# 20MCA241 – Data Science Lab

Lab Report Submitted By

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#### **AJC22MCA-2044**

*In Partial fulfilment for the Award of the Degree Of* 

# MASTER OF COMPUTER APPLICATIONS (MCA TWO YEAR)

[Accredited by NBA]

# APJ ABDUL KALAM TECHNOLOGICAL UNIVERSITY



# AMAL JYOTHI COLLEGE OF ENGINEERING KANJIRAPPALLY

[Affiliated to APJ Abdul Kalam Technological University, Kerala. Approved by AICTE, Accredited by NAAC. Koovappally, Kanjirappally, Kottayam, Kerala – 686518]

2022-2024

# DEPARTMENT OF COMPUTER APPLICATIONS

# AMAL JYOTHI COLLEGE OF ENGINEERING KANJIRAPPALLY



# **CERTIFICATE**

This is to certify that the lab report, "20MCA241 DATA SCIENCE LAB" is the bonafide work of FATHIMA HAZBIN R (AJC22MCA-2044) in partial fulfilment of the requirements for the award of the Degree of Master of Computer Applications under APJ Abdul Kalam Technological University during the year 2023-24.

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Course Code	Course Name	Syllabus Year	L-T-P-C
20MCA241	Data Science Lab	2020	0-1-3-2

#### **VISION**

To promote an academic and research environment conducive for innovation centric technical education.

#### **MISSION**

- MS1 Provide foundations and advanced technical education in both theoretical and applied Computer Applications in-line with Industry demands.
- MS2 Create highly skilled computer professionals capable of designing and innovating real life solutions.
- MS3 Sustain an academic environment conducive to research and teaching focused to generate up-skilled professionals with ethical values.
- MS4 Promote entrepreneurial initiatives and innovations capable of bridging and contributing with sustainable, socially relevant technology solutions.

#### **COURSE OUTCOME**

CO	Outcome	Target
CO1	Use different python packages to perform numerical calculations, statistical computations and data visualization.	60.2
CO2	Use different packages and frameworks to implement regression and classification algorithms.	60.2
CO3	Use different packages and frameworks to implement text classification using SVM and clustering using K-means.	
CO4	Implement convolutional neural network algorithm using Keras framework.	60.2
CO5	Implement programs for web data mining and natural language processing using NLTK.	60.2

#### **COURSE END SURVEY**

CO	Survey Question	Answer Format
CO1	To what extend you are able to use different python packages to perform numerical calculations, statistical computations and data visualization?	Excellent/Very Good/Good/Satisfactory/Poor
	To what extend you are able to use different packages and frameworks to implement regression and classification algorithms?	Excellent/Very Good/Good/Satisfactory/Poor
CO3	To what extend you are able to use different packages and frameworks to implement text classification using SVM and clustering using K-means?	Excellent/Very Good/Good/Satisfactory/Poor
CO4	To what extend you are able to implement convolutional neural network algorithm using Keras framework?	Excellent/Very Good/Good/Satisfactory/Poor
CO5	To what extend you are able to implement programs for web data mining and natural language processing using NLTK?	Excellent/Very Good/Good/Satisfactory/Poor

# **CONTENT**

SL. NO.	LIST OF LAB EXPERIMENTS/EXERCISES	DATE	СО	PAGE NO
1	Program to perform matrix operations. Use numpy as the python library and perform the operation using built in functions.	25-09-23	CO1	1
2	Program to perform single value decomposition using numpy.	29-09-23	CO1	3
3	Program to perform data visualisation using python library matplotlib.	29-09-23	CO1	4
4	Program to implement KNN classification using any standard dataset available in the public domain and find the accuracy of algorithm (Iris Dataset)	10-10-23	CO2	5
5	Program to implement KNN classification using any standard dataset available in the public domain and find the accuracy of algorithm (Load Digits)	10-10-23	CO2	6
6	Program to implement Naïve Bayes Algorithm using any standard dataset available in the public domain and find the accuracy of algorithm (Iris Dataset)	31-10-23	CO2	7
7	Program to implement Naïve Bayes Algorithm using any standard dataset available in the public domain and find the accuracy of algorithm (Breast Cancer Dataset)	31-10-23	CO2	8
8	Given dimensional dataset represented with numpy array.  Write a program to calculate slope and intercept	10-11-23	CO2	10
9	Program to implement simple linear regression using any standard dataset available in the public domain and find r2 score.	07-11-23	CO2	11
10	Program to implement linear and multiple regression techniques using any standard dataset available in the public domain and evaluate its performance	10-11-23	CO2	13
11	Program to implement decision trees using any standard dataset available in the public domain and find the accuracy of the algorithm (Iris Dataset)	03-11-23	CO3	14
12	Program to implement decision trees using any standard dataset available in the public domain and find the accuracy of the algorithm (Breast Cancer Dataset)	03-11-23	CO3	16
13	Program to implement k-means clustering technique using any standard dataset available in the public domain (Iris Dataset)	21-11-23	CO3	18
14	Program to implement k-means clustering technique using any standard dataset available in the public domain (Breast Cancer Dataset)	21-11-23	CO3	20
15	Program to implement text classification using support vector machine.	30-11-23	CO3	23
16	Program on artificial neural network to classify images from any standard dataset in the public domain using Keras framework.	01-12-23	CO4	25

17	Program to implement a simple web crawler using requests library	06-12-23	CO5	27
18	Program to implement a simple web crawler and parse the content using BeautifulSoup.	06-12-23	CO5	28
19	Implement problems on natural language processing - Part of Speech tagging, N-gram & smoothening and Chunking using NLTK	07-12-23	CO5	29

<u>Aim:</u> Program to perform matrix operations. Use Numpy as the python library and perform the operations using built in functions in Numpy.

**CO1:** Use different python packages to perform numerical calculations, statistical computations and data visualization.

```
import numpy as np
def input_matrix(ourmatrix):
  r = int(input(f"Enter the no of rows for {ourmatrix}:"))
  c = int(input(f"Enter the no of columns for {ourmatrix}:"))
  matrix=[]
  print("Enter the elements:")
  for i in range(r):
    r=[]
    for j in range(c):
       elements=int(input(f"enter the element at row\{i+1\},colomn\{j+1\}"))
       r.append(elements)
       matrix.append(r)
  return np.array(matrix)
matrix1=input_matrix("matrix1")
input_matrix(matrix1)
matrix2=input_matrix("matrix2")
input_matrix(matrix2)
```

#### **Result:**

**<u>Aim:</u>** Program to perform single value decomposition(SVD) using python Numpy.

**CO1:** Use different python packages to perform numerical calculations, statistical computations and data visualization.

#### **Procedure**:

# **Output Screenshot**

# **Result:**

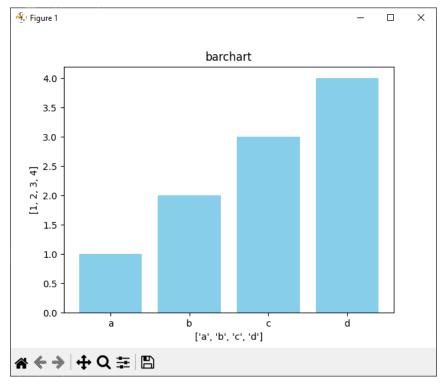
**<u>Aim:</u>** Program to perform data visualisation using the python library matplotlib.

**CO1:** Use different python packages to perform numerical calculations, statistical computations and data visualization.

#### **Procedure**:

```
import matplotlib.pyplot as plt
categories=["a","b","c","d"]
values=[1,2,3,4]
plt.bar(categories,values,color='skyblue')
plt.xlabel(categories)
plt.ylabel(values)
plt.title("barchart")
plt.show()
```

# **Output Screenshot**



# **Result:**

<u>Aim:</u> Program to implement k-NN classification using any standard dataset available in the public domain and find the accuracy of the algorithm(Iris Dataset).

<u>CO2:</u> Use different packages and frameworks to implement regression and classification algorithms.

#### **Procedure**:

```
from sklearn.neighbors import KNeighborsClassifier
from sklearn.model_selection import train_test_split
from sklearn.datasets import load_iris
from sklearn.metrics import accuracy_score
iris=load_iris()
x=iris.data
y=iris.target
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,random_state=42)
knn=KNeighborsClassifier(n_neighbors=7)
knn.fit(x_train,y_train)
print(knn.predict(x_test))
V=knn.predict(x_test)
result=accuracy_score(y_test,V)
print("Accuracy= ",result)
```

# **Output Screenshot**

# **Result:**

<u>Aim:</u> Program to implement k-NN classification using any standard dataset available in the public domain and find the accuracy of the algorithm(Load Digits).

<u>CO2:</u> Use different packages and frameworks to implement regression and classification algorithms.

#### **Procedure**:

```
from sklearn.neighbors import KNeighborsClassifier
from sklearn.datasets import load_digits
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score
digits=load_digits()
x=digits.data
y=digits.target
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,random_state=42)
knn=KNeighborsClassifier(n_neighbors=7)
knn.fit(x_train,y_train)
print((knn.predict(x_test)))
P=knn.predict(x_test)
R=accuracy_score(y_test,P)
print("Accuracy= ",R)
```

# **Output Screenshot**

```
C:\Users\ajcemca\PycharmProjects\pythonProject\venv\Scripts\python.exe C:\Users\ajcemca\PycharmProjects\pythonProject\venv\Hazbi\exp5.py

[6 9 3 7 2 1 5 2 5 2 1 9 4 0 4 2 3 7 8 8 4 3 9 7 5 6 3 5 6 3 4 9 1 4 4 6 9

4 7 6 6 9 1 3 6 1 3 0 6 5 5 1 9 5 6 0 9 0 0 1 0 4 5 2 4 5 7 0 7 5 9 5 5 4

7 0 4 5 5 9 9 0 2 3 8 0 6 4 4 9 1 2 8 3 5 2 9 0 4 4 4 3 5 3 1 3 5 9 4 2 7

7 4 4 1 9 2 7 8 7 2 6 9 4 0 7 2 7 5 8 7 5 7 9 0 6 6 4 2 8 0 9 4 6 9 9 6 9

0 3 5 6 6 0 6 4 3 9 3 4 7 2 9 0 4 5 3 6 5 9 9 8 4 2 1 3 7 7 2 2 3 9 8 0 3

2 2 5 6 9 9 4 1 5 4 2 3 6 4 8 5 9 5 7 8 9 4 8 1 5 4 4 9 6 1 8 6 0 4 5 2 7

4 6 4 5 6 0 3 2 3 6 7 1 5 1 4 7 6 8 8 5 5 1 6 2 8 8 9 5 7 6 2 2 2 3 4 8 8

3 6 0 9 7 7 0 1 0 4 5 1 5 3 6 0 4 1 0 0 3 6 5 9 7 3 5 5 9 9 8 5 3 3 2 0 5

8 3 4 0 2 4 6 4 3 4 5 0 5 2 1 3 1 4 1 1 7 0 1 5 2 1 2 8 7 0 6 4 8 8 5 1 8

4 5 8 7 9 8 6 0 6 2 0 7 9 8 9 5 2 7 7 1 8 7 4 3 8 3 5]

Accuracy= 0.988888888888888889

Process finished with exit code 0
```

#### **Result:**

<u>Aim:</u> Program to implement Naïve Bayes Algorithm using any standard dataset available in the public domain and find the accuracy of the algorithm (Iris Dataset).

**CO2:** Use different packages and frameworks to implement regression and classification algorithms.

#### **Procedure**:

```
from sklearn.naive_bayes import GaussianNB
from sklearn.datasets import load_iris
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score
iris=load_iris()
x=iris.data
y=iris.target
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,random_state=42)
clf=GaussianNB()
clf.fit(x_train,y_train)
print(clf.predict(x_test))
V=clf.predict(x_test)
result=accuracy_score(y_test,V)
print("Accuracy= ",result)
```

# **Output Screenshot**

# **Result:**

<u>Aim:</u> Program to implement Naïve Bayes Algorithm using any standard dataset available in the public domain and find the accuracy of the algorithm (Breast Cancer Dataset).

