#### svm

## November 18, 2024

[21]: # Importing required libraries

```
import pandas as pd
     import numpy as np
     from sklearn.model_selection import train_test_split
     from sklearn.preprocessing import StandardScaler
     from sklearn.svm import SVC
     from sklearn.metrics import confusion_matrix, accuracy_score
[23]: # Loading the dataset
     dataset = pd.read csv('C:/Users/divaa/Downloads/Social Network Ads.csv')
      # Displaying the first few rows of the dataset
     print(dataset.head())
         User ID Gender Age EstimatedSalary Purchased
     0 15624510
                   Male 19
                                        19000
     1 15810944
                 Male
                          35
                                        20000
                                                       0
     2 15668575 Female 26
                                        43000
                                                       0
     3 15603246 Female 27
                                        57000
                                                       0
     4 15804002
                                        76000
                 Male 19
                                                       0
[25]: # Selecting relevant features (Age, EstimatedSalary) and target (Purchased)
     X = dataset[['Age', 'EstimatedSalary']].values
     y = dataset['Purchased'].values
[27]: # Splitting the dataset into training and testing sets
     X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,_
       →random state=0)
[29]: # Feature scaling
     scaler = StandardScaler()
     X train = scaler.fit transform(X train)
     X_test = scaler.transform(X_test)
[31]: # Training the SVM classifier
     model = SVC(kernel='linear') # Using a linear kernel
     model.fit(X_train, y_train)
```

```
[31]: SVC(kernel='linear')
[50]: # Making predictions on the test set
      y_pred = model.predict(X_test)
[52]: # Evaluating the model
      accuracy = accuracy_score(y_test, y_pred)
      conf_matrix = confusion_matrix(y_test, y_pred)
      print(f"Accuracy: {accuracy * 100:.2f}%")
      print("Confusion Matrix:")
      print(conf_matrix)
     Accuracy: 93.75%
     Confusion Matrix:
     [[55 3]
      [ 2 20]]
[35]: from imblearn.over_sampling import SMOTE
      # Applying SMOTE to the training data
      smote = SMOTE(random_state=0)
      X_res, y_res = smote.fit_resample(X_train, y_train)
      # Training the model with the resampled data
      model = SVC(kernel='rbf')
      model.fit(X_res, y_res)
      # Predictions
      y_pred = model.predict(X_test)
      # Evaluating the model
      accuracy = accuracy_score(y_test, y_pred)
      print(f"Accuracy after SMOTE: {accuracy * 100:.2f}%")
     Accuracy after SMOTE: 95.00%
[37]: # Evaluating the model
      accuracy = accuracy_score(y_test, y_pred)
      conf_matrix = confusion_matrix(y_test, y_pred)
      print(f"Accuracy: {accuracy * 100:.2f}%")
      print("Confusion Matrix:")
      print(conf_matrix)
     Accuracy: 95.00%
```

Confusion Matrix:

```
[[55 3]
[ 1 21]]
```

### 0.1 Taking user input for Age and EstimatedSalary

```
age = float(input("Enter age:")) salary = float(input("Enter estimated salary:"))
```

## 1 Scaling the input data

```
user\_input = scaler.transform([[age, salary]])
```

# 2 Making prediction

prediction = model.predict(user\_input)

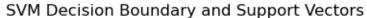
## 3 Displaying the result

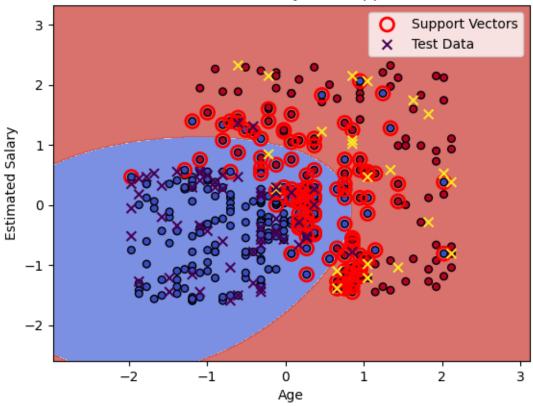
if prediction == 1: print("Prediction: Purchase will be made.") else: print("Prediction: No purchase will be made.")

```
[44]: import matplotlib.pyplot as plt
      import numpy as np
      from sklearn.svm import SVC
      from sklearn.preprocessing import StandardScaler
      from sklearn.model_selection import train_test_split
      # Load the dataset
      dataset = pd.read_csv('C:/Users/divaa/Downloads/Social_Network_Ads.csv')
      # Selecting relevant features (Age, EstimatedSalary) and target (Purchased)
      X = dataset[['Age', 'EstimatedSalary']].values
      y = dataset['Purchased'].values
      # Splitting the dataset into training and testing sets
      X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,_
       →random_state=0)
      # Feature scaling
      scaler = StandardScaler()
      X_train = scaler.fit_transform(X_train)
      X_test = scaler.transform(X_test)
      # Train the SVM model with RBF kernel
      model = SVC(kernel='rbf', C=1, gamma=0.1)
      model.fit(X_train, y_train)
      # Create a mesh grid for plotting the decision boundary
```

```
x_min, x_max = X_train[:, 0].min() - 1, X_train[:, 0].max() + 1
y_min, y_max = X_train[:, 1].min() - 1, X_train[:, 1].max() + 1
xx, yy = np.meshgrid(np.arange(x_min, x_max, 0.01),
                     np.arange(y_min, y_max, 0.01))
# Predict class labels for all points in the grid
Z = model.predict(np.c_[xx.ravel(), yy.ravel()])
Z = Z.reshape(xx.shape)
# Plotting the decision boundary
plt.contourf(xx, yy, Z, alpha=0.75, cmap=plt.cm.coolwarm)
# Plotting the points
plt.scatter(X_train[:, 0], X_train[:, 1], c=y_train, s=30, edgecolor='k', __
 ⇒cmap=plt.cm.coolwarm)
# Plotting the support vectors
plt.scatter(model.support_vectors_[:, 0], model.support_vectors_[:, 1],
            s=100, facecolors='none', edgecolors='red', linewidth=2,__
 ⇔label="Support Vectors")
# Plotting the margin lines
plt.scatter(X_test[:, 0], X_test[:, 1], c=y_test, s=50, edgecolor='k',_
 →marker='x', label='Test Data')
plt.title('SVM Decision Boundary and Support Vectors')
plt.xlabel('Age')
plt.ylabel('Estimated Salary')
plt.legend()
plt.show()
```

C:\Users\divaa\AppData\Local\Temp\ipykernel\_22316\350850842.py:47: UserWarning:
You passed a edgecolor/edgecolors ('k') for an unfilled marker ('x').
Matplotlib is ignoring the edgecolor in favor of the facecolor. This behavior
may change in the future.
 plt.scatter(X\_test[:, 0], X\_test[:, 1], c=y\_test, s=50, edgecolor='k',
marker='x', label='Test Data')





```
[46]: from sklearn.metrics import classification_report, confusion_matrix

# Predictions on the test set
y_pred = model.predict(X_test)

# Confusion Matrix
cm = confusion_matrix(y_test, y_pred)
print("Confusion Matrix:\n", cm)

# Classification Report
print("Classification Report:\n", classification_report(y_test, y_pred))
```

### Confusion Matrix:

[[55 3]

[ 2 20]]

Classification Report:

	precision	recall	f1-score	support
0	0.96	0.95	0.96	58
1	0.87	0.91	0.89	22

```
accuracy 0.94 80 macro avg 0.92 0.93 0.92 80 weighted avg 0.94 0.94 0.94 80
```

```
[75]: import numpy as np
       import pandas as pd
       from sklearn.svm import SVC
       from sklearn.preprocessing import StandardScaler
       from sklearn.model_selection import train_test_split
       # Load the dataset
       dataset = pd.read_csv('C:/Users/divaa/Downloads/Social_Network_Ads.csv')
       # Selecting relevant features (Age, EstimatedSalary) and target (Purchased)
       X = dataset[['Age', 'EstimatedSalary']].values
       y = dataset['Purchased'].values
       # Splitting the dataset into training and testing sets
       X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,__
         →random_state=0)
       # Feature scaling
       scaler = StandardScaler()
       X_train = scaler.fit_transform(X_train)
       X_test = scaler.transform(X_test)
       # Train the SVM model with RBF kernel
       model = SVC(kernel='rbf', C=1, gamma=0.1, probability=True) # Enable_
         ⇔probability estimation
       model.fit(X_train, y_train)
       # Function to predict purchase
       def predict_purchase(age, salary):
            # Scale the input using the same scaler
            input_scaled = scaler.transform([[age, salary]])
            prediction = model.predict(input scaled)[0]
            probability = model.predict_proba(input_scaled)[0][1] # Get the__
         ⇔probability of class 1 (Purchased)
             # Display prediction and probability
            if prediction == 1:
                 print(f"Prediction: Likely to Purchase (Confidence: {probability*100:.

<p
            else:
```

Enter Age: 19

Enter Estimated Salary: 70000

Prediction: Unlikely to Purchase (Confidence: 96.39%)