1. An **AVL Tree** is a special type of Binary Search Tree (BST) that it is *height balanced*. By height balanced I mean that the height of every node’s left and right subtrees differ by at most one. This is enough to guarantee that a AVL tree with \( n \) nodes has a height no worst than \( O(1.44 \log_2 n) \). Therefore, insertions, deletions, and search are worst case \( O(\log_2 n) \). An example of an AVL tree with integer keys is shown below. The height of each node is shown.

![AVL Tree Example](image)

Each AVL-tree node usually stores a *balance factor* in addition to its key and payload. The balance factor keeps track of the relative height difference between its left and right subtrees, i.e., \( \text{height(left subtree)} - \text{height(right subtree)} \).

a) Label each node in the above AVL tree with one of the following *balance factors*:
   - 0 if its left and right subtrees are the same height
   - 1 if its left subtree is one taller than its right subtree
   - -1 if its right subtree is one taller than its left subtree

b) We start a *put* operation by adding the new item into the AVL as a leaf just like we did for Binary Search Trees (BSTs). Add the key 90 to the above tree.

c) Identify the node "closest up the tree" from the inserted node (90) that no longer satisfies the height-balanced property of an AVL tree. This node is called the *pivot node*. Label the pivot node above.

d) Consider the subtree whose root is the pivot node. How could we rearrange this subtree to restore the AVL height balanced property? (Draw the rearranged tree below)
2. Typically, the addition of a new key into an AVL requires the following steps:
   - compare the new key with the current tree node's key (as we did in the \_put function called by the put method in the BST) to determine whether to recursively add the new key into the left or right subtree
   - add the new key as a leaf as the base case(s) to the recursion
   - recursively (updateBalance method) adjust the balance factors of the nodes on the search path from the new node back up toward the root of the tree. If we encounter a pivot node (as in question (c) above) we perform one or two "rotations" to restore the AVL tree's height-balanced property.

For example, consider the previous example of adding 90 to the AVL tree. Before the addition, the pivot node (60) was already -1 ("tall right" - right subtree had a height one greater than its left subtree). After inserting 90, the pivot's right subtree had a height 2 more than its left subtree (balance factor -2) which violates the AVL tree's height-balance property. This problem is handled with a left rotation about the pivot as shown in the following generalized diagram:

![Diagram showing AVL tree with rotations]

a) Assuming the same initial AVL tree (upper, left-hand of above diagram) if the new node would have increased the height of Tc (instead of Tb), would a left rotation about the node B have rebalanced the AVL tree?
b) Before the addition, if the pivot node was already -1 (tall right) and if the new node is inserted into the left subtree of the pivot node's right child, then we must do two rotations to restore the AVL-tree's height-balance property.

Before the addition:  
After the addition, but before first rotation:

After the left rotation at pivot and balance factors adjusted correctly:

After right rotation at F, but before left rotation at pivot:

b) Suppose that the new node was added in $T_c$ instead of $T_r$, then the same two rotations would restore the AVL-tree's height-balance property. However, what should the balance factors of nodes B, D, and F be after the rotations?
Consider the AVLTreeNode class that inherits and extends the TreeNode class to include balance factors.

```python
from tree_node import TreeNode

class AVLTreeNode(TreeNode):
    def __init__(self, key, val, left=None, right=None, parent=None, balanceFactor=0):
        TreeNode.__init__(self, key, val, left, right, parent)
        self.balanceFactor = balanceFactor
```

Now let's consider the partial AVLTree class code that inherits from the BinarySearchTree class:

```python
from avl_tree_node import AVLTreeNode
from binary_search_tree import BinarySearchTree

class AVLTree(BinarySearchTree):
    def __init__(self):
        self.root = None
        self.size = 0

    def __put(self, key, val):
        if self.root:
            self._put(key, val, self.root)
        else:
            self.root = AVLTreeNode(key, val)
            self.size = self.size + 1

    def _put(self, key, val, currentNode):
        if key < currentNode.key:
            if currentNode.hasLeftChild():
                self._put(key, val, currentNode.leftChild)
            else:
                currentNode.leftChild = AVLTreeNode(key, val, parent=currentNode)
                self.updateBalance(currentNode.leftChild)
        elif key > currentNode.key:
            if currentNode.hasRightChild():
                self._put(key, val, currentNode.rightChild)
            else:
                currentNode.rightChild = AVLTreeNode(key, val, parent=currentNode)
                self.updateBalance(currentNode.rightChild)
        else:
            currentNode.payload = val
            self.size = self.size + 1

    def updateBalance(self, node):
        if node.balanceFactor > 1 or node.balanceFactor < -1:
            self.rebalance(node)
        if node.parent is None:
            if node.isLeftChild():
                node.parent.balanceFactor = 1
            elif node.isRightChild():
                node.parent.balanceFactor = -1
            return

    def rebalance(self, node):
        newRoot = rotRoot.rightChild
        rotRoot.rightChild = newRoot.leftChild
        if newRoot.leftChild is None:
            newRoot.leftChild = rotRoot
            newRoot.parent = rotRoot.parent
        elif rotRoot.isRoot():
            self.root = newRoot
        else:
            if rotRoot.isLeftChild():
                rotRoot.parent.leftChild = newRoot
            else:
                rotRoot.parent.rightChild = newRoot
            newRoot.leftChild = rotRoot
            rotRoot.parent = newRoot
            rotRoot.balanceFactor = rotRoot.balanceFactor + 1 - min(rotRoot.balanceFactor, 0)
            newRoot.balanceFactor = newRoot.balanceFactor + 1 + max(rotRoot.balanceFactor, 0)

    def rotateLeft(self, node, rotRoot):
        newRoot = rotRoot.rightChild
        rotRoot.rightChild = newRoot.leftChild
        if newRoot.leftChild is None:
            newRoot.leftChild = rotRoot
            newRoot.parent = rotRoot.parent
        elif rotRoot.isRoot():
            self.root = newRoot
        else:
            if rotRoot.isLeftChild():
                rotRoot.parent.leftChild = newRoot
            else:
                rotRoot.parent.rightChild = newRoot
            newRoot.leftChild = rotRoot
            rotRoot.parent = newRoot
            rotRootbalanceFactor = rotRoot.balanceFactor + 1 - min(rotRoot.balanceFactor, 0)
            newRoot.balanceFactor = newRoot.balanceFactor + 1 + max(rotRoot.balanceFactor, 0)

    def rotateRight(self, node, rotRoot):
        newRoot = rotRoot.leftChild
        rotRoot.leftChild = newRoot.rightChild
        if newRoot.rightChild is None:
            newRoot.rightChild = rotRoot
            newRoot.parent = rotRoot.parent
        elif rotRoot.isRoot():
            self.root = newRoot
        else:
            if rotRoot.isLeftChild():
                rotRoot.parent.leftChild = newRoot
            else:
                rotRoot.parent.rightChild = newRoot
            newRoot.leftChild = rotRoot
            rotRoot.parent = newRoot
            rotRoot.balanceFactor = rotRoot.balanceFactor + 1 - min(rotRoot.balanceFactor, 0)
            newRoot.balanceFactor = newRoot.balanceFactor + 1 + max(rotRoot.balanceFactor, 0)
```

