


```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd

from google.colab import drive
drive.mount('/content/drive')


Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).

dataset=pd.read_csv("/content/drive/MyDrive/Personal/Studies/MSC Data Science Material/SEM2/ML/Practical/data_set/Social_Network_Ads.csv")
```

dataset



	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
...
395	15691863	Female	46	41000	1
396	15706071	Male	51	23000	1
397	15654296	Female	50	20000	1
398	15755018	Male	36	33000	0
399	15594041	Female	49	36000	1



400 rows × 5 columns

```
#Splitting dataset into X and Y
x=dataset.iloc[:,[2,3]].values
y=dataset.iloc[:,4].values

#Splitting X and Y dataset into Train and Test data
from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test=train_test_split(x,y,test_size = 0.25, random_state=0)

x_train
```

```
[ 18, 85000],
[ 35, 79000],
[ 57, 33000],
[ 34, 72000],
[ 49, 39000],
[ 27, 31000],
[ 19, 70000],
[ 39, 79000],
[ 26, 81000],
[ 25, 80000],
[ 28, 85000],
[ 55, 39000],
[ 50, 88000],
[ 49, 88000],
[ 52, 150000],
[ 35, 65000],
[ 42, 54000],
[ 34, 43000],
[ 37, 52000],
[ 48, 30000],
[ 29, 43000],
[ 36, 52000],
[ 27, 54000],
[ 26, 118000]])
```

x_test

```
[ 27, 84000],
[ 35, 20000],
[ 43, 112000],
[ 27, 58000],
[ 37, 80000],
[ 52, 90000],
[ 26, 30000],
[ 49, 86000],
[ 57, 122000],
[ 34, 25000],
[ 35, 57000],
[ 34, 115000],
[ 59, 88000],
[ 45, 32000],
[ 29, 83000],
[ 26, 80000],
[ 49, 28000],
[ 23, 20000],
[ 32, 18000],
[ 60, 42000],
[ 19, 76000],
[ 36, 99000],
[ 19, 26000],
[ 60, 83000],
[ 24, 89000],
[ 27, 58000],
[ 40, 47000],
[ 42, 70000],
[ 32, 150000],
[ 35, 77000],
[ 22, 63000],
[ 45, 22000],
[ 27, 89000],
[ 18, 82000],
[ 42, 79000],
[ 40, 60000],
[ 53, 34000],
[ 47, 107000],
[ 58, 144000],
[ 59, 83000],
[ 24, 55000],
[ 26, 35000],
[ 58, 38000],
[ 42, 80000],
[ 40, 75000],
[ 59, 130000],
[ 46, 41000],
[ 41, 60000],
[ 42, 64000],
[ 37, 146000],
[ 23, 48000],
[ 25, 33000],
[ 24, 84000],
[ 27, 96000],
[ 23, 63000],
[ 48, 33000],
[ 48, 90000],
[ 42, 104000]])
```

y_train

```
array([0, 1, 0, 1, 1, 1, 0, 0, 0, 0, 0, 0, 1, 1, 1, 0, 1, 0, 0, 1, 0, 1,
       0, 1, 0, 0, 1, 1, 1, 1, 0, 1, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0,
       0, 1, 1, 1, 1, 0, 0, 0, 1, 0, 1, 0, 1, 0, 0, 1, 0, 0, 0, 1, 0, 0,
       0, 1, 1, 0, 0, 1, 0, 1, 1, 1, 0, 0, 1, 1, 0, 0, 1, 1, 0, 1, 0, 0,
       1, 1, 0, 1, 1, 1, 0, 0, 0, 0, 0, 1, 0, 0, 1, 1, 1, 1, 1, 0, 1, 1,
       0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0, 1, 0,
       1, 1, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 1, 1, 0, 0, 0, 0, 1, 0, 1, 0,
       0, 0, 1, 0, 0, 0, 0, 1, 1, 1, 0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 1, 0,
       1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 1, 0, 1, 0, 0, 1,
       0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 1, 1, 0, 0, 0,
       0, 0, 0, 1, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 1, 0, 1, 0, 0, 0,
       1, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 1, 0, 1,
       1, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 1, 0, 0, 1, 0, 1, 0, 0, 0, 0, 0,
       0, 1, 1, 1, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0])
```

y_test

```
array([0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1,
       0, 1, 0, 1, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0,
       1, 0, 0, 1, 0, 1, 1, 0, 0, 0, 1, 1, 0, 0, 1, 0, 0, 1, 0, 1, 0, 1,
       0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 1, 1, 1, 0, 0, 0, 1, 1, 0, 1,
       1, 0, 0, 1, 0, 0, 0, 1, 0, 1, 0, 1, 1])
```

#Perform Feature scaling

#in the dataset all values are not in the same range hence we use feature scaling to overcome this problem

#feature scaling helps us normalize data within the range

#fit_transform will fit all the data in the variable

from sklearn.preprocessing import StandardScaler

sc=StandardScaler()

X_Train=sc.fit_transform(x_train)

X_Test=sc.fit_transform(x_test)

X_Train

```
[ 1.17584296, 0.53395707 ],
[ 1.07681071, 0.53395707 ],
[ 1.37390747, 2.331532 ],
[-0.30964085, -0.13288524],
[ 0.38358493, -0.45180983],
[-0.4086731 , -0.77073441],
[-0.11157634, -0.50979612],
[ 0.97777845, -1.14764529],
[-0.90383437, -0.77073441],
[-0.21060859, -0.50979612],
[-1.10189888, -0.45180983],
[-1.20093113, 1.40375139]]
```

X_Test

```
[-0.81070599, 0.42134337],
[-0.10879604, -1.53554892],
[ 0.59311391, 1.27748375],
[-0.81070599, -0.37364412],
[ 0.06668145, 0.2990376 ],
[ 1.3827626 , 0.60480202],
[-0.89844474, -1.2297845 ],
[ 1.11954637, 0.48249625],
[ 1.82145632, 1.58324817],
[-0.19653479, -1.38266671],
[-0.10879604, -0.40422056],
[-0.19653479, 1.36921307],
[ 1.99693381, 0.54364914],
[ 0.7685914 , -1.16863161],
[-0.63522851, 0.39076693],
[-0.89844474, 0.2990376 ],
[ 1.11954637, -1.29093738],
[-1.16166097, -1.53554892],
[-0.37201227, -1.5967018 ],
[ 2.08467255, -0.86286719],
[-1.51261594, 0.17673183],
[-0.0210573 , 0.87999 ],
[-1.51261594, -1.35209027],
[ 2.08467255, 0.39076693],
[-1.07392223, 0.57422558],
[-0.81070599, -0.37364412],
[ 0.32989768, -0.70998498],
[ 0.50537516, -0.00672682],
[-0.37201227, 2.43938854],
[-0.10879604, 0.20730828],
[-1.24939971, -0.22076191],
[ 0.7685914 , -1.47439603],
[-0.81070599, 0.57422558],
[-1.60035469, 0.36019049],
[ 0.50537516, 0.26846116],
[ 0.32989768, -0.31249124],
[ 1.47050135, -1.10747873],
[ 0.94406888, 1.12460154],
[ 1.90919507, 2.25592989],
[ 1.99693381, 0.39076693],
[-1.07392223, -0.46537345],
[-0.89844474, -1.07690229],
[ 1.90919507, -0.98517296],
[ 0.50537516, 0.2990376 ],
[ 0.32989768, 0.14615539],
[ 1.99693381, 1.8278597 ],
[ 0.85633014, -0.89344364],
[ 0.41763642, -0.31249124],
[ 0.50537516, -0.19018547],
[ 0.06668145, 2.31708278],
[-1.16166097, -0.67940854],
[-0.98618348, -1.13805517],
[-1.07392223, 0.42134337],
[-0.81070599, 0.78826068],
[-1.16166097, -0.22076191],
[ 1.03180763, -1.13805517],
[ 1.03180763, 0.60480202],
[ 0.50537516, 1.03287221]]
```

```
#We are done with data pre-processing

#Fit SVM to training set

from sklearn.svm import SVC
classifier=SVC(kernel="rbf", random_state=0)
classifier.fit(X_Train, y_train)
```

▼

SVC

SVC(random_state=0)

```
#Predict the test set result
y_pred= classifier.predict(X_Test)

#Make the confusion matrix

from sklearn.metrics import confusion_matrix, accuracy_score
cm=confusion_matrix(y_test, y_pred)
print(cm)
print("\n Accuracy: ")
accuracy_score(y_test,y_pred)

[[64  4]
 [ 3 29]]

Accuracy:
0.93

#Visualise the test set results

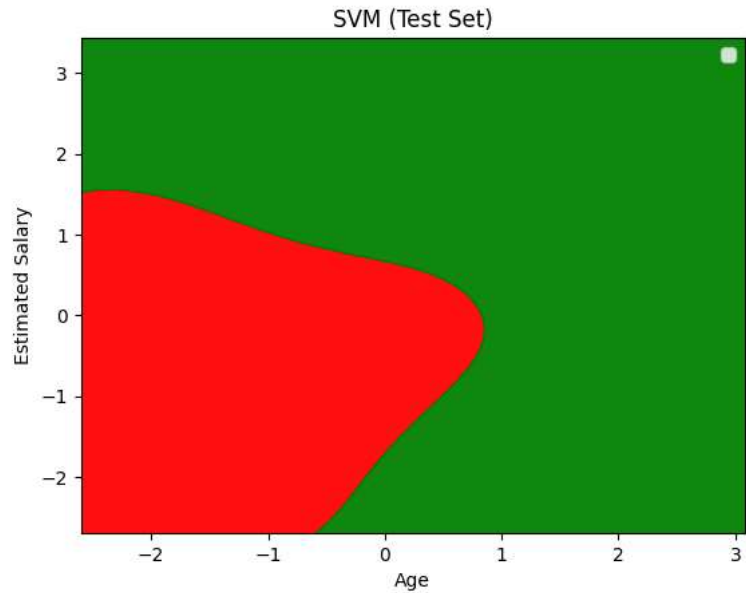
from matplotlib.colors import ListedColormap

X_set, y_set = X_Test, y_test
X1, X2 = np.meshgrid(np.arange(start=X_set[:,0].min() - 1, stop=X_set[:,0].max() + 1, step=0.01),
                     np.arange(start=X_set[:,1].min() - 1, stop=X_set[:,1].max() + 1, step=0.01))
plt.contourf(X1,X2, classifier.predict(np.array([X1.ravel(), X2.ravel()]).T).reshape(X1.shape),
             alpha=0.75, cmap=ListedColormap(('red', 'green')))

plt.xlim(X1.min(), X1.max())
plt.ylim(X2.min(), X2.max())

for i,j in enumerate(np.unique(y_set)):
    plt.contourf(X1, X2, classifier.predict(np.array([X1.ravel(), X2.ravel()]).T).reshape(X1.shape),
                alpha=0.75, cmap=ListedColormap(('red', 'green')))
    plt.title("SVM (Test Set)")
    plt.xlabel("Age")
    plt.ylabel("Estimated Salary")
    plt.legend()
plt.show()
```

WARNING:matplotlib.legend:No artists with labels found to put in legend. Note that



WARNING:matplotlib.legend:No artists with labels found to put in legend. Note that

SVM (Test Set)

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