Date: 22/Feb/2024 EXPERIMENT – 05

SUPPORT VECTOR MACHINE

AIM: To perform support vector machine

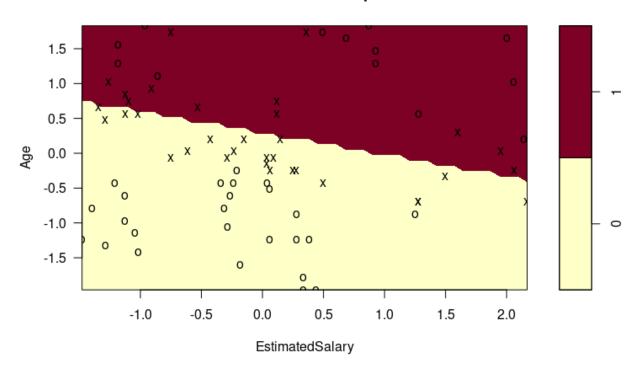
SOFTWARE REQUIRED: RStudio

R CODE:

```
rm(list=ls())
data =read.csv('social.csv' )
library (dplyr)
data=sample n(data, 80)
data=data [3:5]
datasPurchased= factor(data$Purchased, levels=c(0,1))
data TRAIN<-sample n(data,0.9*length(data$Purchased))</pre>
data TEST<-setdiff(data, data TRAIN)</pre>
data TRAIN[-3] <- scale (data TRAIN[-3])</pre>
data TEST[-3] <- scale(data TEST[-3])</pre>
library (e1071)
SVMclassifier =svm(formula=Purchased ~ ., data= data TRAIN, type=
'C-classification', kernel ='linear')
plot(SVMclassifier, data TRAIN)
y p =predict(SVMclassifier, newdata =data TEST[-3])
install.packages('caret')
library (caret)
library (ggplot2)
install.packages('lattice')
library (lattice)
confusionMatrix(table (y p, data TEST$Purchased))
```

OUTPUT:

SVM classification plot



Data			
O data	80 obs. of 3 variables		
O data_TEST	8 obs. of 3 variables		
O data_TRAIN	72 obs. of 3 variables		
SVMclassifier	List of 31		
Values			
datașPurchased	Factor w/ 2 levels "0","1": 1 2 1 2 1 1 2 2 1 2		
y_p	Factor w/ 2 levels "0","1": 1 1 2 1 2 2 1 1		

Confusion Matrix and Statistics

y_p 0 1 0 4 1 1 0 3

Accuracy: 0.875

95% CI: (0.4735, 0.9968)

No Information Rate : 0.5 P-Value [Acc > NIR] : 0.03516

Kappa: 0.75

Mcnemar's Test P-Value : 1.00000

Sensitivity: 1.000
Specificity: 0.750
Pos Pred Value: 0.800
Neg Pred Value: 1.000
Prevalence: 0.500
Detection Rate: 0.500
Detection Prevalence: 0.625
Balanced Accuracy: 0.875

'Positive' Class : 0

•	Age [‡]	EstimatedSalary [‡]	Purchased	÷
1	33	28000		0
2	60	34000		1
3	42	64000		0
4	37	79000		1
5	39	59000		0
6	26	72000		0
7	45	22000		1
8	32	150000		1
9	34	43000		0
10	51	23000		1
11	31	18000		0
12	37	62000		0
13	33	60000		0
14	18	82000		0
15	40	47000		0
16	35	61000		0
17	43	129000		1

```
1 User ID, Gender, Age, Estimated Salary, Purchased
   15624510, Male, 19, 19000, 0
3 15810944, Male, 35, 20000, 0
4 15668575, Female, 26, 43000, 0
 5 15603246, Female, 27, 57000, 0
 6 15804002, Male, 19, 76000, 0
7 15728773, Male, 27, 58000, 0
8 15598044, Female, 27, 84000, 0
9 15694829, Female, 32, 150000, 1
10 15600575, Male, 25, 33000, 0
11 15727311, Female, 35, 65000, 0
12 15570769, Female, 26, 80000, 0
13 15606274, Female, 26, 52000, 0
14 15746139, Male, 20, 86000, 0
15 15704987, Male, 32, 18000, 0
16 15628972, Male, 18,82000,0
17
   15697686, Male, 29, 80000, 0
18 15733883, Male, 47, 25000, 1
19 15617482, Male, 45, 26000, 1
20 15704583, Male, 46, 28000, 1
21 15621083, Female, 48, 29000, 1
   15649487, Male, 45, 22000, 1
   15736760, Female, 47, 49000, 1
23
24 15714658, Male, 48, 41000, 1
25 15599081, Female, 45, 22000, 1
26 15705113, Male, 46, 23000, 1
   15631159, Male, 47, 20000, 1
27
28 15792818, Male, 49, 28000, 1
29 15633531, Female, 47, 30000, 1
```