

Data Structures and Algorithms

Khazhak Galstyan

Session 4:

Binary Search

Searching Problem

The Searching Problem (in a sorted array)

INPUT: a **sorted** list of n elements and **a single value**

1	3	3	3	4	6	6	8
---	---	---	---	---	---	---	---

6

The Searching Problem (in a sorted array)

INPUT: a **sorted** list of n elements and **a single value**

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OUTPUT: where is (or where would be) given value in a given list

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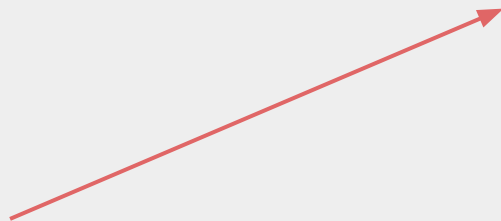
The Searching Problem (in a sorted array)

INPUT: a **sorted** list of n elements and a **single value**

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---	---	---	---	---

6	6	8
---	---	---

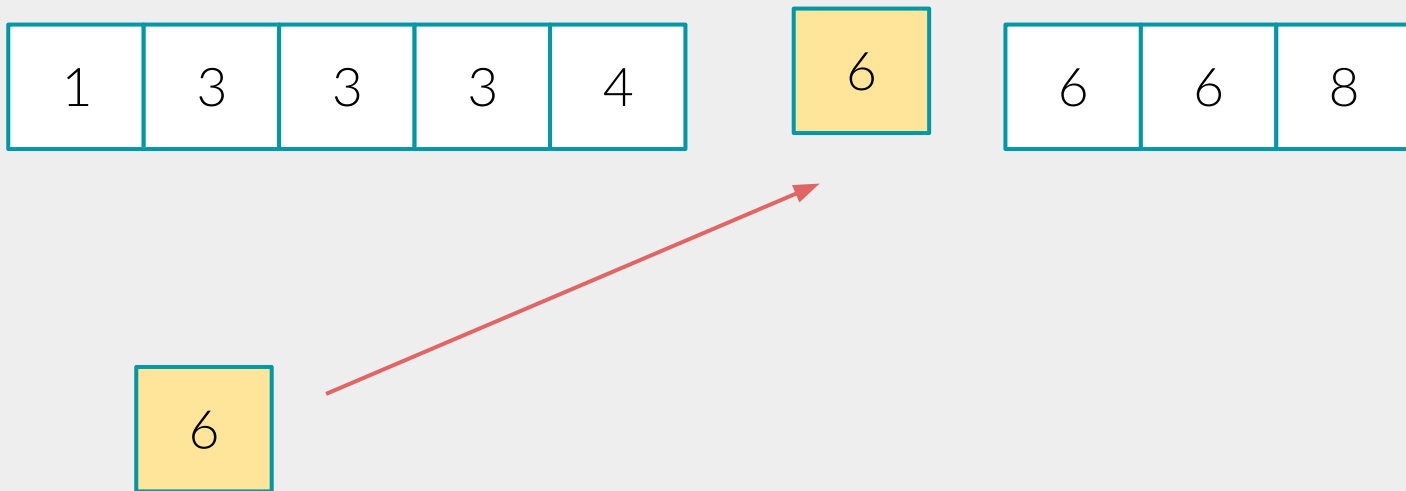
6



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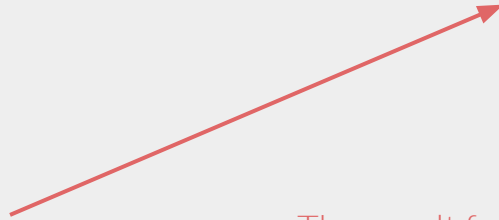
