NAME:- Harshil Agarwal

4

dtype: int64

7.4

0.70

0.00

1.9

0.076

11.0

Reg No:- RA1911003010325

Machine Learning - Core Concepts with Applications

ASSIGNMENT - 5

Classification using Decision Trees and Random Forest

DATASET:- RED WINE QUALITY CLASSIFICATION

Importing Libraries and Reading the Dataset

```
import numpy as np
In [6]:
          import pandas as pd
          import matplotlib.pyplot as plt
          import seaborn as sns
          from sklearn.metrics import accuracy score
          from sklearn.metrics import mean squared error
          from sklearn.model selection import cross val score
          from sklearn.metrics import confusion matrix
          from collections import Counter
          from IPython.core.display import display, HTML
          sns.set style('darkgrid')
         dataset = pd.read csv('winequality-red.csv')
          dataset.head()
            fixed acidity volatile acidity citric acid residual sugar chlorides free sulfur dioxide total sulfur dioxide density
                                                                                                                   pH sulphates alcohol
                                                                                                                                        quality
         0
                    7.4
                                 0.70
                                           0.00
                                                                 0.076
                                                                                   11.0
                                                                                                          0.9978
                                                                                                                 3.51
                                                                                                                                    9.4
                                                                                                                                              5
                                                         1.9
                                                                                                     34.0
                                                                                                                            0.56
                    7.8
                                0.88
                                           0.00
                                                         2.6
                                                                 0.098
                                                                                   25.0
                                                                                                                                     9.8
                                                                                                     67.0
                                                                                                          0.9968
                                                                                                                 3.20
                                                                                                                            0.68
                                                                                                                                              5
         2
                    7.8
                                 0.76
                                           0.04
                                                         2.3
                                                                 0.092
                                                                                   15.0
                                                                                                          0.9970
                                                                                                                 3.26
                                                                                                                            0.65
                                                                                                                                    9.8
                                                                                                                                              5
                                                                                                     54.0
                   11.2
                                0.28
                                           0.56
                                                          1.9
                                                                 0.075
                                                                                   17.0
                                                                                                     60.0
                                                                                                          0.9980
                                                                                                                 3.16
                                                                                                                            0.58
                                                                                                                                    9.8
                                                                                                                                              6
```

Data Visualization and Preprocessing

34.0 0.9978 3.51

0.56

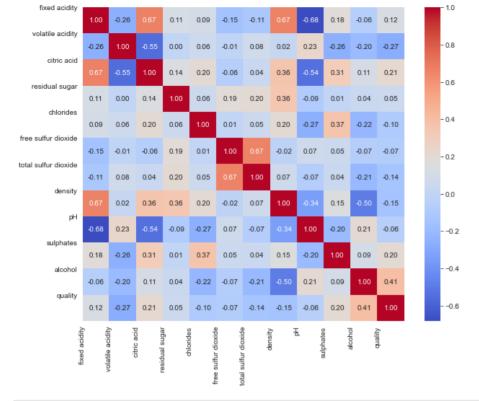
9.4

5

```
dataset.isnull().sum()
Out[8]: fixed acidity
                                0
        volatile acidity
                                0
        citric acid
                                0
        residual sugar
                                0
        chlorides
        free sulfur dioxide
        total sulfur dioxide
        density
        рН
        sulphates
                                0
        alcohol
        quality
```