<u>CO2:</u> Use different packages and frameworks to implement regression and classification algorithms.

```
from sklearn.naive_bayes import GaussianNB
from sklearn.datasets import load_breast_cancer
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score,classification_report
bc=load_breast_cancer()
x=bc.data
y=bc.target
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,random_state=42)
gnb=GaussianNB()
gnb.fit(x_train,y_train)
print(gnb.predict(x_test))
G=gnb.predict(x_test)
result=accuracy_score(y_test,G)
print("Accuracy= ",result)
cr=classification_report(y_test,G)
print("/n Classification Report: ",cr)
```

# **Result:**

<u>Aim:</u> Given a one dimensional data represented with Numpy array. Write a program to calculate slope and intercept.

**CO2:** Use different packages and frameworks to implement regression and classification algorithms.

# **Procedure**:

```
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_squared_error
import numpy as np
x_value = np.array([64,75,68,73,78,82,76,85,71,88]).reshape(-1,1)
y_value = np.array([17,27,15,24,39,44,30,48,19,47])
model=LinearRegression()
model.fit(x_value,y_value)
slope=model.coef_[0]
intercept=model.intercept_
print(f"Slope: {slope}")
print(f"Intercept: {intercept}")
```

# **Output Screenshot**

```
C:\Users\ajcemca\PycharmProjects\pythonProject\venv\Scripts\python.exe C:\Users\ajcemca\PycharmProjects\pythonProject\venv\Hazbi\exp11.py
Slope: 1.6141732283464565
Intercept: -91.6771653543307
Process finished with exit code 0
```

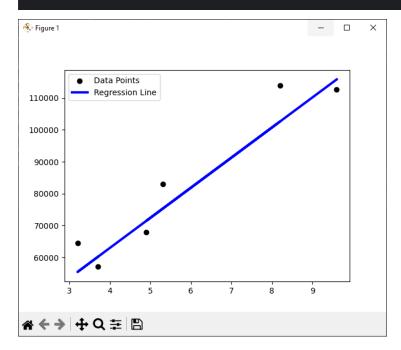
# **Result:**

<u>Aim:</u> Program to implement Simple Linear Regression using any standard dataset available in public domain and find the R2 score.

**CO2:** Use different packages and frameworks to implement regression and classification algorithms.

```
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn.metrics import r2_score
data=pd.read_csv('Salary_Data.csv')
x=data['YearsExperience'].values.reshape(-1,1)
y=data['Salary'].values
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,random_state=42)
LR=LinearRegression()
LR.fit(x_train,y_train)
D=LR.predict(x_test)
r2 = r2\_score(y\_test, D)
print("R2 Score: ", r2)
plt.scatter(x_test,y_test,color='black',label='Data Points')
plt.plot(x_test,D,color='blue', linewidth=3,label='Regression Line')
plt.xlabel='YearsExperience'
plt.ylabel='Salary'
plt.legend()
plt.show()
```

C:\Users\ajcemca\PycharmProjects\pythonProject\venv\Scripts\python.exe C:\Users\ajcemca\PycharmProjects\pythonProject\venv\Hazbi\exp10.py
R2 Score: 0.9024461774180497



# **Result:**

<u>Aim:</u> Program to implement Multiple Linear Regression using any standard dataset available in public domain and evaluate its performance.

**CO2:** Use different packages and frameworks to implement regression and classification algorithms.

#### **Procedure**:

```
import pandas as pd
from sklearn.datasets import fetch california housing
from sklearn.metrics import mean_squared_error
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
california_housing=fetch_california_housing()
df=pd.DataFrame(data=california_housing.data,columns=california_housing.feature_names)
df['Target']=california_housing.target
x=df.drop('Target',axis=1)
y=df['Target']
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,random_state=42)
model = LinearRegression()
model.fit(x train, y train)
predictions = model.predict(x_test)
mse = mean_squared_error(y_test, predictions)
print(f"Mean Squared Error: {mse}")
```

#### **Output Screenshot**

```
C:\Users\ajcemca\PycharmProjects\pythonProject\venv\Scripts\python.exe C:\Users\ajcemca\PycharmProjects\pythonProject\venv\Hazbi\exp12.py
Mean Squared Error: 0.555891598695244

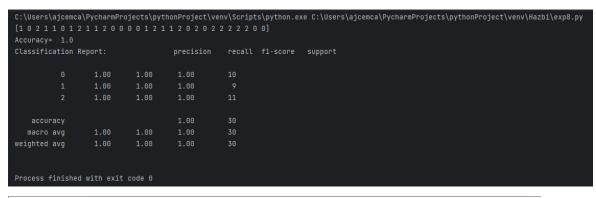
Process finished with exit code 0
```

