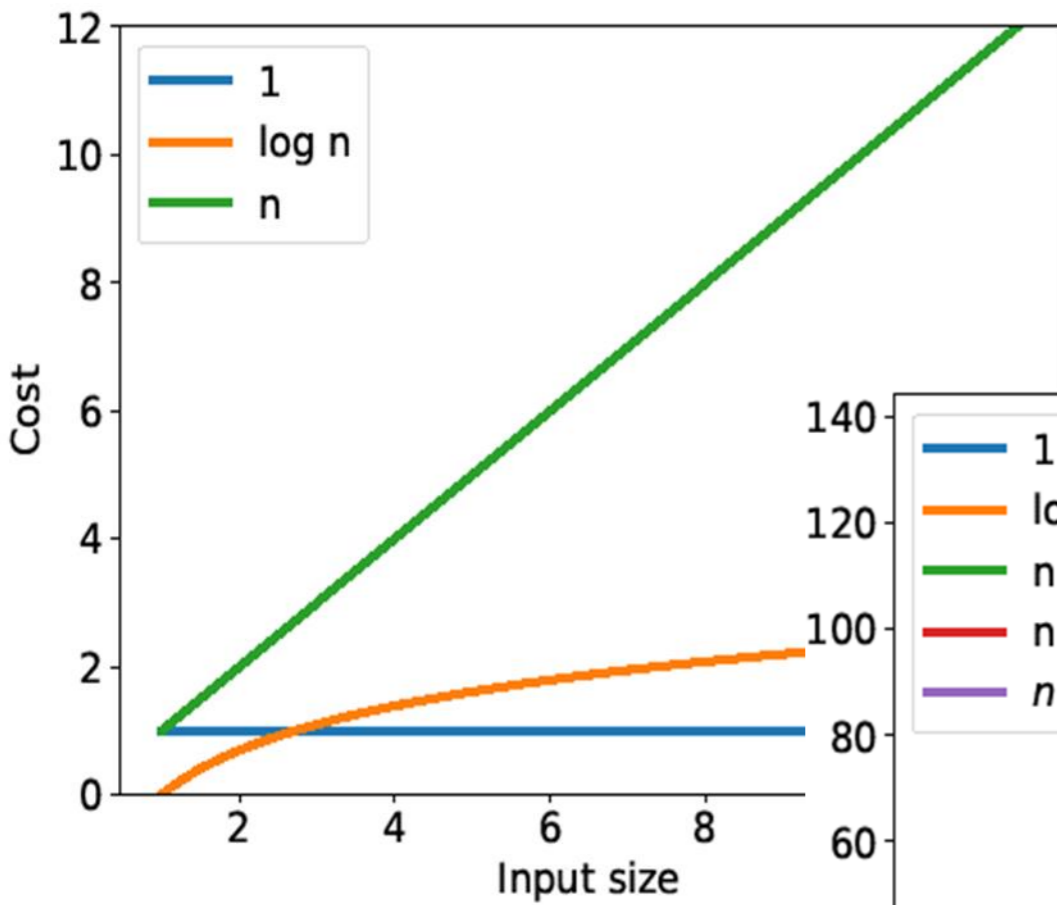
Complexity classes we will use in this course

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

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

- 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 < n; i++) { // at most</code>	n times		
<code> // i < 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>	$O(1)$	$O(n)$	
<code>}</code>			

Complexity of search(x, A, n)

