SVM 3/1/25, 12:22 PM

```
In [ ]: import pandas as pd
         import numpy as np
         from sklearn import svm
         from sklearn.model_selection import train_test_split
         import seaborn as sns
         import matplotlib.pyplot as plt
In [ ]: df=pd.read_csv("winequality-red.csv", sep=';')
In [ ]: df
Out[ ]:
               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
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                       7.8
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                                                                        0.075
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                                                                                                              60.0 0.99800 3.16
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                                     0.600
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                                     0.310
                                                0.47
                                                                3.6
                                                                        0.067
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                                                                                                              42.0 0.99549 3.39
                                                                                                                                      0.66
                                                                                                                                               11.0
                                                                                                                                                          6
```

1599 rows × 12 columns

In []: print(df.isna().sum())

3/1/25, 12:22 PM SVM

```
fixed acidity
      volatile acidity
       citric acid
                               0
      residual sugar
       chlorides
      free sulfur dioxide
       total sulfur dioxide
      density
       рΗ
       sulphates
      alcohol
                               0
      quality
      dtype: int64
In [ ]: X=df.drop(["quality"],axis=1)
In [ ]: y=df["quality"].values
In [ ]: X_train, X_test, y_train, y_test = train_test_split(X, y, train_size=0.8)
        #X_val, X_test, y_val, y_test = train_test_split(X_test, y_test, test_size=0.5)
In [ ]: model = svm.SVC(kernel="linear")
        # pass data and train the model
        model.fit(X_train, y_train)
        # get prediction accuracy
        model.score(X_test, y_test)
Out[]: 0.56875
In [ ]: corr_matrix = df.corr()
        plt.figure(figsize=(10, 6))
        sns.heatmap(corr_matrix, annot=True, fmt=".2f", cmap="coolwarm", linewidths=0.5)
        plt.title("Feature Correlation Heatmap")
        plt.show()
```

3/1/25, 12:22 PM SVM





