1.	
import pandas as pd	2.
import numpy as np	import numpy as np
import matplotlib.pyplot as plt	import pandas as pd
dataset = pd.read_csv("suv_data.csv")	import matplotlib.pyplot as plt
dataset.head()	data = pd.read_csv('Titanic-Dataset.csv')
x = dataset.iloc[:, [2, 3]].values	print(data)
y = dataset.iloc[:, 4].values	x = data.drop('Survived', axis = 1)
bool_series=pd.isnull(dataset["Gender"])	y = data['Survived']
dataset[bool_series]	print(x)
bool_series=pd.notnull(dataset["Gender"])	print(y)
dataset[bool_series] dataset[10:25]	x.drop(['Name', 'Ticket', 'Cabin'],axis = 1, inplace
new_data=dataset.dropna(axis=0,how='any')	=
new_data	True)
dataset.replace(to_replace=np.nan,value=-99)	print(x)
dataset["Gender"].fillna("No	x['Age'] = x['Age'].fillna(x['Age'].mean())
Gender",inplace=True)	print(x)
dataset	x['Embarked'] =
<pre>print("Old data frame length:", len(dataset))</pre>	x['Embarked'].fillna(x['Embarked'].mode()[0])
print("New data frame length:", len(dataset))	print(x)
print("Number of rows with at least 1 NA	x = pd.get_dummies(x, columns = ['Sex',
value:",	'Embarked'],prefix = ['Sex',
len(dataset)-len(new_data))	'Embarked'],drop_first =
Old data frame length: 400	True)
New data frame length: 400	print(x)
Number of rows with at least 1 NA value: 0	from sklearn.model_selection import
new_df1=dataset.fillna(method="ffill")	train_test_split
new_df1	x_train, x_test, y_train, y_test =
new_df3=dataset.dropna(how='all')	train_test_split(x, y,
new_df3	test_size = 0.2, random_state = 0)
	print(x_train)
	print(y_train)
	from sklearn.preprocessing import
	StandardScaler
	std_x = StandardScaler()

 $x_{train} = std_x.fit_{transform}(x_{train})$

x_test = std_x.transform(x_test)

print(x_train)

3.	4.
from pandas import read_csv	import numpy as np
from numpy import set_printoptions	import pandas as pd
from sklearn.model_selection import	import matplotlib.pyplot as plt
train_test_split	import seaborn as sns from
from sklearn.feature_selection import	sklearn.feature_selection
SelectKBest	import chi2
from sklearn.feature_selection import f_classif	df=pd.read_csv('loandata.csv')
from matplotlib import pyplot	df.head()
path=r'diabetes.csv'	from sklearn.preprocessing import LabelEncoder
names=['preg','plas','pres','skin','test','mass','pe	for col in df.columns:
ds','ag	le=LabelEncoder()
e','class']	df[col]=le.fit_transform(df[col])
dataframe=read_csv(path,names=names)	df.head()
dataframe.head()	x=df.iloc[:,0:6]
array=dataframe.values	y=df.iloc[:,-1]
x=array[:,0:8]	f_score=chi2(x,y)
y=array[:,8]	f_score
print(x)	<pre>p_value=pd.Series(f_score[1], index=x.columns)</pre>
print(y)	<pre>p_value.sort_values(ascending=False,inplace=Tr</pre>
x_train,x_test,y_train,y_test,=train_test_split(x,y	ue)
,test_	p_value
size=0.33,random_state=1)	p_value.plot(kind="bar")
fs= SelectKBest(score_func=f_classif,k='all')	plt.xlabel("Features", fontsize=20)
fs.fit(x_train,y_train)	plt.ylabel("p_values", fontsize=20)
x_train_fs=fs.transform(x_train)	plt.title("chi squared test base on p value")
x_test_fs=fs.transform(x_test)	plt.show()
for i in range(len(fs.scores_)):	
print('feature %d:%f'%(i,fs.scores_[i]))	
pyplot.bar([i for i in	
range(len(fs.scores_))],fs.scores_)	
pyplot.show()	

5 import numpy as np import matplotlib.pyplot as plt import pandas as pd import seaborn as sns from sklearn import tree from sklearn import metrics from sklearn.metrics import accuracy_score, classification report from sklearn.datasets import load_iris from sklearn.tree import DecisionTreeClassifier from sklearn.model selection import train_test_split iris = load_iris() iris = sns.load dataset('iris') iris.head() x=iris.iloc[:,:-1] y=iris.iloc[:,-1] x_train, x_test, y_train, y_test = train test split(x,y, test size=0.33, random_state=42) treemodel = DecisionTreeClassifier() treemodel.fit(x train, y train) y_pred = treemodel.predict(x_test) plt.figure(figsize=(20,30)) tree.plot tree(treemodel, filled=True) print(classification_report(y_test, y_pred)) from sklearn.metrics import confusion matrix cm=confusion matrix(y test, y pred) print("Confusion Matrix:") print(cm) from sklearn.metrics import accuracy_score accuracy_score(y_test, y_pred)

import numpy as nm import matplotlib.pyplot as mtp import pandas as pd datasetpd.read_csv('User_data.csv") x= dataset.iloc[:,[2,3]].values print(x) from sklearn.preprocessing import StandardScaler from sklearn.naive bayes import GaussianNB y=dataset.iloc[:,4].values print(y) 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) sc StandardScaler() x train sc.fit transform(x train) x_testsc.fit_transform(x_test) classifier GaussianNB() classifier.fit(x_train, y_train) y pred classifier.predict(x test) from sklearn.metrics import confusion matrix cm confusion_matrix(y_test, y_pred) print("Confusion Matrix:") print(cm)

6.

```
7.
                                                         8. import numpy as np
import numpy as np
                                                         import matplotlib.pyplot as plt
model's performance.
                                                         import pandas as pd
import matplotlib.pyplot as plt
                                                         datasetpd.read_csv('Salary_Data.csv')
import pandas as pd
                                                         x=dataset.iloc[:, :-1],values
dataset pd.read_csv('Social Network Ads.csv')
                                                         y=dataset.iloc[:,-1].values
X dataset.iloc[:, :-1].values
                                                         dataset.head()
ydataset.iloc[:,-1].values
                                                         from sklearn.model_selection import
dataset.head()
                                                         train test split
from sklearn.model_selection import
                                                         x_train,x_test,y_train,y_testtrain_test_split(x,y,t
train_test_split
                                                         est_size=1/3, random_state-0)
X_train, X test, y train, y_test train_test_split(X,
                                                         print(x_train)
                                                         print(x_test)
y, test
size 0.20, random state-42)
                                                         print(y train)
X
                                                         print(y_test)
                                                         from sklearn.linear model import
from sklearn.preprocessing import
                                                         LinearRegression
StandardScaler
                                                         regressor-LinearRegression()
se StandardScaler()
                                                         regressor.fit(x_train.y_train)
X_train sc.fit_transform(X_train)
                                                         y_pred-regressor.predict(x_test)
X_test sc.transform(X_test)
                                                         print(y_test)
print(X_train)
                                                         print(y_pred)
                                                         print(np.concatenate((y_test.reshape(len(y_test)))
from sklearn.neighbors import KNeighbors
Classifier
                                                         ,1),y_
classifier KNeighborsClassifier(n_neighbors 5,
                                                         pred.reshape(len(y_pred),131,13)
metrie
                                                         from sklearn.metrics import
'minkowski', p2)
                                                         mean_squared_error
classifier.fit(X_train, y_train)
                                                         mean mean_squared_error(y_test,y_pred)
print(classifier.predict(sc.transform([[46,28000]])
))
                                                         plt.scatter(x_train,y_train,color='red')
y_pred classifier.predict(X_test)
                                                         plt.plot(x train
print(np.concatenate((y_pred.reshape(len(y_pre
                                                         regressor.predict(x_train),color="blue")
d), 1),
                                                         plt.title('salary vs Experience(Training set)')
y_test.reshape(len(y_test),1)),1))
                                                         plt.xlabel('years of Experience')
from sklearn.metrics import confusion_matrix,
                                                         plt.ylabel('salary)
                                                         plt.show()
accuracy_score
                                                         plt.scatterfx test, y test, color-'red')
cm confusion_matrix(y_test, y_pred)
                                                         plt.plot(x train, regressor.predict(x_train),
print(cm)
                                                         color="blue")
accuracy_score(y_test, y_pred)
                                                         plt.titlet Salary vs Experience (Test set)')
                                                         plt.xlabel("Years of Experience')
                                                         plt.ylabel("Salary)
```

plt.show()

J.	10.
import numpy as np	import numpy as nm
import pandas as pd	import matplotlib.pyplot as mtp
from sklearn.model_selection import	import pandas as pd
train_test_split	dataset = pd.read_csv('Mall_Customers.csv')
from sklearn.linear_model import	x = dataset.iloc[:, [3, 4]].values
LinearRegression	dataset.head()
from sklearn.metrics import	import scipy.cluster.hierarchy as shc
mean_squared_error,	<pre>dendro = shc.dendrogram(shc.linkage(x,</pre>
r2_score	method="ward"))
from sklearn.preprocessing import	mtp.title("Dendrogrma Plot")
OneHotEncoder	mtp.ylabel("Euclidean Distances")
from sklearn.compose import	mtp.xlabel("Customers")
ColumnTransformer	mtp.show()
data = pd.read_csv('50_Startups.csv')	from sklearn.cluster import
df = pd.DataFrame(data)	AgglomerativeClustering
print("Dataset:")	<pre>hc = AgglomerativeClustering(n_clusters=5,</pre>
print(df.head())	metric='euclidean', linkage='ward')
X = df.drop(columns=['Profit'])	y_pred = hc.fit_predict(x)
y = df['Profit']	#visulaizing the clusters
column_transformer = ColumnTransformer(mtp.scatter(x[y_pred == 0, 0], x[y_pred == 0, 1], s
transformers=[('encoder', OneHotEncoder(),	=100, c = 'blue', label = 'Cluster 1')
['State'])], remainder='passthrough')	mtp.scatter(x[y_pred == 1, 0], x[y_pred == 1, 1], s
X = column_transformer.fit_transform(X)	=100, c = 'green', label = 'Cluster 2')
X_train, X_test, y_train, y_test =	mtp.scatter(x[y_pred== 2, 0], x[y_pred == 2, 1], s
train_test_split(X, y,	=100, c = 'red', label = 'Cluster 3')
test_size=0.2, random_state=42)	mtp.scatter(x[y_pred == 3, 0], x[y_pred == 3, 1], s
model = LinearRegression()	=100, c = 'cyan', label = 'Cluster 4')
model.fit(X_train, y_train)	mtp.scatter(x[y_pred == 4, 0], x[y_pred == 4, 1], s
y_pred = model.predict(X_test)	=100, c = 'magenta', label = 'Cluster 5')
mse = mean_squared_error(y_test, y_pred)	mtp.title('Clusters of customers')
r2 = r2_score(y_test, y_pred)	mtp.xlabel('Annual Income (k\$)')
print(f"Mean Squared Error: {mse}")	mtp.ylabel('Spending Score (1-100)')
print(f"R-squared Score: {r2}")	mtp.legend()
print("Coefficients:", model.coef_)	mtp.show()
<pre>print("Intercept:", model.intercept_)</pre>	