# CS 228: Introduction to Data Structures Lecture 17 Wednesday, October 5, 2016

### AbstractCollection<E>

Implementing the Collection interface from scratch would be daunting — just look at the number of methods specified in the javadoc! Fortunately, Java offers us AbstractCollection<E>, a generic abstract class that implements all the methods of Collection<E>, except size() and iterator().

```
public abstract class AbstractCollection<E>
implements Collection<E>
{

   public abstract Iterator<E> iterator();

   public abstract int size();
```

All other methods are implemented based on iterator() and size(). For example, here are isEmpty() (which is true if and only if the collection is empty) and contains().

```
public boolean isEmpty() {
    return size() == 0;
}

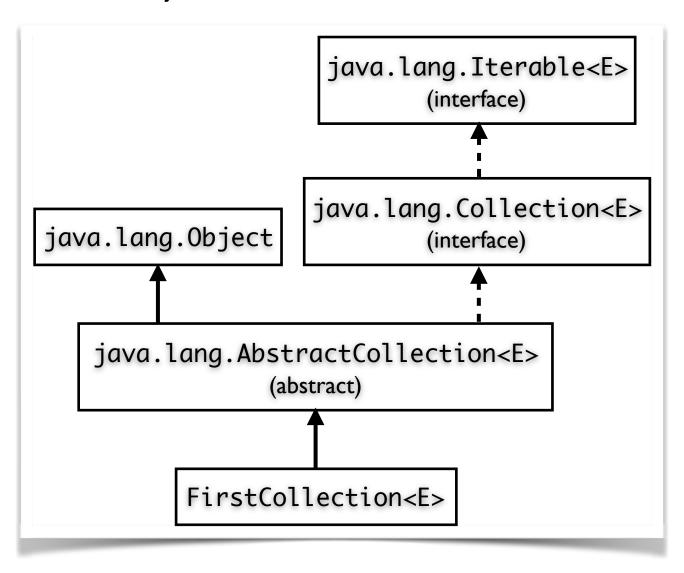
public boolean contains(Object o) {
    Iterator<E> e = iterator();
    if (o==null) {
        while (e.hasNext())
            if (e.next()==null)
                return true;
    } else {
        while (e.hasNext())
            if (o.equals(e.next()))
               return true;
    }
    return false;
}
```

Some methods of Collection are *optional*; i.e., they are not required to be implemented by an implementing class. Optional methods in AbstractCollection are implemented in a simple fashion: Just throw an UnsupportedOperationException, like this:

```
public boolean add(E o)
{
    throw new UnsupportedOperationException;
}
```

AbstractCollection serves as a starting point for concrete implementations of Collection.

The figure on the next page shows the portion of the Java class hierarchy that we will study in the next few weeks. The figure also shows where FirstCollection, our first implementation of the Collection interface, lies within that hierarchy.

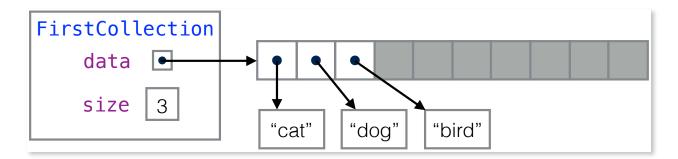


# **An Array-Based Generic Collection**

We start with a simple array-based implementation of Collection<E> called FirstCollection<E>. The code is posted on Blackboard. We will only discuss in detail a few representative methods. You are responsible for carefully reading the posted code.

#### **Basic Structure**

The basic elements of FirstCollection are a data array, which stores the items, and a size field, which indicates how many slots of data are being used.



The class definition begins like this:

There are two constructors. One takes an initialCapacity argument, specifying the initial length of data. The default constructor initializes data to DEFAULT\_SIZE (= 10).

To complete the description of FirstCollection, we need to explain how to implement add(), size(), and iterator().

## add()

Adding is easy: just put the new item in the next available slot at the end of the data array. But what if we run out of space? The code below handles this possibility by invoking a checkCapacity() method, which, if necessary, doubles the capacity of the data array to accommodate the new item.

```
public boolean add(E item)
{
    checkCapacity();
    data[size++] = item;
    return true;
}
```

**checkCapacity()** ensures that there is enough space for a new item. If there isn't, it copies the entire array into a new array double the size of the original one.

An add() takes O(1) time if data is not full. If data is full, though, it takes O(n) time, where n = data. length. Which is more typical?

Notice that the good case occurs much more frequently than the bad case: In a sequence of 161 adds, starting with the default size of 10, the bad case occurs only 5 times: at operations 11, 21, 41, 81, and 161. As the array gets bigger, the time between bad cases increases drastically — the next bad add occurs at operation 321!

**Theorem.** The **total time** for a sequence of n add operations is O(n).

**Proof:** Suppose we start out at size 1 and that n is a multiple of 2, say n is 2<sup>p</sup>. Each time we run out of space, we double the capacity. Then, there are resize operations for sizes 1, 2, 4, 8, 16, ..., 2<sup>p-1</sup>. A resize operation on an

array of size k takes time O(k). So the total cost of all the resize operations is

$$O(1 + 2 + 4 + ... + 2^{p-1}) = O(2^p) = O(n).$$

The theorem implies that the time for n add() operations averages out to O(1) per operation. More formally, we say that the *amortized time* for add() is O(1).