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Import basic Libraries

In [1]:

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

In [2]:

```
df = pd.read_csv("wine_fraud.csv")
df.head()
```

Out[2]:

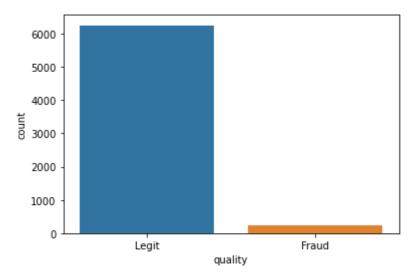
| | fixed acidity | volatile acidity | citric acid | residual sugar | chlorides | free sulfur dioxide | total sulfur dioxide | density | рН | sulphates | alcoh |
|---|------------------|---------------------|----------------|-------------------|-----------|---------------------------|----------------------------|---------|------|-----------|-------|
| 0 | 7.4 | 0.70 | 0.00 | 1.9 | 0.076 | 11.0 | 34.0 | 0.9978 | 3.51 | 0.56 | 9 |
| 1 | 7.8 | 0.88 | 0.00 | 2.6 | 0.098 | 25.0 | 67.0 | 0.9968 | 3.20 | 0.68 | 9 |
| 2 | 7.8 | 0.76 | 0.04 | 2.3 | 0.092 | 15.0 | 54.0 | 0.9970 | 3.26 | 0.65 | 9 |
| 3 | 11.2 | 0.28 | 0.56 | 1.9 | 0.075 | 17.0 | 60.0 | 0.9980 | 3.16 | 0.58 | 9 |
| 4 | 7.4 | 0.70 | 0.00 | 1.9 | 0.076 | 11.0 | 34.0 | 0.9978 | 3.51 | 0.56 | 9 |

In [3]:

```
sns.countplot(x='quality',data=df)
```

Out[3]:

<AxesSubplot:xlabel='quality', ylabel='count'>



Preprocessing

```
In [4]:
```

```
df['type'] = pd.get_dummies(df['type'],drop_first=True)
df['quality']= df['quality'].map({'Legit':0,'Fraud':1})
df.head()
```

Out[4]:

| | fixed acidity | volatile acidity | citric acid | residual sugar | chlorides | free sulfur dioxide | total sulfur dioxide | density | рН | sulphates | alcoh |
|---|------------------|---------------------|----------------|-------------------|-----------|---------------------------|----------------------------|---------|------|-----------|-------|
| 0 | 7.4 | 0.70 | 0.00 | 1.9 | 0.076 | 11.0 | 34.0 | 0.9978 | 3.51 | 0.56 | 9 |
| 1 | 7.8 | 0.88 | 0.00 | 2.6 | 0.098 | 25.0 | 67.0 | 0.9968 | 3.20 | 0.68 | 9 |
| 2 | 7.8 | 0.76 | 0.04 | 2.3 | 0.092 | 15.0 | 54.0 | 0.9970 | 3.26 | 0.65 | 9 |
| 3 | 11.2 | 0.28 | 0.56 | 1.9 | 0.075 | 17.0 | 60.0 | 0.9980 | 3.16 | 0.58 | 9 |
| 4 | 7.4 | 0.70 | 0.00 | 1.9 | 0.076 | 11.0 | 34.0 | 0.9978 | 3.51 | 0.56 | 9 |

Split data into training and testing

```
In [5]:

X = df.drop('quality',axis=1)
y = df['quality']
```

In [6]:

```
from sklearn.model_selection import train_test_split

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.1, random_state=101)
```

In [7]:

```
from sklearn.preprocessing import StandardScaler

scaler = StandardScaler()

scaled_X_train = scaler.fit_transform(X_train)
scaled_X_test = scaler.transform(X_test)
```

Lets move to SVM model

In [8]:

```
from sklearn.svm import SVC
svc = SVC(class_weight='balanced')
```

Using GridSearchCV to run a grid search for the best C and gamma parameters.

```
In [9]:
```

```
from sklearn.model_selection import GridSearchCV

param_grid = {'C':[0.001,0.01,0.1,0.25,0.5,0.75,1]}
grid = GridSearchCV(svc,param_grid)
```

In [10]:

```
grid.fit(scaled_X_train,y_train)
```

Out[10]:

```
In [11]:
```

```
grid.best_params_
```

Out[11]:

{'C': 1}

Displaying Confusion matrix and Classification Report

In [12]:

```
from sklearn.metrics import confusion_matrix,classification_report,plot_confusion_matrix,ac
grid_pred = grid.predict(scaled_X_test)
confusion_matrix(y_test,grid_pred)
```

Out[12]:

```
array([[531, 92],
[ 10, 17]], dtype=int64)
```

In [13]:

```
y_pred = grid.predict(scaled_X_test)
accuracy_score(y_test,y_pred)
```

Out[13]:

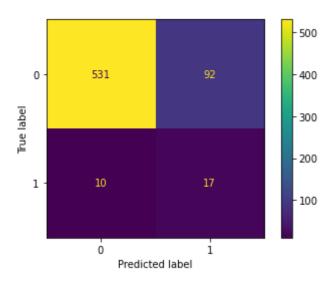
0.8430769230769231

In [14]:

```
plot_confusion_matrix(grid,scaled_X_test,y_test)
```

Out[14]:

<sklearn.metrics._plot.confusion_matrix.ConfusionMatrixDisplay at 0x2990476e
308>

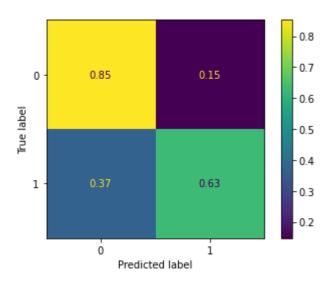


In [15]:

```
# Scaled so highest value=1
plot_confusion_matrix(grid,scaled_X_test,y_test,normalize='true')
```

Out[15]:

<sklearn.metrics._plot.confusion_matrix.ConfusionMatrixDisplay at 0x2990477e
488>



In [16]:

print(classification_report(y_test,grid_pred))

| | precision | recall | f1-score | support | |
|--------------|-----------|--------|----------|---------|--|
| 0 | 0.98 | 0.85 | 0.91 | 623 | |
| 1 | 0.16 | 0.63 | 0.25 | 27 | |
| accuracy | | | 0.84 | 650 | |
| macro avg | 0.57 | 0.74 | 0.58 | 650 | |
| weighted avg | 0.95 | 0.84 | 0.88 | 650 | |

In []: