

In [1]:

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

In [2]:

```
data=pd.read_csv('WINE DATA\\winequality-red.csv',delimiter = ";")
```

In [3]:

data

Out[3]:

| | fixed acidity | volatile acidity | citric acid | residual sugar | chlorides | free sulfur dioxide | total sulfur dioxide | density | pH | sulphates | alcohol | quality |
|------|------------------|---------------------|----------------|-------------------|-----------|------------------------|-------------------------|---------|------|-----------|---------|---------|
| 0 | 7.4 | 0.700 | 0.00 | 1.9 | 0.076 | 11.0 | 34.0 | 0.99780 | 3.51 | 0.56 | 9.4 | 5 |
| 1 | 7.8 | 0.880 | 0.00 | 2.6 | 0.098 | 25.0 | 67.0 | 0.99680 | 3.20 | 0.68 | 9.8 | 5 |
| 2 | 7.8 | 0.760 | 0.04 | 2.3 | 0.092 | 15.0 | 54.0 | 0.99700 | 3.26 | 0.65 | 9.8 | 5 |
| 3 | 11.2 | 0.280 | 0.56 | 1.9 | 0.075 | 17.0 | 60.0 | 0.99800 | 3.16 | 0.58 | 9.8 | 6 |
| 4 | 7.4 | 0.700 | 0.00 | 1.9 | 0.076 | 11.0 | 34.0 | 0.99780 | 3.51 | 0.56 | 9.4 | 5 |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 1594 | 6.2 | 0.600 | 0.08 | 2.0 | 0.090 | 32.0 | 44.0 | 0.99490 | 3.45 | 0.58 | 10.5 | 5 |
| 1595 | 5.9 | 0.550 | 0.10 | 2.2 | 0.062 | 39.0 | 51.0 | 0.99512 | 3.52 | 0.76 | 11.2 | 6 |
| 1596 | 6.3 | 0.510 | 0.13 | 2.3 | 0.076 | 29.0 | 40.0 | 0.99574 | 3.42 | 0.75 | 11.0 | 6 |
| 1597 | 5.9 | 0.645 | 0.12 | 2.0 | 0.075 | 32.0 | 44.0 | 0.99547 | 3.57 | 0.71 | 10.2 | 5 |
| 1598 | 6.0 | 0.310 | 0.47 | 3.6 | 0.067 | 18.0 | 42.0 | 0.99549 | 3.39 | 0.66 | 11.0 | 6 |

1599 rows × 12 columns

In [4]:

```
print("Total quantity of wine quality 3 : ",len([a for a in data['quality'] if a==3]))
print("Total quantity of wine quality 4 : ",len([a for a in data['quality'] if a==4]))
print("Total quantity of wine quality 5 : ",len([a for a in data['quality'] if a==5]))
print("Total quantity of wine quality 6 : ",len([a for a in data['quality'] if a==6]))
print("Total quantity of wine quality 7 : ",len([a for a in data['quality'] if a==7]))
print("Total quantity of wine quality 8 : ",len([a for a in data['quality'] if a==8]))
```

```
Total quantity of wine quality 3 : 10
Total quantity of wine quality 4 : 53
Total quantity of wine quality 5 : 681
Total quantity of wine quality 6 : 638
Total quantity of wine quality 7 : 199
Total quantity of wine quality 8 : 18
```

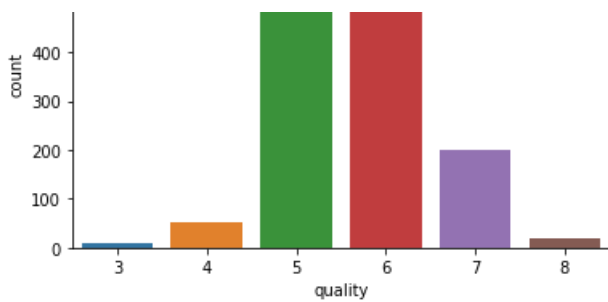
In [5]:

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

Out[5]:

<matplotlib.axes._subplots.AxesSubplot at 0x277cbd20b38>





In [6]:

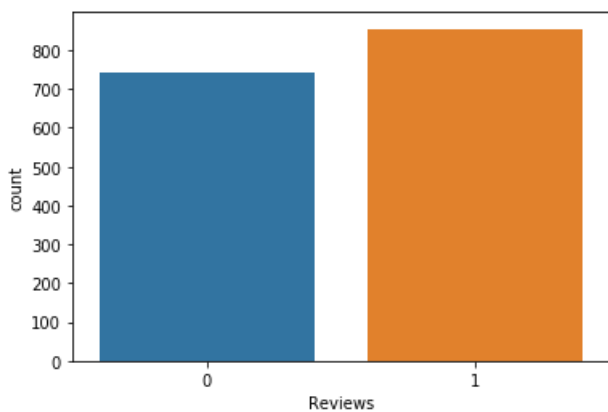
```
reviews = []
for i in data['quality']:
    if i <= 5:
        reviews.append(0)
    else:
        reviews.append(1)
data['Reviews'] = reviews
```

In [7]:

```
sns.countplot(x='Reviews', data=data)
```

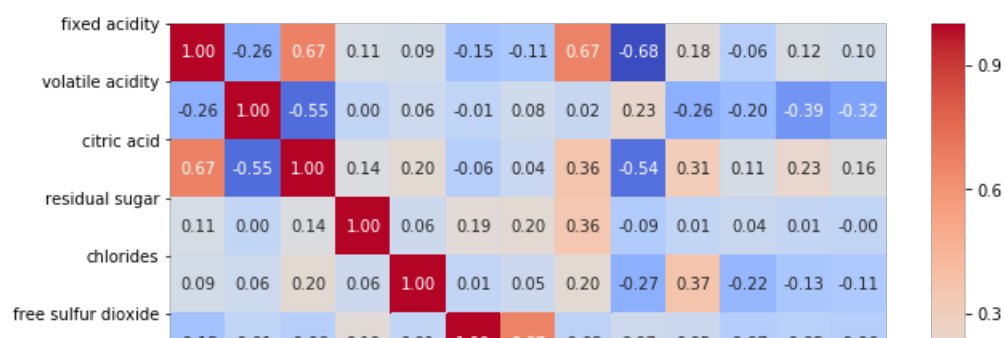
Out[7]:

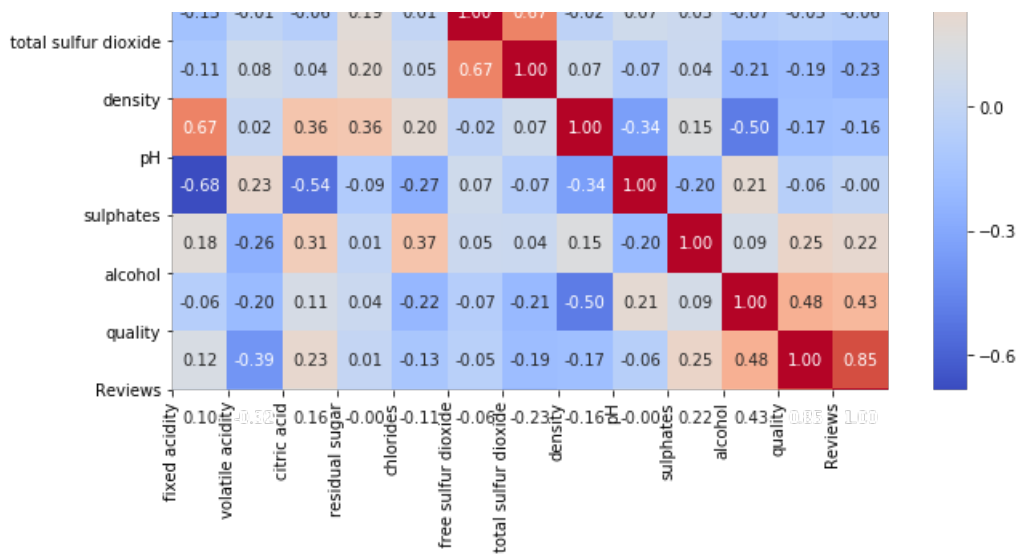
<matplotlib.axes._subplots.AxesSubplot at 0x277d0b62198>



In [8]:

```
corr = data.corr()
#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 [9]:

```
X = data.iloc[:,0:11]
y = data.iloc[:,12]
y_mul = data.iloc[:,11]
```

In [10]:

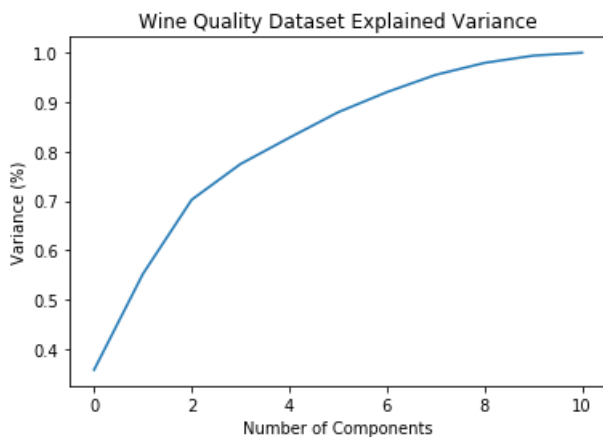
```
from sklearn.preprocessing import MinMaxScaler
minmax = MinMaxScaler()
X_s = minmax.fit_transform(X)
```

In [13]:

```
from sklearn.decomposition import PCA
pca = PCA()
X_s_pca = pca.fit(X_s)
```

In [14]:

```
plt.figure()
plt.plot(np.cumsum(X_s_pca.explained_variance_ratio_))
plt.xlabel('Number of Components')
plt.ylabel('Variance (%)') #for each component
plt.title('Wine Quality Dataset Explained Variance')
plt.show()
```



In [15]:

```
pca = PCA(n_components=9)
X_s_pca = pca.fit_transform(X_s)
```

In [16]:

```
from sklearn.model_selection import train_test_split

X_train,X_test,y_train,y_test = train_test_split(X_s_pca,y,test_size = 0.3,random_state = 42)
```

In [17]:

```
print(X_train.shape)
print(X_test.shape)
```

```
(1119, 9)
(480, 9)
```

In [18]:

```
from sklearn.svm import SVC
from sklearn.metrics import accuracy_score,classification_report,confusion_matrix
```

In [19]:

```
svc_lin = SVC(kernel = 'linear')
svc_lin.fit(X_train, y_train)
y_pred_svc_lin=svc_lin.predict(X_test)
lin_svc_conf_matrix = confusion_matrix(y_test, y_pred_svc_lin)
lin_svc_clf = classification_report(y_test, y_pred_svc_lin)
print(lin_svc_conf_matrix)
print(lin_svc_clf)
```

```
[[164  49]
 [ 87 180]]
```

| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 0.65 | 0.77 | 0.71 | 213 |
| 1 | 0.79 | 0.67 | 0.73 | 267 |
| accuracy | | | 0.72 | 480 |
| macro avg | 0.72 | 0.72 | 0.72 | 480 |
| weighted avg | 0.73 | 0.72 | 0.72 | 480 |

In [20]:

```
svc_rbf = SVC(kernel = 'rbf')
svc_rbf.fit(X_train, y_train)
y_pred_svc_rbf=svc_rbf.predict(X_test)
rbf_svc_conf_matrix = confusion_matrix(y_test, y_pred_svc_rbf)
rbf_svc_clf = classification_report(y_test, y_pred_svc_rbf)
print(rbf_svc_conf_matrix)
print(rbf_svc_clf)
```