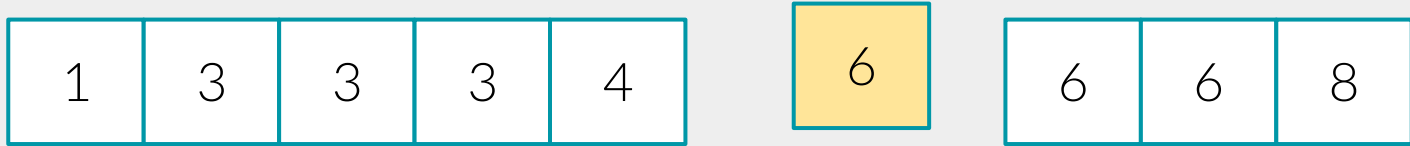
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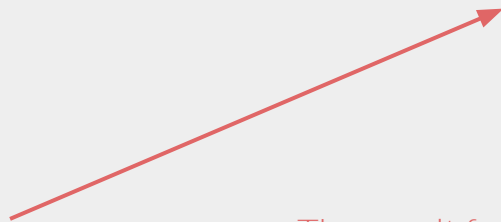
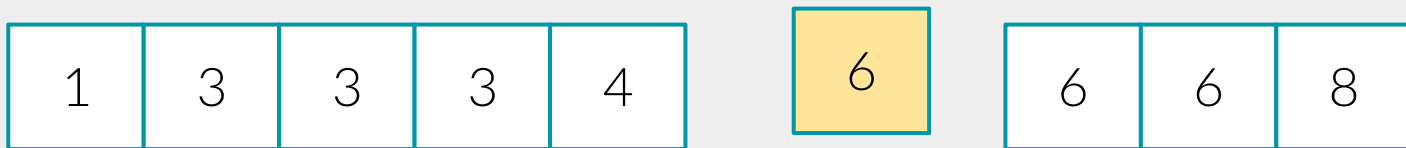


The result for this example will be **5**,
as '**6**' is **5th (0 based)** in the list.

OUTPUT: where is (or where would be) given value in a given list

The Searching Problem (in a sorted array)

INPUT: a **sorted** list of n elements and a **single value**



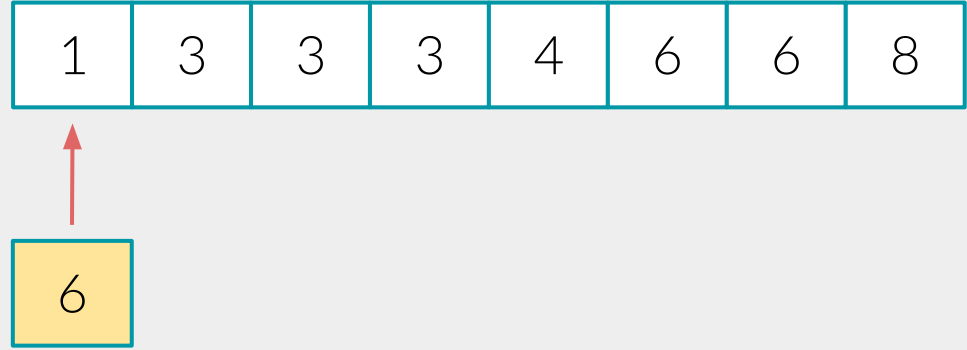
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OUTPUT: where is (or where would be) given value in a given list*

* For equal elements take the smallest possible index (leftmost available position)

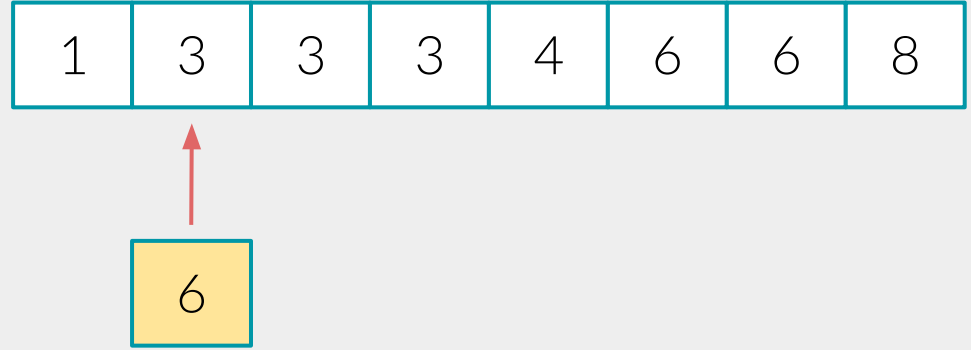
The Searching Problem: Iterative Solution

```
search(A, x):  
    n = length of A  
    i = 0  
    while i < n and A[i] < x:  
        i += 1  
    return i
```



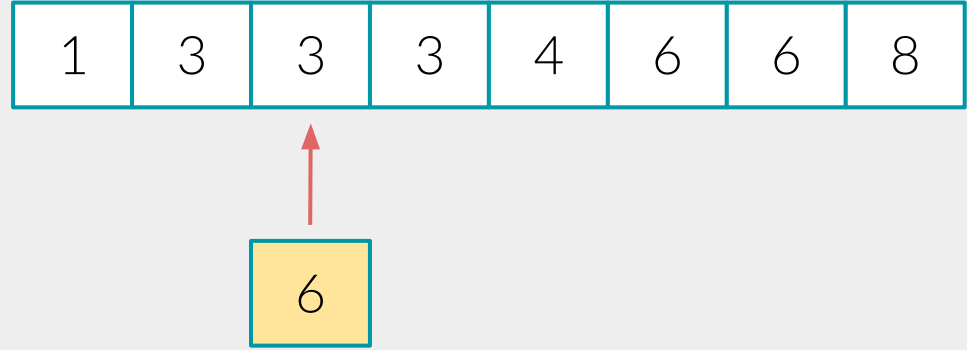
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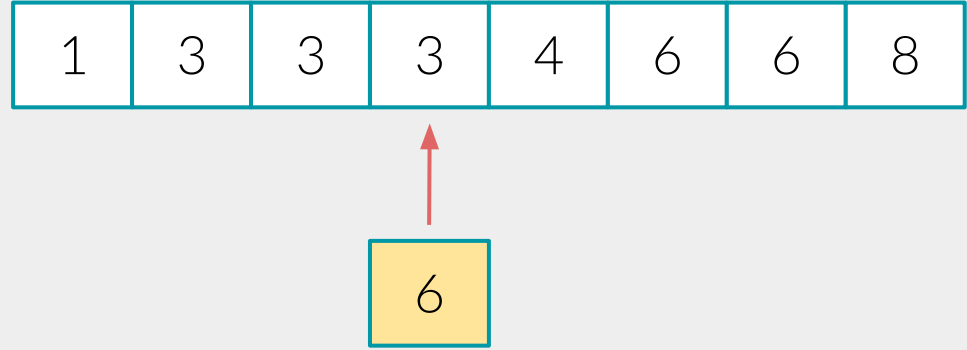
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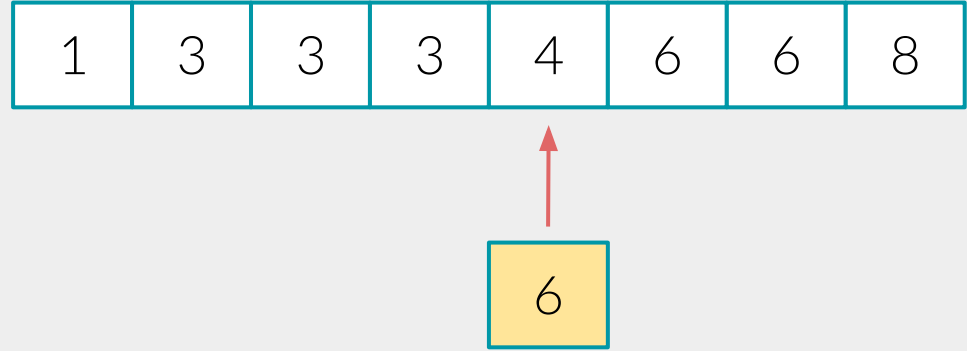
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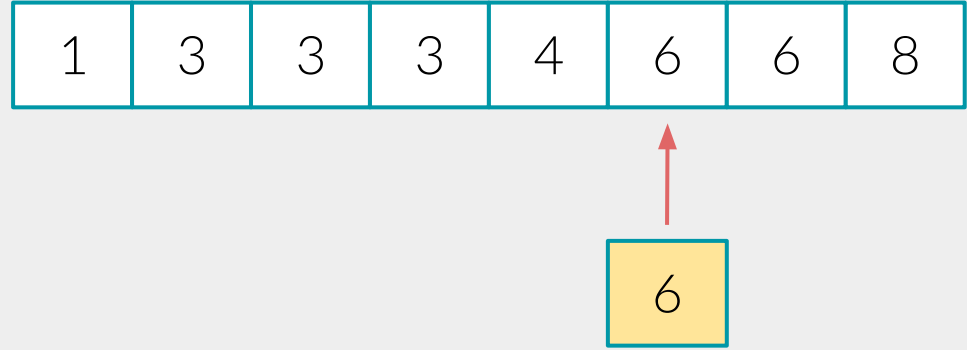
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Complexity

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Complexity
 $O(n)$

Binary Search

Binary Search

Binary Search

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BinarySearch(A, x):  
    if A is empty:  
        return 0  
    n = length of A  
    m = index of the middle element  
    if middle element is >= than x:  
        return BinarySearch(A[:m], x)  
    else:  
        return BinarySearch(A[m:], x) + m + 1
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Binary Search

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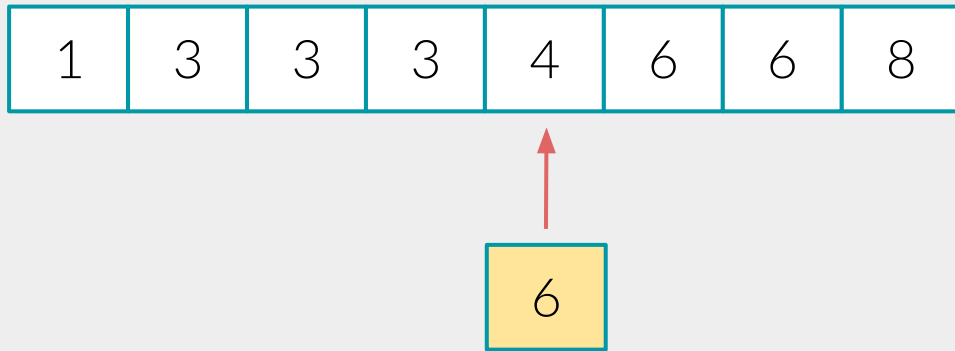
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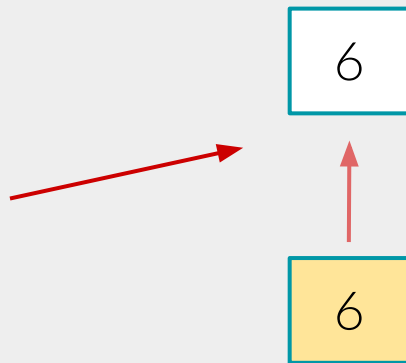
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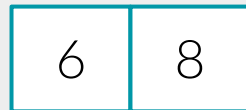
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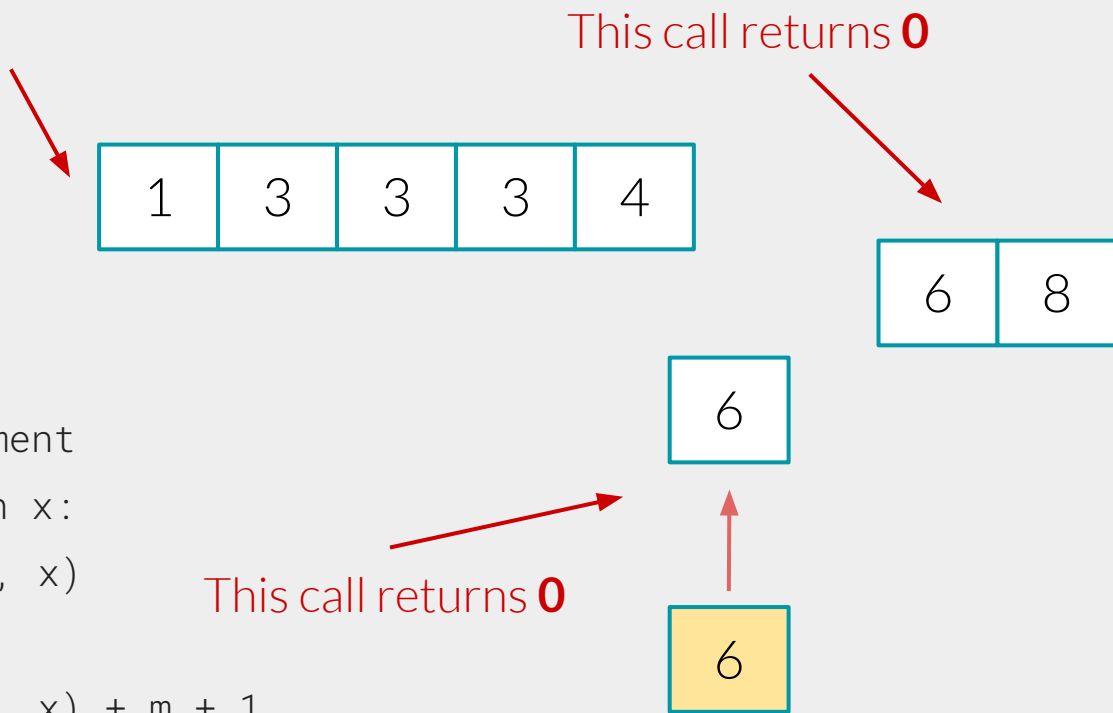
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        return BinarySearch(A[m:], x) + m + 1
```

This (initial) call returns **5**



This call returns **0**



This call returns **0**



Binary Search: Complexity

Substitution Method For Solving Recurrences

MergeSort: Review

```
MergeSort(A):
```

```
    if len(A) <= 1:
```

```
        return A
```

```
    L = MergeSort(A[0:n/2])
```

```
    R = MergeSort(A[n/2:n])
```

```
    return Merge(L, R)
```

Length $n/2$ array



Length $n/2$ array



n operations



Substitution Method: MergeSort Example

Recurrence relation: $T(n) = 2 \times T(n/2) + n$, $T(1) = 1$

Substitution Method: MergeSort Example

Recurrence relation: $T(n) = 2 \times T(n/2) + n$, $T(1) = 1$

First Observations: $T(1) = 1$

Substitution Method: MergeSort Example

Recurrence relation: $T(n) = 2 \times T(n/2) + n$, $T(1) = 1$

First Observations: $T(1) = 1$, $T(2) =$

Substitution Method: MergeSort Example

Recurrence relation: $T(n) = 2 \times T(n/2) + n$, $T(1) = 1$

First Observations: $T(1) = 1$, $T(2) = 2 \times T(1) + 2 = 4$

Substitution Method: MergeSort Example

Recurrence relation: $T(n) = 2 \times T(n/2) + n$, $T(1) = 1$

First Observations: $T(1) = 1$, $T(2) = 2 \times T(1) + 2 = 4$, $T(4) =$

Substitution Method: MergeSort Example

Recurrence relation: $T(n) = 2 \times T(n/2) + n$, $T(1) = 1$

First Observations: $T(1) = 1$, $T(2) = 2 \times T(1) + 2 = 4$, $T(4) = 12$,

Substitution Method: MergeSort Example

Recurrence relation: $T(n) = 2 \times T(n/2) + n$, $T(1) = 1$

First Observations: $T(1) = 1$, $T(2) = 2 \times T(1) + 2 = 4$, $T(4) = 12$, $T(8) = 32$, $T(16) = 80 \dots$

Substitution Method: MergeSort Example

Recurrence relation: $T(n) = 2 \times T(n/2) + n$, $T(1) = 1$

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Our Guess: $T(n) =$

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Our Guess: $T(n) = O(n \log n)$

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Our Guess: $T(n) = O(n \log n)$

Base Case: For $n = 2$, $T(2) = 4 < C \times 2 \times \log(2) = C \times 2$ for any $C > 2$

Substitution Method: MergeSort Example

Recurrence relation: $T(n) = 2 \times T(n/2) + n$, $T(1) = 1$

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Inductive Hypothesis: For all $i < k$, $T(i) = O(i \times \log(i)) \Leftrightarrow T(i) < C \times i \times \log(i)$ for some C

Substitution Method: MergeSort Example

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$$< C \times 2 \times (k/2) \times \log(k/2) + k$$

from inductive hypothesis

Substitution Method: MergeSort Example

Recurrence relation: $T(n) = 2 \times T(n/2) + n$, $T(1) = 1$

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Inductive Step: $T(k) = 2 \times T(k/2) + k$

$< C \times 2 \times (k/2) \times \log(k/2) + k$ from inductive hypothesis

$= C \times k \times (\log(k) - 1) + k$

Substitution Method: MergeSort Example

Recurrence relation: $T(n) = 2 \times T(n/2) + n$, $T(1) = 1$

First Observations: $T(1) = 1$, $T(2) = 2 \times T(1) + 2 = 4$, $T(4) = 12$, $T(8) = 32$, $T(16) = 80 \dots$

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$= C \times k \times \log(k) - C \times k + k$

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Recurrence relation: $T(n) = 2 \times T(n/2) + n$, $T(1) = 1$

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$= C \times k \times (\log(k) - 1) + k$

$= C \times k \times \log(k) - C \times k + k$

$< C \times k \times \log(k)$ as $C > 2 \Rightarrow C \times (k/2) > k$

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Recurrence relation: $T(n) = 2 \times T(n/2) + n$, $T(1) = 1$

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$= C \times k \times (\log(k) - 1) + k$

$= C \times k \times \log(k) - C \times k + k$

$< C \times k \times \log(k)$ as $C > 2 \Rightarrow C \times (k/2) > k$

$= O(k \times \log(k))$

Substitution Method: MergeSort Example

Recurrence relation: $T(n) = 2 \times T(n/2) + n$, $T(1) = 1$

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$< C \times k \times \log(k)$ as $C > 2 \Rightarrow C \times (k/2) > k$

$= O(k \times \log(k))$

So we conclude the proof. We proved that $T(n) < C \times n \times \log(n)$ for $C = 3$ and for all $n > 1$.

Binary Search: Complexity

Recurrence relation: $T(n) =$

Binary Search: Complexity

Recurrence relation: $T(n) = T(n/2) + 1$

Substitution Method

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Substitution Method

Recurrence relation: $T(n) = T(n/2) + 1$, $T(1) = 1$

First Observations: $T(1) = 1$, $T(2) = T(1) + 1 = 2$

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First Observations: $T(1) = 1$, $T(2) = T(1) + 1 = 2$, $T(4) = 3$, $T(8) = 4$

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Our Guess: $T(n) = O(\log n)$

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Base Case: For $n = 2$, $T(2) = 2 < C \times \log(2) = C$ for any $C > 2$

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Inductive Hypothesis: For all $i < k$, $T(i) = O(\log i) \Leftrightarrow T(i) < C \times \log(i)$ for some C

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Inductive Step: $T(k) = T(k/2) + 1$

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Inductive Step: $T(k) = T(k/2) + 1$

$< C \times \log(k/2) + 1$

from inductive hypothesis

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Inductive Step: $T(k) = T(k/2) + 1$

$$< C \times \log(k/2) + 1$$

from inductive hypothesis

$$= C \times (\log(k) - 1) + 1$$

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Inductive Step: $T(k) = T(k/2) + 1$

$$< C \times \log(k/2) + 1$$

from inductive hypothesis

$$= C \times (\log(k) - 1) + 1$$

$$= C \times \log(k) - C + 1$$

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$$< C \times \log(k/2) + 1$$

from inductive hypothesis

$$= C \times (\log(k) - 1) + 1$$

$$= C \times \log(k) - C + 1$$

$$< C \times \log(k)$$

as $C > 2$

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$$= C \times (\log(k) - 1) + 1$$

$$= C \times \log(k) - C + 1$$

$$< C \times \log(k)$$

as $C > 2$

$$= O(\log(k))$$

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$$< C \times \log(k/2) + 1$$

from inductive hypothesis

$$= C \times (\log(k) - 1) + 1$$

$$= C \times \log(k) - C + 1$$

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as $C > 2$

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So we conclude the proof. We proved that $T(n) < C \times \log(n)$ for $C = 3$ and for all $n > 1$.

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$$< C \times \log(k/2) + 1$$

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$$= C \times (\log(k) - 1) + 1$$

$$= C \times \log(k) - C + 1$$

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as $C > 2$

$$= O(\log(k))$$

So we conclude the proof. We proved that $T(n) < C \times \log(n)$ for $C = 3$ and for all $n > 1$.

Which means that $T(n) = O(\log n)$!

Binary Search: Complexity

Recurrence relation: $T(n) = T(n/2) + 1$

Binary Search: Complexity

Recurrence relation: $T(n) = T(n/2) + 1$

Complexity: $O(\log n)$

Binary Search: Use Examples

Find number of occurrences of a value in a sorted array.

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Binary Search Trees!

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Find number of occurrences of a value in a sorted array.

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Searching for a word in a dictionary (a real one).

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Literally everywhere.