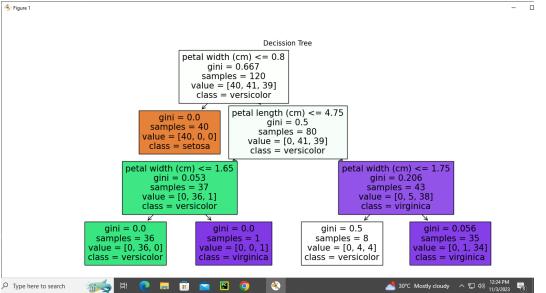
# **Result:**

<u>Aim:</u> Program to implement decision trees using any standard dataset available in the public domain and find the accuracy of the algorithm(Iris Dataset).

**CO3:** Use different packages and frameworks to implement text classification using SVM and clustering using k-means.

```
from sklearn.datasets import load_iris
from sklearn.model_selection import train_test_split
from sklearn.tree import DecisionTreeClassifier,plot_tree
from sklearn.metrics import accuracy_score,classification_report
from matplotlib import pyplot as plt
iris=load_iris()
x=iris.data
y=iris.target
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,random_state=42)
dt=DecisionTreeClassifier(max_depth=3)
dt.fit(x_train,y_train)
print(dt.predict(x_test))
D=dt.predict(x test)
result=accuracy_score(y_test,D)
print("Accuracy= ",result)
cr=classification_report(y_test,D)
print("Classification Report: ",cr)
plt.figure(figsize=(15,20))
plot_tree(dt,filled=True,feature_names=iris.feature_names,class_names=iris.target_names)
plt.title("Decission Tree")
plt.show()
```



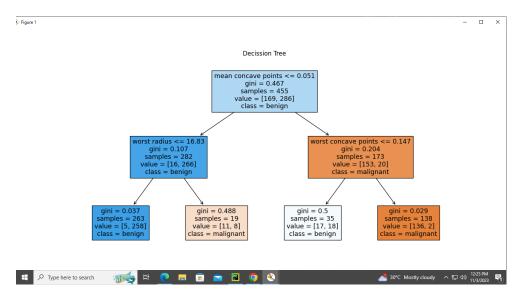


#### **Result:**

<u>Aim:</u> Program to implement decision trees using any standard dataset available in the public domain and find the accuracy of the algorithm(Breast Cancer Dataset).

<u>CO3:</u> Use different packages and frameworks to implement text classification using SVM and clustering using k-means

```
from sklearn.model_selection import train_test_split
from sklearn.tree import DecisionTreeClassifier,plot_tree
from sklearn.metrics import accuracy_score,classification_report
from matplotlib import pyplot as plt
bc=load_breast_cancer()
x=bc.data
y=bc.target
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,random_state=42)
dt=DecisionTreeClassifier(max_depth=2)
dt.fit(x_train,y_train)
print(dt.predict(x_test))
D=dt.predict(x_test)
result=accuracy_score(y_test,D)
print("Accuracy= ",result)
cr=classification_report(y_test,D)
print("Classification Report: ",cr)
plt.figure(figsize=(15,10))
plot_tree(dt,filled=True,feature_names=bc.feature_names,class_names=bc.target_names)
plt.title("Decission Tree")
plt.show()
```

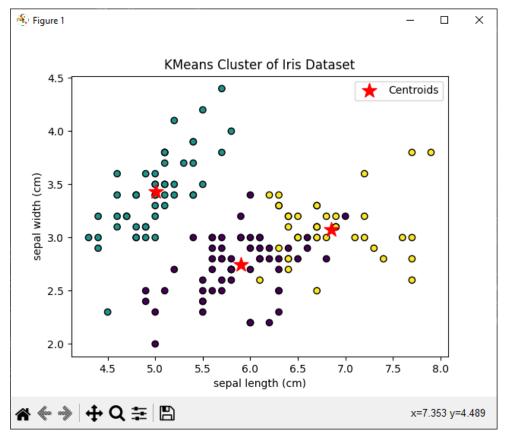


# **Result:**

<u>Aim:</u> Program to implement k-means clustering technique using any standard dataset available in the public domain (Iris Dataset).

<u>CO3:</u> Use different packages and frameworks to implement text classification using SVM and clustering using k-means

```
from sklearn.datasets import load_iris
from sklearn.cluster import KMeans
import matplotlib.pyplot as plt
iris = load_iris()
x = iris.data
y = iris.target
kmeans = KMeans(n clusters=3, random state=42)
kmeans.fit(x)
cluster_labels = kmeans.labels_
print(cluster_labels)
centroids = kmeans.cluster centers
print(centroids)
plt.scatter(x[:, 0], x[:, 1], c=cluster_labels, cmap='viridis', marker='o', edgecolors='black')
plt.scatter(centroids[:, 0], centroids[:, 1], marker="*", s=200, c='red', label='Centroids')
plt.xlabel(iris.feature_names[0])
plt.ylabel(iris.feature_names[1])
plt.title('KMeans Cluster of Iris Dataset')
plt.legend()
plt.show()
```



# **Result:**

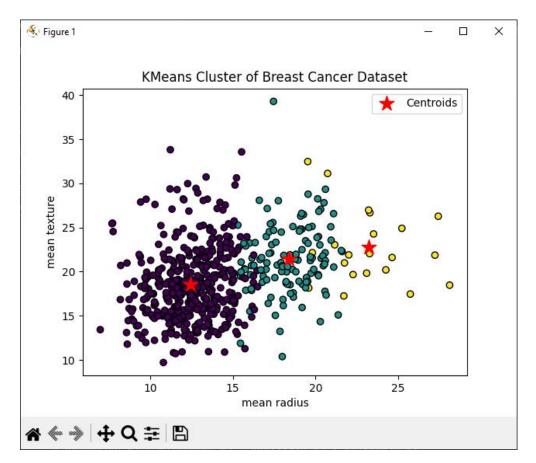
<u>Aim:</u> Program to implement k-means clustering technique using any standard dataset available in the public domain (Breast Cancer Dataset).

<u>CO3:</u> Use different packages and frameworks to implement text classification using SVM and clustering using k-means

```
from sklearn.datasets import load_breast_cancer
from sklearn.cluster import KMeans
import matplotlib.pyplot as plt
bc = load_breast_cancer()
x = bc.data
y = bc.target
kmeans = KMeans(n clusters=3, random state=42)
kmeans.fit(x)
cluster_labels = kmeans.labels_
print(cluster_labels)
centroids = kmeans.cluster centers
print(centroids)
plt.scatter(x[:, 0], x[:, 1], c=cluster_labels, cmap='viridis', marker='o', edgecolors='black')
plt.scatter(centroids[:, 0], centroids[:, 1], marker="*", s=200, c='red', label='Centroids')
plt.xlabel(bc.feature_names[0])
plt.ylabel(bc.feature_names[1])
plt.title('KMeans Cluster of Breast Cancer Dataset')
plt.legend()
plt.show()
```

```
[1\ 1\ 1\ 0\ 1\ 0\ 1\ 0\ 0\ 0\ 1\ 1\ 1\ 0\ 0\ 0\ 1\ 1\ 2\ 0\ 0\ 0\ 0\ 2\ 1\ 1\ 0\ 1\ 1\ 1\ 1\ 0\ 1\ 1\ 1\ 1\ 0
0 0 0 0 0 0 0 0 1 1 1 1 1 0]
```

```
[[1.24468918e+01 1.85046588e+01 8.03803294e+01 4.86458118e+02
  9.48157176e-02 9.04881882e-02 6.08800016e-02 3.25494682e-02
  1.77951765e-01 6.35771765e-02 3.00681647e-01 1.21837294e+00
  2.12940400e+00 2.32080188e+01 7.17541647e-03 2.33490235e-02
  2.84143873e-02 1.05452329e-02 2.07137600e-02 3.75171835e-03
 1.38918094e+01 2.45948235e+01 9.09125412e+01 6.04658353e+02
 1.29845529e-01 2.21074000e-01 2.14822228e-01 8.97035082e-02
  2.82468471e-01 8.32831059e-02]
 [1.83820325e+01 2.14148780e+01 1.21238537e+02 1.05796098e+03
 1.00221870e-01 1.40414797e-01 1.58604959e-01 9.06387805e-02
 1.91033333e-01 6.06883740e-02 6.40926016e-01 1.20443577e+00
  4.50100813e+00 7.53708943e+01 6.57197561e-03 3.09567967e-02
  4.08994309e-02 1.53647805e-02 2.00486992e-02 3.93508943e-03
  2.22162602e+01 2.86411382e+01 1.47833333e+02 1.52278862e+03
 1.39408780e-01 3.45692358e-01 4.26761789e-01 1.81023984e-01
  3.15549593e-01 8.64585366e-02]
 [2.32147619e+01 2.27285714e+01 1.55066667e+02 1.70276190e+03
 1.05001429e-01 1.73405714e-01 2.44971429e-01 1.35852381e-01
 1.88309524e-01 5.93747619e-02 1.13901429e+00 1.25883333e+00
  8.19842857e+00 1.81798571e+02 7.06723810e-03 3.64780952e-02
  4.95609524e-02 1.62100000e-02 1.99633333e-02 3.84780952e-03
  2.95500000e+01 3.02228571e+01 2.00490476e+02 2.70328571e+03
  1.42195238e-01 3.90485714e-01 5.27814286e-01 2.29571429e-01
  2.94823810e-01 8.26404762e-02]]
```



