## **NAME: D TONIA SREE**

**REG NO.: 22MID0346** 

## Task 7:

## CODE:

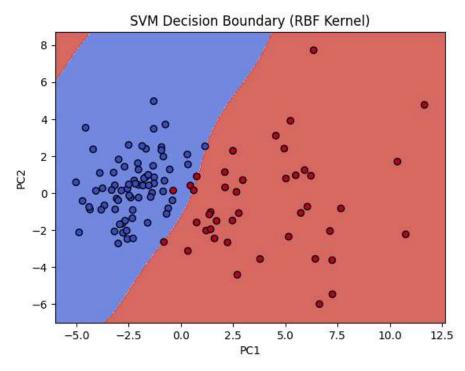
```
import numpy as np
    import pandas as pd
    import matplotlib.pyplot as plt
    from sklearn.svm import SVC
    from sklearn.model_selection import train_test_split, GridSearchCV, cross_val_score
    from sklearn.preprocessing import StandardScaler
    from sklearn.decomposition import PCA
    from sklearn.metrics import classification_report, confusion_matrix
    df = pd.read_csv('breast-cancer.csv')
    print(df.head())
    # Step 2: Preprocess Data
    # Drop any unnamed columns if present
    df = df.loc[:, ~df.columns.str.contains('^Unnamed')]
    # Check target column - assume it's named "diagnosis" or similar
    if 'diagnosis' in df.columns:
        df['diagnosis'] = df['diagnosis'].map({'M': 1, 'B': 0}) # Malignant = 1, Benign = 0
    # Split features and target
    X = df.drop('diagnosis', axis=1)
    y = df['diagnosis']
    # Normalize features
    scaler = StandardScaler()
    X_scaled = scaler.fit_transform(X)
    # Train/test split
    X_train, X_test, y_train, y_test = train_test_split(X_scaled, y, test_size=0.2, random_state=42)
```

```
# Step 3: Train SVM (Linear)
svm_linear = SVC(kernel='linear', C=1)
svm_linear.fit(X_train, y_train)
print("Linear SVM Accuracy:", svm_linear.score(X_test, y_test))
# Step 4: Train SVM (RBF Kernel)
svm_rbf - SVC(kernel-'rbf', C-1, gamma-'scale')
svm_rbf.fit(X_train, y_train)
print("RBF SVM Accuracy:", svm_rbf.score(X_test, y_test))
# Step 5: Visualization using PCA (2D projection)
pca = PCA(n_components=2)
X_pca = pca.fit_transform(X_scaled)
# Retrain on PCA-reduced data for visualization
X_train_pca, X_test_pca, y_train_pca, y_test_pca = train_test_split(X_pca, y, test_size=0.2, random_state=42)
svm_vis = SVC(kernel='rbf', C=1, gamma='scale')
svm_vis.fit(X_train_pca, y_train_pca)
# Plot decision boundary
def plot_decision_boundary(X, y, model):
   h = .82
    x_{min}, x_{max} = X[:, \theta].min() - 1, X[:, \theta].max() + 1
   y_min, y_max = X[:, 1].min() - 1, X[:, 1].max() + 1
   xx, yy = np.meshgrid(np.arange(x_min, x_max, h),
                         np.arange(y_min, y_max, h))
   Z = model.predict(np.c_[xx.ravel(), yy.ravel()])
    Z - Z.reshape(xx.shape)
```

```
plt.contourf(xx, yy, Z, cmap=plt.cm.coolwarm, alpha=0.8)
    plt.scatter(X[:,\;\theta],\;X[:,\;1],\;c=y,\;cmap=plt.cm.coolwarm,\;edgecolors='k')
    plt.xlabel('PC1')
    plt.ylabel('PC2')
    plt.title('SVM Decision Boundary (RBF Kernel)')
    plt.show()
plot_decision_boundary(X_test_pca, y_test_pca, svm_vis)
# Step 6: Hyperparameter Tuning with GridSearchCV
param_grid = {
    'C': [0.1, 1, 10],
    'gamma': ['scale', 0.1, 1, 10],
    'kernel': ['rbf']
grid = GridSearchCV(SVC(), param_grid, refit=True, verbose=2, cv=5)
grid.fit(X_train, y_train)
print("Best parameters:", grid.best_params_)
print("Best estimator accuracy:", grid.best_estimator_.score(X_test, y_test))
# Step 7: Cross-Validation
cv_scores = cross_val_score(grid.best_estimator_, X_scaled, y, cv=5)
print("Cross-validation scores:", cv_scores)
print("Mean CV accuracy:", cv_scores.mean())
```