```
[[160  53]
 [ 71 196]]
```

| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 0.69 | 0.75 | 0.72 | 213 |
| 1 | 0.79 | 0.73 | 0.76 | 267 |
| accuracy | | | 0.74 | 480 |
| macro avg | 0.74 | 0.74 | 0.74 | 480 |
| weighted avg | 0.75 | 0.74 | 0.74 | 480 |

In [21]:

```
svc_poly = SVC(kernel = 'poly')
svc_poly.fit(X_train, y_train)
y_pred_svc_poly = svc_poly.predict(X_test)
poly_svc_conf_matrix = confusion_matrix(y_test, y_pred_svc_poly)
poly_svc_clf = classification_report(y_test, y_pred_svc_poly)
```

```
print(poly_svc_conf_matrix)
print(poly_svc_clf)
```

```
[[161  52]
 [ 68 199]]
```

| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 0.70 | 0.76 | 0.73 | 213 |
| 1 | 0.79 | 0.75 | 0.77 | 267 |
| accuracy | | | 0.75 | 480 |
| macro avg | 0.75 | 0.75 | 0.75 | 480 |
| weighted avg | 0.75 | 0.75 | 0.75 | 480 |

In [22]:

```
svc_sigmoid = SVC(kernel = 'sigmoid')
svc_sigmoid.fit(X_train, y_train)
y_pred_svc_sigmoid=svc_sigmoid.predict(X_test)
sigmoid_svc_conf_matrix = confusion_matrix(y_test, y_pred_svc_sigmoid)
sigmoid_svc_clf = classification_report(y_test, y_pred_svc_sigmoid)
print(sigmoid_svc_conf_matrix)
print(sigmoid_svc_clf)
```

```
[[120  93]
 [ 86 181]]
```

| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 0.58 | 0.56 | 0.57 | 213 |
| 1 | 0.66 | 0.68 | 0.67 | 267 |
| accuracy | | | 0.63 | 480 |
| macro avg | 0.62 | 0.62 | 0.62 | 480 |
| weighted avg | 0.63 | 0.63 | 0.63 | 480 |

In [30]:

```
from sklearn.model_selection import cross_val_score
clf = SVC(kernel='linear', C=1)
scores = cross_val_score(clf, X_s_pca, y, cv=5)
print(scores)
print("Accuracy: %0.2f (+/- %0.2f)" % (scores.mean(), scores.std() * 2))
```

```
[0.696875  0.69375  0.78125  0.753125  0.7460815]
Accuracy: 0.73 (+/- 0.07)
```

In [29]:

```
from sklearn.model_selection import cross_val_score
clf = SVC(kernel='rbf', C=1)
scores = cross_val_score(clf, X_s_pca, y, cv=5)
print(scores)
print("Accuracy: %0.2f (+/- %0.2f)" % (scores.mean(), scores.std() * 2))
```

```
[0.715625  0.709375  0.790625  0.753125  0.73981191]
Accuracy: 0.74 (+/- 0.06)
```

In [28]:

```
from sklearn.model_selection import cross_val_score
clf = SVC(kernel='poly',C=1)
scores = cross_val_score(clf, X_s_pca, y, cv=5)
print(scores)
print("Accuracy: %0.2f (+/- %0.2f)" % (scores.mean(), scores.std() * 2))
```

```
[0.684375  0.715625  0.765625  0.728125  0.72727273]
Accuracy: 0.72 (+/- 0.05)
```

In [27]:

```
from sklearn.model_selection import cross_val_score
clf = SVC(kernel='sigmoid',C=1)
scores = cross_val_score(clf, X_s_pca, y, cv=5)
print(scores)
print("Accuracy: %0.2f (+/- %0.2f)" % (scores.mean(), scores.std() * 2))
```

```
[0.684375  0.6       0.678125  0.625       0.58934169]
Accuracy: 0.64 (+/- 0.08)
```

In []: