

Dugga/Partial Exam Solutions

Data Structures (DAT037/DAT036)

2014-12-08

1. The complexity of the piece of code is $O(\log n)$. Below are the complexities of the individual elements:

- Loop: $O(\log n)$. Loop goes from 1 to n , with increments doubling in size for each iteration, ie. $\log n$ times.
 - add: $O(1)$. This is repeated $\log n$ times, so the complexity of the whole loop is $O(\log n)$.
- toArray: $O(\log n)$. toArray is linear in the number of elements. In this case there are $\log n$ elements in the dynamic array that is constructed in the loop.
- insert_sort: $O(\log n)$. insert_sort is quadratic in general case, but linear **for presorted input**. In this case, it is given a sorted array of $\log n$ elements.

We have three consecutive $\log n$ terms, thus the complexity of the whole code is $O(\log n)$.

Note that it is incorrect to say that insertion sort is $O(n^2)$, even if that is the general case. In this case we see clearly that the input is presorted, and we must use that information in deciding what is the complexity of that particular piece of code.

2. We are using the following representation for linked lists:

```
List {Node head;}
Node {int info; Node next;}
```

The algorithm below traverses the list twice, first time pushing each item into a stack, second time popping from the stack and comparing the two items. It returns false immediately if any two elements are not same. If it has successfully compared the whole list to the elements in the stack, it returns true.

Algorithm 1 isPalindrome

Input: List a

Output: Boolean

```
 $b \leftarrow$  empty stack
 $nd \leftarrow a.head$ 
while  $nd$  is not null do
     $b.push(nd)$ 
     $nd \leftarrow nd.next$ 
end while
 $nd \leftarrow a.head$ 
while  $nd$  is not null do
     $prev \leftarrow b.pop()$ 
    if  $nd$  is not  $prev$  then
        return false
    end if
     $nd = nd.next$ 
end while
return true
```

The complexity of this code is $O(n)$. We have the following elements:

- Create stack: $O(1)$.
- Get head of the list for first traversal: $O(1)$.
- First while loop: $O(n)$. Traverses the list and performs two actions which are $O(1)$: push and accessing the *next* node of the current node.
- Get head of the list for second traversal: $O(1)$.
- Second while loop: $O(n)$. Traverses the list and performs three $O(1)$ actions: pop, comparison and accessing the *next* node.

The algorithm goes through the list twice, both in $O(n)$ time, and two consecutive $O(n)$ operations is in total just $O(n)$.

3. We use the following representation for binary search trees:

```
Tree {Node root;}
Node {E info; Node left; Node right;}
```

E must be a type that has an ordering. In the pseudocode below we are using \geq and \leq ; in Java, we would compare with `compareTo` or using a comparator defined for that type. In Haskell, we would define `Ord` and `Eq` instances for the type, and we would be able to use the operators $>$, $<$, $==$.

The solution below traverses the tree and keeps track of the minimum and maximum values, updating them as the algorithm descends to the subtrees. We define first a recursive helper function that takes a node and two values of type E :

Algorithm 2 isBSTUtil

Input: Node $node$, E min , E max

Output: Boolean

if $node$ is leaf then

return true

end if

if $node.info \leq min$ or $node.info \geq max$ then

return false

end if

$leftIsBST \leftarrow isBSTUtil(node.left, min, node.info)$

$rightIsBST \leftarrow isBSTUtil(node.right, node.info, max)$

return $leftIsBST$ and $rightIsBST$

The final algorithm takes the helper function and starts from the root of the tree, min and max initialised as the minimum and maximum value of the data type that is stored in the tree.

Algorithm 3 isBST

Input: Tree $tree$

Output: Boolean

return $isBSTUtil(tree.root, E.MIN_VALUE, E.MAX_VALUE)$

The time complexity of this algorithm is $O(n)$. It traverses the tree once and does only operations which take constant time: accessing the fields of the nodes, comparisons and recursive function calls.

Space complexity is between $O(\log n)$ and $O(n)$, depending how balanced the tree is. There is no additional data used; instead, the space complexity comes from the recursive step. The function call stack accumulates the function calls until the algorithm reaches the base case, where the tree is a leaf. In the

case of a balanced BST, the algorithm needs $\log n$ steps to reach to a leaf. In the worst case scenario, where the tree is a chain of length n , the algorithm needs n steps to reach a leaf.