

EX.NO:1	DEMONSTRATE AGGREGATION
DATE:	

AIM:

To write a program to understand aggregate functions in python.

APPROACH:

Step 1: Open Jupiter notebook, create a new file ex01.

Step 2: Import Pandas library from python.

Step 3: Create a data with dictionary datatype in a variable and convert it into dataframe.

Step 4: Select two or more attributes and perform aggregation using sum, average, min, max and mean.

Step 5: Execute the program and display the output.

PROGRAM:

```
import pandas as pd
technologies = {
    'Courses':["Spark","PySpark","Hadoop","Python","PySpark","Spark"],
    'Fee': [20000,25000,26000,22000,24000,3000],
    'Duration':['30day','40days','35days','40days','60days','60days'],
    'Discount':[1000,2300,1200,2500,2000,2000]}
df = pd.DataFrame(technologies)
print("DataFrame:")
print(df)
print()

# Using Aggregate Functions Sum
print("Aggregate Functions sum on Fee and Discount: ")
result = df[['Fee','Discount']].aggregate('sum')
print(result)
print()

# Using Aggregate Function Average
print("Aggregate Functions average on Fee and Discount: ")
result = df[['Fee','Discount']].aggregate('average')
print(result)
print()

# Using Aggregate Function MinMax
result = df.groupby('Courses')['Fee'].aggregate(['min','max'])
print(result)
print()

# Using Aggregate Function Mean
print("Aggregate Functions mean value on Fee and Discount: ")
result = df[["Fee","Discount"]].mean()
print(result)
```

OUTPUT:

```
↳ DataFrame:
   Courses  Fee Duration Discount
0   Spark 20000   30day    1000
1 PySpark 25000   40days    2300
2   Hadoop 26000   35days    1200
3   Python 22000   40days    2500
4 PySpark 24000   60days    2000
5   Spark  3000   60days    2000

Aggregate Functions sum on Fee and Discount:
Fee          120000
Discount      11000
dtype: int64

Aggregate Functions average on Fee and Discount:
10916.666666666666

      min    max
Courses
Hadoop  26000  26000
PySpark 24000  25000
Python  22000  22000
Spark   3000   20000

Aggregate Functions mean value on Fee and Discount:
Fee          20000.000000
Discount      1833.333333
dtype: float64
```

RESULT:

EX.NO:1(a)	DISPLAY AGGREGATES AND VALUES
DATE:	

AIM:

To write a Python program that calculates and displays the mean, median, and sum of a given list of numbers.

ALGORITHM:**PROGRAM:**

OUTPUT:

CRITERIA	MAX.MARKS	MARKS OBTAINED
AIM & DESCRIPTION	15	
VIVA	10	
TOTAL	25	

RESULT:

EX.NO:2	DEMONSTRATE INDEXING AND SORTING
DATE:	

AIM:

To write a program to understand indexing and sorting in dataframes using python.

APPROACH:

Step 1: Open Jupiter notebook, create a new file ex02.

Step 2: Import Pandas library from python.

Step 3: Create a data with dictionary datatype in a variable and convert it into dataframe.

Step 4: Using sort_values() the dataframe is sorted.

Step 5: Using reset_index() the sorted dataframe is indexed.

Step 6: Execute the program and display the output.

PROGRAM:

```
import pandas as pd
technologies = {
    'Courses':["Spark","PySpark","Hadoop","Python","PySpark","Spark"],
    'Fee': [20000,25000,26000,22000,24000,3000],
    'Duration':['30day','40days','35days','40days','60days','60days'],
    'Discount':[1000,2300,1200,2500,2000,2000]
}
#before sorting and indexing
df = pd.DataFrame(technologies)
print(df)
print()

# indexing
print("Indexed dataframe:")
df.index.name = 'Index'
print(df)
print()
# after sorting
print("Sorted dataframe: ")
sorted_df = df.sort_values(by = ["Courses"])
print(sorted_df)
print()
```

OUTPUT:

```
↳ Courses      Fee Duration Discount
0    Spark    20000    30day    1000
1  PySpark    25000    40days    2300
2    Hadoop    26000    35days    1200
3    Python    22000    40days    2500
4  PySpark    24000    60days    2000
5    Spark     3000    60days    2000
```

Indexed dataframe:

```
      Courses      Fee Duration Discount
Index
0      Spark    20000    30day    1000
1  PySpark    25000    40days    2300
2    Hadoop    26000    35days    1200
3    Python    22000    40days    2500
4  PySpark    24000    60days    2000
5      Spark     3000    60days    2000
```

Sorted dataframe:

```
      Courses      Fee Duration Discount
Index
2    Hadoop    26000    35days    1200
1  PySpark    25000    40days    2300
4  PySpark    24000    60days    2000
3    Python    22000    40days    2500
0      Spark    20000    30day    1000
5      Spark     3000    60days    2000
```

RESULT:

EX.NO:2(A)	DATAFRAME OPERATIONS
DATE:	

AIM:

To write a Python program to create a DataFrame from a dictionary, then index it by a specific column and sort the DataFrame based on that column.

ALGORITHM:**PROGRAM:**

OUTPUT:

CRITERIA	MAX.MARKS	MARKS OBTAINED
AIM & DESCRIPTION	15	
VIVA	10	
TOTAL	25	

RESULT:

EX.NO:3	DEMONSTRATE HANDLING OF MISSING DATA
DATE:	

AIM:

To write a program to handle missing data in a CSV file using python.

APPROACH:

1. Load the Dataset.
2. Read the dataset from a CSV file into a Pandas DataFrame.
3. Identify Missing Data.
4. Handle Missing Data.
5. Analyze the Cleaned Data.
6. Output and Result.

PROGRAM:

```
import pandas as pd

df = pd.read_csv('your_dataset.csv')

missing_data = df.isnull()
data_info = df.info()

df.fillna(df.mean(), inplace=True)

print("Cleaned Dataset:")
print(df)

print("Missing Data Information:")
print(missing_data)
print("Dataset Information:")
print(data_info)
```

OUTPUT:

Output of Dataset Information:

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 4 entries, 0 to 3

Data columns (total 3 columns):

#	Column	Non-Null	Count	Datatype
0	1	Yes	4	Float64
1	2	Yes	3	Int32
2	3	yes	4	Int16

dtypes: float64(3)

memory usage: 224.0 bytes

RESULT:

EX.NO:3(a)	DATAFRAME REFINING
DATE:	

AIM:

To Write a Python program that reads a CSV file into a DataFrame and displays the count of missing values for each column in the DataFrame.

ALGORITHM:**PROGRAM:**

OUTPUT:

CRITERIA	MAX.MARKS	MARKS OBTAINED
AIM & DESCRIPTION	15	
VIVA	10	
TOTAL	25	

RESULT:

EX.NO:4**DATE:****DEMONSTRATE HIERARCHICAL INDEXING****AIM:**

To write a program to understand hierarchical indexing in csv file using python.

APPROACH:

- Step 1: Open Jupiter notebook, create a new file ex04.
- Step 2: Import Pandas library from python.
- Step 3: Upload the csv file in the current directory.
- Step 4: Read the content of csv file using read_csv().
- Step 5: Set multiple levels of indexing using function set_index().
- Step 6: Execute the program and display the output.

PROGRAM:

```
import pandas as pd
```

```
df = pd.read_csv('homelessness.csv')  
print(df.head()) # first 10 rows from dataframe  
print()
```

```
# using the pandas set_index() function.  
indexed_df = df.set_index(['region', 'state', 'individuals'])
```

```
print(indexed_df.head())
```

OUTPUT:

```
↗  
   region      state  individuals  family_members  state_pop  
0  East South Central  Alabama      2570           864    4887681  
1      Pacific      Alaska      1434           582     735139  
2      Mountain      Arizona      7259          2606    7158024  
3  West South Central  Arkansas      2280           432    3009733  
4      Pacific  California    109008          20964   39461588
```

```
      region      state  individuals  family_members  state_pop  
East South Central  Alabama      2570           864    4887681  
Pacific      Alaska      1434           582     735139  
Mountain      Arizona      7259          2606    7158024  
West South Central  Arkansas      2280           432    3009733  
Pacific      California    109008          20964   39461588
```

RESULT:

EX.NO:4(a)	DATAFRAME INDEXING AND DISPLAYING
DATE:	

AIM:

To Write a Python program to read a CSV file with hierarchical index columns and display the data using hierarchical indexing.

ALGORITHM:**PROGRAM:**

OUTPUT:

CRITERIA	MAX.MARKS	MARKS OBTAINED
AIM & DESCRIPTION	15	
VIVA	10	
TOTAL	25	

RESULT:

EX.NO:5	DEMONSTRATE USAGE OF PIVOT TABLE
DATE:	

AIM:

To write a program to create a pivot table of dataframe using python.

APPROACH:

Step 1: Open Jupiter notebook, create a new file ex05.

Step 2: Import Pandas library from python.

Step 3: Import Numpy library from python.

Step 4: Create a data with dictionary datatype in a variable and convert it into dataframe.

Step 5: A pivot table is created using pivot_table() function.

Step 6: Basic aggregation operations are performed on the pivot table.

Step 7: Execute the program and display the output.

PROGRAM:

```
import pandas as pd
```

```
import numpy as np
```

```
#Pandas pivot tables provide a powerful tool to perform these analysis techniques with python.
```

```
#Creating own dataframe or load a dataset
```

```
df = pd.DataFrame({"A": ["burger", "burger", "burger", "burger", "burger", "fries", "fries", "fries", "fries"],
```

```
                  "B": ["one", "one", "one", "two", "two", "one", "one", "two", "two"],
```

```
                  "C": ["small", "large", "large", "small", "small", "large", "small", "small", "large"],
```

```
                  "D": [1, 2, 2, 3, 3, 4, 5, 6, 7],
```

```
                  "E": [2, 4, 5, 5, 6, 6, 8, 9, 9]})
```

```
print("Dataframe:")
```

```
print(df)
```

```
print()
```

```
#CREATING BASIC PIVOT TABLE
```

```
print("BASIC PIVOT TABLE")
```

```
table = pd.pivot_table(df, index=['A', 'B'])
```

```
print(table)
```

```
#SOME BASIC AGGREGATE OPERATION USAGE USING PIVOT
```

```
print("SOME BASIC AGGREGATE OPERATIONS")
```

```
table = pd.pivot_table(df, index=['A', 'B'], aggfunc=np.sum)
```

```
print(table)
```

```
print()
```

```
#mean
```

```
table = pd.pivot_table(df, values=['D', 'E'], index=['A', 'C'],
```

```
                      aggfunc={'D': np.mean,
```

```
                      'E': np.mean})
```

```
print(table)
```

```
print(0)
```



```
#multiple aggregations with pivot
table = pd.pivot_table(df, values=['D', 'E'], index=['A', 'C'],
                        aggfunc={'D': np.mean,
                                'E': [min, max, np.mean]})
print(table)
```

OUTPUT:

↳ Dataframe:

	A	B	C	D	E
0	burger	one	small	1	2
1	burger	one	large	2	4
2	burger	one	large	2	5
3	burger	two	small	3	5
4	burger	two	small	3	6
5	fries	one	large	4	6
6	fries	one	small	5	8
7	fries	two	small	6	9
8	fries	two	large	7	9

BASIC PIVOT TABLE

		D		E	
A	B				
burger	one	1.666667		3.666667	
	two	3.000000		5.500000	
fries	one	4.500000		7.000000	
	two	6.500000		9.000000	

SOME BASIC AGGREGATE OPERATIONS

		D		E	
A	B				
burger	one	5	11		
	two	6	11		
fries	one	9	14		
	two	13	18		

		D		E	
A	C				
burger	large	2.000000		4.500000	
	small	2.333333		4.333333	
fries	large	5.500000		7.500000	
	small	5.500000		8.500000	

		D		E	
		mean	max	mean	min
burger	large	2.000000	5.0	4.500000	4.0
	small	2.333333	6.0	4.333333	2.0
fries	large	5.500000	9.0	7.500000	6.0
	small	5.500000	9.0	8.500000	8.0

RESULT:

EX.NO:5(a)	PIVOT TABLE CREATION
DATE:	

AIM:

To Write a Python program to create a pivot table from a given DataFrame, aggregating values based on specified rows and columns.

ALGORITHM:**PROGRAM:**

OUTPUT:

CRITERIA	MAX.MARKS	MARKS OBTAINED
AIM & DESCRIPTION	15	
VIVA	10	
TOTAL	25	

RESULT:

EX.NO:6a	DEMONSTRATE USE OF EVAL()
DATE:	

AIM:

To write a program to perform eval operation using python.

APPROACH:

Step 1: Open Jupiter notebook, create a new file ex06.

Step 2: Import Pandas library from python.

Step 3: Import NumPy library from python.

Step 4: Get the expression as input from user.

Step 5: Get the values of the variables used in the expression from the user.

Step 6: Then the expression is evaluated using eval() function.

Step 7: Execute the program and display the output.

PROGRAM:

```
import numpy as np
import pandas as pd

# expression to be evaluated
expr = input("Enter the function(in terms of a and b):")

# variable used in expression
a = int(input("Enter the value of a:"))
b = int(input("Enter the value of b:"))

# evaluating expression
res= eval(expr)

# printing evaluated result
print("Result=", res)
```

OUTPUT:

```
☞ Enter the function(in terms of a and b):(a*a)+(b*b)+2*a*b
Enter the value of a:2
Enter the value of b:4
Result= 36
```

RESULT:

EX.NO:6b	DEMONSTRATE USE OF QUERY()
DATE:	

AIM:

To write a program to perform query operation using python.

APPROACH:

Step 1: Open Jupiter notebook, create a new file ex06.

Step 2: Import Pandas library from python.

Step 3: Import NumPy library from python

Step 4: Create a data with dictionary datatype in a variable and convert it into dataframe.

Step 5: Write a query according to the given statement and process using query() function.

Store the result set in a separate variable.

Step 6: Execute the program and display the output.

PROGRAM:

```
import numpy as np
import pandas as pd
```

```
dataFrame = pd.DataFrame({'Name': ['RACHEL', 'MONICA', 'PHOEBE','ROSS',
'CHANDLER', 'JOEY'],
                          'Age': [30, 35, 37, 23, 24, 30],
                          'Salary': [100000, 93000, 88000, 120000, 94000, 95000],
                          'JOB': ['DESIGNER', 'CHEF', 'MASUS', 'PALENTOLOGY', 'IT',
'ARTIST']})
print('Dataframe:')
print(dataFrame)
print()
```

```
#Basic query
print("To print age below 30")
res=dataFrame.query('Age<30')
print(res)
print()
```

```
#multiple query
print("To print salary <=100000 and age<40 and job name starts with c")
res=dataFrame.query('Salary <= 100000 & Age < 40 & JOB.str.startswith("C").values')
print(res)
```

OUTPUT:

☞ Dataframe:

	Name	Age	Salary	JOB
0	RACHEL	30	100000	DESIGNER
1	MONICA	35	93000	CHEF
2	PHOEBE	37	88000	MASUS
3	ROSS	23	120000	PALENTOLOGY
4	CHANDLER	24	94000	IT
5	JOEY	30	95000	ARTIST

To print age below 30

	Name	Age	Salary	JOB
3	ROSS	23	120000	PALENTOLOGY
4	CHANDLER	24	94000	IT

To print salary <=100000 and age<40 and job name starts with c

	Name	Age	Salary	JOB
1	MONICA	35	93000	CHEF

RESULT:

EX.NO:6(c)	EVALUATE USER INPUTS
DATE:	

AIM:

To Write a Python program that uses the eval() function to perform a mathematical operation on user-input numbers.

ALGORITHM:**PROGRAM:**

OUTPUT:

CRITERIA	MAX.MARKS	MARKS OBTAINED
AIM & DESCRIPTION	15	
VIVA	10	
TOTAL	25	

RESULT:

EX.NO:7a	DEMONSTRATE SCATTER PLOT
DATE:	

AIM:

To write a program to display the dataset in a scatter plot using python.

APPROACH:

Step 1: Open Jupiter notebook, create a new file ex07.

Step 2: Import pyplot from matplotlib.

Step 3: Create the coordinates values for x and y axis.

Step 4: Using scatter() function set x and y as parameters.

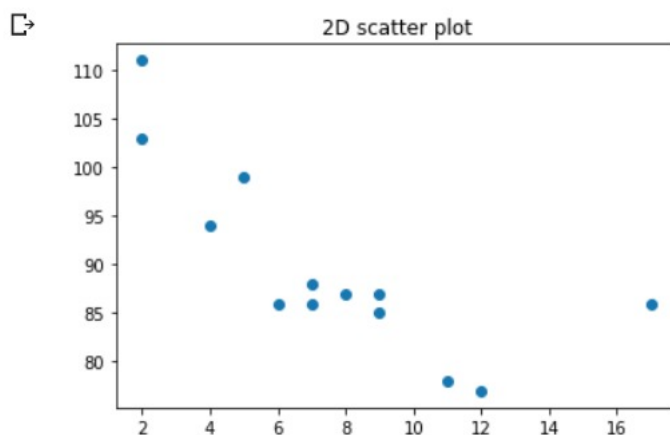
Step 5: Using title() to set the title and show() to display the graph.

Step 6: Execute the program and display the output.

PROGRAM:

```
import matplotlib.pyplot as plt
from mpl_toolkits import mplot3d
import numpy as np
#A scatter plot is a diagram where each value in the data set is represented by a dot.
#The Matplotlib module has a method for drawing scatter plots
#BASIC 2D X,Y PLOT
x = [5,7,8,7,2,17,2,9,4,11,12,9,6]
y = [99,86,87,88,111,86,103,87,94,78,77,85,86]

plt.scatter(x, y)
plt.title("2D scatter plot")
plt.show()
```

OUTPUT:**RESULT:**

EX.NO:7b	DEMONSTRATE 3D PLOTTING
DATE:	

AIM:

To write a program to display the dataset in a 3D plot using python.

APPROACH:

Step 1: Open Jupiter notebook, create a new file ex07.

Step 2: Import pyplot from matplotlib.

Step 3: Create 3 coordinates x, y, z and set the start value and size of the points which should be plotted.

Step 4: Create figure of the graph and axes projection using the function figure() and axes().

Step 5: Using scatter3D() function set x, y and z as parameters.

Step 6: Using title() to set the title and show() to display the graph.

Step 7: Execute the program and display the output.

PROGRAM:

```
import matplotlib.pyplot as plt
from mpl_toolkits import mplot3d
import numpy as np
```

```
#3D scatter plot is used for 3d representation and is created by using ax.scatter3D()
```

```
# Creating dataset
```

```
z = np.random.randint(100, size =(50))
```

```
x = np.random.randint(80, size =(50))
```

```
y = np.random.randint(60, size =(50))
```

```
# Creating figure
```

```
fig = plt.figure(figsize = (10, 7))
```

```
ax = plt.axes(projection ="3d")
```

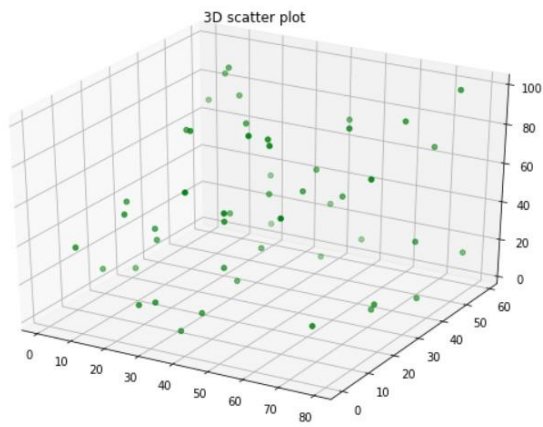
```
# Creating plot
```

```
ax.scatter3D(x, y, z, color = "green")
```

```
plt.title("3D scatter plot")
```

```
plt.show()
```

OUTPUT:



RESULT:

EX.NO:7(c)	EVALUATE USER QUERIES
DATE:	

AIM:

To Write a Python program that queries a given DataFrame based on user-defined conditions and displays the filtered results.

ALGORITHM:**PROGRAM:**

OUTPUT:

CRITERIA	MAX.MARKS	MARKS OBTAINED
AIM & DESCRIPTION	15	
VIVA	10	
TOTAL	25	

RESULT:

EX.NO:8	IMPLEMENT AN ANALYTIC APPLICATION FOR INSTAGRAM TO DEMONSTRATE THE NUMBER OF LIKES, EMOTIONS.
DATE:	

AIM:

To write a program to implement an analytic application for instagram to demonstrate the number of likes, emotions using python.

APPROACH:

- Step 1: Open Jupiter notebook, create a new file ex08.
- Step 2: Import all necessary libraries to perform data analysis and visualization.
- Step 3: Load the dataset into the current directory and read using read_csv().
- Step 4: Group the associated attributes of the dataset and create a chart to get the pictorial representation of the data.
- Step 5: Analyze the dataset using linear model and get the distribution of values over the attributes.
- Step 6: Display a graphical representation of the distribution acquired from the analysis.
- Step 7: Execute the program and display the output.

PROGRAM:

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import plotly.express as px
from wordcloud import WordCloud, STOPWORDS, ImageColorGenerator
from sklearn.model_selection import train_test_split
from sklearn.linear_model import PassiveAggressiveRegressor

data = pd.read_csv(r"Instagram.csv", encoding = 'latin1')
print(data.head())

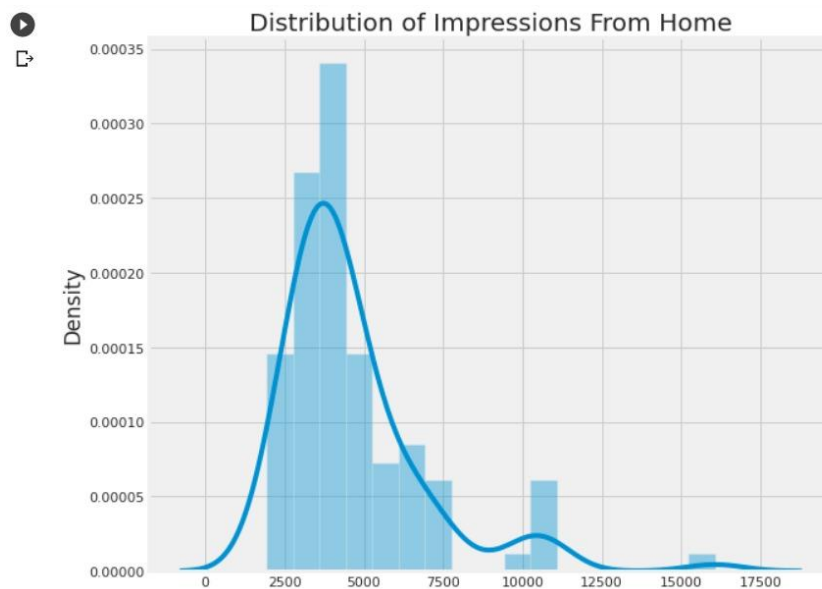
plt.figure(figsize=(10, 8))
plt.style.use('fivethirtyeight')
plt.title("Distribution of Impressions From Home")
sns.distplot(data['From Home'])
plt.show()
plt.figure(figsize=(10, 8))
plt.title("Distribution of Impressions From Hashtags")
sns.distplot(data['From Hashtags'])
plt.show()
plt.figure(figsize=(10, 8))
plt.title("Distribution of Impressions From Explore")
sns.distplot(data['From Explore'])
plt.show()
home = data["From Home"].sum()
hashtags = data["From Hashtags"].sum()
explore = data["From Explore"].sum()
```

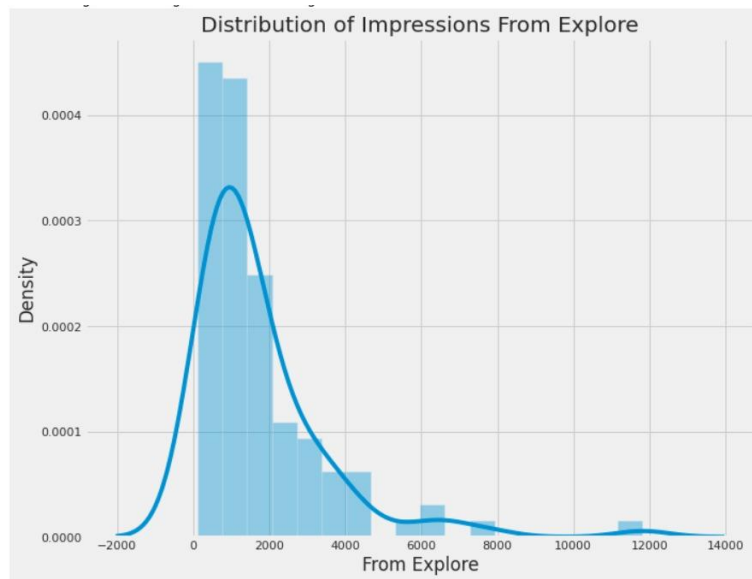
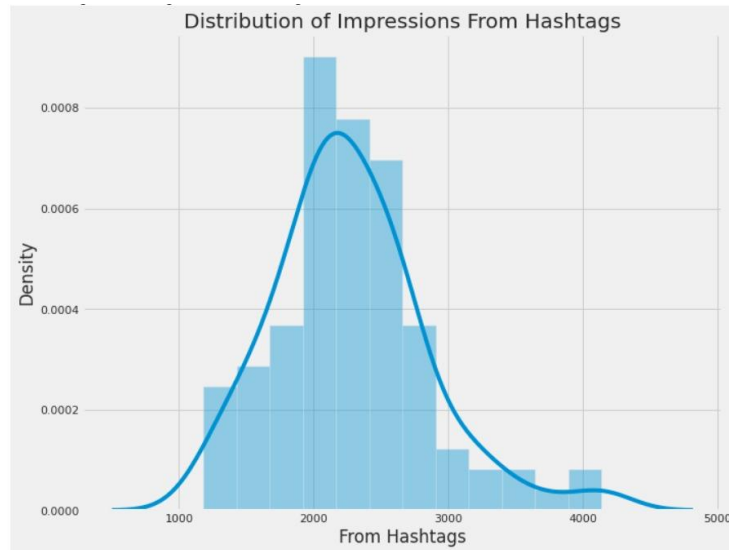
```
other = data["From Other"].sum()

labels = ['From Home', 'From Hashtags', 'From Explore', 'Other']
values = [home, hashtags, explore, other]

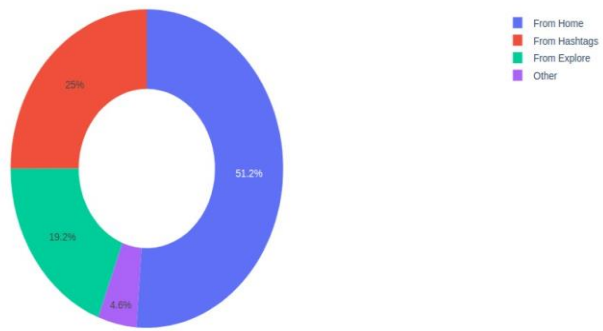
fig = px.pie(data, values=values, names=labels,
              title='Impressions on Instagram Posts From Various Sources', hole=0.5)
fig.show()
```

OUTPUT:





Impressions on Instagram Posts From Various Sources



RESULT:

EX.NO:8(a)	DATAFRAME EXTRACTION
DATE:	

AIM:

To Write a Python program to read a dataset, extract two numeric columns, and create a scatter plot to visualize their relationship.

ALGORITHM:**PROGRAM:****OUTPUT:**

CRITERIA	MAX.MARKS	MARKS OBTAINED
AIM & DESCRIPTION	15	
VIVA	10	
TOTAL	25	

RESULT:

EX.NO:9	IMPLEMENT AN ANALYTIC APPLICATION FOR TWITTER TO DEMONSTRATE SENTIMENT ANALYSIS AND ENTITY RECOGNITION.
DATE:	

AIM:

To write a program to implement an analytic application for twitter to demonstrate Sentiment Analysis and Entity Recognition using python.

APPROACH:

- Step 1: Open Jupiter notebook, create a new file ex09.
- Step 2: Import all necessary libraries to perform data analysis and visualization.
- Step 3: Load the dataset into the current directory and read using read_csv().
- Step 4: Group the associated attributes of the dataset and create a chart to get the pictorial representation of the data.
- Step 5: Perform data cleaning by altering the non-value added attributes.
- Step 6: Import machine learning model of sentiment analysis.
- Step 7: Perform data processing by re, stemming, vectorization and split the dataset into training and testing.
- Step 8: Get the analytics using sentiment analysis model and gather the total distribution.
- Step 9: Display a graphical representation of the distribution acquired from the analysis.
- Step 10: Execute the program and display the output.

PROGRAM:

```
import numpy as np
import pandas as pd
import time
import re

from nltk.corpus import stopwords
from nltk.tokenize import sent_tokenize, word_tokenize
from sklearn.feature_extraction.text import CountVectorizer
from sklearn.feature_extraction.text import TfidfVectorizer

from nltk.stem.porter import PorterStemmer
from nltk.stem import WordNetLemmatizer

#preprocessing
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import LabelEncoder

# machine learning

from sklearn.naive_bayes import GaussianNB
from sklearn.naive_bayes import MultinomialNB
from sklearn.tree import DecisionTreeClassifier
from sklearn.ensemble import RandomForestClassifier
from sklearn.linear_model import LogisticRegression
from sklearn.svm import LinearSVC
```

```

from sklearn.metrics import classification_report

import seaborn as sns
sns.set(style='whitegrid')
pd.set_option('display.max_columns',None)

headers=['Tweet_ID','Entity','Sentiment','Tweet_content']

train_df=pd.read_csv('twitter_training.csv', sep=',', names=headers)

valid_df=pd.read_csv('twitter_validation.csv', sep=',', names=headers)

train_df['Sentiment'].value_counts()
sns.countplot(x=train_df['Sentiment'])

train_df= train_df.drop_duplicates()

train_df.Sentiment.value_counts().plot(kind='pie',
autopct='%1.0f%%',figsize=(5,5),colors=["red", "yellow", "green", 'blue'])

Twitter_sentiment = train_df.groupby(['Entity', 'Sentiment']).Sentiment.count().unstack()
Twitter_sentiment.plot(kind='bar',figsize=(20,20))

# SENTIMENTAL ANALYSIS #

# encoder for target feature
from sklearn import preprocessing
lb = preprocessing.LabelEncoder()
train_df['Sentiment']=lb.fit_transform(train_df['Sentiment'])

train_df.dropna(axis=0, inplace=True)
tweet_train = train_df["Tweet_content"]
tweet_valid=valid_df["Tweet_content"]
target=train_df['Sentiment']

# Step (1): Remove Additional Letter such as
REPLACE_WITH_SPACE = re.compile("(@)")
SPACE = " "

def preprocess_reviews(reviews):
    reviews = [REPLACE_WITH_SPACE.sub(SPACE, line.lower()) for line in reviews]

    return reviews

reviews_train_clean = preprocess_reviews(tweet_train)
reviews_valid_clean = preprocess_reviews(tweet_valid)

# Step (2): Remove Stop Words

```

```

english_stop_words = stopwords.words('english')
def remove_stop_words(corpus):
    removed_stop_words = []
    for review in corpus:
        removed_stop_words.append(' '.join([word for word in review.split() if word not in
english_stop_words]))
    return removed_stop_words

no_stop_words_train = remove_stop_words(reviews_train_clean)
no_stop_words_valid = remove_stop_words(reviews_valid_clean)

# Step(3) : Stemming
def get_stemmed_text(corpus):
    stemmer = PorterStemmer()

    return [' '.join([stemmer.stem(word) for word in review.split()]) for review in corpus]

stemmed_reviews_train = get_stemmed_text(no_stop_words_train)
stemmed_reviews_test = get_stemmed_text(no_stop_words_valid)

# Step(4) : TF-IDF
tfidf_vectorizer = TfidfVectorizer()
tfidf_vectorizer.fit(stemmed_reviews_train)
X = tfidf_vectorizer.transform(stemmed_reviews_train)
X_test = tfidf_vectorizer.transform(stemmed_reviews_test)

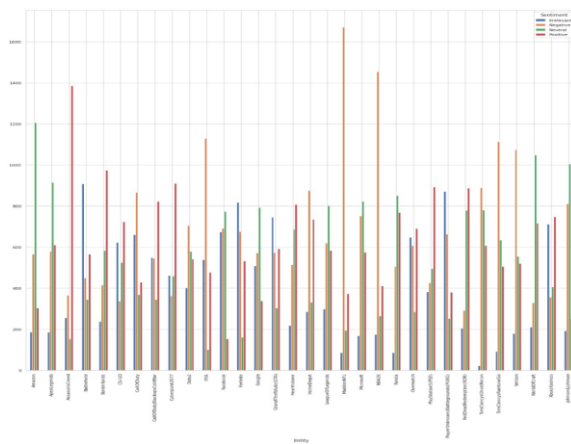
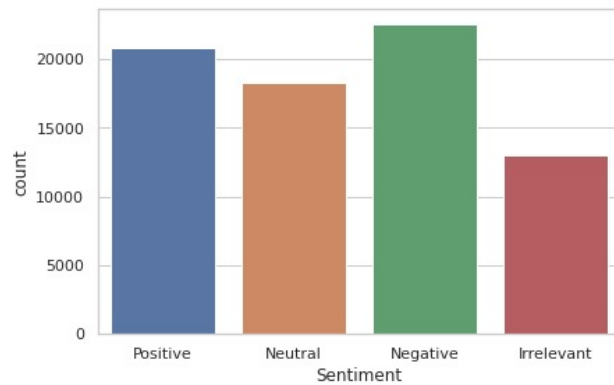
# Step(5) : Splitting Data
X_train, X_val, y_train, y_val = train_test_split(X, target, train_size = 0.75)

# Step(6) : Machine Learning Model
text_classifier = RandomForestClassifier(n_estimators=500, random_state=0)
text_classifier.fit(X_train, y_train)

y_pred=text_classifier.predict(X_val)
print(classification_report(y_val,y_pred))

```

OUTPUT:



	precision	recall	f1-score	support
0	0.96	0.88	0.92	723
1	0.92	0.94	0.93	1127
2	0.97	0.89	0.93	867
3	0.87	0.95	0.91	1071
accuracy			0.92	3788
macro avg	0.93	0.92	0.92	3788
weighted avg	0.92	0.92	0.92	3788

RESULT:

EX.NO:9(a)	3D PLOTTING
DATE:	

AIM:

To Write a Python program to create a 3D plot using a dataset with three numeric columns, demonstrating a three-dimensional visualization.

ALGORITHM:**PROGRAM:**

OUTPUT:

CRITERIA	MAX.MARKS	MARKS OBTAINED
AIM & DESCRIPTION	15	
VIVA	10	
TOTAL	25	

RESULT:

EX.NO:10	SOCIAL MEDIA ANALYSIS
DATE:	

AIM:

To Write a Python program that collects and analyzes Instagram data from a given csv, including the number of likes and emotions, then displays the results.

ALGORITHM:**PROGRAM:**



OUTPUT:

CRITERIA	MAX.MARKS	MARKS OBTAINED
AIM & DESCRIPTION	15	
VIVA	10	
TOTAL	25	

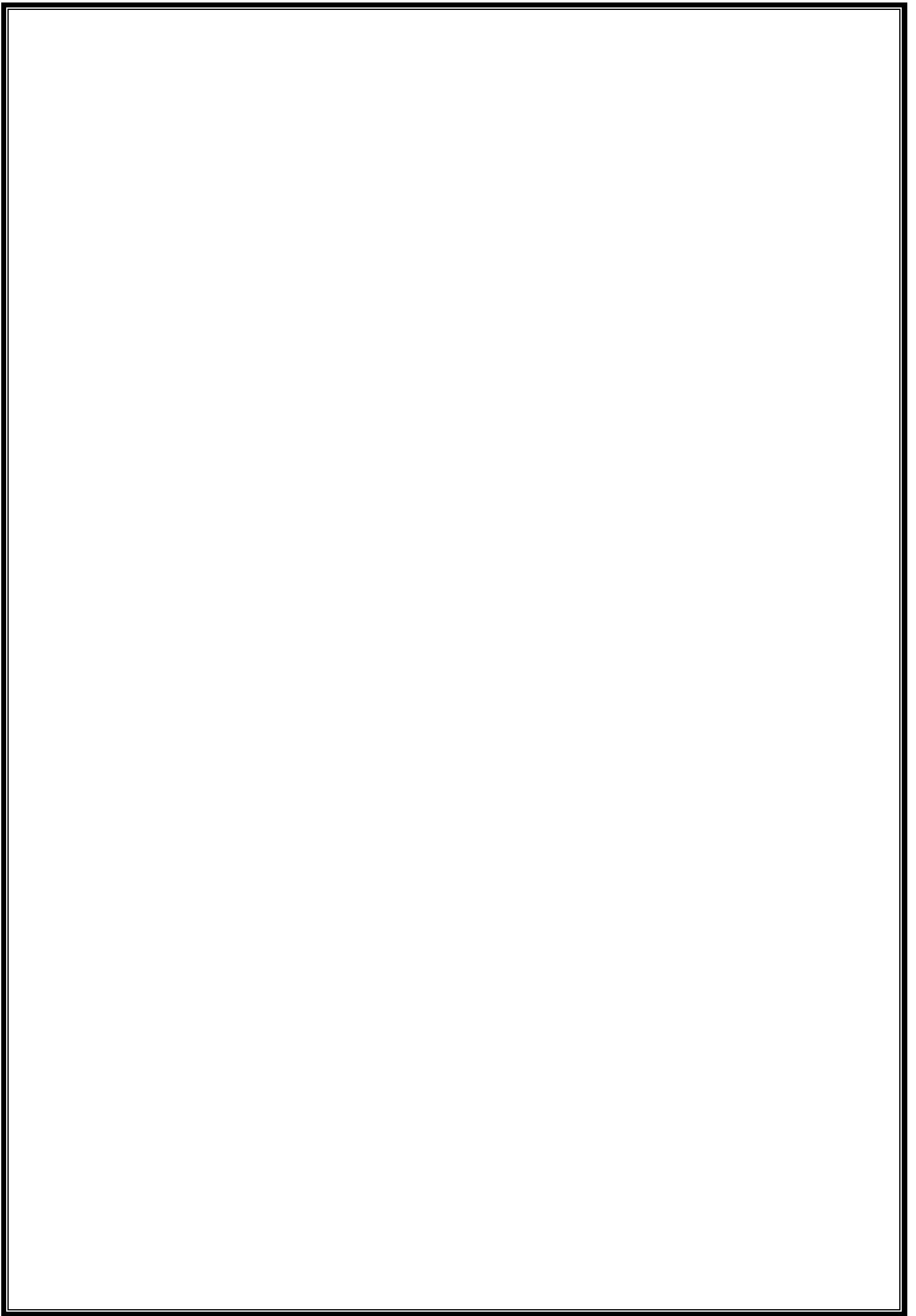
RESULT:

EX.NO:11	SENTIMENT ANALYSIS
DATE:	

AIM:

To Write a Python program that performs sentiment analysis and entity recognition on a set of Twitter data from a given csv, then displays the sentiment scores.

ALGORITHM:**PROGRAM:**



OUTPUT:

CRITERIA	MAX.MARKS	MARKS OBTAINED
AIM & DESCRIPTION	15	
VIVA	10	
TOTAL	25	

RESULT: