(ECE 418) Machine Learning Project

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Importing Libraries

```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
import seaborn as sns

from warnings import filterwarnings
filterwarnings(action='ignore')
```

Loading Dataset

```
wine = pd.read_csv("/content/winequality-red.csv")
print("Successfully Imported Data!")
wine.head()
```

Successfully Imported Data!

	fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfur dioxide	density	рН	S
0	7.4	0.70	0.00	1.9	0.076	11.0	34.0	0.9978	3.51	
1	7.8	0.88	0.00	2.6	0.098	25.0	67.0	0.9968	3.20	
2	7.8	0.76	0.04	2.3	0.092	15.0	54.0	0.9970	3.26	
3	11.2	0.28	0.56	1.9	0.075	17.0	60.0	0.9980	3.16	

Description and Finding Null Values

```
volatile acidity
citric acid
residual sugar
                        0
chlorides
                        0
free sulfur dioxide
total sulfur dioxide
density
                        0
рΗ
sulphates
                        0
alcohol
                        0
quality
                        0
dtype: int64
```

wine.groupby('quality').mean()

	fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfur dioxide	den
quality								
3	8.360000	0.884500	0.171000	2.635000	0.122500	11.000000	24.900000	0.997
4	7.779245	0.693962	0.174151	2.694340	0.090679	12.264151	36.245283	0.996
5	8.167254	0.577041	0.243686	2.528855	0.092736	16.983847	56.513950	0.997
6	8.347179	0.497484	0.273824	2.477194	0.084956	15.711599	40.869906	0.996
7	8.872362	0.403920	0.375176	2.720603	0.076588	14.045226	35.020101	0.996

Feature Selection

```
# Create Classification version of target variable
wine['goodquality'] = [1 if x >= 7 else 0 for x in wine['quality']]# Separate feature vari
X = wine.drop(['quality','goodquality'], axis = 1)
Y = wine['goodquality']

# See proportion of good vs bad wines
wine['goodquality'].value_counts()

0     1382
     1     217
     Name: goodquality, dtype: int64
```

X # attributes which are responsible for classification

	fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfur dioxide	density	рН
0	7.4	0.700	0.00	1.9	0.076	11.0	34.0	0.99780	3.51
1	7.8	0.880	0.00	2.6	0.098	25.0	67.0	0.99680	3.20
2	7.8	0.760	0.04	2.3	0.092	15.0	54.0	0.99700	3.26
3	11.2	0.280	0.56	1.9	0.075	17.0	60.0	0.99800	3.16
4	7.4	0.700	0.00	1.9	0.076	11.0	34.0	0.99780	3.51
1594	6.2	0.600	0.08	2.0	0.090	32.0	44.0	0.99490	3.45
(Y)									
0 1 2 3 4	0 0 0 0 0								
(1 2 3 4 1594 (Y)	acidity 0 7.4 1 7.8 2 7.8 3 11.2 4 7.4 1594 6.2 (Y) 6 0 1 0 2 0 3 0 4 0	acidity acidity 0 7.4 0.700 1 7.8 0.880 2 7.8 0.760 3 11.2 0.280 4 7.4 0.700 1594 6.2 0.600 (Y) 2 0 3 0 4 0	acidity acidity acid 0 7.4 0.700 0.00 1 7.8 0.880 0.00 2 7.8 0.760 0.04 3 11.2 0.280 0.56 4 7.4 0.700 0.00 1594 6.2 0.600 0.08 (Y) 0 0 0 0 1 0 0 0 0 2 0 0 0 0 3 0 0 0 0 4 0 0 0 0 5 0 0 0 0 6 0 0 0 0 6 0 0 0 0 6 0 0 0 0 7 0 0 0 0 8 0 0 0<	acidity acidity acid sugar 0 7.4 0.700 0.00 1.9 1 7.8 0.880 0.00 2.6 2 7.8 0.760 0.04 2.3 3 11.2 0.280 0.56 1.9 4 7.4 0.700 0.00 1.9 1594 6.2 0.600 0.08 2.0 (Y) 0 0 0 0 0 1 0 0 0 0 0 0 0 3 0 </td <td>acidity acidity acid sugar chlorides 0 7.4 0.700 0.00 1.9 0.076 1 7.8 0.880 0.00 2.6 0.098 2 7.8 0.760 0.04 2.3 0.092 3 11.2 0.280 0.56 1.9 0.075 4 7.4 0.700 0.00 1.9 0.076 1594 6.2 0.600 0.08 2.0 0.090 (Y) 0</td> <td> 11xed acidity acidity acid sugar chlorides sulfur dioxide 0</td> <td> Tixed acidity Citric acid Sugar Chlorides Sulfur dioxide </td> <td> Tixed acidity Citric acid Sugar Chlorides Sulfur dioxide Sulfur dioxide Chlorides Chlori</td>	acidity acidity acid sugar chlorides 0 7.4 0.700 0.00 1.9 0.076 1 7.8 0.880 0.00 2.6 0.098 2 7.8 0.760 0.04 2.3 0.092 3 11.2 0.280 0.56 1.9 0.075 4 7.4 0.700 0.00 1.9 0.076 1594 6.2 0.600 0.08 2.0 0.090 (Y) 0	11xed acidity acidity acid sugar chlorides sulfur dioxide 0	Tixed acidity Citric acid Sugar Chlorides Sulfur dioxide	Tixed acidity Citric acid Sugar Chlorides Sulfur dioxide Sulfur dioxide Chlorides Chlori

Name: goodquality, Length: 1599, dtype: int64

Feature Importance

```
from sklearn.linear_model import LogisticRegression
model = LogisticRegression()

from sklearn.ensemble import ExtraTreesClassifier
classifiern = ExtraTreesClassifier()
classifiern.fit(X,Y)
score = classifiern.feature_importances_
print(score)

[0.07556531 0.09627568 0.0986353 0.07197105 0.06738145 0.06913588]
```

0.0876835 0.08411249 0.06778224 0.10949398 0.17196312]

Splitting Dataset

```
from sklearn.model_selection import train_test_split
X_train, X_test, Y_train, Y_test = train_test_split(X,Y,test_size=0.3,random_state=7)
```

LogisticRegression:

```
from sklearn.linear_model import LogisticRegression
model = LogisticRegression()
model.fit(X train,Y train)
Y_pred = model.predict(X_test)
from sklearn.metrics import accuracy score, confusion matrix
print("Accuracy Score:",accuracy_score(Y_test,Y_pred))
     Accuracy Score: 0.872916666666667
confusion_mat = confusion_matrix(Y_test,Y_pred)
print(confusion_mat)
     [[399 18]
      [ 43 20]]
Using KNN:
from sklearn.neighbors import KNeighborsClassifier
model = KNeighborsClassifier(n_neighbors=3)
model.fit(X train,Y train)
y_pred = model.predict(X_test)
from sklearn.metrics import accuracy_score
print("Accuracy Score:",accuracy_score(Y_test,y_pred))
     Accuracy Score: 0.872916666666667
Using SVM:
from sklearn.svm import SVC
model = SVC()
model.fit(X_train,Y_train)
pred_y = model.predict(X_test)
from sklearn.metrics import accuracy_score
print("Accuracy Score:",accuracy score(Y test,pred y))
     Accuracy Score: 0.86875
Using Decision Tree:
from sklearn.tree import DecisionTreeClassifier
model = DecisionTreeClassifier(criterion='entropy',random state=7)
model.fit(X train,Y train)
y_pred = model.predict(X_test)
from sklearn.metrics import accuracy score
print("Accuracy Score:",accuracy_score(Y_test,y_pred))
```

Accuracy Score: 0.8645833333333334

Using Random Forest:

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