# **Result:**

**<u>Aim:</u>** Program to implement test classification using Support Vector Machine.

**<u>CO3:</u>** Use different packages and frameworks to implement text classification using SVM and clustering using k-means.

```
from sklearn.datasets import fetch 20newsgroups
from sklearn.model_selection import train_test_split
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.svm import SVC
from sklearn.metrics import accuracy_score,classification_report
categories=['alt.atheism','soc.religion.christian','comp.graphics','sci.med']
twenty_train=fetch_20newsgroups(subset='train',categories=categories,shuffle=True,random_sta
te=42)
vectorizer=TfidfVectorizer()
x_train_tfidf=vectorizer.fit_transform(twenty_train.data)
y_train=twenty_train.target
x_train,x_test,y_train,y_test=train_test_split(x_train_tfidf,y_train,test_size=0.3,random_state=42
)
svm_classifier=SVC(kernel='linear',random_state=42)
svm_classifier.fit(x_train,y_train)
predictions=svm classifier.predict(x test)
accuracy=accuracy_score(y_test,predictions)
classification=classification_report(y_test,predictions,target_names=twenty_train.target_names)
print("Accuracy: ",accuracy)
print("Classification Report: ",classification)
new_data=["I have a question about computer graphics","This is a medical related topic"]
x_new_tfidf=vectorizer.transform(new_data)
new_predictions=svm_classifier.predict(x_new_tfidf)
for i,text in enumerate(new data):
```

predicted\_category=twenty\_train.target\_names[new\_predictions[i]]
print("Predicted Cate",predicted\_category)

# **Output Screenshot**

		honProject	t\venv\Script:	s\python.e	xe C:\User	s\ajcemca\PycharmProjects\pythonProject\venv\Hazbi\exp15.py
Accuracy: 0.9646017699115	5044					
Classification Report:			precision	recall	f1-score	support
alt.atheism	0.98	0.95	0.96	129		
comp.graphics	0.92	0.99	0.96	169		
sci.med	0.98	0.96	0.97	189		
soc.religion.christian	0.97	0.96	0.97	191		
accuracy			0.96	678		
macro avg	0.97	0.96	0.96	678		
weighted avg	0.97	0.96	0.96	678		
Predicted Cate comp.graphi	ics					
Predicted Cate sci.med						
Process finished with exit code 0						

# **Result:**

<u>Aim:</u> Program on artificial neural network to classify images from any standard dataset in the public domain using Keras framework.

