

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:

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

Problem: is x in array A ?

input: x, A

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

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

$x=5, A = \langle 4, 3, 5, 10, 5 \rangle$. output: T bc $A[3] = 5$

Solution 1:

notation to say that the array is indexed 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

$::=$ is assignment

$O(1)$ $\sum_{i=1}^n 1 = n$

But if $A[i] = x$, then linear search has a runtime of $O(1)$.

So runtime depends not just on input size, but what the input is!

When runtime depends on the specific input, we could be

1. optimistic — best case
2. pessimistic — worst case
3. neither — avg case

easier to define
← guarantee for
all inputs
← $O(\cdot)$ works for all

Def Worst-case RT of an alg is $\max_{\text{over all possible inputs of size } n} T(n)$, where $T(n)$ is RT of alg on input of size n

intuitively: what is the input that makes the runtime as bad as possible?

ex for linear search, worst-case input is $A[1..n]$ where $\forall i \in \{1, 2, \dots, n\}$ s.t. $A[i] \neq x$.

So worst-case runtime is $\mathcal{O}(n)$.

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 := ⌊ $\frac{lo+hi}{2}$ ⌋
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
```

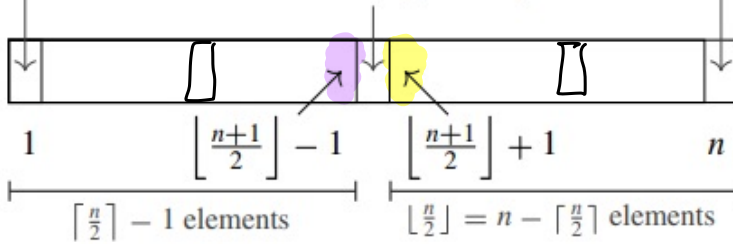
$\mathcal{O}(1)$ how many times?

It's not obvious how many times the loop executes.

initially, $lo = 1$

initially, $hi = n$

For $lo = 1$ and $hi = n$,
middle is set to $\lfloor \frac{n+1}{2} \rfloor = \lceil \frac{n}{2} \rceil$.



First loop: considering $\{i \in \mathbb{Z} : 1 \leq i \leq n\}$

Second loop: $\{i : lo \leq i \leq hi\}$, and its size is halved

⋮

$|\{i : lo \leq i \leq hi\}|$ gets halved each iteration.

we can only do $\log_2 n$ halvings before the size of the set is 1.

