

## 10.8

# Binary Search

# Binary Search in Sorted Arrays

**Input** Sorted array  $A$  of  $n$  numbers and number  $x$

**Goal** Is  $x$  in  $A$ ?

```
BinarySearch( $A[a..b]$ ,  $x$ ):  
    if ( $b - a < 0$ ) return NO  
     $mid = A[\lfloor (a + b)/2 \rfloor]$   
    if ( $x = mid$ ) return YES  
    if ( $x < mid$ )  
        return BinarySearch( $A[a..\lfloor (a + b)/2 \rfloor - 1]$ ,  $x$ )  
    else  
        return BinarySearch( $A[\lfloor (a + b)/2 \rfloor + 1..b]$ ,  $x$ )
```

Analysis:  $T(n) = T(\lfloor n/2 \rfloor) + O(1)$ .  $T(n) = O(\log n)$ .

**Observation:** After  $k$  steps, size of array left is  $n/2^k$

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# Another common use of binary search

- 1 **Optimization version:** find solution of best (say minimum) value
- 2 **Decision version:** is there a solution of value at most a given value  $v$ ?

Reduce optimization to decision (may be easier to think about):

- 1 Given instance  $I$  compute upper bound  $U(I)$  on best value
- 2 Compute lower bound  $L(I)$  on best value
- 3 Do binary search on interval  $[L(I), U(I)]$  using decision version as black box
- 4  $O(\log(U(I) - L(I)))$  calls to decision version if  $U(I), L(I)$  are integers

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# Example

- ① **Problem:** shortest paths in a graph.
- ② **Decision version:** given  $G$  with non-negative integer edge lengths, nodes  $s, t$  and bound  $B$ , is there an  $s$ - $t$  path in  $G$  of length at most  $B$ ?
- ③ **Optimization version:** find the length of a shortest path between  $s$  and  $t$  in  $G$ .

**Question:** given a black box algorithm for the decision version, can we obtain an algorithm for the optimization version?

# Example continued

**Question:** given a black box algorithm for the decision version, can we obtain an algorithm for the optimization version?

- 1 Let  $U$  be maximum edge length in  $G$ .
- 2 Minimum edge length is  $L$ .
- 3  $s$ - $t$  shortest path length is at most  $(n - 1)U$  and at least  $L$ .
- 4 Apply binary search on the interval  $[L, (n - 1)U]$  via the algorithm for the decision problem.
- 5  $O(\log((n - 1)U - L))$  calls to the decision problem algorithm sufficient. Polynomial in input size.



# THE END

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# (for now)