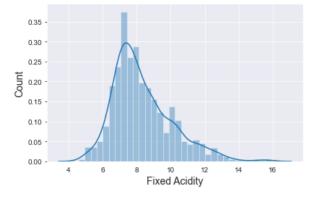
```
In [9]:
          bins = (2, 6.5, 8)
          labels = ['bad', 'good']
          dataset['quality'] = pd.cut(x = dataset['quality'], bins = bins, labels = labels)
          dataset['quality'].value counts()
In [10]:
Out[10]: bad
                  1382
         good
                  217
         Name: quality, dtype: int64
In [11]: from sklearn.preprocessing import LabelEncoder
          labelencoder y = LabelEncoder()
          dataset['quality'] = labelencoder y.fit transform(dataset['quality'])
          dataset.head()
In [12]:
            fixed acidity volatile acidity citric acid residual sugar chlorides free sulfur dioxide total sulfur dioxide density
                                                                                                              pH sulphates alcohol quality
         0
                    7.4
                                0.70
                                          0.00
                                                        1.9
                                                               0.076
                                                                                11.0
                                                                                                 34.0 0.9978 3.51
                                                                                                                       0.56
                                                                                                                               9.4
                                                                                                                                        0
          1
                    7.8
                                0.88
                                          0.00
                                                        2.6
                                                               0.098
                                                                                25.0
                                                                                                 67.0 0.9968 3.20
                                                                                                                       0.68
                                                                                                                               9.8
                                                                                                                                        0
          2
                    7.8
                                0.76
                                          0.04
                                                       2.3
                                                               0.092
                                                                                15.0
                                                                                                 54.0 0.9970 3.26
                                                                                                                       0.65
                                                                                                                               9.8
                                                                                                                                        0
          3
                   11.2
                                0.28
                                          0.56
                                                        1.9
                                                               0.075
                                                                                 17.0
                                                                                                 60.0 0.9980 3.16
                                                                                                                       0.58
                                                                                                                                        0
                                                                                                                               9.8
          4
                    7.4
                                0.70
                                          0.00
                                                        1.9
                                                               0.076
                                                                                 11.0
                                                                                                 34.0 0.9978 3.51
                                                                                                                       0.56
                                                                                                                               9.4
                                                                                                                                        0
          corr = dataset.corr()
In [13]:
          #Plot figsize
          fig, ax = plt.subplots(figsize=(10, 8))
          #Generate Heat Map, allow annotations and place floats in map
          sns.heatmap(corr, cmap='coolwarm', annot=True, fmt=".2f")
          #Apply xticks
          plt.xticks(range(len(corr.columns)), corr.columns);
          #Apply yticks
          plt.yticks(range(len(corr.columns)), corr.columns)
          #show plot
```

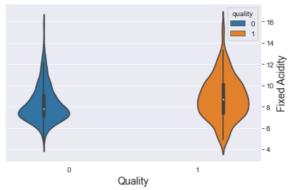
plt.show()



```
In [14]: f, axes = plt.subplots(1,2,figsize=(14,4))
sns.distplot(dataset['fixed acidity'], ax = axes[0])
axes[0].set_xlabel('Fixed Acidity', fontsize=14)
axes[0].set_ylabel('Count', fontsize=14)
axes[0].yaxis.tick_left()

sns.violinplot(x = 'quality', y = 'fixed acidity', data = dataset, hue = 'quality', ax = axes[1])
axes[1].set_xlabel('Quality', fontsize=14)
axes[1].set_ylabel('Fixed Acidity', fontsize=14)
axes[1].yaxis.set_label_position("right")
axes[1].yaxis.set_label_position("right")
axes[1].yaxis.tick_right()
```

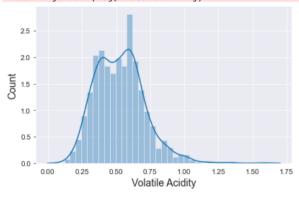


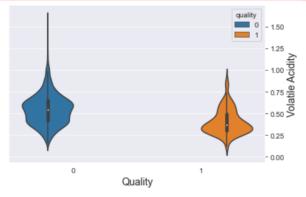


```
In [15]: f, axes = plt.subplots(1,2,figsize=(14,4))

sns.distplot(dataset['volatile acidity'], ax = axes[0])
axes[0].set_xlabel('Volatile Acidity', fontsize=14)
axes[0].set_ylabel('Count', fontsize=14)
axes[0].yaxis.tick_left()

sns.violinplot(x = 'quality', y = 'volatile acidity', data = dataset, hue = 'quality', ax = axes[1])
axes[1].set_xlabel('Quality', fontsize=14)
axes[1].set_ylabel('Volatile Acidity', fontsize=14)
axes[1].yaxis.set_label_position("right")
axes[1].yaxis.tick_right()
```





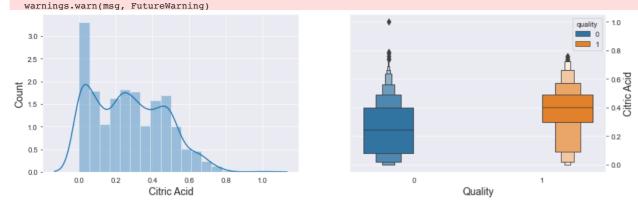
```
In [16]: f, axes = plt.subplots(1,2,figsize=(14,4))

sns.distplot(dataset['citric acid'], ax = axes[0])
axes[0].set_xlabel('Citric Acid', fontsize=14)
axes[0].set_ylabel('Count', fontsize=14)
axes[0].yaxis.tick_left()

sns.boxenplot(x = 'quality', y = 'citric acid', data = dataset, hue = 'quality', ax = axes[1])
axes[1].set_xlabel('Quality', fontsize=14)
axes[1].set_ylabel('Citric Acid', fontsize=14)
axes[1].yaxis.set_label_position("right")
axes[1].yaxis.tick_right()

plt.show()
```

/Users/harshilagarwal/opt/anaconda3/lib/python3.8/site-packages/seaborn/distributions.py:2551: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

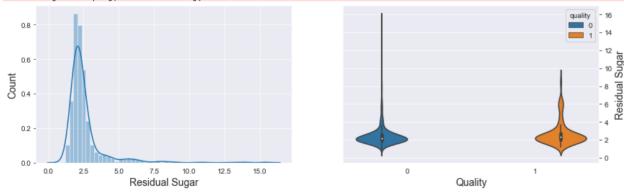


```
In [17]: f, axes = plt.subplots(1,2,figsize=(14,4))

sns.distplot(dataset['residual sugar'], ax = axes[0])
axes[0].set_xlabel('Residual Sugar', fontsize=14)
axes[0].set_ylabel('Count', fontsize=14)
axes[0].yaxis.tick_left()

sns.violinplot(x = 'quality', y = 'residual sugar', data = dataset, hue = 'quality', ax = axes[1])
axes[1].set_xlabel('Quality', fontsize=14)
axes[1].set_ylabel('Residual Sugar', fontsize=14)
axes[1].yaxis.set_label_position("right")
axes[1].yaxis.tick_right()

plt.show()
```



```
In [18]: f, axes = plt.subplots(1,2,figsize=(14,4))

sns.distplot(dataset['chlorides'], ax = axes[0])
axes[0].set_xlabel('Chlorides', fontsize=14)
axes[0].set_ylabel('Count', fontsize=14)
axes[0].yaxis.tick_left()

sns.violinplot(x = 'quality', y = 'chlorides', data = dataset, hue = 'quality', ax = axes[1])
axes[1].set_xlabel('Quality', fontsize=14)
axes[1].set_ylabel('Chlorides', fontsize=14)
axes[1].yaxis.set_label_position("right")
axes[1].yaxis.tick right()
```

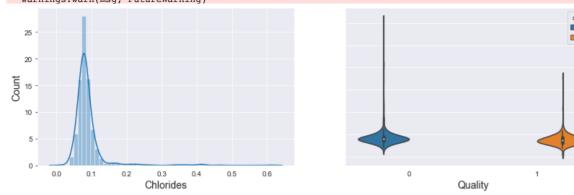
```
plt.show()
```

/Users/harshilagarwal/opt/anaconda3/lib/python3.8/site-packages/seaborn/distributions.py:2551: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)

-0.4 sepido

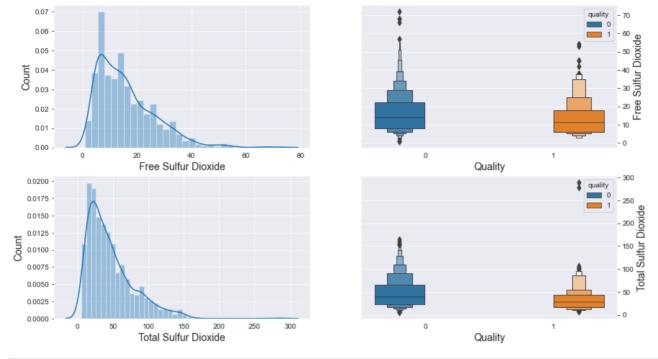
-0.2 -0.1



```
f, axes = plt.subplots(2,2,figsize=(14,8))
sns.distplot(dataset['free sulfur dioxide'], ax = axes[0,0])
axes[0,0].set xlabel('Free Sulfur Dioxide', fontsize=14)
axes[0,0].set ylabel('Count', fontsize=14)
axes[0,0].yaxis.tick left()
sns.boxenplot(x = 'quality', y = 'free sulfur dioxide', data = dataset, hue = 'quality', ax = axes[0,1])
axes[0,1].set xlabel('Quality', fontsize=14)
axes[0,1].set ylabel('Free Sulfur Dioxide', fontsize=14)
axes[0,1].yaxis.set label position("right")
axes[0,1].yaxis.tick right()
sns.distplot(dataset['total sulfur dioxide'], ax = axes[1,0])
axes[1,0].set xlabel('Total Sulfur Dioxide', fontsize=14)
axes[1,0].set ylabel('Count', fontsize=14)
axes[1,0].yaxis.tick left()
sns.boxenplot(x = 'quality', y = 'total sulfur dioxide', data = dataset, hue = 'quality', ax = axes[1,1])
axes[1,1].set xlabel('Quality', fontsize=14)
axes[1,1].set ylabel('Total Sulfur Dioxide', fontsize=14)
axes[1,1].yaxis.set label position("right")
axes[1,1].yaxis.tick right()
plt.show()
```

/Users/harshilagarwal/opt/anaconda3/lib/python3.8/site-packages/seaborn/distributions.py:2551: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)

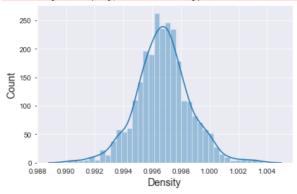


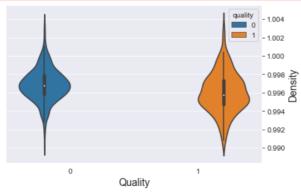
```
In [20]: f, axes = plt.subplots(1,2,figsize=(14,4))

sns.distplot(dataset['density'], ax = axes[0])
axes[0].set_xlabel('bensity', fontsize=14)
axes[0].set_ylabel('Count', fontsize=14)
axes[0].yaxis.tick_left()

sns.violinplot(x = 'quality', y = 'density', data = dataset, hue = 'quality', ax = axes[1])
axes[1].set_xlabel('Quality', fontsize=14)
axes[1].set_ylabel('Density', fontsize=14)
axes[1].yaxis.set_label_position("right")
axes[1].yaxis.tick_right()

plt.show()
```



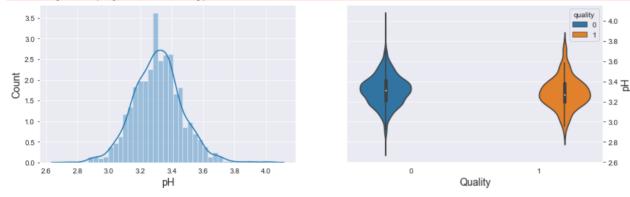


```
sns.distplot(dataset['pH'], ax = axes[0])
axes[0].set_xlabel('pH', fontsize=14)
axes[0].set_ylabel('Count', fontsize=14)
axes[0].yaxis.tick_left()

sns.violinplot(x = 'quality', y = 'pH', data = dataset, hue = 'quality', ax = axes[1])
axes[1].set_xlabel('Quality', fontsize=14)
axes[1].set_ylabel('pH', fontsize=14)
axes[1].yaxis.set_label_position("right")
axes[1].yaxis.tick_right()
```

/Users/harshilagarwal/opt/anaconda3/lib/python3.8/site-packages/seaborn/distributions.py:2551: FutureWarning: `distplot` is a deprecated function and will be removed in a future versio n. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

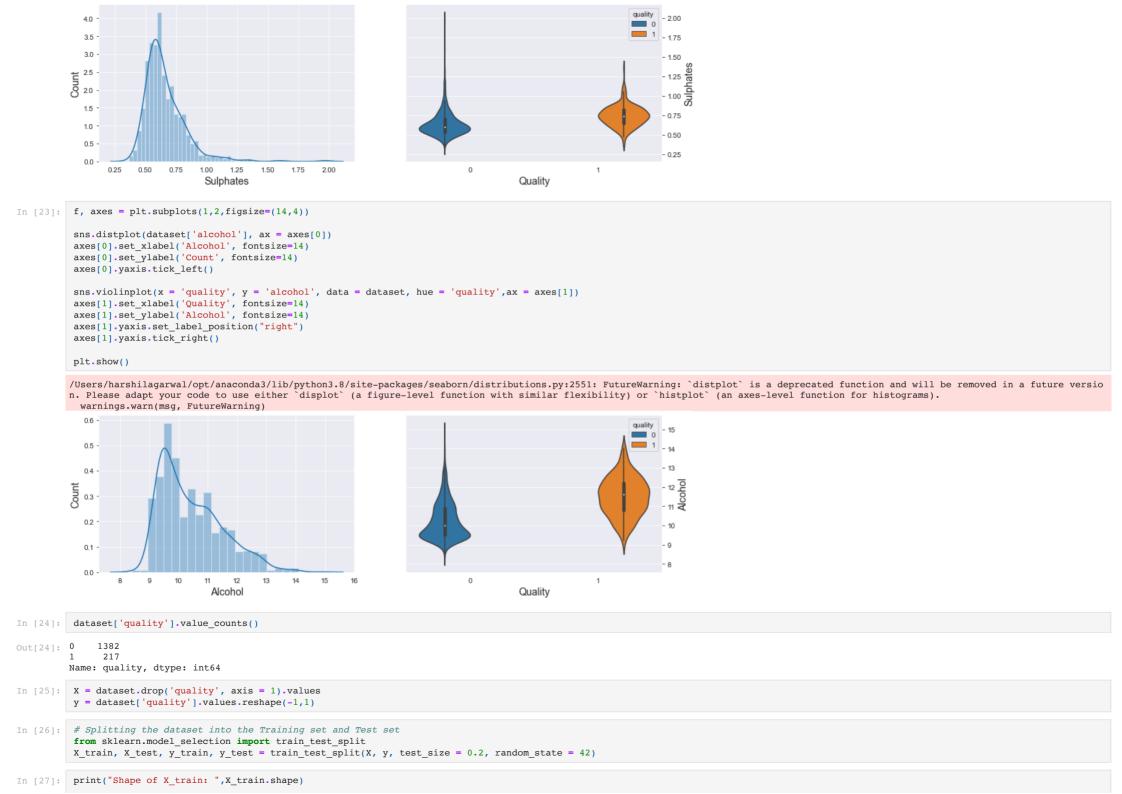
warnings.warn(msq, FutureWarning)



```
In [22]: f, axes = plt.subplots(1,2,figsize=(14,4))

sns.distplot(dataset['sulphates'], ax = axes[0])
axes[0].set_xlabel('Sulphates', fontsize=14)
axes[0].set_ylabel('Count', fontsize=14)
axes[0].yaxis.tick_left()

sns.violinplot(x = 'quality', y = 'sulphates', data = dataset, hue = 'quality', ax = axes[1])
axes[1].set_xlabel('Quality', fontsize=14)
axes[1].set_ylabel('Sulphates', fontsize=14)
axes[1].yaxis.set_label_position("right")
axes[1].yaxis.tick_right()
```



```
print("Shape of X_test: ", X_test.shape)
print("Shape of Y_terin: ", Y_test.shape)
print("Shape of Y_test.", Y_test.shape)

Shape of X_test: (329, 11)
Shape of Y_terin: (1279, 12)
Shape of Y_test: (320, 11)
Shape of Y_test: (320, 1)

Classification Models

In [28]: # Feature Scaling
from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
X_train_scaled = sc.fit_transform(X_test)

Decision Tree Classification

In [29]: # Fitting classifier to the Training set
from sklearn.tree import DecisionTreeClassifier
```

```
from sklearn.tree import DecisionTreeClassifier
          classifier dt = DecisionTreeClassifier(criterion = 'gini', max features=6, max leaf nodes=400, random state = 33)
          classifier dt.fit(X train scaled, y train.ravel())
Out[29]: DecisionTreeClassifier(max_features=6, max_leaf_nodes=400, random_state=33)
In [30]: # Predicting Cross Validation Score
          cv dt = cross val score(estimator = classifier dt, X = X train scaled, y = y train.ravel(), cv = 10)
          print("CV: ", cv_dt.mean())
          y pred dt train = classifier dt.predict(X train scaled)
          accuracy_dt_train = accuracy_score(y_train, y_pred_dt_train)
          print("Training set: ", accuracy dt train)
          y pred dt test = classifier dt.predict(X test scaled)
          accuracy dt test = accuracy score(y test, y pred dt test)
          print("Test set: ", accuracy dt test)
         CV: 0.8960014763779528
         Training set: 1.0
         Test set: 0.878125
          confusion matrix(y test, y pred dt test)
Out[31]: array([[252, 21],
                [ 18, 29]])
          tp dt = confusion matrix(y test, y pred dt test)[0,0]
In [32]:
          fp_dt = confusion_matrix(y_test, y_pred_dt_test)[0,1]
          tn dt = confusion matrix(y test, y pred dt test)[1,1]
          fn_dt = confusion_matrix(y_test, y_pred_dt_test)[1,0]
```

Random Forest Classification

Fitting Random Forest Classification to the Training set

```
y pred rf train = classifier rf.predict(X train scaled)
          accuracy rf train = accuracy score(y train, y pred rf train)
          print("Training set: ", accuracy rf train)
          y pred rf test = classifier rf.predict(X test scaled)
          accuracy rf test = accuracy score(y test, y pred rf test)
          print("Test set: ", accuracy rf test)
         CV: 0.9140194389763779
         Training set: 1.0
         Test set: 0.9125
         confusion matrix(y test, y pred rf test)
Out[35]: array([[267, 6],
                [ 22, 25]])
In [36]: tp_rf = confusion_matrix(y_test, y_pred rf test)[0,0]
          fp rf = confusion matrix(y test, y pred rf test)[0,1]
          tn rf = confusion_matrix(y_test, y_pred_rf_test)[1,1]
          fn rf = confusion matrix(y test, y pred rf test)[1,0]
```

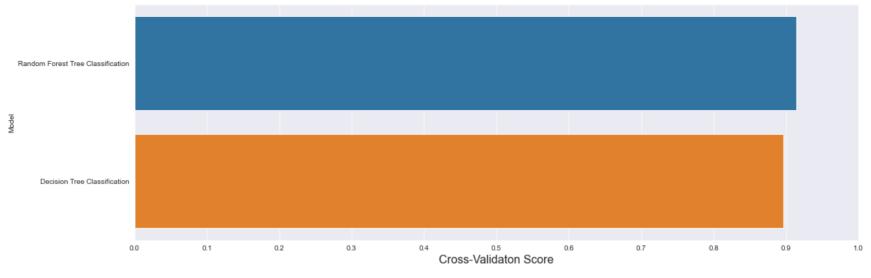
OBSERVATION

38]:		Model	True Positive	False Positive	True Negative	False Negative	Accuracy(training)	Accuracy(test)	Cross-Validation
	0	Decision Tree Classification	252	21	29	18	1.0	0.878125	0.896001
	1	Random Forest Tree Classification	267	6	25	22	1.0	0.912500	0.914019

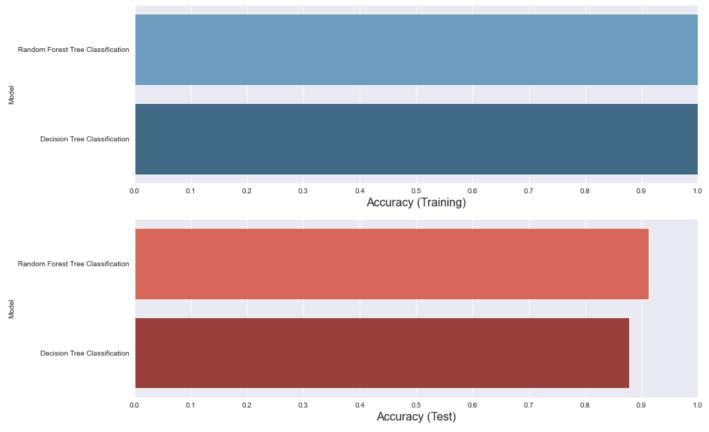
```
In [39]: f, axe = plt.subplots(1,1, figsize=(18,6))

predict.sort_values(by=['Cross-Validation'], ascending=False, inplace=True)

sns.barplot(x='Cross-Validation', y='Model', data = predict, ax = axe)
#axes[0].set(xlabel='Region', ylabel='Charges')
axe.set_xlabel('Cross-Validaton Score', size=16)
axe.set_ylabel('Model')
axe.set_xlim(0,1.0)
axe.set_xlim(0,1.0)
axe.set_xticks(np.arange(0, 1.1, 0.1))
plt.show()
```



```
In [40]: f, axes = plt.subplots(2,1, figsize=(14,10))
          predict.sort values(by=['Accuracy(training)'], ascending=False, inplace=True)
          sns.barplot(x='Accuracy(training)', y='Model', data = predict, palette='Blues_d', ax = axes[0])
          #axes[0].set(xlabel='Region', ylabel='Charges')
          axes[0].set_xlabel('Accuracy (Training)', size=16)
          axes[0].set ylabel('Model')
          axes[0].set xlim(0,1.0)
          axes[0].set_xticks(np.arange(0, 1.1, 0.1))
          predict.sort values(by=['Accuracy(test)'], ascending=False, inplace=True)
          sns.barplot(x='Accuracy(test)', y='Model', data = predict, palette='Reds d', ax = axes[1])
          #axes[0].set(xlabel='Region', ylabel='Charges')
          axes[1].set_xlabel('Accuracy (Test)', size=16)
          axes[1].set_ylabel('Model')
          axes[1].set xlim(0,1.0)
          axes[1].set_xticks(np.arange(0, 1.1, 0.1))
          plt.show()
```



```
In [41]: predict.sort_values(by=(['Accuracy(test)']), ascending=True, inplace=True)

f, axe = plt.subplots(1,1, figsize=(24,8))
    sns.barplot(x = predict['Model'], y=predict['False Positive'] + predict['False Negative'], ax = axe)
    axe.set_xlabel('Model', size=20)
    axe.set_ylabel('False Observations', size=20)

plt.show()
```

