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# **ArraySet and Binary Search**

## **Lecture 16**

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# Menu

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- Cost of ArraySet operations
- Binary Search
- Cost of SortedArraySet with Binary Search

# ArrayList costs: Summary

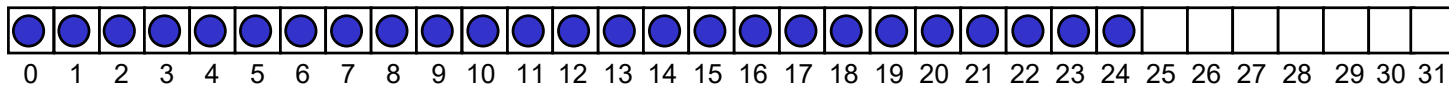
- get  $O(1)$
- set  $O(1)$
- remove  $O(n)$
- add (at i)  $O(n)$  (worst and average)  
(have to shift up  
may have to double capacity)
- add (at end)  $O(1)$  (most of the time)  
(when doubles cap)  $O(n)$  (worst)  
 $O(1)$  (amortised average)  
(if doubled each time)

# ArraySet costs

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## What about ArraySet:

- Order is not significant
  - ⇒ add() can choose to put a new item anywhere. where?
  - ⇒ can reorder when removing an item. how?
- **Duplicates not allowed.**
  - ⇒ must check if item already present before adding



# ArraySet algorithms

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Contains(value):

- search through array,
- if value equals item
- return true
- return false

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Costs?

Add(value):

- if not contains(value),
- place value at end, (doubling array if necessary)
- increment size

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Remove(value):

- search through array
- if value equals item
- replace item by item at end. (why?)
- decrement size
- return

# ArraySet costs

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## Costs:

- contains, add, remove:  $O(n)$
- 

## Question:

- How can we speed up the search?

# Making ArraySet faster.

All the cost is in the searching:

- Searching for “Eel”

8								
Bee	Dog	Ant	Fox	Hen	Gnu	Eel	Cat	
0	1	2	3	4	5	6	7	8

- but if sorted...

8								
Ant	Bee	Cat	Dog	Eel	Fox	Gnu	Hen	

# Making ArraySet faster

- Binary Search: Finding “Eel”

8								
Ant	Bee	Cat	Dog	Eel	Fos	Gnu	Pig	
0	1	2	3	4	5	6	7	8

- If the items are sorted (“ordered”), then we can search fast
  - Look in the middle:
    - if item is middle item  $\Rightarrow$  return
    - if item is before middle item  $\Rightarrow$  look in left half
    - if item is after middle item  $\Rightarrow$  look in right half



# Divide and Conquer

One of the **best-known** algorithm design techniques.

Idea:

- A problem instance is divided into several smaller instances of the same problem, ideally of about same size
- The smaller instances are solved, typically recursively
- The solutions for the smaller instances are combined to get a solution to the original problem

# Binary Search

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```

private boolean contains(Object item){
    Comparable<E> value = (Comparable<E>) item;

    int low = 0;                                // min possible index of item
    int high = count-1;                          // max possible index of item
                                                // item in [low .. high] (if present)

    while (low <= high){
        int mid = (low + high) / 2;
        int comp = value.compareTo(data[mid]);
        if (comp == 0)                          // item is present
            return true;
        if (comp < 0)                            // item in [low .. mid-1]
            high = mid - 1;                      // item in [low .. high]
        else                                    // item in [mid+1 .. high]
            low = mid + 1;                       // item in [low .. high]
    }
    return false; // item in [low .. high] and low > high,
                  // therefore item not present
}

```

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low
mid
high

- |   |   |   |   |   |   |   |   |   |   |    |    |    |    |    |    |    |    |    |    |    |    |    |    |    |    |    |    |    |    |    |    |  |
|---|---|---|---|---|---|---|---|---|---|----|----|----|----|----|----|----|----|----|----|----|----|----|----|----|----|----|----|----|----|----|----|--|
|   |   |   |   |   |   |   |   |   |   |    |    |    |    |    |    |    |    |    |    |    |    |    |    |    |    |    |    |    |    |    |    |  |
| 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 12 | 13 | 14 | 15 | 16 | 17 | 18 | 19 | 20 | 21 | 22 | 23 | 24 | 25 | 26 | 27 | 28 | 29 | 30 | 31 |  |

Iteration	Size of range	Cost of iteration
1	n	
2		
k	1	

# Time complexity

Let  $T(n)$  denote the time complexity of binary search algorithm on  $n$  numbers.

$$T(n) = \begin{cases} 1 & \text{if } n=1 \\ T(n/2) + 1 & \text{otherwise} \end{cases}$$

We call this formula a recurrence.

# Recurrence

A recurrence is an equation or inequality that describes a function in terms of its value on smaller inputs.

E.g., 
$$T(n) = \begin{cases} 1 & \text{if } n = 1 \\ T(n/2) + 1 & \text{if } n > 1 \end{cases}$$

To solve a recurrence is to derive *asymptotic bounds* on the solution

# $\text{Log}_2(n)$ or $\log(n)$

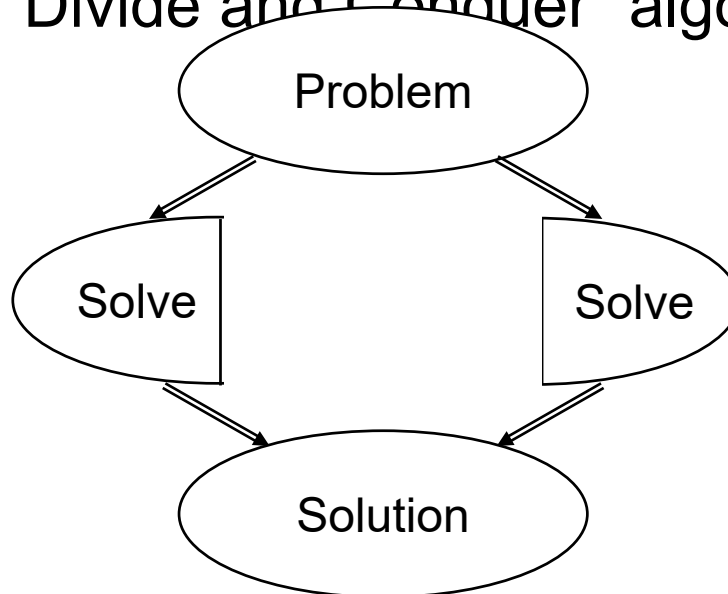
The number of times you can divide a set of  $n$  things in half.

$\log(1000) = 10$ ,  $\log(1,000,000) = 20$ ,  $\log(1,000,000,000) = 30$

Every time you double  $n$ , you add one step to the cost! (why?)

- $\log(2n) = \log(n) + \log 2 = \log(n) + 1$
- Arises all over the place in analysing algorithms

Especially “Divide and Conquer” algorithms:



# Substitution method

$$T(n) = \begin{cases} 1 & \text{if } n=1 \\ T(n/2) + 1 & \text{otherwise} \end{cases}$$

Make a guess,  $T(n) \leq 2 \log n$

We prove statement by MI.

Base case? When  $n=1$ , statement is FALSE!

$$\text{L.H.S} = T(1) = 1 \quad \text{R.H.S} = c \log 1 = 0 < \text{L.H.S}$$

Yet, when  $n=2$ ,

$$\text{L.H.S} = T(2) = T(1)+1 = 2$$

$$\text{R.H.S} = 2 \log 2 = 2$$

$$\text{L.H.S} \leq \text{R.H.S}$$

# Substitution method

$$T(n) = \begin{cases} 1 & \text{if } n=1 \\ T(n/2) + 1 & \text{otherwise} \end{cases}$$

Make a guess,  $T(n) \leq 2 \log n$

We prove statement by MI.

Assume true for all  $n' < n$  [assume  $T(n/2) \leq 2 \log (n/2)$ ]

$$\begin{aligned} T(n) &= T(n/2) + 1 \\ &\leq 2 \log (n/2) + 1 \quad \leftarrow \text{by hypothesis} \\ &= 2(\log n - 1) + 1 \quad \leftarrow \log(n/2) = \log n - \log 2 \\ &< 2 \log n \end{aligned}$$

**i.e.,  $T(n) \leq 2 \log n$**

..



## More Example

Prove that  $T(n) = \begin{cases} 1 & \text{if } n=1 \\ 2T(n/2) + n & \text{otherwise} \end{cases}$  is  $O(n \log n)$

**Guess:**  $T(n) \leq 2 n \log n$

## More Example

Prove that  $T(n) = \begin{cases} 1 & \text{if } n=1 \\ 2T(n/2) + n & \text{otherwise} \end{cases}$  is  $O(n \log n)$

**Guess:**  $T(n) \leq 2 n \log n$

Assume true for all  $n' < n$  [assume  $T(n/2) \leq 2 (n/2) \log(n/2)$ ]

$$\begin{aligned} T(n) &\leq 2 (2 (n/2) \log (n/2) ) + n \\ &= 2 n (\log n - 1) + n \\ &= 2 n \log n - 2n + n \\ &\leq 2 n \log n \end{aligned}$$

For the base case when  $n=2$ ,  
 L.H.S =  $T(2) = 2T(1)+2 = 4$ ,  
 R.H.S =  $2 * 2 \log 2 = 4$   
 L.H.S  $\leq$  R.H.S

**i.e.,  $T(n) \leq 2 n \log n$**

# ArraySet with Binary Search

## ArraySet: unordered

- All cost in the searching:  $O(n)$ 
  - contains:  $O(n)$
  - add:  $O(n)$
  - remove:  $O(n)$

## SortedArraySet: with Binary Search

- Binary Search is fast:  $O(\log(n))$ 
  - contains:  $O(\log(n))$
  - add:
  - remove:
- All the cost is in keeping it sorted!!!!

# Making SortedArraySet fast

- If you have to call `add()` and/or `remove()` many items, then `SortedArraySet` is no better than `ArraySet`!
  - Both  $O(n)$
  - Either pay to search
  - Or pay to keep it in order
- If you only have to construct the set once, and then many calls to `contains()`, then `SortedArraySet` is much better than `ArraySet`.
  - `SortedArraySet contains()` is  $O(\log(n))$
- But, how do you construct the set fast?
  - Adding each item, one at a time

# Alternative Constuctor

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- Sort the items all at once

```
public SortedArraySet(Collection<E> col){  
    // Make space  
    count=col.size();  
    data = (E[]) new Object[count];  
  
    // Put items from collection into the data array.  
    col.toArray(data);  
  
    // sort the data array.  
    Arrays.sort(data);  
}
```

- How do you sort?

# Summary

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- Cost of ArraySet operations
- Binary Search
- Cost of SortedArraySet with Binary Search

# Readings

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- [Mar07] Read 4.3
- [Mar13] Read 4.3