

Recall that in CS:

- a problem is defined precisely by its input + desired output
- a solution is an algorithm that takes any acceptable input and computes the desired output via a sequence of steps that a computer can perform, described in pseudocode that a human can understand.

Algorithms analysis involves

1. analyzing whether an alg. is correct - for all valid inputs, does it output what it should? (later courses)
2. analyzing the algorithm's runtime, so that we can
  - know how fast it is as input size grows
  - compare algs(later courses will also deal w/ other forms of computational complexity)

We analyze runtime by:  $\frac{n^2}{2} = O(n^2)$

1. Finding a function that counts the # of primitive operations (arithmetic ops, boolean ops, fetching variable values, etc) in terms of input
2. Using big O notation, bc we only care how alg scales + want to compare to other algs

ex for  $i=1$  to  $i=n$  do  
 for  $j=1$  to  $j=n$  do  
 for  $k=1$  to  $k=n$  do  
 sum = sum + 1  $O(1)$  ]  $\sum_{k=1}^n 1 = n$

$$f(n) = n^3 = O(n^3)$$

ex while  $n > 1$  do  
 $n = n - 1$  ]  $O(1)$  ]  $n$  times  
 do  $O(1)$ .

$$O(n)$$

ex while  $n > 1$  do  
 $n = n/2$  ]  $O(1)$  ]  $\log_2 n$  times  
 do  $O(1)$

Suppose  $n = 32 = 2^5$ .  $\log_2 32 = 5$

1st loop	$n = 16$	$\log_2 n$
2nd	$n = 8$	
3rd	$n = 4$	
4th	$n = 2$	
5th	$n = 1$	

$$O(\log n)$$

Problem: is  $x$  in array  $A$ ?

input:  $x, A$

output: T if  $\exists i: A[i]$  equals  $x$ ; F otherwise

ex  $x = 5, A = \langle 4, 3, 2, 10 \rangle$  F

$x = 5, A = \langle 4, 3, 2, 10, 5 \rangle$  T b.c.  $A[5] = 5$   
 1 2 3 4 5

$\leftarrow$  indexed from 1 to n  
**linearSearch**( $A[1 \dots n], x$ ):

**Input:** an array  $A[1 \dots n]$  and an element  $x$

**Output:** is  $x$  in the (possibly unsorted) array  $A$ ?

1 **for**  $i := 1$  to  $n$ :

2     **if**  $A[i] = x$  **then**

3         **return** True

4 **return** False

$$\left. \begin{array}{l} 1 \\ 2 \\ 3 \end{array} \right\} O(1) \left. \begin{array}{l} \\ \\ \end{array} \right\} \sum_{i=1}^n 1 = n = \underline{O(n)}$$

if  $A[i] = x$ , 1 time do  $O(1)$  prim. ops.

if  $\exists i: A[i] = x$ ,  $n$  times do  $O(1)$  prim. ops.

runtime depends on not just  $n$ , but what the input is!

When runtime depends on input, we could be:

1. optimistic — best case

2. pessimistic — worst case

3. neither — average case

easier to define

guarantee for all inputs

$O(\cdot)$  works for all

Def Worst-case runtime of an alg. is

max  
over all possible  
input of size  $n$

$T(n)$ , where  $T(n)$  is  
RT of alg on  
input of size  $n$ .

# Solution #2:

**binarySearch**( $A[1 \dots n], x$ ):

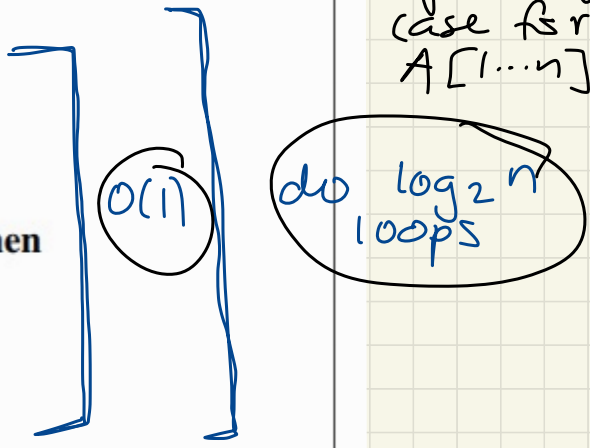
**Input:** a sorted array  $A[1 \dots n]$ ; an element  $x$

**Output:** is  $x$  in the (sorted) array  $A$ ?

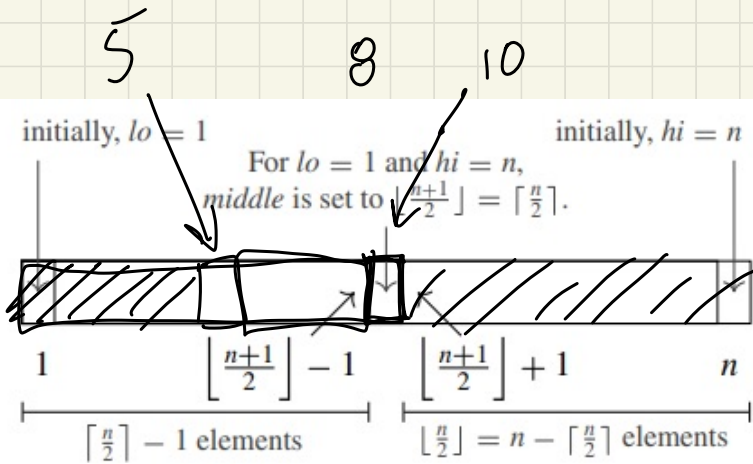
```

1 lo := 1
2 hi := n
3 while lo ≤ hi:
4     middle := ⌊ (lo+hi) / 2 ⌋
5     if A[middle] = x then
6         return True
7     else if A[middle] ≥ x then
8         hi := middle - 1
9     else
10        lo := middle + 1
11 return False
    
```

① give  $f$  expressing # of prim. ops in worst case for  $A[1 \dots n]$ .



Worst case input:  $\exists i : A[i] = x$



$$f = 3 + (\log_2 n) 20 = O(\log n)$$