

## MLA Assignment 3

Aim:-

Implement SVM for performing classification and find its accuracy on the given data.

Title:-

Support Vector Machine.

Theory:-

Q1) What is Support Vector Machine?

→ Support Vector Machine

Support Vector Machine, or SVM is a supervised machine learning algorithm used for classification. It is sometimes also used for regression. SVM finds a hyperplane that creates a boundary between types of data. In 2-dimensional spaces, this hyperplane is nothing but a line.

In SVM we plot each data item in the dataset in an  $N$ -dimensional space, where  $N$  is the number of features in the data. Next, find the optimal hyperplane to separate the data. SVM is used for both binary and multi-class classification problems.

To perform SVM on multi-class problems, we can create a binary classifier for each class of the data. The two results of each classifier will be-

- i) The data points belong to that class OR
- ii) The data point doesn't belong to that class.

Q2) Explain Support Vectors.

→ Support vectors are data points that are closer to the hyperplane and influence the position and orientation of the hyperplane. Using these support vectors, we maximize the margin of the classifier. Deleting the support vectors will change the position of the hyperplane. These are the points that help us build our SVM.

Q3) Differentiate between Hard margin and soft margin SVM.

→	Hard margin SVM	Soft Margin SVM
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a) Used if data is linearly separable.

If data is not linearly separable we use Soft margin SVM

b) Used for proper classification with no misclassified data.

Used for better model generalization.

Q4

Q4) Explain in brief gamma and regularization <sup>parameters.</sup> ~~process.~~

→ Gamma-parameter-

Gamma is used when we use the Gaussian RBF kernel. If you use linear or polynomial kernel then you only need C hyperparameter.

Gamma is a hyperparameter which we have to set before training model. Gamma decides that how much curvature we want in a decision boundary.



High Gamma means more curvature and less low gamma means less curvature. What value of gamma one should use totally depends upon the data.

Regularization parameter -

The regularization parameter (often termed as  $C$  parameter in python's sklearn library) tells the SVM optimization how much you want to avoid misclassifying each training example.

For large values of  $C$ , the optimization will choose a smaller margin hyperplane if that hyperplane does a better job of getting all the training points classified correctly. Conversely, a very small value of  $C$  will cause the optimizer to look for a larger-margin separating hyperplane, even if that hyperplane misclassifies more points.

In [1]:

```
#Support Vector Machine(SVM)
```

In [2]:

```
#importing required libraries  
from sklearn import svm, metrics  
from sklearn.model_selection import train_test_split  
import numpy as np  
import pandas as pd  
import matplotlib.pyplot as plt  
from mlxtend.plotting import plot_decision_regions
```

In [3]:

```
#loading the dataset  
data=pd.read_csv("SVM_DataSet_1.csv")  
data.head()
```

Out [3]:

	User ID	Gender	Age	EstimatedSalary	Purchased
0	15624510	Male	19	19000	0
1	15810944	Male	35	20000	0
2	15668575	Female	26	43000	0
3	15603246	Female	27	57000	0
4	15804002	Male	19	76000	0

In [4]:

```
#Splitting the dataset into features and labels  
X = data.iloc[:, [2, 3]].values  
y = data.iloc[:, 4].values
```

In [5]:

```
#Splitting the dataset into training and testing set  
X_train, X_test, y_train, y_test=train_test_split(X, y, train_size=0.8)
```

In [6]:

```
#SVM model(Kernel=Linear)  
linear_model=svm.SVC(kernel='linear')  
linear_model.fit(X_train, y_train)
```

```
#prediction  
prediction=linear_model.predict(X_test)
```

```

print("Actual set:\n", y_test)
print("Predictions:\n", prediction)

#Confusion matrix
confusion=metrics.confusion_matrix(y_test, prediction)
print("Confusion matrix:\n", confusion)

#classification report
report=metrics.classification_report(y_test, prediction)
print("Classification report:\n", report)

#plot
plot_decision_regions(X=X, y=y, clf=linear_model)
plt.xlabel("Age", size=14)
plt.ylabel("EstimatedSalary", size=14)
plt.title('SVM Decision Region Boundary', size=16)

```

```

Actual set:
[0 1 1 0 0 1 0 0 0 1 0 0 0 0 0 0 0 0 0 0 1 0 1 1 1 1 0 1 0 0 1 0 0 0 0 1 0
 1 0 0 0 1 0 1 0 1 1 1 0 0 0 0 0 1 0 1 0 0 1 0 0 0 0 0 0 0 0 0 0 1 1 0 0 0 1
 1 0 0 0 0 1]
Predictions:
[0 1 1 1 0 0 0 0 0 1 0 0 0 0 0 0 0 0 0 0 1 0 1 1 1 1 0 1 1 0 1 0 1 0 0 1 0
 1 0 1 0 1 0 1 0 1 1 1 0 0 0 0 0 1 0 1 0 0 1 0 0 0 0 0 0 0 0 0 0 1 1 0 0 0 1
 0 0 0 0 0 1]
Confusion matrix:
[[52  4]
 [ 2 22]]
Classification report:
      precision    recall  f1-score   support

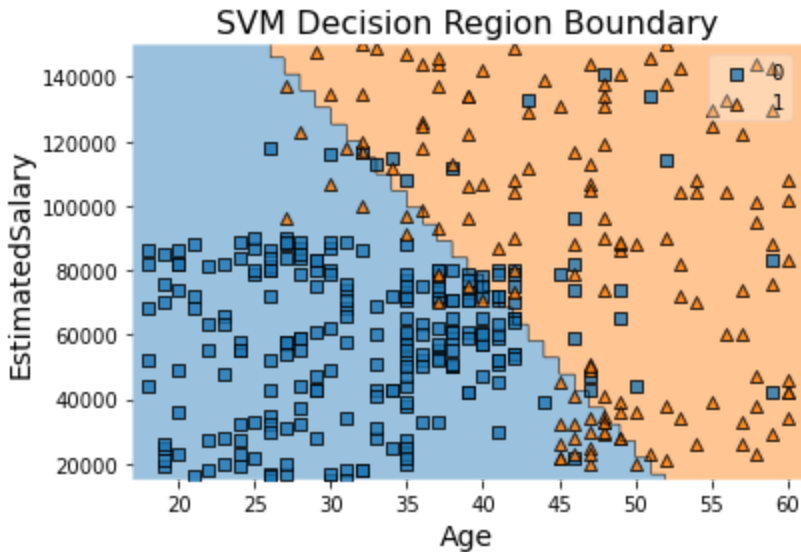
     0       0.96     0.93     0.95         56
     1       0.85     0.92     0.88         24

 accuracy          0.93         80
 macro avg       0.90     0.92     0.91         80
 weighted avg     0.93     0.93     0.93         80

```

Out [6] :

Text(0.5, 1.0, 'SVM Decision Region Boundary')



In [7]:

```
#SVM model(kernel=rbf)
rbf_model=svm.SVC()
rbf_model.fit(X_train, y_train)

#prediction
prediction=rbf_model.predict(X_test)
print("Actual set:\n", y_test)
print("Predictions:\n", prediction)

#Confusion matrix
confusion=metrics.confusion_matrix(y_test, prediction)
print("Confusion matrix:\n", confusion)

#classification report
report=metrics.classification_report(y_test, prediction)
print("Classification report:\n", report)

#plot
plot_decision_regions(X=X, y=y, clf=rbf_model)
plt.xlabel("Age", size=14)
plt.ylabel("EstimatedSalary", size=14)
plt.title('SVM Decision Region Boundary', size=16)
```

Actual set:

```
[0 1 1 0 0 1 0 0 0 1 0 0 0 0 0 0 0 0 0 0 1 0 1 1 1 0 1 0 0 1 0 0 0 0 1 0
 1 0 0 0 1 0 1 0 1 1 1 0 0 0 0 0 1 0 1 0 0 1 0 0 0 0 0 0 0 0 0 0 0 1 1 0 0 0 1
 1 0 0 0 0 1]
```

Predictions:

```
[0 0 1 0 0 0 0 0 0 1 0 0 0 0 0 0 0 0 0 0 0 1 0 0 1 0 0 0 1 0 1 0 0 0 0 1 0]
```

```
1 0 0 0 0 0 1 0 1 0 0 0 0 0 0 1 0 0 0 0 1 0 0 0 0 0 0 0 0 0 0 0 1 0 0 0 1
0 0 0 0 0 0]
```

Confusion matrix:

```
[[55  1]
```

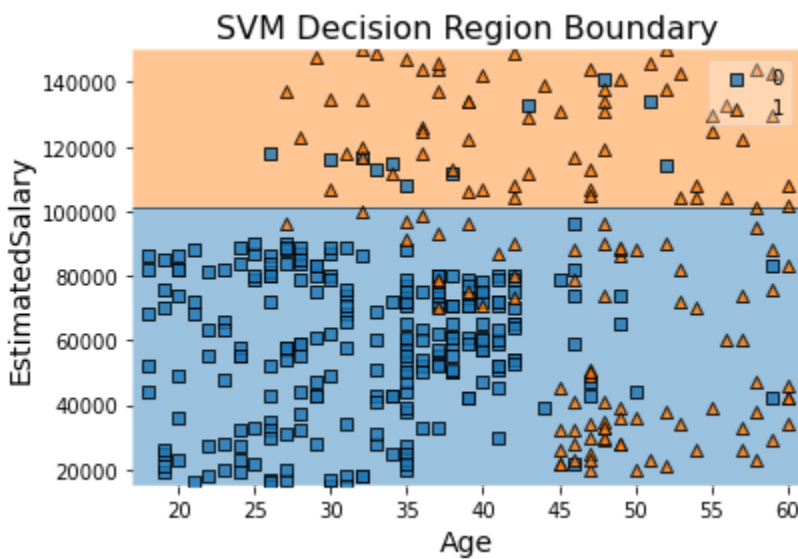
```
[11 13]]
```

Classification report:

	precision	recall	f1-score	support
0	0.83	0.98	0.90	56
1	0.93	0.54	0.68	24
accuracy			0.85	80
macro avg	0.88	0.76	0.79	80
weighted avg	0.86	0.85	0.84	80

Out[7]:

Text(0.5, 1.0, 'SVM Decision Region Boundary')



In [11]:

```
#SVM model(kernel=poly)
poly_model=svm.SVC(kernel='poly', C=4, gamma=0.00000001, max_iter=1e8)
poly_model.fit(X_train, y_train)

#prediction
prediction=poly_model.predict(X_test)
print("Actual set:\n", y_test)
print("Predictions:\n", prediction)

#Confusion matrix
confusion=metrics.confusion_matrix(y_test, prediction)
```

```

print("Confusion matrix:\n", confusion)

#classification report
report=metrics.classification_report(y_test, prediction)
print("Classification report:\n", report)

#plot
plot_decision_regions(X=X, y=y, clf=rbf_model)
plt.xlabel("Age", size=14)
plt.ylabel("EstimatedSalary", size=14)
plt.title('SVM Decision Region Boundary', size=16)

```

C:\Users\Maitreya\anaconda3\envs\tensorflow\lib\site-packages\sklearn\svm\\_base.py:255: ConvergenceWarning: Solver terminated early (max\_iter=1000000000). Consider pre-processing your data with StandardScaler or MinMaxScaler.  
 warnings.warn('Solver terminated early (max\_iter=%i).'

Actual set:  
 [0 1 1 0 0 1 0 0 0 1 0 0 0 0 0 0 0 0 0 0 1 0 1 1 1 1 0 1 0 0 1 0 0 0 0 1 0  
 1 0 0 0 1 0 1 0 1 1 0 0 0 0 0 0 1 0 1 0 0 1 0 0 0 0 0 0 0 0 0 0 0 1 1 0 0 0 1  
 1 0 0 0 0 1]

Predictions:  
 [0 0 1 0 0 0 0 0 0 1 0 0 0 0 0 0 0 0 0 0 0 1 0 0 1 0 0 0 1 0 1 0 0 0 0 1 0  
 1 0 0 0 1 0 1 0 1 1 0 0 0 0 0 0 1 0 1 0 0 1 0 0 0 0 0 0 0 0 0 0 1 1 0 0 0 1  
 0 0 0 0 0 0]

Confusion matrix:  
 [[55 1]  
 [ 7 17]]

Classification report:  
 precision recall f1-score support

0	0.89	0.98	0.93	56
1	0.94	0.71	0.81	24

accuracy			0.90	80
macro avg	0.92	0.85	0.87	80
weighted avg	0.90	0.90	0.90	80

Out[11]:

Text(0.5, 1.0, 'SVM Decision Region Boundary')



