In [2]: **#Data Pre-processing Step** # importing libraries import numpy as nm import matplotlib.pyplot as mtp import pandas as pd from sklearn import metrics #importing datasets data_set= pd.read_csv('user_data.csv') #Extracting Independent and dependent Variable x= data_set.iloc[:, [2,3]].values y= data_set.iloc[:, 4].values # Splitting the dataset into training and test set. 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) #feature Scaling from sklearn.preprocessing import StandardScaler st x= StandardScaler() x_train= st_x.fit_transform(x_train) x_test= st_x.transform(x_test) In [3]: data_set Out[3]: User ID Gender Age EstimatedSalary Purchased **0** 15624510 19 19000 0 Male **1** 15810944 35 20000 0 Male **2** 15668575 Female 26 43000 0 **3** 15603246 Female 57000 0 27 **4** 15804002 Male 19 76000 0 **395** 15691863 Female 46 41000 1 **396** 15706071 23000 1 Male 51 **397** 15654296 Female 20000 1 33000 0 **398** 15755018 36 Male **399** 15594041 Female 36000 1 400 rows × 5 columns In [4]: x_test Out[4]: array([[-0.80480212, 0.50496393], [-0.01254409, -0.5677824], [-0.30964085, 0.1570462], [-0.80480212, 0.27301877], [-0.30964085, -0.5677824], [-1.10189888, -1.43757673], [-0.70576986, -1.58254245], [-0.21060859, 2.15757314], [-1.99318916, -0.04590581], [0.8787462 , -0.77073441], [-0.80480212, -0.59677555], [-1.00286662, -0.42281668], -0.11157634, -0.42281668], [0.08648817, 0.21503249], [-1.79512465, 0.47597078], [-0.60673761, 1.37475825], [-0.11157634, 0.21503249], [-1.89415691, 0.44697764], [1.67100423, 1.75166912], -0.30964085, -1.37959044], -0.30964085, -0.65476184] 0.8787462 , 2.15757314], 0.28455268, -0.53878926], 0.8787462 , 1.02684052], [-1.49802789, -1.20563157], [1.07681071, 2.07059371], [-1.00286662, 0.50496393], [-0.90383437, 0.30201192], [-0.11157634, -0.21986468], [-0.60673761, 0.47597078], [-1.6960924 , 0.53395707], [-0.11157634, 0.27301877], [1.86906873, -0.27785096], [-0.11157634, -0.48080297],[-1.39899564, -0.33583725], [-1.99318916, -0.50979612], [-1.59706014, 0.33100506], [-0.4086731 , -0.77073441], [-0.70576986, -1.03167271], [1.07681071, -0.97368642], [-1.10189888, 0.53395707], [0.28455268, -0.50979612], [-1.10189888, 0.41798449], [-0.30964085, -1.43757673], [0.48261718, 1.22979253], [-1.10189888, -0.33583725], [-0.11157634, 0.30201192], [1.37390747, 0.59194336], [-1.20093113, -1.14764529], [1.07681071, 0.47597078], [1.86906873, 1.51972397], [-0.4086731 , -1.29261101], [-0.30964085, -0.3648304], [-0.4086731 , 1.31677196], [2.06713324, 0.53395707], [0.68068169, -1.089659], [-0.90383437, 0.38899135], [-1.20093113, 0.30201192], [1.07681071, -1.20563157], [-1.49802789, -1.43757673], [-0.60673761, -1.49556302], [2.1661655 , -0.79972756], [-1.89415691, 0.18603934], [-0.21060859, 0.85288166], [-1.89415691, -1.26361786], [2.1661655 , 0.38899135], [-1.39899564, 0.56295021], [-1.10189888, -0.33583725], 0.18552042, -0.65476184], 0.38358493, 0.01208048], [-0.60673761, 2.331532], [-0.30964085, 0.21503249], [-1.59706014, -0.19087153], [0.68068169, -1.37959044], [-1.10189888, 0.56295021], [-1.99318916, 0.35999821], 0.38358493, 0.27301877], 0.18552042, -0.27785096], 1.47293972, -1.03167271], [0.8787462 , 1.08482681], 1.96810099, 2.15757314], [2.06713324, 0.38899135], [-1.39899564, -0.42281668], [-1.20093113, -1.00267957], 1.96810099, -0.91570013], 0.38358493, 0.30201192], 0.18552042, 0.1570462], 2.06713324, 1.75166912], [0.77971394, -0.8287207], [0.28455268, -0.27785096], [0.38358493, -0.16187839], [-0.11157634, 2.21555943], [-1.49802789, -0.62576869], [-1.29996338, -1.06066585], [-1.39899564, 0.41798449], [-1.10189888, 0.76590222], [-1.49802789, -0.19087153], [0.97777845, -1.06066585], [0.97777845, 0.59194336], [0.38358493, 0.99784738]]) In [5]: y_test Out[5]: array([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, 1, 1, 0, 0, 0, 0, 0, 0, 1, 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, 1, 1], dtype=int64) In [6]: from sklearn.svm import SVC # "Support vector classifier" classifier = SVC(kernel='linear', random_state=0) classifier.fit(x_train, y_train) Out[6]: SVC(kernel='linear', random_state=0) In [7]: #Predicting the test set result y_pred= classifier.predict(x_test) y_pred Out[7]: array([0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, $0,\ 1,\ 0,\ 1,\ 0,\ 0,\ 0,\ 0,\ 0,\ 1,\ 0,\ 0,\ 0,\ 0,\ 0,\ 0,\ 1,\ 0,\ 0,\ 0,$ 1, 0, 0, 1, 0, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 1, 1, 1, 1, 0, 0, 1, 1, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 1], dtype=int64) In [8]: #Creating the Confusion matrix from sklearn.metrics import confusion_matrix cm= confusion_matrix(y_test, y_pred) In [9]: *#Visualizing the training set result:* from matplotlib.colors import ListedColormap x_{set} , $y_{set} = x_{train}$, y_{train} x1, $x2 = nm.meshgrid(nm.arange(start = x_set[:, 0].min() - 1, stop = x_set[:, 0].max() + 1, step = 0.01),$ $nm.arange(start = x_set[:, 1].min() - 1, stop = x_set[:, 1].max() + 1, step = 0.01))$ mtp.contourf(x1, x2, classifier.predict(nm.array([x1.ravel(), x2.ravel()]).T).reshape(x1.shape), alpha = 0.75, cmap = ListedColormap(('red', 'green'))) mtp.xlim(x1.min(), x1.max()) mtp.ylim(x2.min(), x2.max()) for i, j in enumerate(nm.unique(y_set)): $mtp.scatter(x_set[y_set == j, 0], x_set[y_set == j, 1],$ c = ListedColormap(('red', 'green'))(i), label = j) mtp.title('SVM classifier (Training set)') mtp.xlabel('Age') mtp.ylabel('Estimated Salary') mtp.legend() mtp.show() *c* argument looks like a single numeric RGB or RGBA sequence, which should be avoided as value-mapping will have precedence in case its length matches with * x* & *y*. Please use the *color* keyword-argument or provide a 2-D array with a single row if you intend to specify the same RGB or RGBA value for all point *c* argument looks like a single numeric RGB or RGBA sequence, which should be avoided as value-mapping will have precedence in case its length matches with * x* & *y*. Please use the *color* keyword-argument or provide a 2-D array with a single row if you intend to specify the same RGB or RGBA value for all point SVM classifier (Training set) 3 2 Estimated Salary 0 -1-2 Age In [10]: #Visulaizing the test set result from matplotlib.colors import ListedColormap x_{set} , $y_{set} = x_{test}$, y_{test} x1, $x2 = nm.meshgrid(nm.arange(start = x_set[:, 0].min() - 1, stop = x_set[:, 0].max() + 1, step = 0.01),$ $nm.arange(start = x_set[:, 1].min() - 1, stop = x_set[:, 1].max() + 1, step = 0.01))$ mtp.contourf(x1, x2, classifier.predict(nm.array([x1.ravel(), x2.ravel()]).T).reshape(x1.shape), alpha = 0.75, cmap = ListedColormap(('red', 'green'))) mtp.xlim(x1.min(), x1.max()) mtp.ylim(x2.min(), x2.max()) for i, j in enumerate(nm.unique(y_set)): $mtp.scatter(x_set[y_set == j, 0], x_set[y_set == j, 1],$ c = ListedColormap(('red', 'green'))(i), label = j) mtp.title('SVM classifier (Test set)') mtp.xlabel('Age') mtp.ylabel('Estimated Salary') mtp.legend() mtp.show() *c* argument looks like a single numeric RGB or RGBA sequence, which should be avoided as value-mapping will have precedence in case its length matches with * x* & *y*. Please use the *color* keyword-argument or provide a 2-D array with a single row if you intend to specify the same RGB or RGBA value for all point *c* argument looks like a single numeric RGB or RGBA sequence, which should be avoided as value-mapping will have precedence in case its length matches with * x* & *y*. Please use the *color* keyword-argument or provide a 2-D array with a single row if you intend to specify the same RGB or RGBA value for all point SVM classifier (Test set) 3 2 Estimated Salary -1 -2 -2 -1 0 1 Age In [15]: accuracy = metrics.accuracy_score(y_test,y_pred) report = metrics.classification_report(y_test,y_pred) cm = metrics.confusion_matrix(y_test,y_pred) print("Classification report:") print("Accuracy: ", accuracy) print(report) print("Confusion matrix:") print(cm) Classification report: Accuracy: 0.9 recall f1-score precision support 0 0.97 0.89 0.93 68 0.92 0.75 0.83 32 1 accuracy 0.90 100 0.86 macro avg 0.91 0.88 100 0.90 0.90 0.90 100 weighted avg Confusion matrix: [[66 2]

[8 24]]

In []: