

# Algorithms and Data Structures

## Searching Abstract Data Structures, Symbol Tables

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Arrays

Symbol Tables

List-based

Array-based

## Arrays

### Symbol Tables

List-based

Array-based

Arrays are by nature of fixed length. How can we make them expandable and still have direct memory access?

When adding a new element to a full array we could:

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2. Copy the array to an array  $m$  elements bigger
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- 2.
- 3.

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3.  $O(1)$  - how can that be?



Money in the bank

Balance: 4 we hope that is enough to pay for future expansions

array is full

7

9

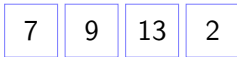
13

2

Money in the bank

Balance:  $4 - 0 = 4$  (creating new array considered free here)

array is full



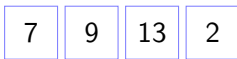
create new array



Money in the bank

Balance:  $4 - 4 = 0$  (using 1 per copy)

array is full



4 copies

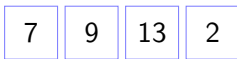
create new array



Money in the bank

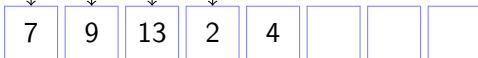
Balance:  $0 + 3 - 1 = 2$  (charging 3 for an insert, using 1)

array is full



4 copies

create new array

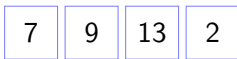


add element 4

Money in the bank

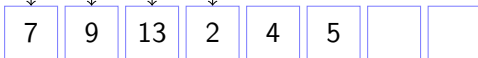
Balance:  $2 + 3 - 1 = 4$  (charging 3 for an insert, using 1)

array is full



4 copies

create new array



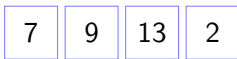
add element 4

add element 5

Money in the bank

Balance:  $4 + 3 - 1 = 6$  (charging 3 for an insert, using 1)

array is full



4 copies

create new array



add element 4

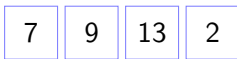
add element 5

add element 8

Money in the bank

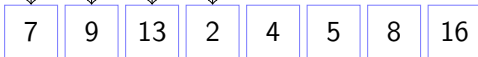
Balance:  $6 + 3 - 1 = 8$  (charging 3 for an insert, using 1)

array is full



4 copies

create new array



add element 4

add element 5

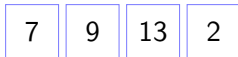
add element 8

add element 16

Money in the bank

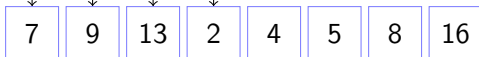
Balance: 8 enough to pay for 8 copies

array is full



4 copies

create new array



add element 4

add element 5

add element 8

add element 16

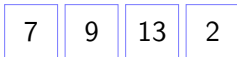
$$O(3) = O(1)$$



Constant payload

Payload: 0

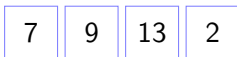
array is full



Constant payload

Payload: 0 (creating new array considered free here)

array is full



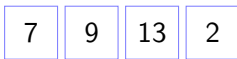
create new array



Constant payload

Payload:  $1 + 1 = 2$  (1 for copying and 1 for inserting)

array is full



create new array

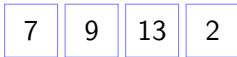


copy 7 insert 4

Constant payload

Payload:  $1 + 1 = 2$  (1 for copying and 1 for inserting)

array is full



create new array



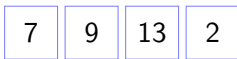
copy 7 insert 4

copy 9 insert 5

Constant payload

Payload:  $1 + 1 = 2$  (1 for copying and 1 for inserting)

array is full



create new array



copy 7 insert 4

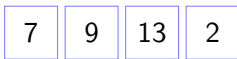
copy 9 insert 5

copy 13 insert 8

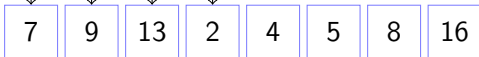
Constant payload

Payload:  $1 + 1 = 2$  (1 for copying and 1 for inserting)

array is full



create new array



copy 7 insert 4

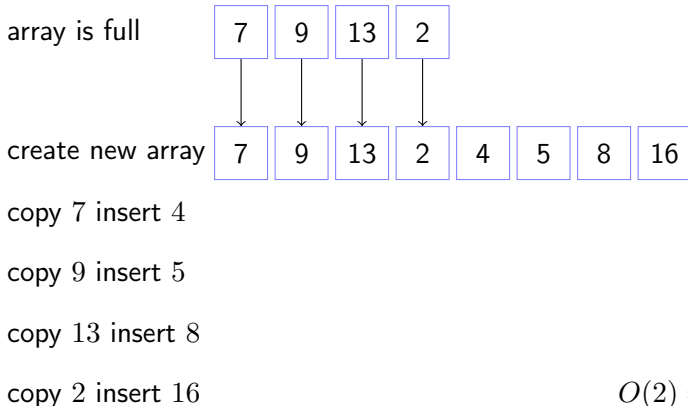
copy 9 insert 5

copy 13 insert 8

copy 2 insert 16

Constant payload

Payload: 8 in total for 4 insertions



- What would the complexity (big-O) be if we:
  - Triple the array size instead of doubling it?
  - Only made the new array 50% bigger?
- Bearing in mind that most modern memory is paged<sup>1</sup>, consider why doubling the array size is not such a bad idea?

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<sup>1</sup>typically in  $2^n$  sized pages



1. Create a Java class `FlexibleArray` that uses the “Constant payload” algorithm.

```
public class FlexibleArray<T> {  
    ...  
    public T get(int index) { ... }  
    public void set(int index, T element) { ... }  
    public void add(T element) { ... }  
    public int size() { ... }  
}
```

**Note** that to create a new array of type `T` you must:

```
private T[] arrayOfT = (T[])new Object[1000];
```

2. Measure the time it takes to add 10.000, 100.000, and 1.000.000 elements.
3. Measure Javas build-in `ArrayList` with the same data.

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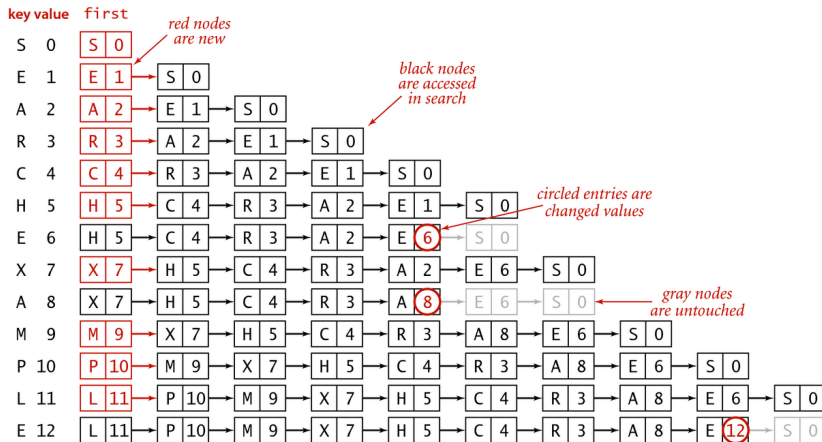
## Map in Java

- Lookup a value based on a key
- No `null` keys
- No `null` values
- No duplicate keys

```
public interface SymbolTable<K,V> {  
    void put(K key, V value);  
    V get(K key);  
    int size();  
    Iterable<K> keys();  
    default void delete(K key) { put(key, null); }  
    default boolean contains(K key) {  
        return get(key) != null;  
    }  
    default boolean isEmpty() { return size() == 0; }  
}
```

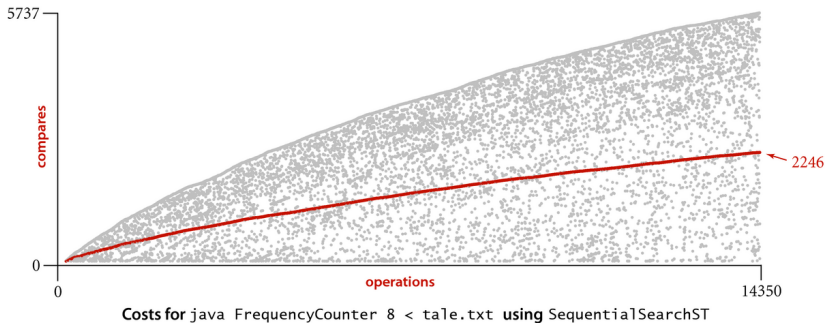
## OrderedMap in Java

```
public interface
    OrderedSymbolTable<K extends Comparable<K>, V>
    extends SymbolTable<K,V> {
    K min();
    K max();
    K floor(K key);
    K ceiling(K key);
    int rank(K key);
    K select(int rank);
    void deleteMin();
    void deleteMax();
    int size(K low, K high);
    Iterable<K> keys(K low, K high);
}
```



Trace of linked-list ST implementation for standard indexing client

## Costs



**Q: go to the book (Ch.3.1, subsect. 'Performance Client') and figure out what the input argument '8' does here? Discuss in Plenum.**

		keys[]												vals[]									
key	value	0	1	2	3	4	5	6	7	8	9	N		0	1	2	3	4	5	6	7	8	9
S	0	S										1	0										
E	1	E	S									2	1	0									
A	2	A	E	S								3	2	1	0								
R	3	A	E	R	S							4	2	1	3	0							
C	4	A	C	E	R	S						5	2	4	1	3	0						
H	5	A	C	E	H	R	S					6	2	4	1	5	3	0					
E	6	A	C	E	H	R	S					6	2	4	6	5	3	0					
X	7	A	C	E	H	R	S	X				7	2	4	6	5	3	0	7				
A	8	A	C	E	H	R	S	X				7	8	4	6	5	3	0	7				
M	9	A	C	E	H	M	R	S	X			8	8	4	6	5	9	3	0	7			
P	10	A	C	E	H	M	P	R	S	X		9	8	4	6	5	9	10	3	0	7		
L	11	A	C	E	H	L	M	P	R	S	X	10	8	4	6	5	11	9	10	3	0	7	
E	12	A	C	E	H	L	M	P	R	S	X	10	8	4	12	5	11	9	10	3	0	7	
		A	C	E	H	L	M	P	R	S	X			8	4	12	5	11	9	10	3	0	7

entries in red were inserted

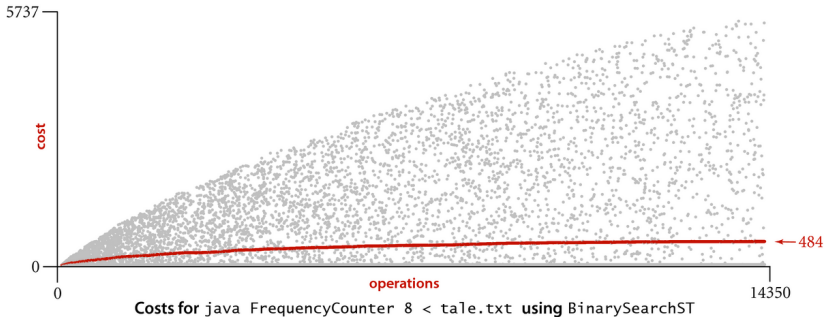
entries in gray did not move

entries in black moved to the right

circled entries are changed values

Trace of ordered-array ST implementation for standard indexing client

## Costs



Q: What happened here? Look at that decrease in cost...  
Discuss in plenum.



## Costs Comparison

Unordered sequential search vs. Ordered array binary search.

algorithm (data structure)	worst-case cost (after $N$ inserts)		average-case cost (after $N$ random inserts)		efficiently support ordered operations?
	search	insert	search hit	insert	
<i>sequential search</i> ( <i>unordered linked list</i> )	$N$	$N$	$N/2$	$N$	no
<i>binary search</i> ( <i>ordered array</i> )	$\lg N$	$2N$	$\lg N$	$N$	yes

Cost summary for basic symbol-table implementations

Q: What about insert (or `put()`)? What is *its* complexity?  
And can we achieve better performance? Discuss in plenum,  
'tales.txt' and '.

## Pros & Cons in Searching

underlying data structure	implementation	pros	cons
<i>linked list (sequential search)</i>	SequentialSearchST	best for tiny STs	slow for large STs
<i>ordered array (binary search)</i>	BinarySearchST	optimal search and space, order-based ops	slow insert
<i>binary search tree</i>	BST	easy to implement, order-based ops	no guarantees space for links
<i>balanced BST</i>	RedBlackBST	optimal search and insert, order-based ops	space for links
<i>hash table</i>	SeparateChainingHashST LinearProbingHashST	fast search/insert for common types of data	need hash for each type no order-based ops space for links/empty

Pros and cons of symbol-table implementations

Q: how can we get better e.g. `put()` performance? Need more than linked lists and binary searches=> Datastructures.

NB: These last exercises are optional, but will definitely help you to understand the linked listst, search complexity issues better.

- Add to the Algorithm 3.1 in the Book (seq. search in unordered list), the methods mentioned in the text, namely
  - `size()`,
  - `keys()`,so essentially exercise 3.1.5, yet, without the *eager* `delete()`.
- Exercise 3.1.6 of the Book.
- (ex. 3.1.10 of the book) What's frequent most word of 10+ letters in '*Tale of two cities*' (ex. 3.1.10 of the book)?