## **OUTPUT:**

```
id diagnosis radius_mean texture_mean perimeter_mean area_mean \
→ 0
       842302 M 17.99
                                  10.38 122.80 1001.0
       842517
                    М
                            20.57
                                        17.77
                                                     132.90
                                                               1326.0
   2 84300903
                   М
                           19.69
                                       21.25
                                                    130.00
                                                              1203.0
                                       20.38
   3 84348301
                   М
                           11.42
                                                     77.58
                                                               386.1
   4 84358402
                    M
                            20.29
                                        14.34
                                                     135.10
                                                              1297.0
      smoothness mean compactness mean concavity mean concave points mean \
                                      0.3001
                       0.27760
   0
          0.11840
                                                            0.14710
   1
             0.08474
                            0.07864
                                          0.0869
                                                            0.07017
             0.10960
                           0.15990
                                          0.1974
                                                            0.12790
   2
   3
             0.14250
                            0.28390
                                          0.2414
                                                            0.10520
   4
             0.10030
                            0.13280
                                           0.1980
                                                            0.10430
      ... radius_worst texture_worst perimeter_worst area_worst \
                                   184.60
                                                  2019.0
                       17.33
   0 ... 25.38
   1
                24.99
                             23.41
                                          158.80
                                                     1956.0
      ...
                                          152.50
   2 ...
               23.57
                            25.53
                                                     1709.0
                             26.50
                                           98.87
                                                      567.7
   3 ...
                14.91
                22.54
                             16.67
                                          152.20
   4 ...
                                                     1575.0
      smoothness_worst compactness_worst concavity_worst concave points_worst \
                                      0.7119
   0
                       0.6656
              0.1622
                                                                0.2654
   1
              0.1238
                              0.1866
                                             0.2416
                                                                0.1860
              0.1444
                              0.4245
                                             0.4504
                                                                0.2430
              0.2098
                                             0.6869
                               0.8663
                                                                0.2575
   3
   4
              0.1374
                               0.2050
                                             0.4000
                                                                 0.1625
      symmetry_worst fractal_dimension_worst
   0
             0.4601
   1
             0.2750
                                 0.08902
             0.3613
                                 0.08758
             0.6638
                                 0 17300
   2
   4
             0.2364
                                 0.07678
   [5 rows x 32 columns]
    Linear SVM Accuracy: 0.956140350877193
   RBF SVM Accuracy: 0.9824561403508771
```



```
Fitting 5 folds for each of 12 candidates, totalling 60 fits
[CV] END ......C=0.1, gamma=scale, kernel=rbf; total time=
                         0.05
0.05
0.05
0.05
0.05
0.05
0.05
0.05
0.05
0.05
0.05
0.05
[CV] END ......C=0.1, gamma=10, kernel=rbf; total time=
                         0.05
0.05
0.05
[CV] END ......C=0.1, gamma=10, kernel=rbf; total time=
                         0.05
0.05
0.05
[CV] END ......C=1, gamma=scale, kernel=rbf; total time=
                         0.05
0.05
0.05
0.05
[CV] END ......C=1, gamma=0.1, kernel=rbf; total time=
                         0.05
[CV] END ......C=1, gamma=0.1, kernel=rbf; total time=
[CV] END ......C=1, gamma=1, kernel=rbf; total time=
                         0.05
[CV] END ......C=1, gamma=1, kernel=rbf; total time=
                         0.05
[CV] END ......C=1, gamma=1, kernel=rbf; total time=
[CV] END ......C=1, gamma=1, kernel=rbf; total time=
                         0.05
0.05
[CV] END ......C=1, gamma=10, kernel=rbf; total time=
                         0.05
```

```
[CV] END ......C=1, gamma=1, kernel=rbf; total time=
[CV] END ......C=1, gamma=1, kernel=rbf; total time=
[CV] END ......C=1, gamma=10, kernel=rbf; total time=
                                    0.05
[CV] END ......C=1, gamma=10, kernel=rbf; total time=
                                    0.05
[CV] END ......C=1, gamma=10, kernel=rbf; total time=
                                    0.05
[CV] END .....C=10, gamma=10, kernel=rbf; total time=
[CV] END .....C=10, gamma=scale, kernel=rbf; total time=
                                    0.05
                                    0.05
0.05
[CV] END ......C=10, gamma=scale, kernel=rbf; total time=
[CV] END ......C=10, gamma=scale, kernel=rbf; total time=
                                    0.05
9.95
0.05
0.05
0.05
[CV] END ......C=10, gamma=1, kernel=rbf; total time=
                                    0.05
0.05
0.0s
0.05
[CV] END ......C=10, gamma=1, kernel=rbf; total time=
                                    0.05
[CV] END ......C=10, gamma=10, kernel=rbf; total time=
                                    0.05
[CV] END ......C=10, gamma=10, kernel=rbf; total time=
[CV] END ......C=10, gamma=10, kernel=rbf; total time=
                                    0.05
[CV] END ......C=10, gamma=10, kernel=rbf; total time=
                                    0.05
0.05
Best parameters: {'C': 1, 'gamma': 'scale', 'kernel': 'rbf'}
Best estimator accuracy: 0.9824561403508771
Cross-validation scores: [0.97368421 0.95614035 1. 0.96491228 0.97345133]
```

Mean CV accuracy: 0.9736376339077782