**# NOTE:** You will complete rotateRight in Lab.
c) Trace the code for `myAVL.put(90, None)` by updating the below diagram:

Consider balance factor formulas for `rotateLeft`. We know: $newBal(B) = h_d - h_c$ and $oldBal(B) = h_d - (1+\max(h_c, h_e))$
$newBal(D) = 1 + \max(h_d, h_c) - h_e$ and $oldBal(D) = h_c - h_e$

Before left rotation:

After left rotation at pivot:

Consider: $newBal(B) - oldBal(B) = (h_d - (1 + \max(h_c, h_e)))$
$newBal(B) - oldBal(B) = h_d - h_a - h_d + (1 + \max(h_c, h_e))$
$newBal(B) - oldBal(B) = 1 + \max(h_c, h_e) - h_c$
$newBal(B) - oldBal(B) = 1 + \max(h_c, h_e) - h_c$
$newBal(B) = oldBal(B) + 1 + \max(0, -oldBal(D))$
$newBal(B) = oldBal(B) + 1 - \min(0, oldBal(D))$, so

\[
rotRoot.balanceFactor = \min(\text{rotRoot.balanceFactor} + 1 - \min(\text{newRoot.balanceFactor}, 0), 0)
\]
3. Complete the below figure which is a "mirror image" to the figure on page 2, i.e., inserting into the pivot's left child's left subtree. Include correct balance factors after the rotation.

b) Complete the below figure which is a "mirror image" to the figure on page 3, i.e., inserting into the pivot's left child's right subtree. Include correct balance factors after the rotation.
1. BST, AVL trees, and hash tables can all be used to implement a dictionary ADT.

<table>
<thead>
<tr>
<th></th>
<th>Items added in sorted order</th>
<th>Items added in random order</th>
<th>Order did not matter (Hash table sizes $2^{15} = 32K$)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>BST</td>
<td>AVL Tree</td>
<td>BST</td>
</tr>
<tr>
<td>Total add/put time</td>
<td>47.785</td>
<td>0.205</td>
<td>0.119</td>
</tr>
<tr>
<td>Total search time</td>
<td>38.100</td>
<td>0.060</td>
<td>0.079</td>
</tr>
<tr>
<td>Height of resulting tree</td>
<td>9,999</td>
<td>13</td>
<td>30</td>
</tr>
</tbody>
</table>

a) The puts of these 10,000 randomly ordered items into the BST took 0.119 seconds and 0.195 seconds into the AVL tree. Why did the BST puts take less time even though the final height was 30 vs. a final AVL tree height of 15? *extra time in AVL to do rebalancing and rotation(s)*

b) With a very, very poor hash function or very, very bad choice of keys, then all keys could hash to the same home address.

- What would be the worst-case big-oh of open-address hashing with quadratic probing? $O(n)$
- What would be the worst-case big-oh of chaining using a linked list at each home address (i.e., ChainingDict)? $O(n)$
- What would be the worst-case big-oh of chaining using an AVL tree at each home address? $O(\log_2 n)$

![AVL Tree Diagram]

2. The data structures we have discussed so far are all in-memory, i.e., data is stored in main/RAM memory. Data can also be stored on secondary storage in a file (e.g., moiveData.txt file). Currently, most secondary storage consists of hard-disks.

a) Complete the following table comparing main/RAM memory vs. hard-disk:

<table>
<thead>
<tr>
<th>Criteria</th>
<th>Main/RAM memory</th>
<th>Hard-disk Drive</th>
<th>Solid-State Drive</th>
</tr>
</thead>
<tbody>
<tr>
<td>Size on a typical desktop computer</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Average access time</td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

b) Which criterion seems to be the most important difference between the main and secondary memories?