



School of Computing
Department of Computer Science & Engineering
(Artificial Intelligence and Machine Learning)

ACADEMIC YEAR 2025-26 (SUMMER SEMESTER)

LAB RECORD NOTEBOOK

10212CA214 - DATA VISUALIZATION

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VTU.NO: 94958

REG.NO: 93V ECL 0055

BRANCH: CSE (AIML)

YEAR/SEM: 3rd | 5th

SLOT: 812 L



School of Computing
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BONAFIDE CERTIFICATE

NAME : Shaik Mohammed Shahid

BRANCH : CSE(AIML)

VTU NO. : 24858

REG.NO. : 93UEC10055

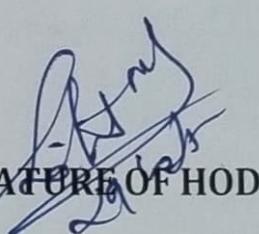
YEAR/SEM : 3rd / 5th

SLOT NO. : S₁₂ L₁

Certified that this is a bonafide record of work done by above student in the "**10212CA214 - DATA VISUALIZATION LABORATORY**" during the year 2025-2026 (Summer Semester).

R.T. Thakur
29/10/25

SIGNATURE OF LAB HANDLING FACULTY


Dr. A. S. Jayaraman

SIGNATURE OF HOD

Submitted for the Semester Practical Examination held on **03.11.25** at
Vel Tech Rangarajan Dr.Sagunthala R&D Institute of Science and Technology.

INTERNAL EXAMINER

EXTERNAL EXAMINER

INDEX

NAME: Mohammed Shahid

ROLLNO.: _____

STD.: _____

DIV./ SEC.: _____

SUBJECT: _____

INDEX

Date	Title	Page No.	Marks	Faculty Signature
21/07/25	Exploration of Data Visualization Tools like Tableau, Python libraries, D3.js	1	16	Dr 21/7
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04/08/25	To visualize and perform Bivariate analysis using continuous and categorical data	9	16	Dr 4/8
11/08/25	To visualize and perform Multivariate analysis using Multiple variables involving Multiple measures	12	16	Dr 11/8
18/08/25	To design and perform visualization for Trees	15	16	Dr 18/8
25/08/25	To design and perform visualization for Graphs and Networks	19	16	Dr 25/8
08/09/25	To generate insight using Text Network Analysis and Visualization	22	17	Dr 8/9
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13/10/25	Use Case: Earthquake and Geospatial Data Analysis	32	17	Dr 13/10
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Completed

Total Marks: 184/220

R.T. T. Sarker
Signature of Faculty

The oriented Data Analysis
to identify Systemic.

Mango Plant Disease Visualization

Exploration of Data Visualization Tools - Tableau / Python/G

- * connecting Data Set

- * Preparation of Data

Aim: To Visualize the distribution of different mango Plant disease and Explore their metadata

Using Python or other visualization tools.

Algorithm:

① Import Dataset:

load the dataset containing folders or records of mango Plant images classified by disease condition.

② Extract metadata:

- * List all image files from each disease class folder. E.g:- Anthracnose, Sooty mold.

- * Create a structured table with attributes like, image name, class label

Data set Table for mango plants:

Plant name	Type	height (cm)	Sunlight	water Needs	Health (color)
MangoTree1	fruit plant	150	high	Medium	Antrac nose
MangoTree2	fruit plant	140	high	high	Sooty mold
MangoTree3	fruit plant	130	Medium	Medium	Powdery mildew
MangoTree4	fruit plant	160	high	low	Healthy
MangoTree5	fruit plant	155	Medium	high	Blebaks
Tree 6	fruit plant	145	low	Medium	Black spot
Tree 7	fruit plant	135	high	Medium	Sooty mold
Tree 8	fruit plant	125	Medium	low	Healthy
Tree 9	fruit plant	165	high	high	Powdery mildew
Tree 10	fruit plant	120	low	Medium	Antrac

Preprocess and clean Data:

Ensure only valid image formats are used
(Jpg, png)

Create Summary Table:

Count the number of images in each class.

Display Sample records in tabular format

Visualization (optional):

Use bar charts to show the number of images

Per disease class using Python

Program:

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

VEL TECH	
EX No.	15
PERFORMANCE (5)	5
RESULT AND ANALYSIS (5)	5
VIVA VOCE (5)	2
RECORD (5)	4
TOTAL (20)	14
SIGN WITH DATE	①

Result: Thus, to visualization the distribution of different mango plant disease and explore their meta data using Python or other visualization tools was executed successfully.

Aim: To Visualize and Perform Univariate analysis Using continuous and categorical data - Bar chart, Pie chart.

Algorithm:

1. Choose Data sets:

- * Select dataset with categorical and continuous attributes.

2. Identify Data types:

- * Distinguish between categorical and continuous attributes.

3. construct Scatter, lines, strip, swarm plots:-

- * Utilize appropriate plots for continuous data analysis considering relationships, trends, & distributions.

4. Interpret visualization:

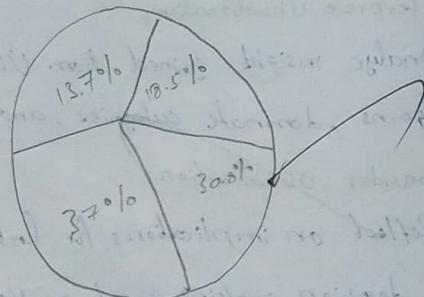
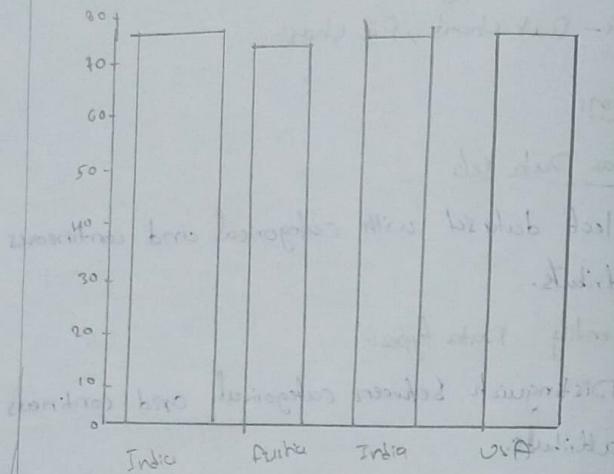
- * Analyze insights gained from visualization, identify patterns, dominate categories, and correlations.

5. consider visualization:

- * Reflect on implications for further exploration or decision-making based on the analysis.

6. Document findings:

- * Summarize finding and observations for reporting or presentation purposes.



Programs

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
df = pd.read_csv("/content/sentimentdata.csv")
platform_likes_top5 = df.groupby('Platform')
    ('Likes').sum().sort_values(ascending=False).head(5)
```

```
plt.figure(figsize=(6,6))
```

```
platform_likes_top5.plot(kind='pie', autopct='%1.1f %%', startangle=140, colors=['skyblue'])
```

```
plt.title('Top 5 Platforms by Total Likes')
```

```
plt.ylabel()
```

```
plt.show()
```

```
top5_countries = df.nlargest(5, 'Likes')
```

Create a choropleth chart

```
plt.figure(figsize=(10,6))
```

```
plt.bar(top5_countries['country'], top5_countries['Likes'], color=['skyblue', 'lightcoral', 'lightgreen', 'lightsalmon', 'lightsteelblue']).
```

```
plt.ylabel('Total Likes')
```

```
plt.xlabel('country')
```

```
plt.title('Top 5 countries by total Likes')
```

```
plt.show).
```

data = h

'Text': ['enjoying a beautiful day at the park!']

'Retweets': [5, 8, 2, 20, 8, 12],

'Likes': [30, 10, 40, 15, 25]

3

df = pd. Data Frame(data)

sns.set(style = "whitegrid")

plt.figure(figsize=(10, 6))

sns.scatterplot(x = 'Retweets', y = 'Likes',
data = df, color = 'blue', alpha = 0.7)

Total = 256

Aim: To Perform Univariate analysis on a dataset containing emp details by identifying categorical and continuous data attributes.

Algorithm:

1. Start

2. Import necessary Python libraries:

Pandas, and matplotlib.pyplot

3. Load the dataset from a csv file.

4. Categorical Data Analysis

5. Add appropriate titles and axis labels

to each plot.

6. End.

Program:

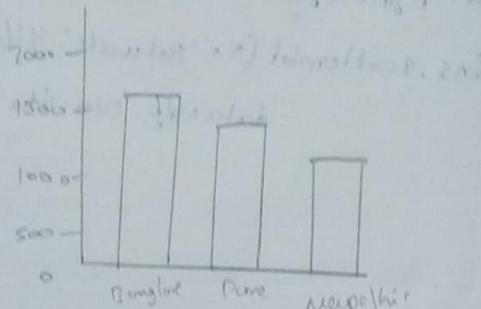
```
import Pandas as pd
```

```
import matplotlib.pyplot as plt
```

```
file_Path = 'Employee.csv'
```

```
df = pd.read_csv(file_Path)
```

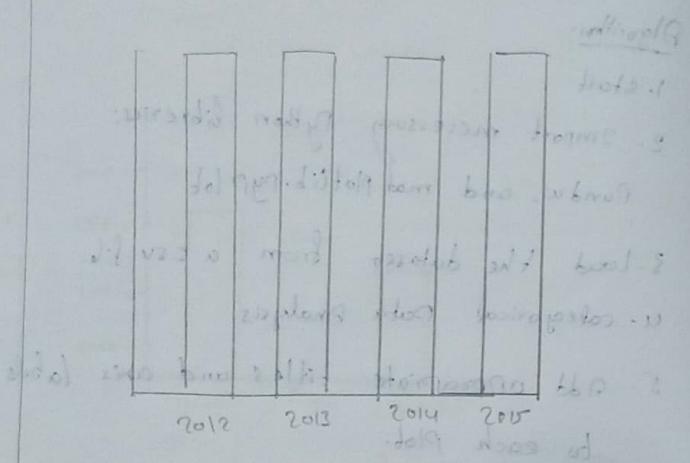
```
categorical_column = 'Gender'
```



OKX



↳ how to explore categorical variables of multiple geographical states over time
↳ understand what maximum how



↳ how to explore frequency
↳ how to explore distribution
↳ (20.20.30.40) = discrete
↳ (20.20.30) = continuous
↳ understand distribution

create a pie chart
plt.pie (category_counts, labels=category_counts)
plt.title ('Pie chart of h categorical-column')
plt.show()
import pandas as pd
import matplotlib.pyplot as plt
file_path = 'Employee.csv'
df = pd.read_csv(file_path)
categorical_column = 'city'
import pandas as pd
import matplotlib.pyplot as plt
continuous_column = 'Age'
plt.hist (df[continuous_column], bins=10,
bins=bins
plt.title ('{continuous_column} Distribution')
plt.xlabel (continuous_column)
plt.show()
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
file_path = 'Employee.csv'
df = pd.read_csv(file_path)
continuous_column = 'joining year'

```

sns.rugplot(dt[continuous_column], height=0.5)
plt.title(f'Rugplot of {continuous_column}')
pltx.table(continuous_column)
plt.show()

```

(1) 28/11

VELTECH	
EX No.	2
PERFORMANCE (5)	5
RESULT AND ANALYSIS (5)	5
VIVA VOCE (5)	
RECORD (5)	12+4=16
TOTAL (20)	
SIGN WITH DATE	(@)

Result: Thus visualizing and performing univariate analysis using continuous and categorical data - Bar chart, Pie chart done successfully.

Task-3

4/8/25

To Visualize and Perform Bivariate analysis
Using continuous and categorical data categorical
vs. categorical stacked Bar chart, Grouped Bar
chart, Segment Bar chart, Mosaic plots.

Aim: To Perform Bivariate Analysis on the
student Alcohol consumption dataset by
visualizing relationships b/w categorical and
continuous variations using various statistical plots.

Algorithm:

- 1) Select Dataset choose a dataset containing both categorical and continuous variables.
- 2) Diferentiative variable: Identify categorical and variables
- 3) continuous vs continuous: Generate scatter plots with fit lines to explore.
- 4) categorical vs continuous: construct barchart for ~~summarize~~ ~~statisticas~~ statistics.
- 5) Interpretation: Analyze visualization for insight into relationship and patterns, drawing conclusions for further analysis or decision making.

Program!

```
import pandas as pd
```

```
import matplotlib.pyplot as plt
```

```
df = pd.read_csv('laptop_clean_dataset.csv')
```

```
industry = df['Industry']
```

```
income = df['Income']
```

```
plt.figure(figsize=(10, 6))
```

```
plt.bar(industry, income, color='skyblue')
```

```
plt.xlabel('Industry')
```

```
plt.ylabel('Income')
```

```
plt.show()
```

```
import pandas as pd
```

```
import seaborn as sns
```

```
import matplotlib.pyplot as plt
```

```
df = pd.read_csv('laptop_clean_dataset.csv')
```

```
categorical_column = 'Ethnicity'
```

```
plt.xlabel(categorical_column)
```

```
plt.ylabel('continuous column')
```

```
plt.show()
```

```
import pandas as pd
```

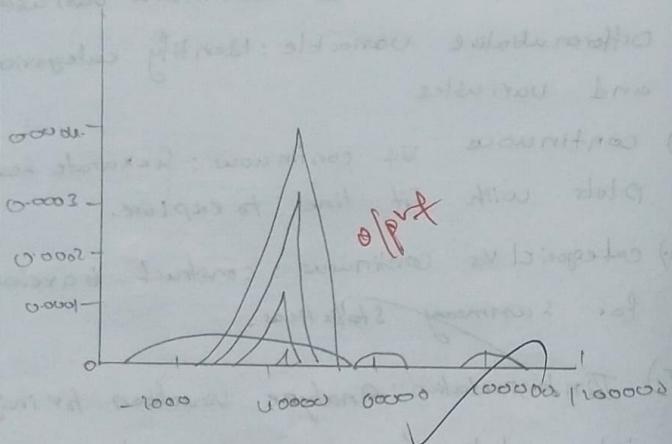
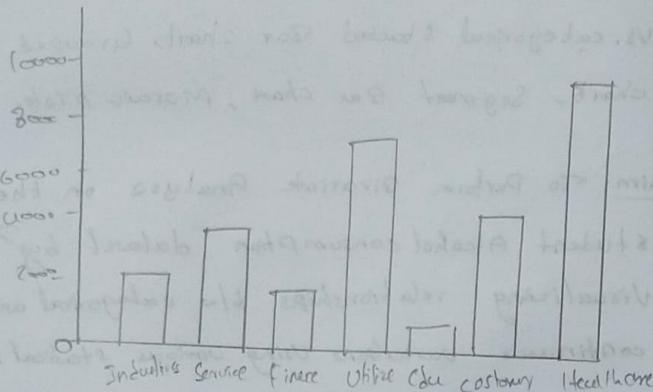
```
import seaborn as sns
```

```
import matplotlib.pyplot as plt
```

```
df = pd.read_csv('laptop_clean_dataset.csv')
```

```
categorical_column = 'Ethnicity'
```

```
continuous_column = 'Income'
```



plt.show()

```
import pandas as pd
```

```
from JoyDay import Joyplot.
```

```
import matplotlib.pyplot as plt
```

```
df = pd.read_csv('counstcat/clean_dataset.csv')
```

```
categorical_column = 'Industry'.
```

```
continuous_column = 'Income'
```

```
import pandas as pd
```

```
import seaborn as sns
```

```
import matplotlib.pyplot as plt.
```

```
df = pd.read_csv('content/PY tasks.csv')
```

```
plt.title('scatter.plot with Fit Line')
```

```
plt.xlabel('Age')
```

```
plt.ylabel('Income')
```

```
plt.show()
```

VEL TECH	
EX No.	3
PERFORMANCE (5)	5
RESULT AND ANALYSIS (5)	5
VIVA VOCE (5)	5
RECORD (5)	7/4
TOTAL (20)	12+4=16
SIGN WITH DATE	12/12/2023

AT8
Result: To Visualize and Perform

Bivariate Analysis Using continuous and categorical data is successfully implemented.

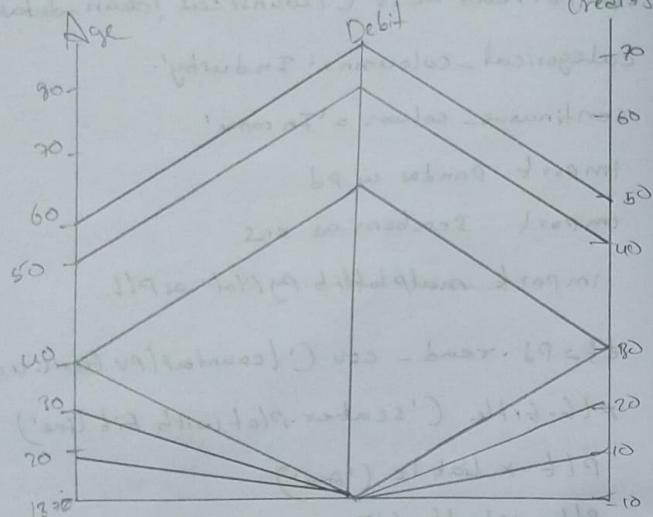
output

base salary, dream

target target profit, need

base salary of different dreams

(which model, technique?) your base - b (credit score)



Age Tech	
10	EX-AN
20	EX-AN
30	EX-AN
40	EX-AN
50	EX-AN
60	EX-AN
70	EX-AN
80	EX-AN
90	EX-AN

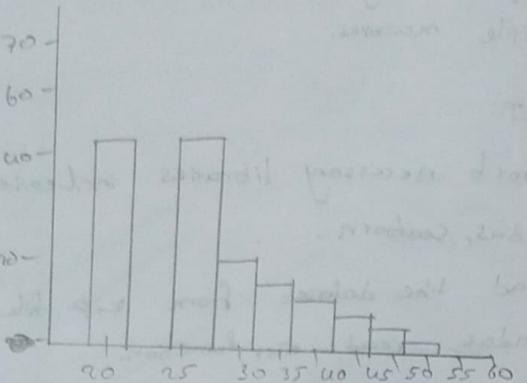
obj

model and solution or illus
are various part equivalent statements
between different & what happens

Aim: To visualize and perform multivariate analysis using multiple variable involving multiple measures.

Algorithm:

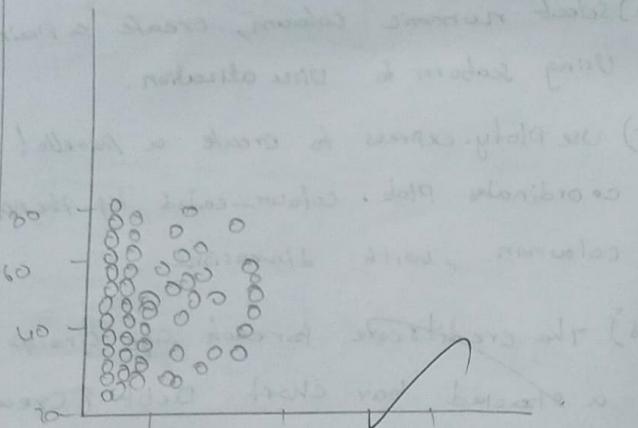
- 1) Import necessary libraries including Numpy, Pandas, Seaborn.
- 2) Load the dataset from zip file using Pandas, read_csv function.
- 3) Print the loaded dataset to inspect its structure.
- 4) Select numeric columns, create a DataFrame. Using seaborn to visualization.
- 5) Use Plotly Express to create a parallel coordinates Plot, colour-coded by the Approval column, with dimensions.
- 6) The credit score for each Age group create a stacked bar chart Debt & Create score for each group & display the chart.



Program:

```

import numpy as np
import pandas as pd
df = pd.read_csv("laptop_fair.csv")
print(df)
import seaborn as sns
import matplotlib.pyplot as plt
numeric = subset = df[['Age', 'Debt', 'Years Employed']]
plt.show()
import pandas as pd
import plotly.express as px
attributes = ["Age", "Debt", "credit_score"]
fig = px.parallel_coordinates(numeric, color = "Approved", title = "Approved", show_colorbar = True)
plt.plot("Age", "Dept.", color = "Dept.colour" = "blue")
plt.xlabel("Age")
plt.ylabel("Dept.value")
plt.title("Line Graph", Age, Debt, credit_score)
plt.legend()
plt.grid(True)
plt.show().
    
```



grouped = data.plot (kind = 'bar', stacked = True, $\alpha(0.2)$)

plt.xlabel('Age')

plt.ylabel('Value')

plt.title('Stacked Bar chart : Debt & Creditor
by Age')

plt.legend(title = ('Attribute'))

plt.show()

VEL TECH	
EX No.	H
PERFORMANCE (5)	5
RESULT AND ANALYSIS (5)	5
VIVA VOCE (5)	2
RECORD (5)	A
TOTAL (15)	12 $\frac{A}{H}$
SIGN WITH DATE	Q

Result: To visualize & Perform multiple analysis Using multing variable involving multiple measures done successfully.

Aim: construct a tree map display on a real-world dataset.

Algorithm:

- 1.) Install the Sque Package using Pipify
- 2) Import necessary libraries such as Pandas, matplotlib.pyplot & squarify.
- 3) Read the data set from `content USA.csv`
- 4) Group the Data Frame df
- 5) Create the figure size 12×12 inches for the tree map
- 6) Use `squarify.plot()` to create tree map plot.
- 7) Set transparency with `alpha = 0.4` for visualization.
- 8) Turn off the axis
- 9) Display plot using `plt.show()`.

out per

			Nissan
Cadillac	Ford		Mercedes
		Porsche	
brown	black	tan	blue
white	white	grey	green
black	black	black	black
orange	chevrolet	gmc	

Program!

Pip install squarify

import Pandas as pd

```
import numpy as np
```

```
import matplotlib.pyplot as plt
```

import seaborn as sns

import squaring.

```
df = pd.read_csv('C:/constUSA - cars - data  
setL.csv')
```

```
print(df)
```

brand-total price = df + group by ("brand")

```
plt.figure(figsize=(12,12))
```

plt-axis (off)

```
plt.title('Tree map of total price by brand!')
```

plt.show() .

TASK-5b

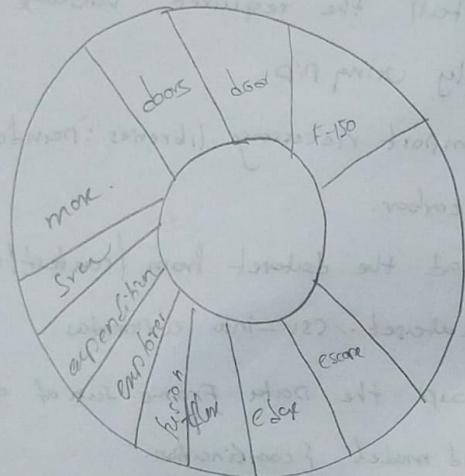
Aim: Build a sunburst display Using above Program dataset.

Algorithm:

- 1.) Install the required package 'squarify'{
Plotly using pip}
- 2.) Import necessary libraries : 'pandas'; 'numpy';
'seaborn'.
- 3.) Read the dataset from /content/usa-cars-
dataset.csv - into a pandas
- 4.) Group the Data Frame sum of df brand
and model {combination}.
- 5.) Display the plot using fig.show();



output:



Program:

Pip install squarify

Pip install plotly

```
import pandas as pd
```

```
import numpy as np
```

```
import matplotlib.pyplot as plt
```

```
import seaborn as sns
```

import security.

impor. Plotly express as P_x

```
df = pd.read_csv('content/usa-care  
-dataset.csv')
```

Print(df)

sun burst - data = df . group by (brand'model)

`sun(b, reset - index())`

Fig. shows

VEL TECH	
EX No.	5
PERFORMANCE (5)	5
RESULT AND ANALYSIS (5)	2
VIVA VOCE (5)	3
RECORD (5)	3
TOTAL (20)	15 + 3 = 18
SIGN WITH DATE	18

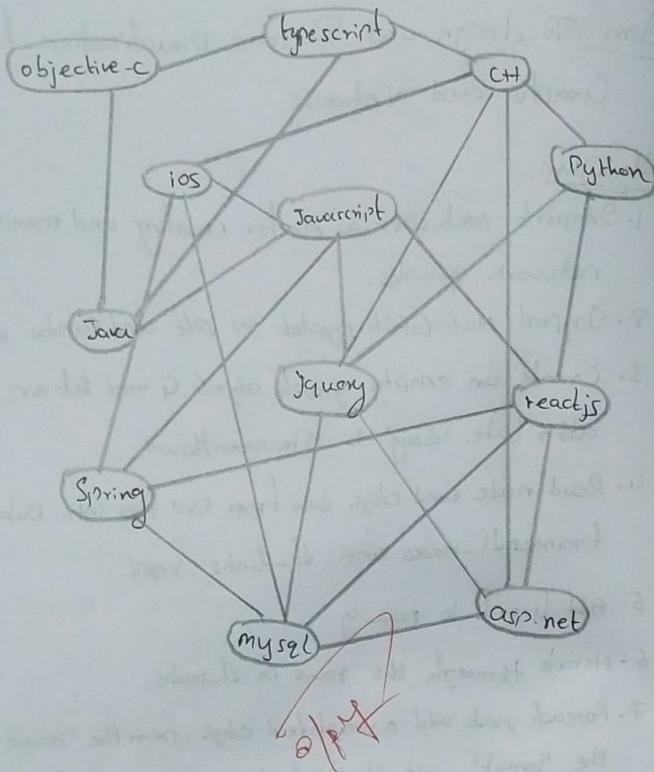
Result: To design & perform visualizations for trees is successfully done.

Aim: To design and Perform visualization for Graphs and Networks.

Algorithm:

1. Import Networkx as nx for creating and manipulating network graphs.
2. Import Matplotlib.pyplot as plt and Pandas as pd.
3. Create an empty graph object G and set an attribute 'day' to "Stackoverflow".
4. Read node and edge data from csv files into Data frames df-nodes and df-links , resp.
5. Add Graph to node g.
6. Iterate through the rows in df-nodes.
7. For each node add a weighted edge from the 'source' to the 'target' with the 'value' as the weight.
8. Iterate through the rows in df-links.
9. Define visualization options like edge color, width, node labels, and font weight.
10. Determine colors for nodes based on their 'group' attribute.
11. Set the edge colour to a specific color.
12. Show the graph using plt.show()
13. End of the algorithm.

output



code:

```

import networkx as nx
import matplotlib.pyplot as plt
import Pandas as pd

g = nx.Graph(day = "stackoverflow")
df_nodes = pd.read_csv('stack-network-nodes.csv')
df_links = pd.read_csv('stack-network-links.csv')
for index, row in df_nodes.iterrows():
    G.add_weighted_edges_from([(row['source'], row['target'],
                                row['value'])])
plt.figure(figsize=(15,15))
option = h edgecolor: "#FFDDEA",
'edge_color': '#FFDDEA',
'width': 1.5,
'with_labels': True,
'font_weight': 'regular',
g
colors = [color_map[G.nodes[node]]["group"] for node
          in G]
sizes = [G.nodes[node]["node_size"]**2.5 for node in G]
nx.draw(G, node_color=colors, node_size=sizes,
        pos=nx.spring_layout(G, k=1.5, iterations=15),
        **options)
ax = plt.gca()
cxt.collections[0].set_edgecolor("#555555")
plt.show()

```

VEL TECH	
EX No.	6
PERFORMANCE (5)	5
RESULT AND ANALYSIS (5)	5
VIVA VOCE (5)	4
RECORD (5)	11
TOTAL (20)	17 + 4 = 18
WITH DATE	10/10/18

Mr
Jest & S

Result: Thus, designing and performing Visualization
for Graphs and Networks done successfully.

Aim: To generate insight from text data using Text network Analysis and Visualization tool Such as Tagcloud and the word cloud Package in Python, by creating word clouds and graphical Visualization to identify frequently occurring terms and their importance.

Algorithm

1. Import necessary libraries.
wordcloud

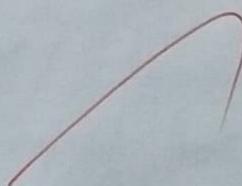
Pandas
matplotlib.pyplot

2. Create a wordcloud object with specific parameters; Set the width and height of the word cloud image. Provide a list of stopword to exclude common words.

3. Display the word cloud image:

Use Alt + rshome to display the word cloud. Turn off the axis labels. Adjust the layout to minimize padding.

4. End of Algorithm.



new dot not much higher storage or imp.
but not always true regular storage tools
cannot handle large file but good to do?
building from scratch from scratch of matplotlib
and covers all

effective library of visualization
Visualization methods all from
Covers
for design
and concern front-end
of graphics
is analysing web
Using developer tools
Various Exploratory Particularity and tool
~~data~~

some tools have off building -
prop. build tools off products of software. No
standard keep all layout stable even off the

graphlib

networkx to build all

Program

from wordcloud import wordcloud
import matplotlib.pyplot as plt
details: "Data visualization is the visual and
interactive exploration and graphic representation
of data of any type, this course covers data
visualization concepts, practice, tools particularly
for analysing and presenting business data!"

```
wordcloud = wordcloud(width=800, height=800,
                      background_color='white',
                      figsize=(5, 5), facecolor='none')
plt.imshow(wordcloud)
plt.axis("off")
plt.tight_layout(pad=0)
plt.show()
```

~~Qn~~

Result: To generate insight from the data using
text network analysis and visualization tool
such as Tagger and the word cloud package
is done successfully.

Task-7b

Aim: To generate the word tree graph that shows how words in a text are connected providing insights into word relationships.

Algorithm:

- * Import text analysis libraries (nltk, network, matplotlib)
- * Load and Prepare the dataset.
- * Select the root word (example "data")
- * Build before tree by finding words that appear before after the root word
- * Create a graph structure using network data visualization
- * Create a tree graph using matplotlib.

Program:

```
import nltk  
import matplotlib.pyplot as plt  
import networkx as nx  
from nltk.corpus import stopwords  
from nltk.tokenize import word_tokenize  
nltk.download('punkt')  
nltk.download('stopwords')
```

text = Data Analysis is import Data Visualization. Clean, on

tokens = word_tokenize(text.lower())
 stop_words = set(stopwords.words('english'))
 bigrams = list(nltk.bigrams(fit.lemmatized_tokens))

read_word = "data"

G = nx.DiGraph()

for w, w2 in bigrams

plt.figure(figsize=(0,

nx.draw_networkx(G, with_labels=True, node_color='skyblue')

plt.title("Word Tree Graph for rootword :

(root-word)", font_size=10)

plt.show()

VELTECH	
EX No.	+
PERFORMANCE (5)	5
RESULT AND ANALYSIS (5)	5
VIVA VOCE (5)	02
RECORD (5)	2
TOTAL (20)	12
MAN WITH DATE	1/1/2018

Result: Thus, the generate right insight using network

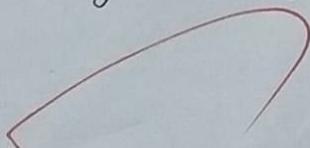
Analysis & visualization utilizing word tree is done & executed successfully.

Aim: To analyse and visualize Tamil Nadu Cities Population data across census years

Using Spatial and geospatial methods, revealing trends, growth patterns, and clusters with Python, QGIS, or Tableau.

Algorithm:

1. Data Preparation: Download Population statistics and spatial boundary data for TN.
2. Spatial Join: Link city/town population data to geographic features align with tabular data.
3. Mapping: Map population values using choropleth maps and plot city/town Point/areas.
4. Spatial Analysis: calculate growth rates for cities from census years.
5. Visualization: Generate maps showing population distribution and growth clusters.



output

Name	Status	District	Pop_1991	2001	Growth	Latitude
Chennai	Municipal Corp	Chennai	3,861,896	6,136,461	839,691	13.082
Coimbatore	MC	Coimbatore	816,821	930,822	214,000	11.078
Madurai	MC	Madurai	477,856	628,869	89,009	9.922
Salem	MC	Salem	490,280	696,760	330,882	11.613
Tiruchirappadi	HAC	Tiruchirappalli	711,104	752,066	25,783	10.716

Program :

```
import pandas as pd
```

```
import geopandas as gpd
```

```
import matplotlib.pyplot as plt
```

```
pop_df = pd.read_csv('cities-in-tamil-nadu-population.csv')
```

```
cities_gdf = gpd.read_file('tamilnad-cities-shapefile.shp')
```

```
merged_gdf = cities_gdf.merge(pop_df, left_on='Name',  
                               right_on='Name')
```

```
merged_gdf[['growth']] = merged_gdf[['2011-03-01']] - merged_gdf[['1991-03-01']]
```

```
merged_gdf.plot(column='growth', cmap='OrRd',  
                 legend=True)
```

```
plt.title('Population of Tamil Nadu Cities (2011 census)')  
plt.show()
```

```
merged_gdf.plot(column='growth', cmap='YlGnBu',  
                 legend=True)
```

```
plt.title('Population Growth of Tamil Nadu Cities  
(1991-2011)')  
plt.show()
```

VEL TECH	
EX No.	8
PERFORMANCE (5)	5
RESULT AND ANALYSIS (5)	5
VIVA VOCE (5)	2
RECORD (5)	3
TOTAL (20)	17
SIGN WITH DATE	15/10/2023

Result: Thus, analyzing and visualizing TN cities population data across census years using statistical and geographical methods, revealing trends, growth patterns, clusters done successfully.

Task:- To analyze and Visualization Time oriented data.

Aim: To analyze and Visualize Time oriented Data Using line graph, trend lines, area chart

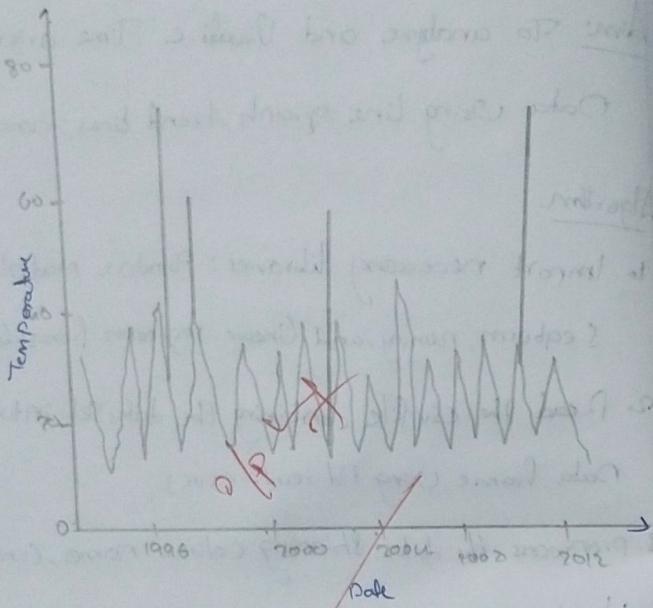
Algorithm:

1. Import necessary libraries: Pandas, matplotlib, Seaborn, numpy, and Linear regression from sklearn.
2. Read the csv file containing the dataset into a Data frame Using pd.read_csv().
3. Preprocess the data. Stripping column names, converting 'datetime-utc' column to datetime format, and setting it as the index.
4. Plot the temperature data over time Using sns.lineplot
5. Customize Plot by adding a title, labels for x & y axis enabling grid lines, rotating x-axis labels plot using plt.show()

Program:

```
import pandas as pd  
import matplotlib.pyplot as plt  
import seaborn as sns  
import numpy as np
```

output



```

from sklearn.linear_model import LinearRegression
data = pd.read_csv('I:/content/festive.csv')
data['datetime-utc'] = pd.to_datetime(data['date-utc'])
plt.figure(figsize=(10,6))
sns.lineplot(data=data, x=data.index, y='tempm')
plt.title('Line Graph of Temperature over Time', data='Temperature')

```

plt.fig.layout()

plt.show()

Trend Lines

```

import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.linear_model import LinearRegression

```

data = pd.read_csv('I:/content/festive.csv')

data['tempm'].fillna(data['tempm'].mean(),
inplace=True).

x = data['index'].values.reshape(-1, 1)

y = data['tempm'].values

model = LinearRegression()

model.fit(x, y)

slope = model.coef_[0]

intercept = model.intercept

plt.scatter(data['index'], data['tempm'], color='blue')

```
label = 'Temperature Data')
```

```
plt.title('Temp Trend Line')
```

```
plt.ylabel('Temp (°C)')
```

```
plt.grid(True)
```

Area chart

```
import pandas as pd
```

```
import matplotlib.pyplot as plt
```

```
data = pd.read_csv('content/test set.csv')
```

```
data['date'] = pd.to_datetime(data['date'], infer_datetime_format=True)
```

```
plt.figure(figsize=(10,6))
```

```
plt.xlabel('Date')
```

```
plt.ylabel('Temp')
```

```
plt.savefig()
```

```
plt.show().
```

VEL TECH	
EX NO.	9
PERFORMANCE (5)	5
RESULT AND ANALYSIS (5)	5
VIVA VOCE (5)	3
RECORD (5)	5
TOTAL (20)	18
SIGN WITH DATE	12/12/2023

Result: 29/12

Result: Thus, we successfully implemented the line graph of trend line, Area chart by Time oriented Data and out is verified.

USE CASE – 1

Earthquake and Geospatial Data Analysis

Batch Mates:

1. VTU24801 JINKALA SHASHIDHAR
2. VTU24802 SHAIK JAKEER HUSSAIN
3. VTU24858 SHAIK MOHAMMED SHAHID
4. VTU24867 KUSUMARAJU BUKARI BABA
5. VTU24905 SHAIK RUBIYA
6. VTU24923 MALLU RUDRA TEJESHWAR REDDY
7. VTU25011 RAAVI NAVADEEP CHOWDARY
8. VTU25104 DONGALA JAYANDAR
9. VTU25106 MADAKKA GARI KRISHNA KOWSHIK
10. VTU25112 SEELAM NAVEEN

Aim:

To analyze and visualize earthquake occurrences using geospatial data.

The goal is to identify regions with the highest seismic activity, visualize earthquake magnitudes on maps,

and observe temporal patterns to support disaster risk assessment and preparedness.

Algorithm:

1. Load the dataset (`earthquake_data.csv`) containing Date, Latitude, Longitude, Magnitude, Depth, Region.
2. Convert **Date** into datetime format and handle missing or invalid entries.
3. Filter earthquakes based on magnitude thresholds (e.g., ≥ 4.0 for moderate and above).
4. Compute:
 - o Average magnitude per region
 - o Total earthquakes per region
 - o Maximum depth and strongest quake recorded
5. Visualize data using:
 - o Geographic Scatter Map (Latitude vs Longitude) color-coded by magnitude
 - o Bar chart showing top earthquake-prone regions
 - o Line plot showing monthly earthquake counts

OUTPUT:

Yearly Average Performance (Descending)

Sales	70.30
Finance	57.68
Marketing	54.05
R&D	49.08
HR	46.83

Best and Worst Month per Department

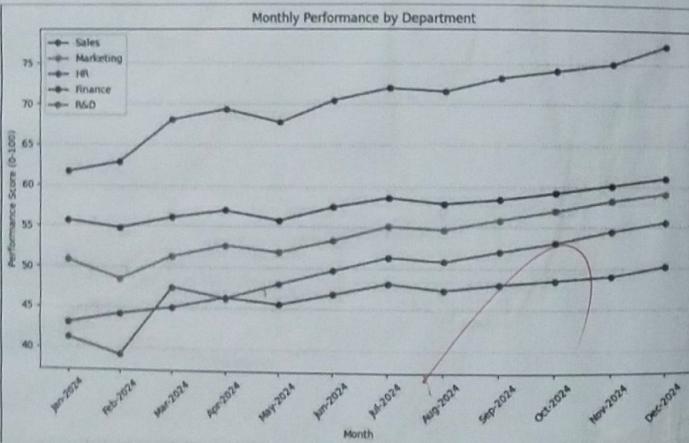
Sales: Best = Dec-2024, Worst = Jan-2024

Marketing: Best = Dec-2024, Worst = Feb-2024

HR: Best = Dec-2024, Worst = Jan-2024

Finance: Best = Dec-2024, Worst = Feb-2024

R&D: Best = Dec-2024, Worst = Feb-2024



6. Generate summary metrics (average magnitude, deepest quake, most active region).
7. Export visualizations and summary results as CSV/PNG files if needed.

Program:

```
import pandas as pd
import matplotlib.pyplot as plt
import plotly.express as px
```

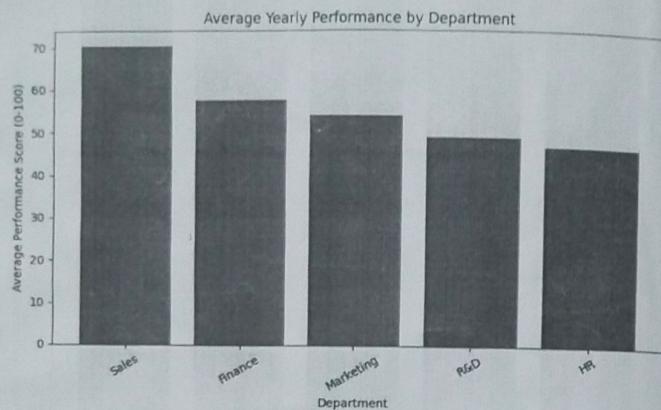
```
df = pd.read_csv("earthquake_data.csv")
```

```
df['Date'] = pd.to_datetime(df['Date'], errors='coerce')
df['Magnitude'] = pd.to_numeric(df['Magnitude'], errors='coerce')
df = df.dropna(subset=['Latitude', 'Longitude', 'Magnitude'])
```

```
avg_mag = df['Magnitude'].mean().round(2)
max_mag = df['Magnitude'].max()
most_active_region = df['Region'].value_counts().idxmax()
```

```
print("\n==== Earthquake Data Summary ===")
print(f"Average Magnitude : {avg_mag}")
print(f"Strongest Earthquake : {max_mag}")
print(f"Most Active Region : {most_active_region}")
```

```
fig = px.scatter_geo(df, lat='Latitude', lon='Longitude', color='Magnitude',
                     title='Global Earthquake Distribution by Magnitude',
                     color_continuous_scale='Reds', size='Magnitude', projection='natural earth')
fig.show()
```



```

region_counts = df['Region'].value_counts().head(10)

plt.figure(figsize=(10,6))
plt.bar(region_counts.index, region_counts.values, color='orange')

```

```

plt.title("Top 10 Most Earthquake-Prone Regions")
plt.xlabel("Region")
plt.ylabel("Number of Earthquakes")
plt.xticks(rotation=45, ha='right')
plt.grid(axis='y', linestyle='--', alpha=0.4)
plt.tight_layout()
plt.show()

```

```

monthly = df.groupby(df['Date'].dt.to_period('M')).size()

monthly.plot(kind='line', figsize=(10,5), color='green')
plt.title("Monthly Earthquake Occurrence Trend")
plt.xlabel("Month")
plt.ylabel("Number of Earthquakes")
plt.grid(True)
plt.tight_layout()
plt.show()

```

VELTECH	
EX. No.	UCI
PERFORMANCE (5)	5
SIMULATION AND ANALYSIS (5)	5
AVOCET (5)	2
3D (5)	5
DATE	13/10/15

By
13/10

Result:

Thus, earthquake occurrences were successfully analyzed and visualized using geospatial data. The analysis revealed regions with the highest seismic activity, identified the strongest earthquakes, and showcased temporal patterns in frequency, enabling better understanding of global earthquake behavior and aiding disaster management planning.

USE CASE – 2

Performance of Different Company Departments Over Year

Batch Mates :

- | | |
|--------------|------------------------------|
| 1. VTU24801 | JINKALA SHASHIDHAR |
| 2. VTU24802 | SHAIK JAKEER HUSSAIN |
| 3. VTU24858 | SHAIK MOHAMMED SHAHID |
| 4. VTU24867 | KUSUMARAJU BUKARI BABA |
| 5. VTU24905 | SHAIK RUBIYA |
| 6. VTU24923 | MALLU RUDRA TEJESHWAR REDDY |
| 7. VTU25011 | RAAVI NAVADEEP CHOWDARY |
| 8. VTU25104 | DONGALA JAYANDAR |
| 9. VTU25106 | MADAKKA GARI KRISHNA KOWSHIK |
| 10. VTU25112 | SEELAM NAVEEN |

Aim:

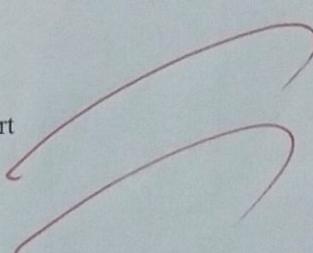
To analyze and visualize yearly department performance using time-series plots and summary metrics to identify trends, best/worst months, and overall strengths or weaknesses.

Algorithm: 1. Define departments and the time dimension (months for the year).

2. Acquire or generate monthly performance metrics for each department (normalized scores, KPIs, etc.).
3. Build a tidy DataFrame: rows = months, columns = departments.
4. Compute derived metrics: yearly average per department, month-over-month changes, best/worst months.
5. Visualize:
 - Line chart: performance over months (one line per department).
 - Bar chart: average yearly performance per department.
6. Export charts and data for reporting (PNG/CSV/PDF).

Program:

```
import numpy as np import
pandas as pd import
matplotlib.pyplot as plt
```



Output:

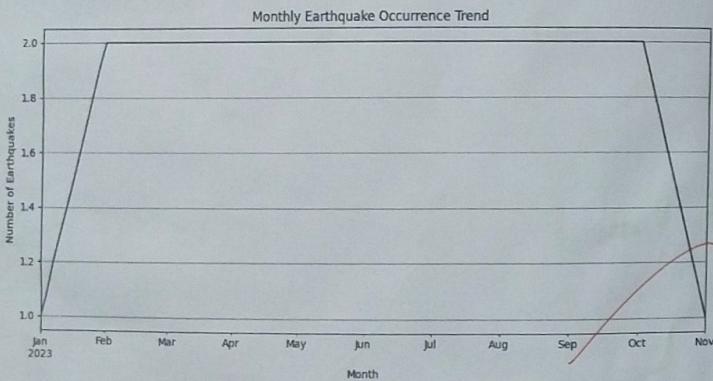
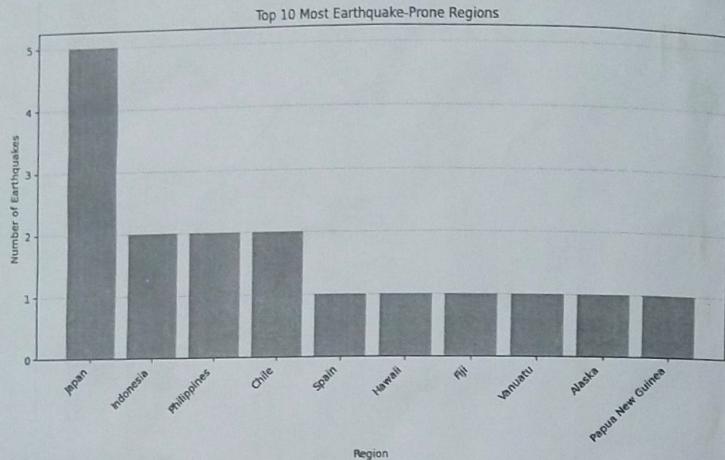
== Earthquake Data Summary ==

Average Magnitude : 5.98

Strongest Earthquake : 8.3

Most Active Region : Japan

4.555.566.577.58MagnitudeGlobal Earthquake Distribution by Magnitude



```
df=pd.read_csv("/content/drive/MyDrive/ColabNotebooks/department_performance_2024.csv")
df.set_index('Month', inplace=True)
yearly_avg = df.mean().round(2).sort_values(ascending=False)

best_month = df.idxmax()
worst_month = df.idxmin()
monthly_changes = df.diff().round(2)

print("\n==== Yearly Average Performance (Descending) ====")
print(yearly_avg)
print("==== Best and Worst Month per Department ====")
for dept in df.columns:
    print(f'{dept}: Best = {best_month[dept]}, Worst = {worst_month[dept]}')  
37
```

```
plt.figure(figsize=(10,6))
for dept in df.columns:
    plt.plot(df.index, df[dept], marker='o', linewidth=2, label=dept)
plt.title("Monthly Performance by Department")
plt.xlabel("Month") plt.ylabel("Performance Score (0-100)")
plt.xticks(rotation=45) plt.grid(axis='y', linestyle='--', alpha=0.4)
plt.legend() plt.tight_layout() plt.show()
```

```
plt.figure(figsize=(8,5))
plt.bar(yearly_avg.index, yearly_avg.values)
plt.title("Average Yearly Performance by Department")
plt.xlabel("Department") plt.ylabel("Average Performance Score (0-100)")
plt.xticks(rotation=30)
plt.tight_layout() plt.show()
```

VEL TECH	
EX No.	UC 2
PERFORMANCE (5)	5
RESULT AND ANALYSIS (5)	5
VAVOCES (5)	2
RD (5)	2
DATE	18/10/2023

(1) 27/10

Result:

Thus, the yearly department performance was successfully analyzed and visualized, revealing key trends, best and worst months, and insights for data-driven decisions.