```
In [2]: import numpy as np
        from sklearn.svm import SVC
        X = np.array([[3,4],[1,4],[2,3],[6,-1],[7,-1],[5,-3]])
        y = np.array([-1,-1, -1, 1, 1, 1])
        clt=SVC(C=1e4 , kernel='linear')
        clt.fit(X,y)
        print("Coeff: ",clt.coef_)
        print("Intercepts: ",clt.intercept_)
        print("Support: ",clt.support_)
        print("Supp Vectors: ",clt.support_vectors_)
        print("N supp: ",clt.n_support_)
        Coeff: [[ 0.25 -0.25]]
        Intercepts: [-0.75]
        Support: [2 3]
        Supp Vectors: [[ 2. 3.]
         [ 6. -1.]]
        N supp: [1 1]
In [4]: x1=clt.support_vectors_[0]
        x2=clt.support_vectors_[1]
In [5]: import matplotlib.pyplot as plt
        plt.scatter(X[:,0],X[:,1])
        plt.scatter(x1[0],x1[1],color='red')
        plt.scatter(x2[0],x2[1],color='red')
        plt.show()
           4
           3
           2
           1
           0
          -1
          -2
```

```
In [7]: import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split,GridSearchCV
from sklearn.metrics import accuracy_score,confusion_matrix,classification_report
from sklearn.svm import SVC
```

-3

1

2

3

```
In [8]: data=pd.read_csv('glass.csv')
data
```

Out[8]:

```
ld
             RI
                                             Ca
                   Na
                        Mg
                             ΑI
                                    Si
                                         Κ
                                                  Ba Fe Type
      1 1.52101 13.64 4.49
                           1.10 71.78 0.06
                                            8.75
                                                0.00
                                                      0.0
 1
      2 1.51761 13.89 3.60 1.36 72.73 0.48 7.83 0.00 0.0
                                                             1
 2
      3 1.51618 13.53 3.55 1.54 72.99 0.39 7.78 0.00 0.0
 3
      4 1.51766 13.21 3.69 1.29 72.61 0.57 8.22 0.00 0.0
      5 1.51742 13.27 3.62 1.24 73.08 0.55 8.07 0.00 0.0
                                    ...
209 210 1.51623 14.14 0.00 2.88 72.61 0.08 9.18 1.06 0.0
                                                             7
210 211 1.51685 14.92 0.00 1.99 73.06 0.00 8.40 1.59 0.0
211 212 1.52065 14.36 0.00 2.02 73.42 0.00 8.44 1.64 0.0
212 213 1.51651 14.38 0.00 1.94 73.61 0.00 8.48 1.57 0.0
213 214 1.51711 14.23 0.00 2.08 73.36 0.00 8.62 1.67 0.0
```

214 rows × 11 columns

```
In [14]: X=data.drop('Type',axis=1)
Y=data.Type
```

In [15]: X.head()

Out[15]:

```
        Id
        RI
        Na
        Mg
        AI
        Si
        K
        Ca
        Ba
        Fe

        0
        1
        1.52101
        13.64
        4.49
        1.10
        71.78
        0.06
        8.75
        0.0
        0.0

        1
        2
        1.51761
        13.89
        3.60
        1.36
        72.73
        0.48
        7.83
        0.0
        0.0

        2
        3
        1.51618
        13.53
        3.55
        1.54
        72.99
        0.39
        7.78
        0.0
        0.0

        3
        4
        1.51766
        13.21
        3.69
        1.29
        72.61
        0.57
        8.22
        0.0
        0.0

        4
        5
        1.51742
        13.27
        3.62
        1.24
        73.08
        0.55
        8.07
        0.0
        0.0
```

```
In [16]: x_train, x_test, y_train,y_test=train_test_split(X,Y)
```

```
In [17]: lin_model=SVC(kernel='linear')
lin_model.fit(x_train,y_train)
y_pred=lin_model.predict(x_test)
```

```
In [18]: print("Accuracy ",accuracy_score(y_test,y_pred))
         print(" Confusion matrix \n",confusion_matrix(y_test,y_pred))
         print("Classification report ",classification_report(y_test,y_pred))
         Accuracy 0.9629629629629
          Confusion matrix
           [[16 0 0 0 0 0]
           [218 0 0 0 0]
                            0]
          [0 0 3
                     0
                        0
          [ 0
              0 0 3 0 0]
          [000050]
          [000007]]
         Classification report
                                                precision
                                                             recall f1-score
                                                                                 support
                     1
                             0.89
                                       1.00
                                                  0.94
                                                              16
                     2
                             1.00
                                       0.90
                                                  0.95
                                                              20
                     3
                             1.00
                                       1.00
                                                  1.00
                                                               3
                     5
                             1.00
                                       1.00
                                                  1.00
                                                               3
                     6
                             1.00
                                       1.00
                                                  1.00
                                                               5
                     7
                                                               7
                             1.00
                                       1.00
                                                  1.00
                                                  0.96
                                                              54
             accuracy
                             0.98
                                       0.98
                                                  0.98
                                                              54
            macro avg
         weighted avg
                             0.97
                                       0.96
                                                  0.96
                                                              54
In [19]: model1=SVC(kernel='sigmoid',gamma=0.0001)
         model2=SVC(kernel='poly',degree=3)
         model3=SVC(kernel='rbf')
In [20]: model1.fit(x_train,y_train)
         model2.fit(x_train,y_train)
         model3.fit(x_train,y_train)
Out[20]:
          ▼ SVC
In [21]: | ypred1=model1.predict(x_test)
         ypred2=model2.predict(x_test)
         ypred3=model3.predict(x_test)
         print(accuracy_score(y_test,ypred1))
         print(accuracy_score(y_test,ypred2))
         print(accuracy_score(y_test,ypred3))
         0.5925925925925926
         0.8148148148148148
         0.7592592592593
In [22]: params_grid = [{'kernel': ['rbf'], 'gamma': [1e-3, 1e-4],
                               'C': [1, 10, 100, 1000]},
                              {'kernel': ['linear'], 'C': [1, 10, 100, 1000]},
                              {'kernel': ['poly'], 'gamma': [1e-3, 1e-4],'degree':[2,3,4]},
{'kernel': ['sigmoid'], 'C': [1, 10, 100, 1000],'gamma': [1e-3, 1e-4,1e-5]}
```

```
svm_model = GridSearchCV(SVC(), params_grid, cv=5)
In [23]:
         svm_model.fit(x_train,y_train)
         C:\Users\Admin\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\model_selecti
         on\_split.py:700: UserWarning: The least populated class in y has only 4 members, which is les
         s than n splits=5.
           warnings.warn(
Out[23]:
         → GridSearchCV
           ▶ estimator: SVC
                ▶ SVC
In [24]: print('Best score for training data:', svm_model.best_score_,"\n")
         # View the best parameters for the model found using grid search
         print('Best C:',svm_model.best_estimator_.C,"\n")
         print('Best Kernel:',svm_model.best_estimator_.kernel,"\n")
         print('Best Gamma:',svm_model.best_estimator_.gamma,"\n")
         final_model = svm_model.best_estimator_
         Y_pred = final_model.predict(x_test)
         Best score for training data: 0.9875
         Best C: 1.0
         Best Kernel: poly
         Best Gamma: 0.001
In [25]: |print(accuracy_score(y_test,Y_pred))
         0.944444444444444
In [ ]:
```