**CO4:** Implement convolutional neural network algorithm using Keras framework.

```
import tensorflow as tf
from tensorflow.keras.datasets import mnist
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, Flatten
from tensorflow.keras.utils import to_categorical
# Load the MNIST dataset
(X_train, y_train), (X_test, y_test) = mnist.load_data()
# Normalize pixel values to be between 0 and 1
X_{train} = X_{train} / 255.0
X_{\text{test}} = X_{\text{test}} / 255.0
# Flatten the images (convert 28x28 images to 1D vectors)
X train = X train.reshape(-1, 28 * 28)
print(X_train)
X_{\text{test}} = X_{\text{test.reshape}}(-1, 28 * 28)
print(X_train)
# One-hot encode the target labels
y_train = to_categorical(y_train)
y_test = to_categorical(y_test)
print(y_test)
# Create a simple feedforward neural network model
model=Sequential([
Dense(128, activation='relu', input_shape=(28 * 28,)),
Dense(68, activation='relu'),
Dense(10, activation='softmax')
```

```
])
model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
model.fit(X_train,y_train, epochs=5, batch_size=32, validation_split=0.2)
loss, accuracy= model.evaluate(X_test,y_test)
print(accuracy)
```

#### **Result:**

**<u>Aim:</u>** Program to implement a simple web crawler using requests library.

**CO5:** Implement programs for web data mining and natural language processing using NLTK.

#### **Procedure:**

```
import requests
def simple_scraper(url):
    response=requests.get(url)
    if response.status_code==200:
        print("Content:")
        print(response.text)
    else:
        print("Failed to fetch the page. Status code:", response.status_code)
url_to_scrap="https://ajce.in"
simple_scraper(url_to_scrap)
```

# **Output Screenshot**

# **Result:**

**<u>Aim:</u>** Program to implement a simple web crawler and parse the content using BeautifulSoup.

**CO5:** Implement programs for web data mining and natural language processing using NLTK.

#### **Procedure:**

```
import requests
from bs4 import BeautifulSoup

def simple_scraper(url):
    response=requests.get(url)
    if response.status_code==200:
        soup=BeautifulSoup(response.content, 'html.parser')
        print("Title:",soup.title.string)
        print("Content:")
        print(soup.get_text())
    else:
        print("Failed to fetch the page. Status code:", response.status_code)
url_to_scrap="https://ajce.in"
simple_scraper(url_to_scrap)
```

# **Output Screenshot**

```
C:\Users\ajcemca\PycharmProjects\pythonProject\venv\Scripts\python.exe C:\Users\ajcemca\PycharmProjects\pythonProject\venv\Hazbi\soup.py
Title: Amal Jyothi College of Engineering (Autonomous)
Content:

Amal Jyothi College of Engineering (Autonomous)
```

# **Result:**

**<u>Aim:</u>** Program to implement a simple web crawler using requests library.

**CO5:** Implement programs for web data mining and natural language processing using NLTK.

```
import nltk
nltk.download('brown')
from nltk.tokenize import word_tokenize
from nltk.util import ngrams
from nltk.corpus import brown
from nltk.chunk import RegexpParser
sentence = "The quick brown fox jumps over the lazy dog"
tokens = word_tokenize(sentence)
print(tokens)
pos_tags = nltk.pos_tag(tokens)
print("Part-of-speech Tagging:")
print(pos_tags)
text = brown.words(categories='news')[:1000]
bigrams = list(ngrams(text, 2))
freq_dist = nltk.FreqDist(bigrams)
print("\nN-gram Analysis(Bigrams with Smoothing):")
for bigram in bigrams:
print(f"{bigram}:{freq_dist[bigram]}")
tagged_sentence = nltk.pos_tag(word_tokenize("The quick brown fox jumps over the lazy dog"))
grammar = r"NP: {<DT>?<JJ>*<<math>NN>}"
cp = RegexpParser(grammar)
result = cp.parse(tagged_sentence)
print("\nChunking with Regular Expression and POS tags:")
print(result)
```

```
['The', 'quick', 'brown', 'fox', 'jumps', 'over', 'the', 'lazy', 'dog']
Part-of-speech Tagging:
[('The', 'DT'), ('quick', 'JJ'), ('brown', 'NN'), ('fox', 'NN'), ('jumps', 'VBZ'), ('over', 'IN'), ('the', 'DT'), ('lazy', 'JJ'), ('dog', 'NN')]

Chunking with Regular Expression and POS tags:
(S
    (NP The/DT quick/JJ brown/NN)
    (NP fox/NN)
    jumps/VBZ
    over/IN
    (NP the/DT lazy/JJ dog/NN))

Process finished with exit code 0
```

# **Result:**