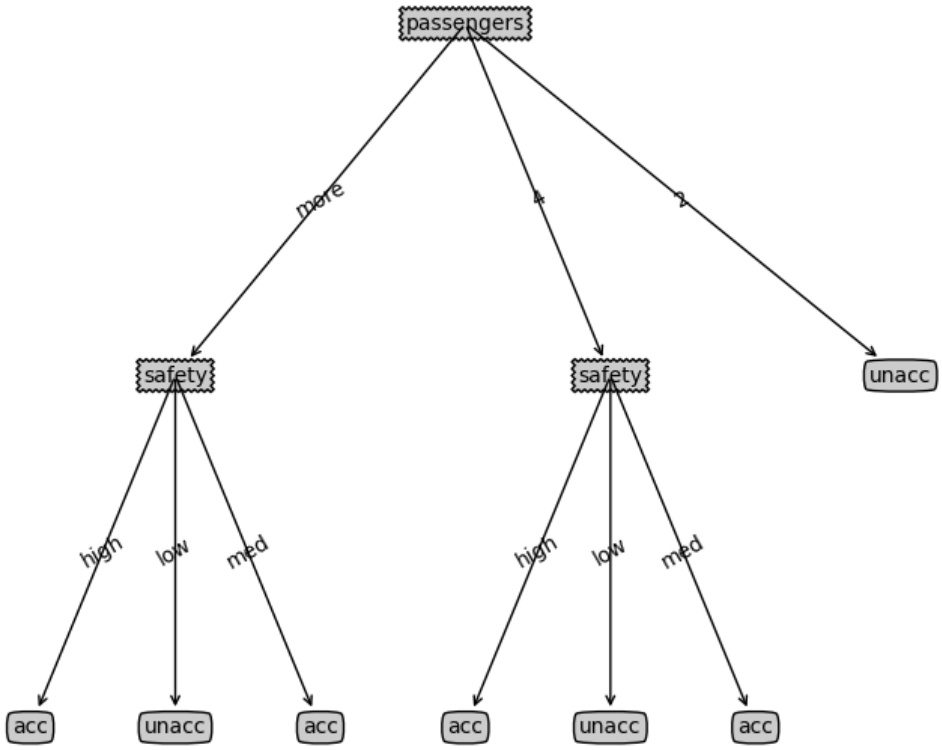
Assignment 4 - Report

1.) The tree drawn based on the output obtained from the Car Dataset using the algorithm.



2.) The code of the functions that were implemented:

```
import treeplot
import numpy as np

# Function to implement Gini Index Calculations:

def gini_index(Splitting_Feature):
    """
    Given the observations of a Feature, calculating the GINI index (measure for impurity)
    """

    observations = list()

    for unq_values in np.unique(Splitting_Feature):
        y_count = 0
        for values in Splitting_Feature:
            if values == unq_values:     # example: Unique Attribute Values for Labels: Yes, No
                y_count = y_count + 1
        observations.append(y_count)

    if(len(np.unique(Splitting_Feature)) != 1):
        n = sum(observations)
        p1 = observations[0]/n
        p2 = observations[1]/n
        gini_ind = (1.0 - ((p1**2)+(p2**2)))

    else:
        n = sum(observations)
        p1 = observations[0]/n
        gini_ind = (1.0 - p1**2)

    return gini_ind


def chooseBestFeature(dataSet):
    '''
    choose best feature to split based on Gini index

    Parameters
    ----------
    dataSet: 2-D list
        [n_sampels, m_features + 1]
```

```
    the last column is class label

Returns
------------------
bestFeatId: int
    index of the best feature
'''

#TODO

classlabels = list()
classlabels = [row[len(dataSet[0])-1] for row in dataSet]

gini_classlabels = gini_index(classlabels)

# Initialization
InfoGain = list()
bestFeatId = 999
bestInfoGain = -1

for index in range(len(dataSet[0])-1):
    feature = list()      # contains attribute values
    gini_ind = list()     # to store the required gini index values accordingly

    for row in dataSet:
        feature.append(row[index])

    n = len(feature)      # no. of values

    for fval in np.unique(feature):      # Consider only unique attribute values
        value = list()     # contains unique attribute values
        value.append(fval)
        subset1 = list()      # to find the subset based on the given axis and feature values
        value = set(value)

        for row in dataSet:
            if value.issubset(row):
                subset1.append(row[len(dataSet[0])-1])

        n1 = len(subset1)     # no. of values in subset

        gini = gini_index(subset1)
        gini_ind.append((n1/n)* gini)
```

```
        gini_feature = sum(gini_ind)
        InfoGain.append(gini_classlabels - gini_feature)

    bestFeatId = InfoGain.index(max(InfoGain))
    bestInfoGain = max(InfoGain)

    # Find best gain and corresponding feature ID
    return bestFeatId


def stopCriteria(dataSet):
    '''
    Criteria to stop splitting:
    1) if all the classe labels are the same, then return the class label;
    2) if there are no more features to split, then return the majority label of the subset.

    Parameters
    ----------
    dataSet: 2-D list
        [n_sampels, m_features + 1]
        the last column is class label

    Returns
    -------
    assignedLabel: string
        if satisfying stop criteria, assignedLabel is the assigned class label;
        else, assignedLabel is None
    '''
    assignedLabel = None
    # TODO

    no_of_columns = len(dataSet[0])
    classlabels = []
    classlabels = [row[no_of_columns - 1] for row in dataSet]


    # A set cannot have duplicates.
    # So if all the elements in the original list are identical,
    # the set will have just one element.
    if len(set(classlabels)) == 1:
        assignedLabel = classlabels[0]

    # Finding Feature Space:
    bestFeatId = chooseBestFeature(dataSet)
```

```
    features = [index for index in range(len(dataSet[0])-1) if index != bestFeatId]

    if len(features) == 0:  # Implies: no more features to split
        assignedLabel = max(set(classlabels), key = classlabels.count)   # Finding Mode of
the classlabels list - Python: Naive Approach

    return assignedLabel
```

Tree:

Output: