

PHASE 4 : DEVELOPMENT PART 2

WEBSITE TRAFFIC ANALYSIS

PRIYADARSHINI ENGINEERING COLLEGE

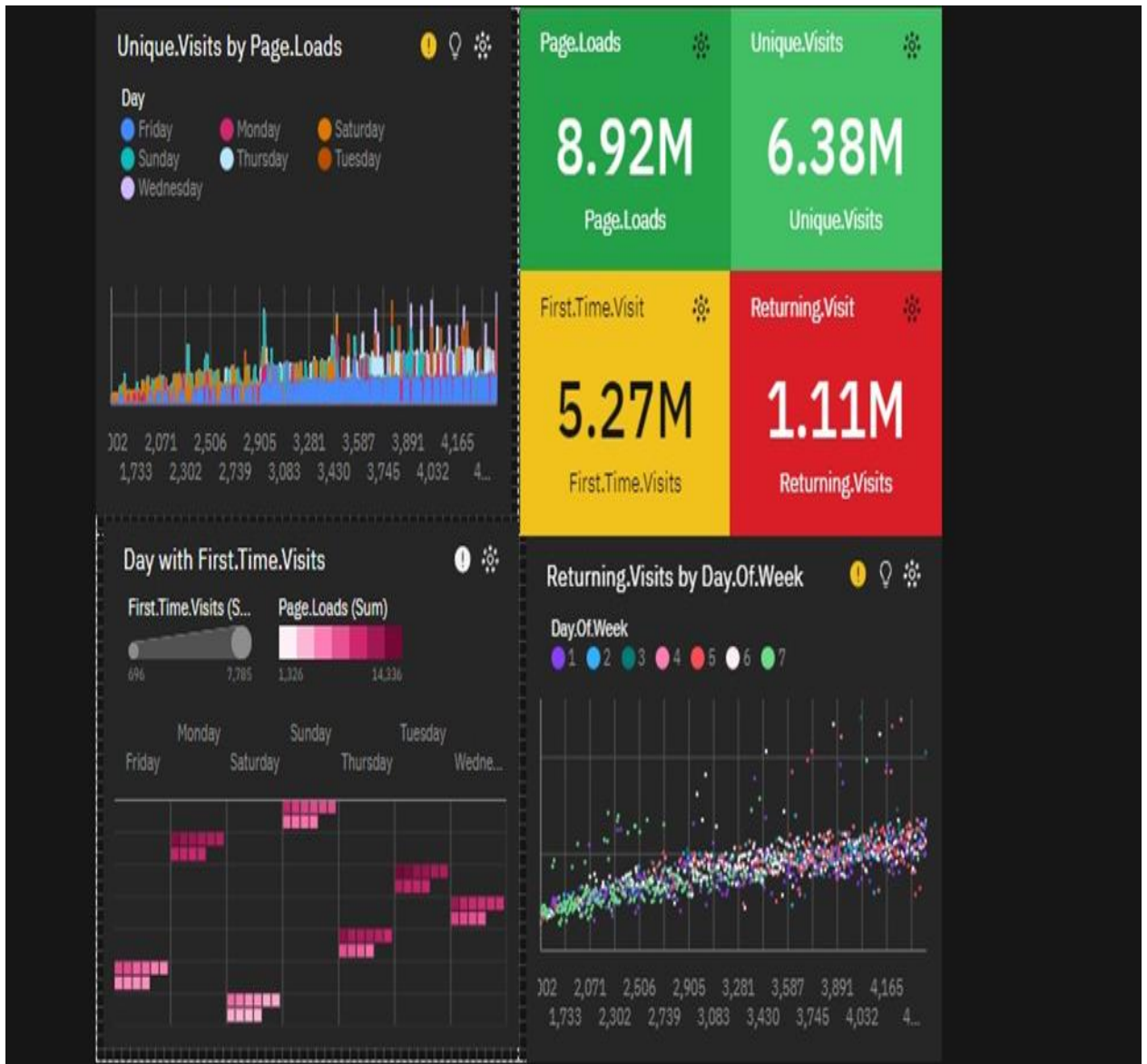
Ram Prasath/511921104055

Website traffic analysis is the process of collecting, examining, and interpreting data related to the visitors and interactions on a website. It provides invaluable insights into user behavior, preferences, and trends, helping organizations make informed decisions, optimize their online presence, and enhance user experiences.

Abstract:

This project aims to analyze website traffic data for insights into user behavior, popular pages, and traffic sources. It involves data collection, visualization using IBM Cognos, and Python for advanced analysis. The goal is to optimize user experiences and enhance website performance.

Data Exploration



Page.Loads has a strong weekly trend. The largest values typically occur on **Tuesday**, whereas the smallest values on **Saturday**.

Based on the current forecasting, **Page.Loads** may reach **nearly four thousand** by **Date 2021-10-27**.

Over all **dates**, the average of **First.Time.Visits** is **almost 2500**.

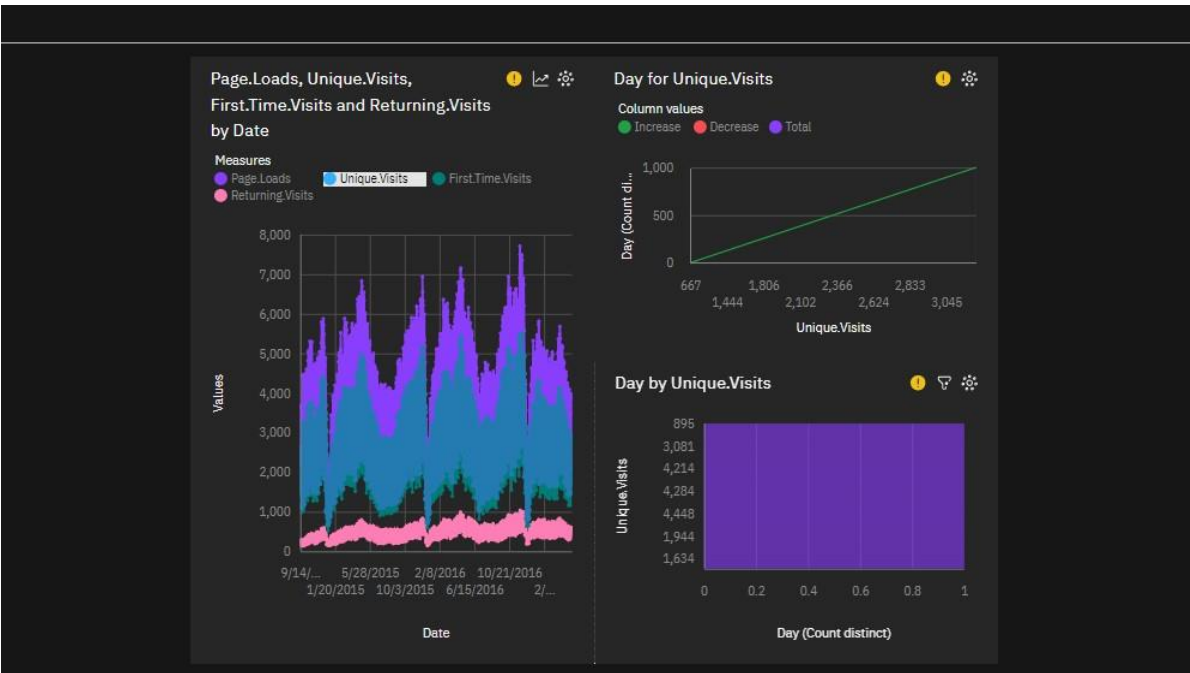
Across all **dates**, the average of **Page.Loads** is **over four thousand**.

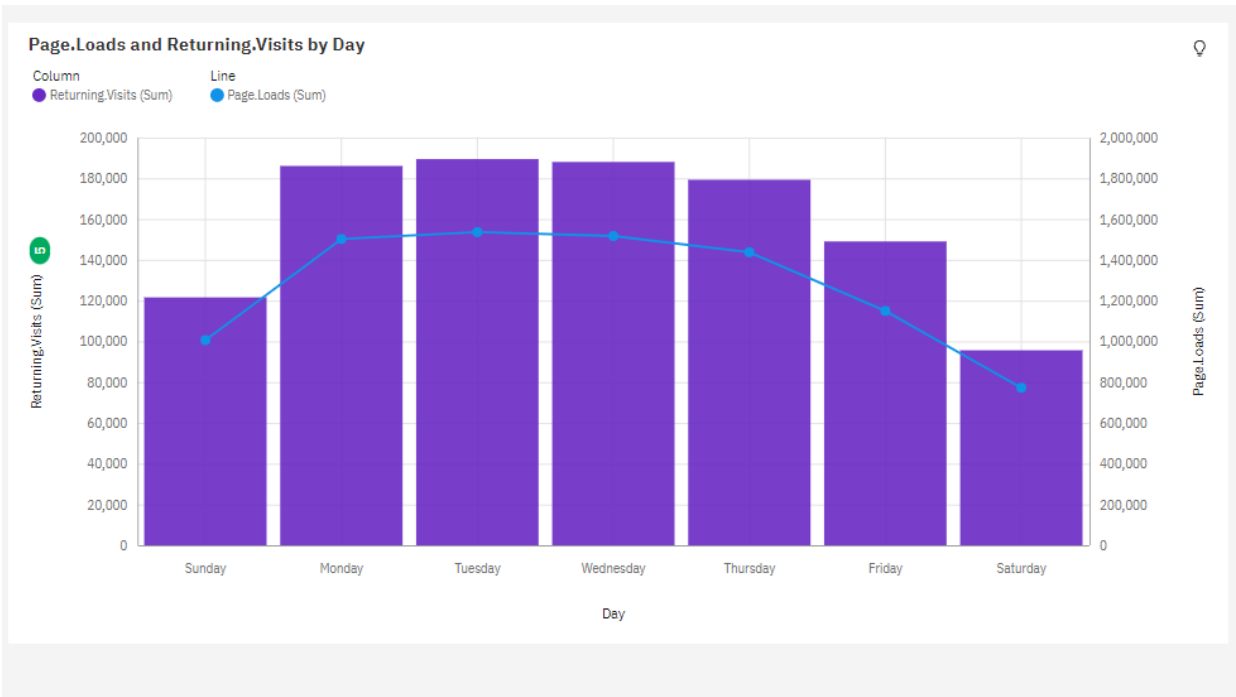
Over all **dates**, the average of **Returning.Visits** is **511.8**.

Over all **dates**, the average of **Unique.Visits** is **nearly three thousand**.

The total number of results for **First.Time.Visits**, across all **dates**, is **over two thousand**.

The total number of results for **Page.Loads**, across all **dates**, is **over two thousand**.





Across all **days**, the sum of **Returning.Visits** is over 1.1 million.

Returning.Visits ranges from almost 96 thousand, when **Day** is Saturday, to over 189 thousand, when **Day** is Tuesday.

Returning.Visits is unusually low when **Day** is Saturday.

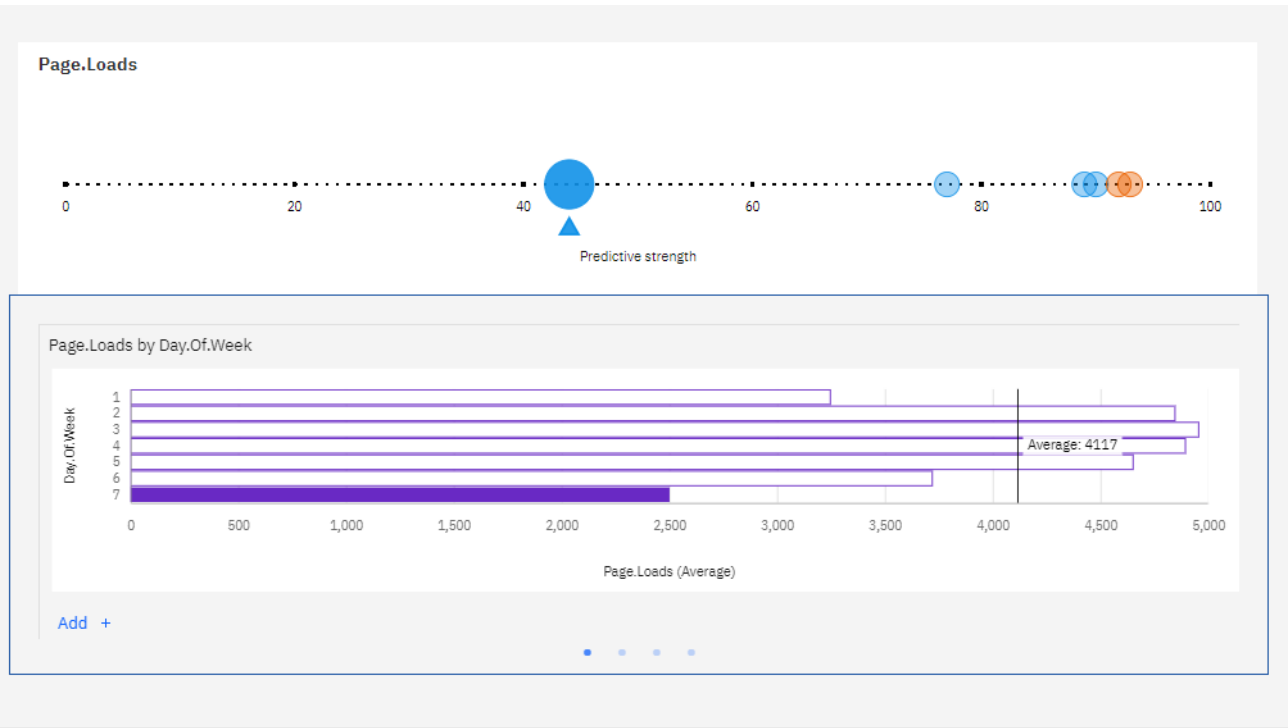
For **Returning.Visits**, the most significant values of **Day** are Tuesday, Wednesday, Monday, Thursday, and Friday, whose respective **Returning.Visits** values add up to almost 892 thousand, or 80.4 % of the total.

Across all **days**, the sum of **Page.Loads** is over 8.9 million.

Page.Loads ranges from nearly 773 thousand, when **Day** is Saturday, to over 1.5 million, when **Day** is Tuesday.

Page.Loads is unusually low when **Day** is Saturday.

For **Page.Loads**, the most significant values of **Day** are Tuesday, Wednesday, Monday, Thursday, and Friday, whose respective **Page.Loads** values add up to over 7.1 million, or 80.1 % of the total.



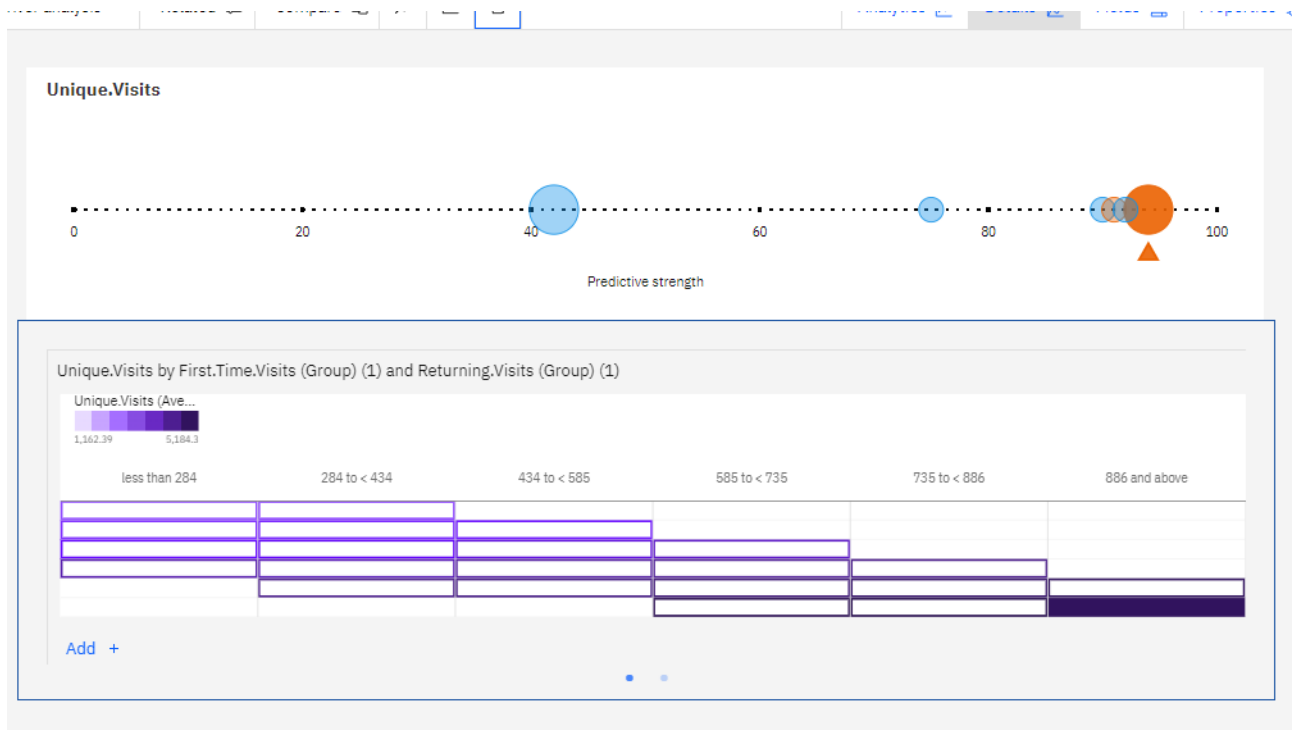
Across all values of **Day.Of.Week**, the average of **Page.Loads** is over four thousand.

The average values of **Page.Loads** range from over 2500, occurring when **Day.Of.Week** is 7, to nearly five thousand, when **Day.Of.Week** is 3.

Day.Of.Week moderately affects **Page.Loads** (44%).

Page.Loads is unusually low when **Day.Of.Week** is 7.

1 (14.3 %), 2 (14.3 %), 3 (14.3 %), and 4 (14.3 %) are the most frequently occurring categories of **Day.Of.Week** with a combined count of 1240 items with **Page.Loads** values (57.2 % of the total).



First.Time.Visits (Group) (3) strongly affects **Unique.Visits** (94%).

Unique.Visits is most unusual when **First.Time.Visits (Group) (3)** is 3934 and above and less than 1205.

Returning.Visits (Group) (2) strongly affects **Unique.Visits** (76%).

Unique.Visits is unusually high when **Returning.Visits (Group) (2)** is 886 and above.

Over all values of **First.Time.Visits (Group) (3)** and **Returning.Visits (Group) (2)**, the average of **Unique.Visits** is nearly three thousand.

The average values of **Unique.Visits** range from over a thousand to over five thousand.

First.Time.Visits (Group) (3) and **Returning.Visits (Group) (2)** strongly affect **Unique.Visits** (96%).

Unique.Visits is unusually high when the combination of **First.Time.Visits (Group) (3)** and **Returning.Visits (Group) (2)** is 3934 and above and 886 and above.

1887 to < 2569 is the most frequently occurring category of **First.Time.Visits (Group) (3)** with a count of 666 items with **Unique.Visits** values (30.7 % of the total).

434 to < 585 is the most frequently occurring category of **Returning.Visits (Group) (2)** with a count of 734 items with **Unique.Visits** values (33.9 % of the total).

There is no significant impact of **Returning.Visits (Group) (2)** on the relationship between **First.Time.Visits (Group) (3)** and **Unique.Visits**.

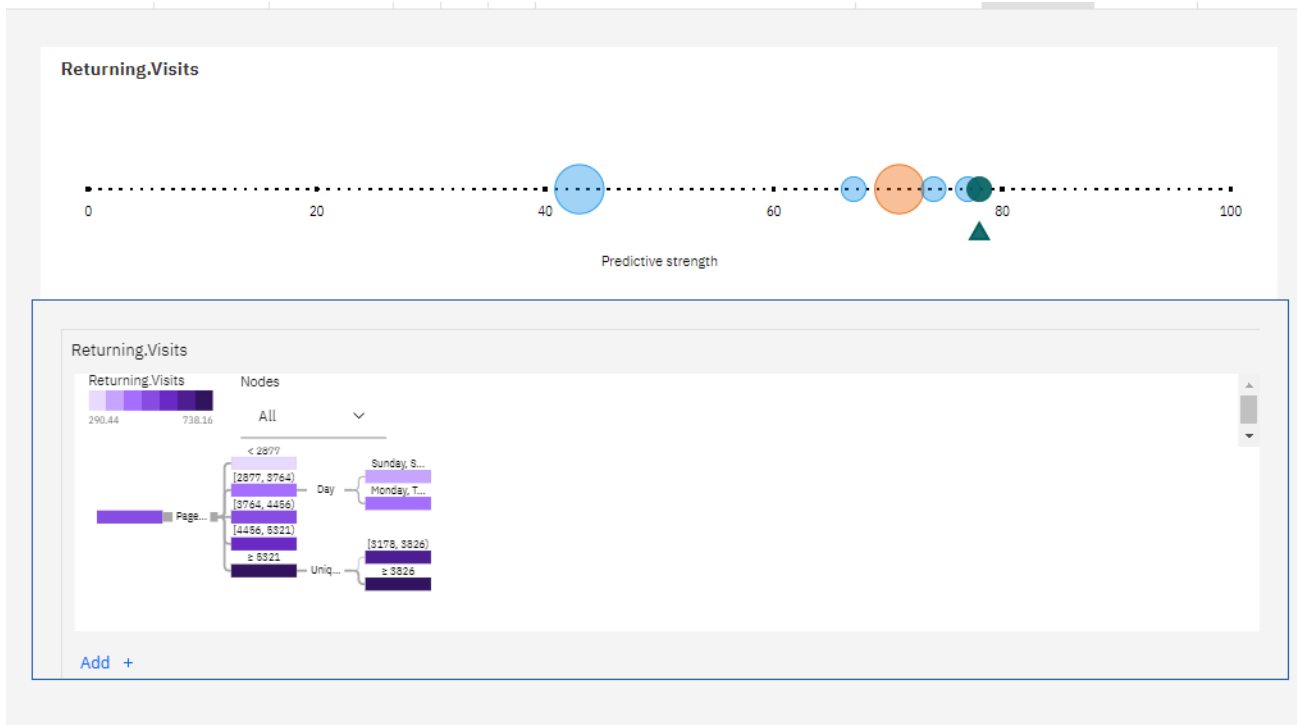


Unique.Visits is unusually high when the combination of **First.Time.Visits (Group) (3)** and **Returning.Visits (Group) (2)** is 3934 and above and 886 and above.

1887 to < 2569 is the most frequently occurring category of **First.Time.Visits (Group) (3)** with a count of 666 items with **Unique.Visits** values (30.7 % of the total).

434 to < 585 is the most frequently occurring category of **Returning.Visits (Group) (2)** with a count of 734 items with **Unique.Visits** values (33.9 % of the total).

There is no significant impact of **Returning.Visits (Group) (2)** on the relationship between **First.Time.Visits (Group) (3)** and **Unique.Visits**.



Page.Loads, Unique.Visits, and Day predict **Returning.Visits** with a strength of 78.1%.

Page.Loads is the most significant predictor of **Returning.Visits** being three times better than any other field.

```
In [38]: import numpy as np
import pandas as pd
import warnings
warnings.filterwarnings('ignore')
import datetime
from datetime import date

import seaborn as sns
import matplotlib.pyplot as plt
%matplotlib inline
sns.set_style("whitegrid")

# import chart_studio.plotly as py
import plotly.express as px

from plotly.offline import download_plotlyjs, init_notebook_mode, plot, iplot
init_notebook_mode(connected=True)

import plotly.graph_objects as go

from sklearn.model_selection import train_test_split, cross_val_score, GridSearchCV
from sklearn.metrics import accuracy_score
from sklearn.svm import SVR
from sklearn.linear_model import LinearRegression
from sklearn.tree import DecisionTreeRegressor
# from prophet import Prophet

In [40]: df=pd.read_csv('D:/daily-website-visitors.csv')
```

localhost:8888/notebooks/Untitled1.ipynb?kernel_name=python3

UPDATE Read the [migration plan](#) to Notebook 7 to learn about the new features and the actions to take if you are using extensions - Please note that updating to Notebook 7 might break some of your extensions. Don't show anymore

jupyter Untitled1 Last Checkpoint: 13 hours ago (autosaved) Logout

File Edit View Insert Cell Kernel Widgets Help Trusted Python 3 (ipykernel)

In [40]:

```
df=pd.read_csv('D:/daily-website-visitors.csv')

df.rename(columns = {'Day.Of.Week':'day_of_week'
                    , 'Page.Loads':'page_loads'
                    , 'Unique.Visits':'unique_visits'
                    , 'First.Time.Visits':'first_visits'
                    , 'Returning.Visits':'returning_visits'}, inplace = True)

df=df.replace(' ','',regex=True)

df['page_loads']=df['page_loads'].astype(int)
df['unique_visits']=df['unique_visits'].astype(int)
df['first_visits']=df['first_visits'].astype(int)
df['returning_visits']=df['returning_visits'].astype(int)

df
```

Out[40]:

Row	Day	day_of_week	Date	page_loads	unique_visits	first_visits	returning_visits
0	1	Sunday	9/14/2014	2146	1582	1430	152
1	2	Monday	9/15/2014	3621	2528	2297	231
2	3	Tuesday	9/16/2014	3698	2630	2352	278
3	4	Wednesday	9/17/2014	3667	2614	2327	287
4	5	Thursday	9/18/2014	3316	2366	2130	236
...
2162	2163	Saturday	8/15/2020	2221	1696	1373	323

localhost:8888/notebooks/Untitled1.ipynb?kernel_name=python3

UPDATE Read the [migration plan](#) to Notebook 7 to learn about the new features and the actions to take if you are using extensions - Please note that updating to Notebook 7 might break some of your extensions. Don't show anymore

jupyter Untitled1 Last Checkpoint: 13 hours ago (autosaved) Logout

File Edit View Insert Cell Kernel Widgets Help Trusted Python 3 (ipykernel)

2167 rows x 8 columns

In [41]:

```
df.isna().sum()
```

Out[41]:

```
Row      0
Day       0
day_of_week  0
Date      0
page_loads  0
unique_visits  0
first_visits  0
returning_visits  0
dtype: int64
```

In [42]:

```
df.duplicated().sum()
```

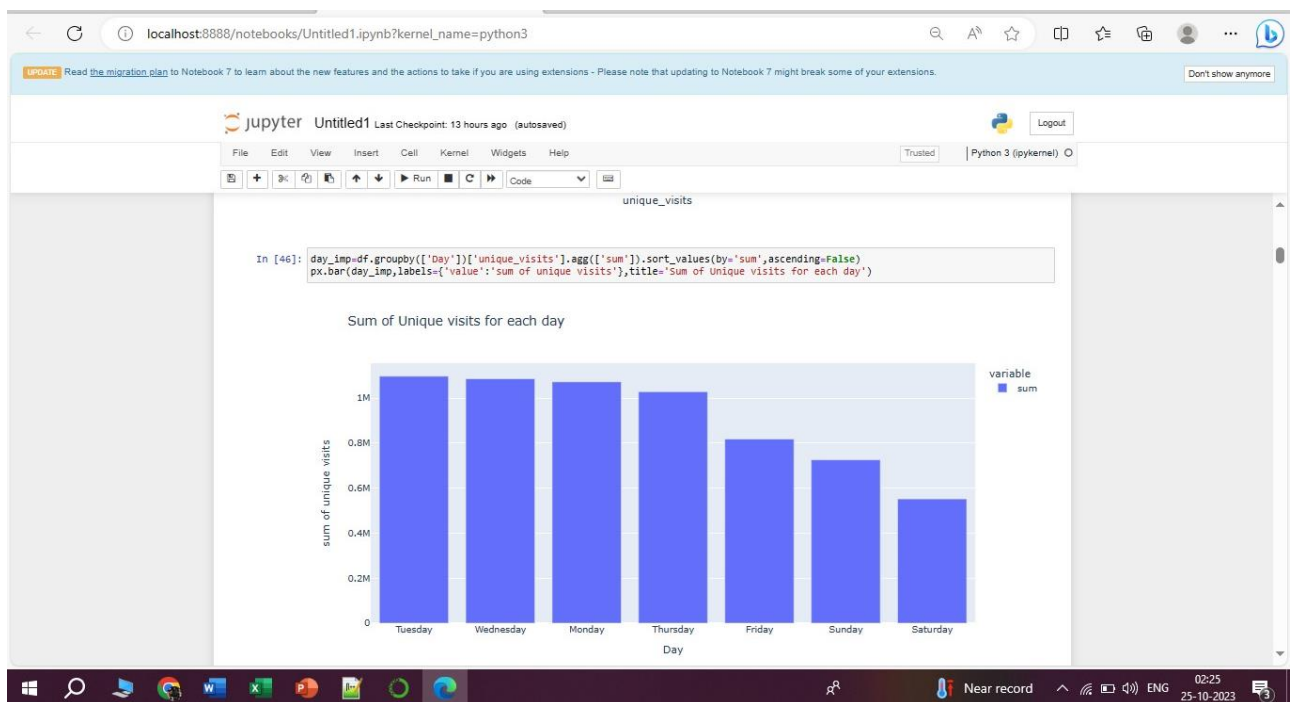
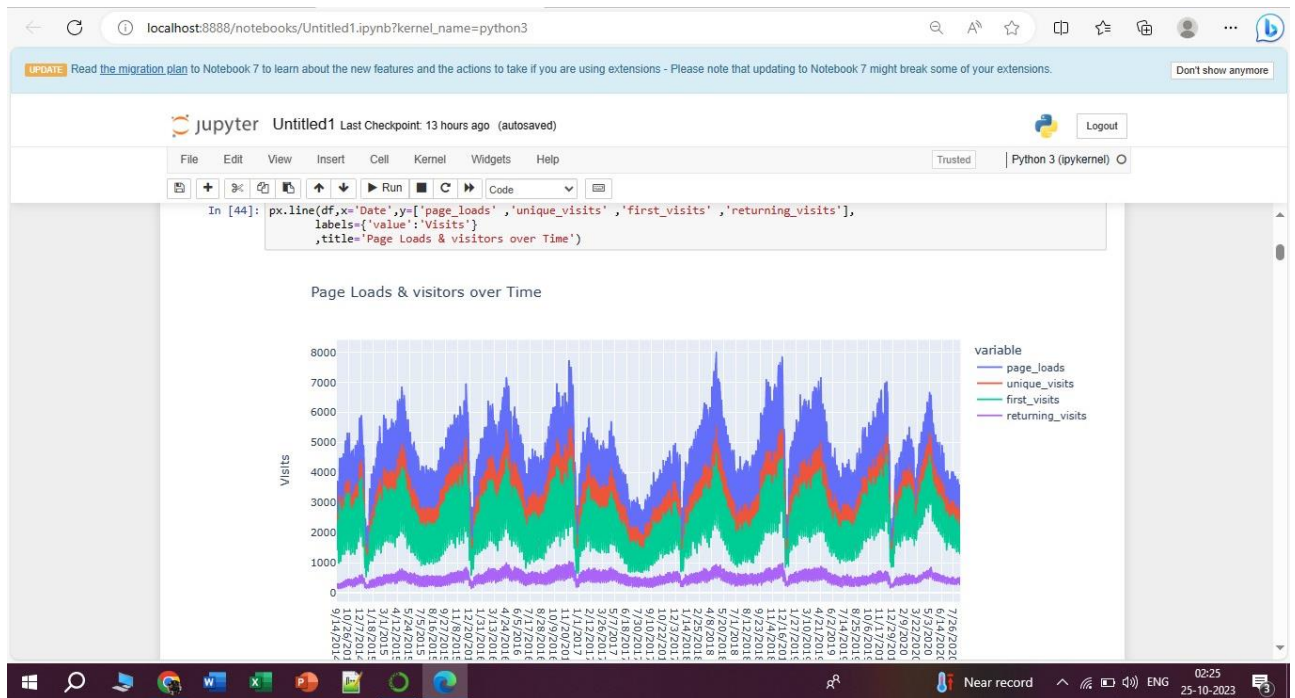
Out[42]:

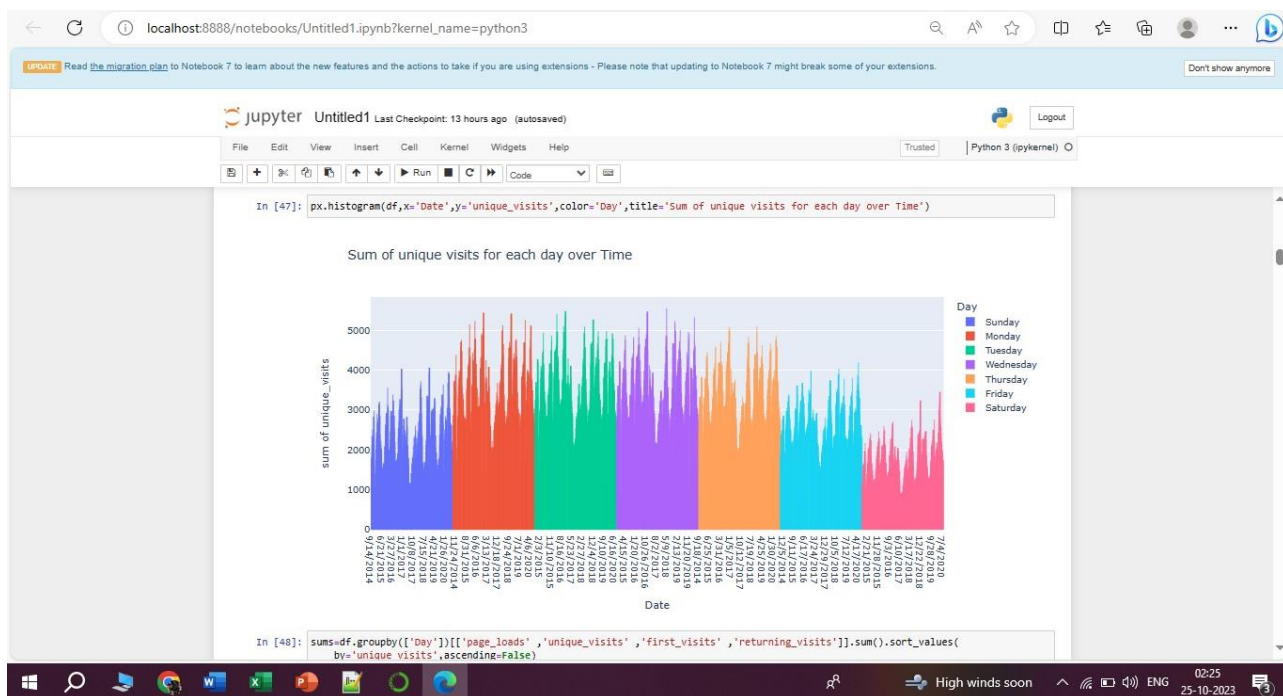
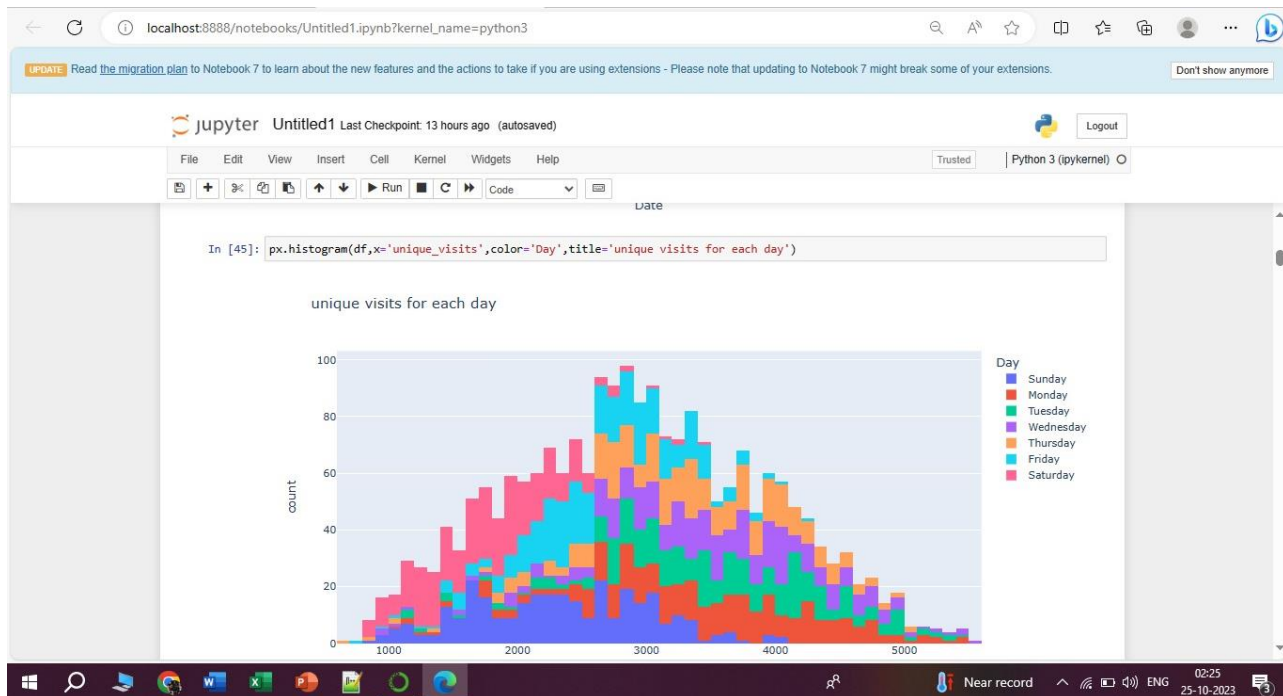
```
0
```

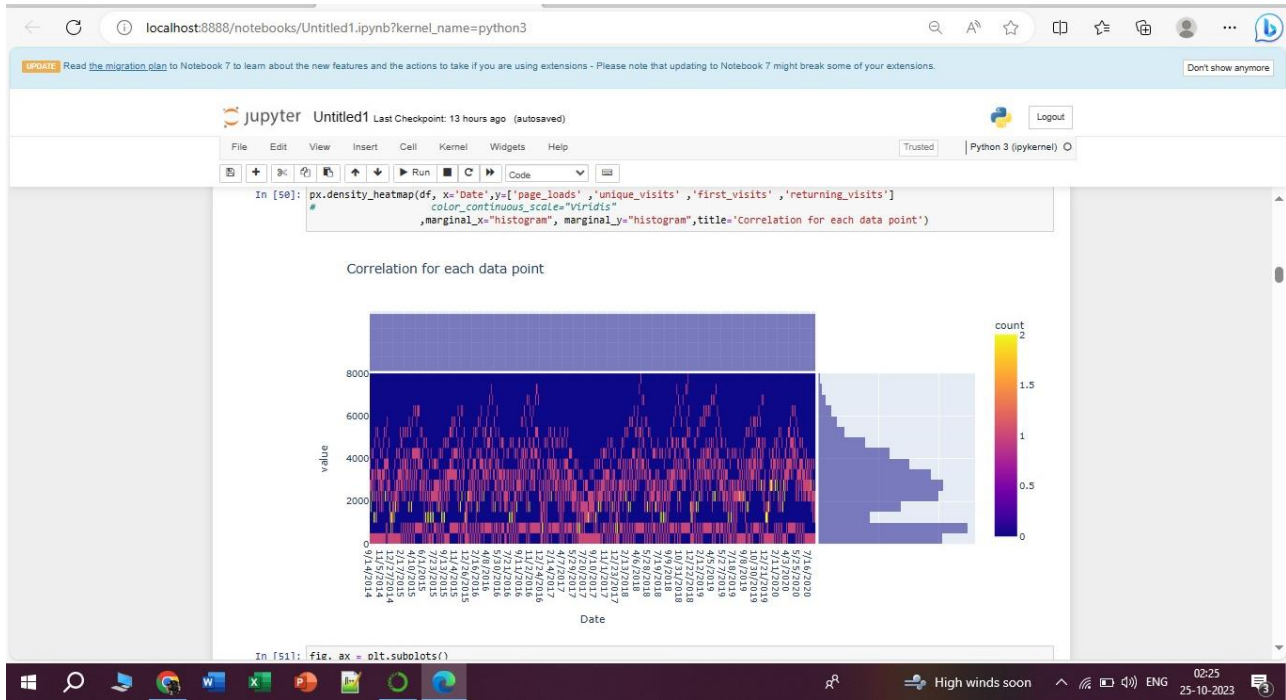
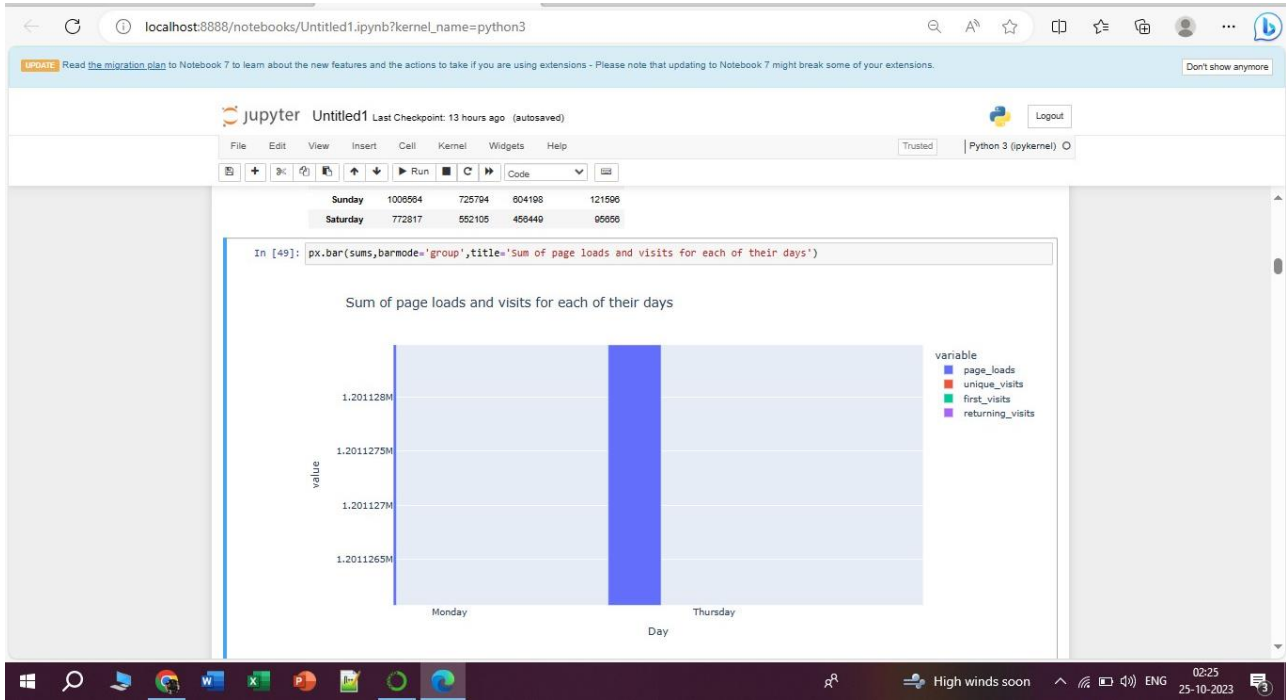
In [43]:

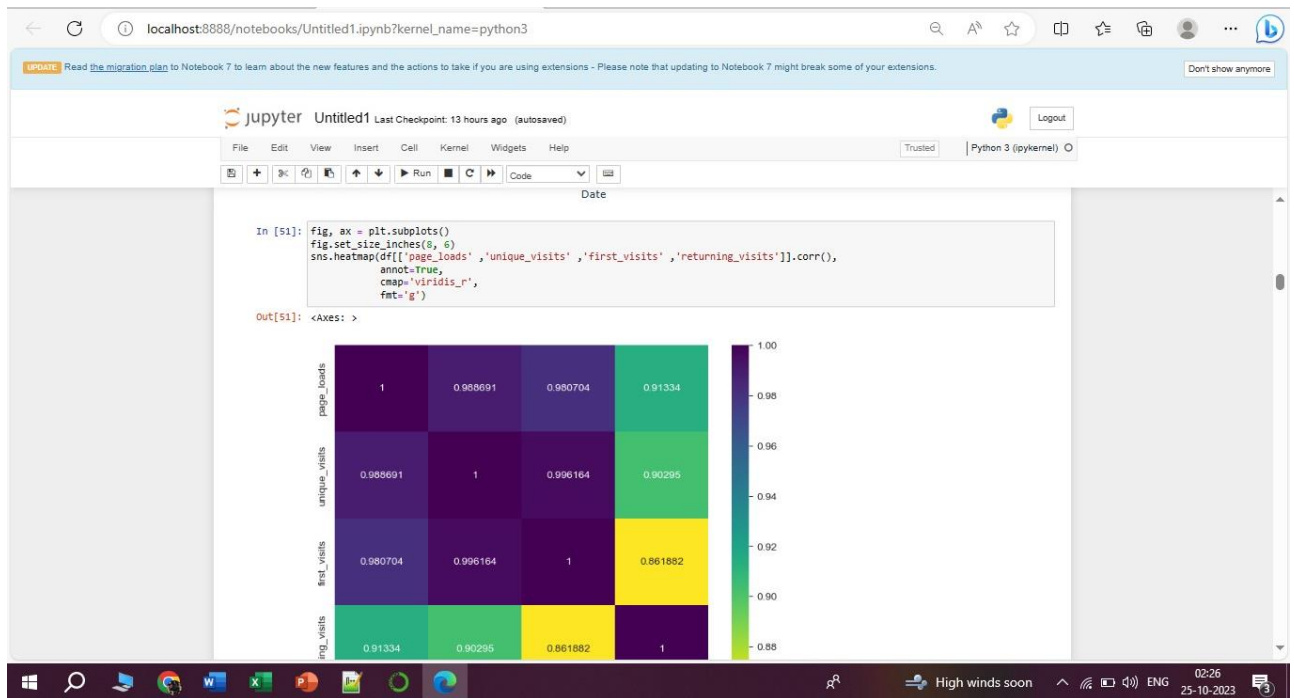
```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2167 entries, 0 to 2166
Data columns (total 8 columns):
#   Column      Non-Null Count  Dtype
---  -
0   Row         2167 non-null  int64
1   Day         2167 non-null  object
2   day_of_week 2167 non-null  int64
3   Date        2167 non-null  object
4   page_loads   2167 non-null  int32
```









Conclusion

Website traffic analysis using IBM Cognos Analytics allows businesses to gain valuable insights into the performance of their websites. This analysis helps in making data-driven decisions to enhance user experience, optimize content, and improve marketing strategies. IBM Cognos Analytics provides robust tools for tracking and visualizing web traffic data, enabling organizations to monitor key metrics, detect trends, and make informed decisions to drive business success. It offers the ability to create interactive reports and dashboards, making it easier for teams to collaborate and act on the insights derived from website traffic data, ultimately leading to improved online performance and user engagement.