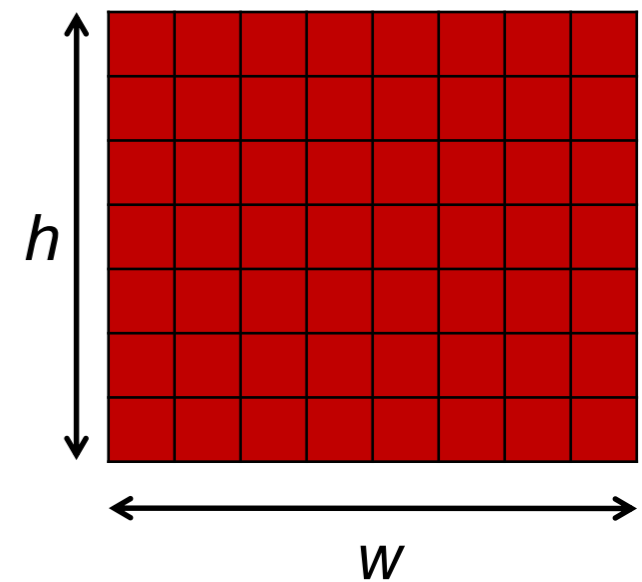
Big-O

- 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;  
}
```

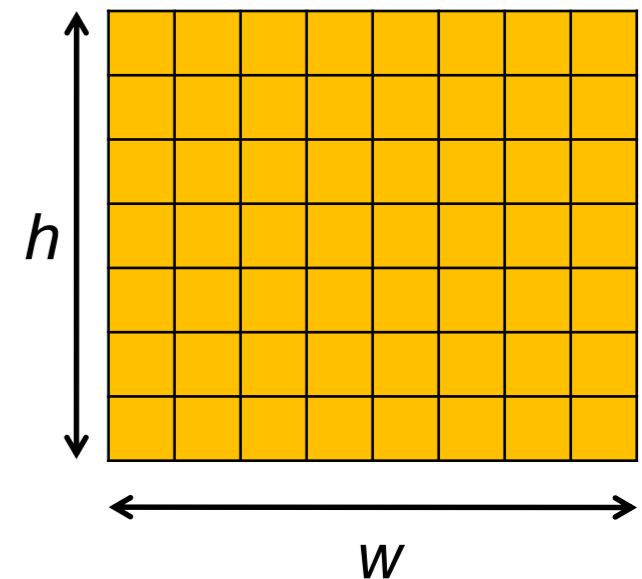
Big-O

An input may have multiple size parameters

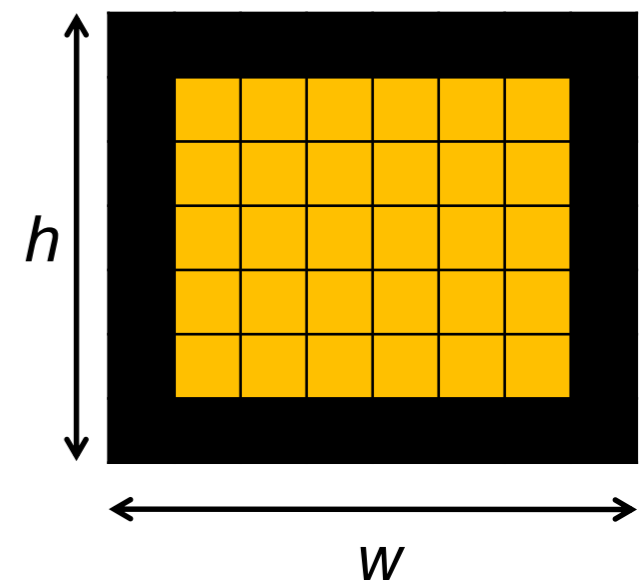


- For example, an image has a width w and a height h

- 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



Towards a Better Search

Algorithms vs. Problems

- Linear search has cost $O(n)$

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

- But this is only one of the many algorithms to search an array

- 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 if we make (reasonable) assumptions?

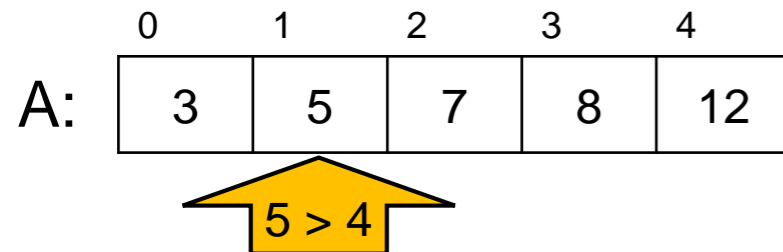
- Say, the array is **sorted**

Searching Sorted Arrays

A is sorted

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

x: 4



● 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;
}
```

If $A[i]$ is not equal to x and not larger than x then it must be smaller than x

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

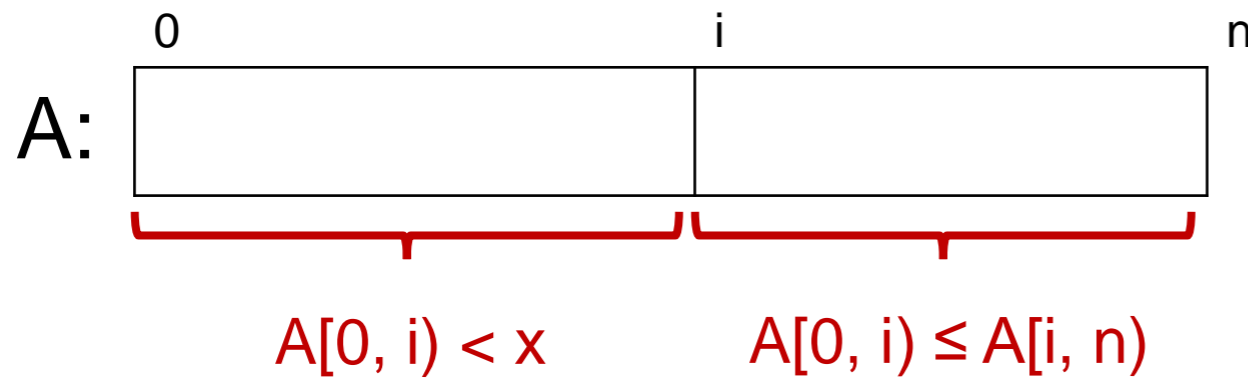
- Is it correct?

- 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)`

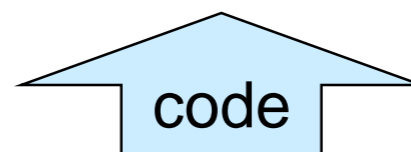
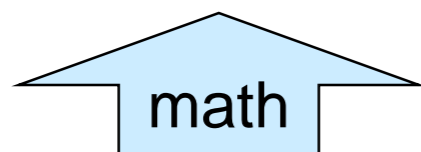
`ge_seg(x, A, lo, hi)`

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

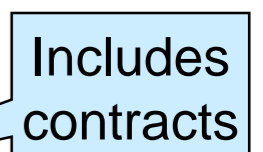
`le_segs(A, lo1, hi1, B, lo2, hi2)`

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

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

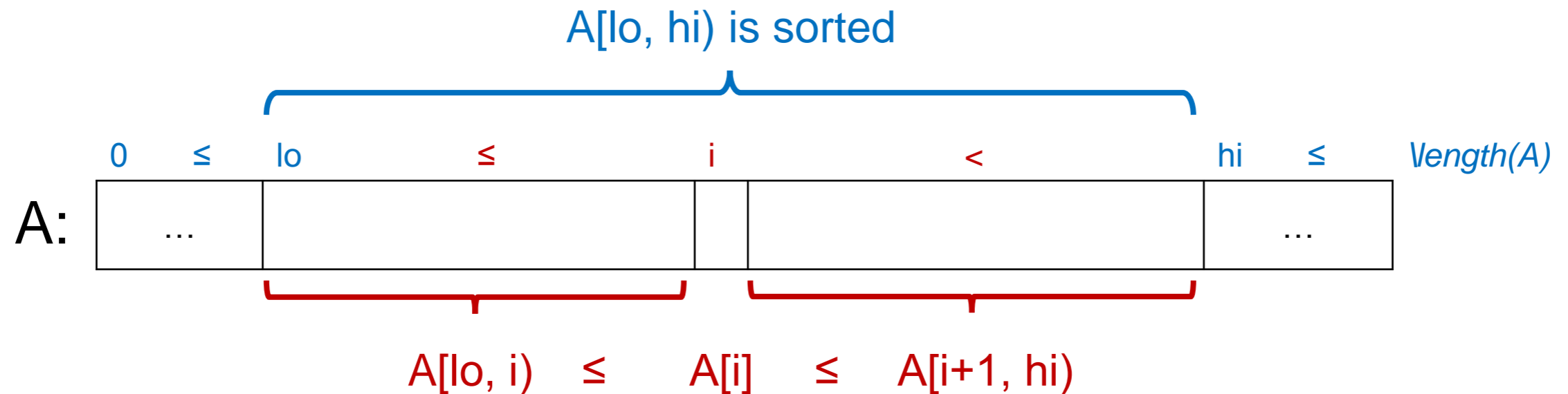


See **arrayutil.c0** file



Reasoning about Sorted Arrays

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



- for every element $A[i]$

- $lo \leq i < hi$

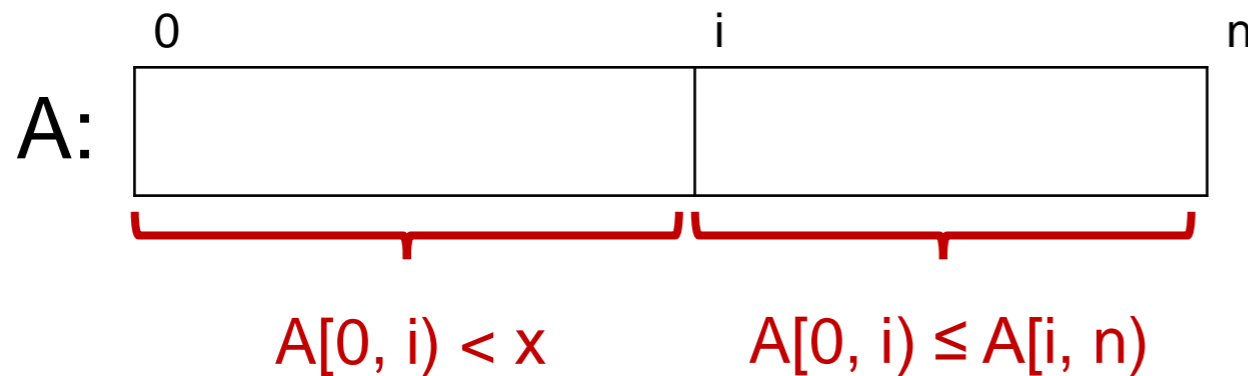
$A[lo, hi)$ can't be empty

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

- $A[lo, i)$ 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 $A[i,n)$

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

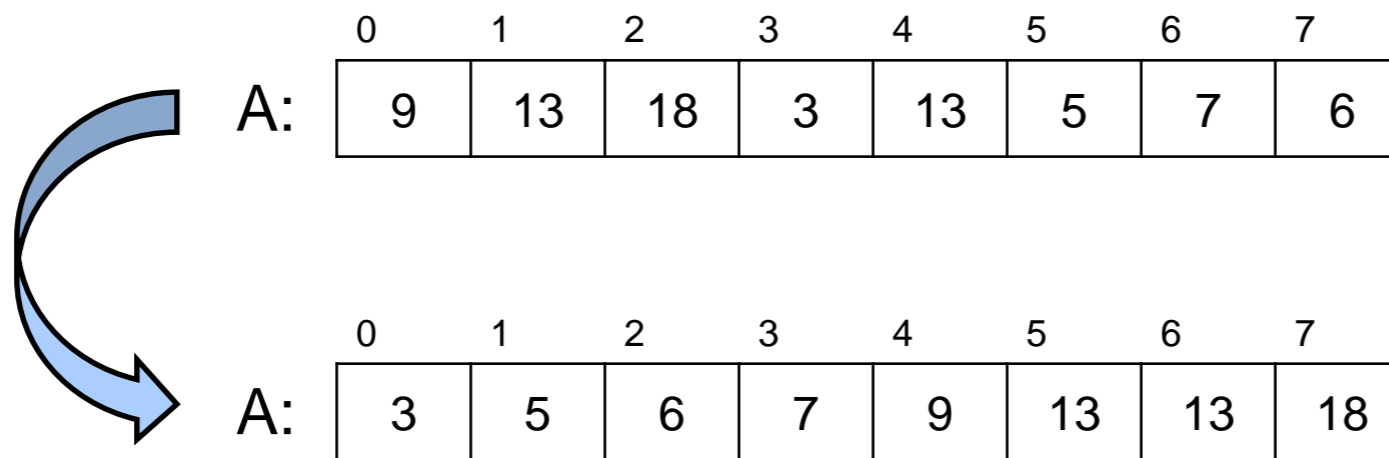
```
1. int search(int x, int[] A, int n)
2. //@requires n == \length(A);
3. //@requires is_sorted(A, 0, n);
4. /*@ensures (\result == -1 && !is_in(x, A, 0, n))
5.           || (0 <= \result && \result < n &&
6.              A[\result] == x);          @*/
7. {
8.   for (int i = 0; i < n; i++)
9.     //@loop_invariant 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. }
```

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

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

```
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); @*/ ;
```

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);
  }
}
```

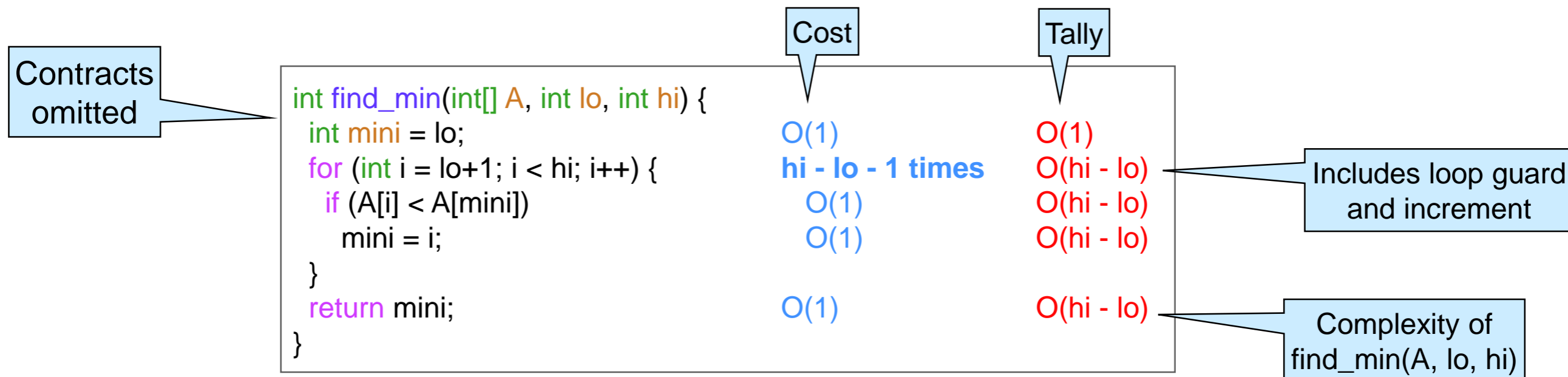
for every index i from lo to hi

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

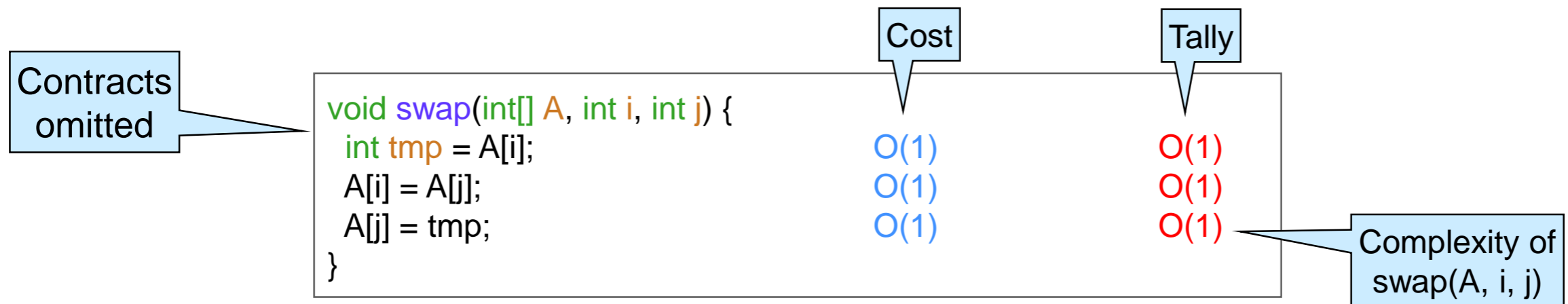


Cost: $O(hi - lo)$

- 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)`

Contracts omitted

```
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);
  }
}
```

Cost

$hi - lo$ times

$O(hi - i)$
 $O(1)$

Tally

$O(hi - lo)$

$O(\sum_{i=lo}^{hi} (hi - i))$
 $O(\sum_{i=lo}^{hi} (hi - i))$

Includes loop guard and increment

Complexity of `sort(A, lo, hi)`

But not in simplest and tightest form!

- 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



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

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)`

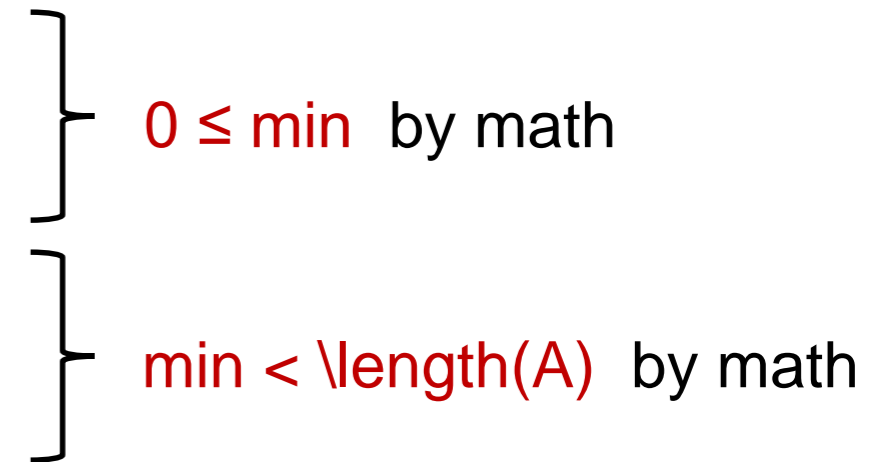
➤ **To show:** $0 \leq \text{min} < \text{length}(A)$

A. $0 \leq i$ by lines 2 and 6

B. $i \leq \text{min}$ by postconditions of `find_min`

C. $\text{min} < \text{hi}$ by postconditions of `find_min`

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



```
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);
```

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



```
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: $\text{is_sorted}(A, lo, hi)$
- 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);
    }
}
```

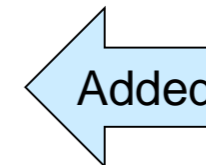


- Candidate loop invariants
 - $lo \leq i \ \&\& \ i \leq hi$
 - $\text{is_sorted}(A, lo, i)$
 - $\text{le_segs}(A, lo, i, A, i, hi)$

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:** `is_sorted(A, lo, 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. $i \geq hi$ by line 5 (negation of loop guard)
- B. $i \leq hi$ by line 6 (LI 1)
- C. $i = hi$ by math on A, B
- D. `is_sorted(A, lo, 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
 $A[lo, i] \leq A[i, hi]$

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

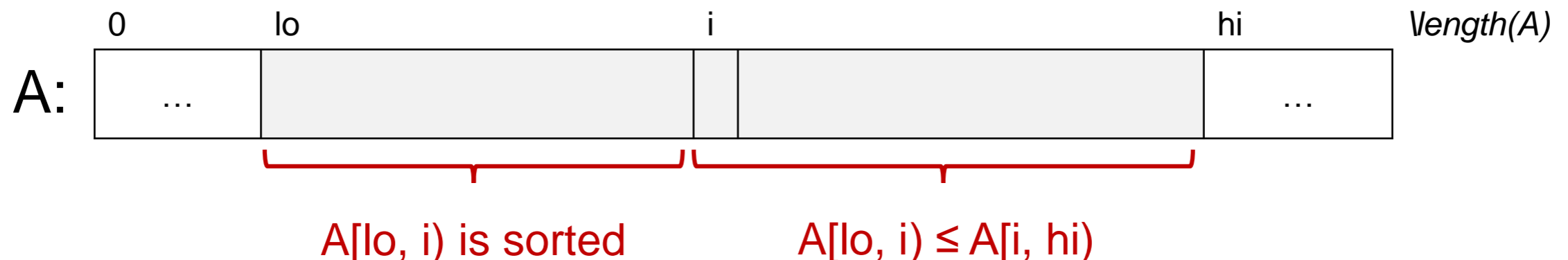
assumption

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;
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. $i' = i + 1$

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

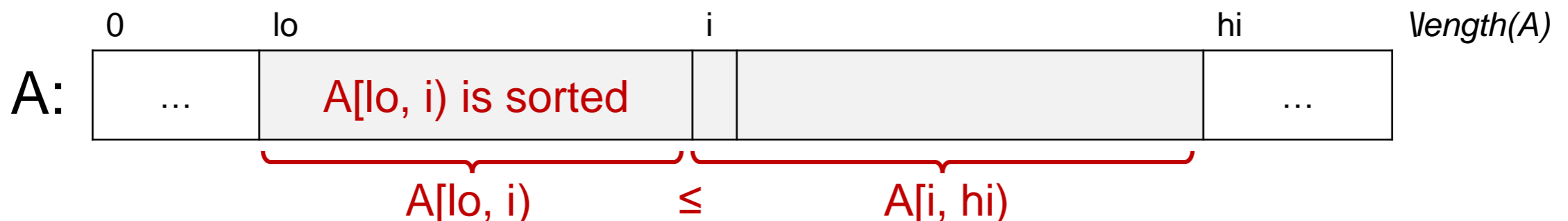
F. $A[lo, i) \leq A[i]$

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

