Website traffic analysis

Phase 4:

Topic: Use IBM Cognos to create interactive dashboards and reports that display insights such as popular pages, traffic sources, and user engagement metrics.

- Use Python libraries like Pandas and Matplotlib to perform more complex analyses on the data, such as time series analysis, user segmentation, or machine learning-based predictions.

Introduction:

- Website traffic refers to the volume of visitors or users who access a particular website during a given period. It is a fundamental metric in web analytics, providing valuable insights into a website's performance and the behavior of its audience. Understanding website traffic is essential for businesses, organizations, and individuals seeking to optimize their online presence and achieve various goals, such as increasing brand visibility, driving conversions, or improving user experience.

Program:

```
import numpy as np

import pandas as pd
```

```
import pandas_profiling

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 cufflinks as cf

import plotly.express as px

from plotly.offline import download_plotlyjs,
init_notebook_mode, plot, iplot

init_notebook_mode(connected=True)

cf.go_offline()

import pandas_profiling

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
```

```
import xgboost as xg

df=pd.read_csv('../input/daily-website-visitors/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 1:

Row	Day	day_of_week	Date	page_loads	unique_visit	first_visits	
0	1	Sunday	1	9/14/2014	2146	1582	1430	152
1	2	Monday	2	9/15/2014	3621	2528	2297	231
2	3	Tuesday	3	9/16/2014	3698	2630	2352	278
3	4	Wednesday	4	9/17/2014	3667	2614	2327	287
4	5	Thursday	5	9/18/2014	3316	2366	2130	236
...	...	... ...	...	...	... ...	...	...	
2162	2163	Saturday	7	8/15/2020	2221	1696	1373	323
2163	2164	Sunday	1	8/16/2020	2724	2037	1686	351

In 2:

df.isna().sum()

out 2:

```
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 3:

df.duplicated().sum()

out 3:

0

In 4:

df.info()

out 4:

```
<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   int64
```

5 unique_visits 2167 non-null int64

6 first_visits 2167 non-null int64

7 returning_visits 2167 non-null int64

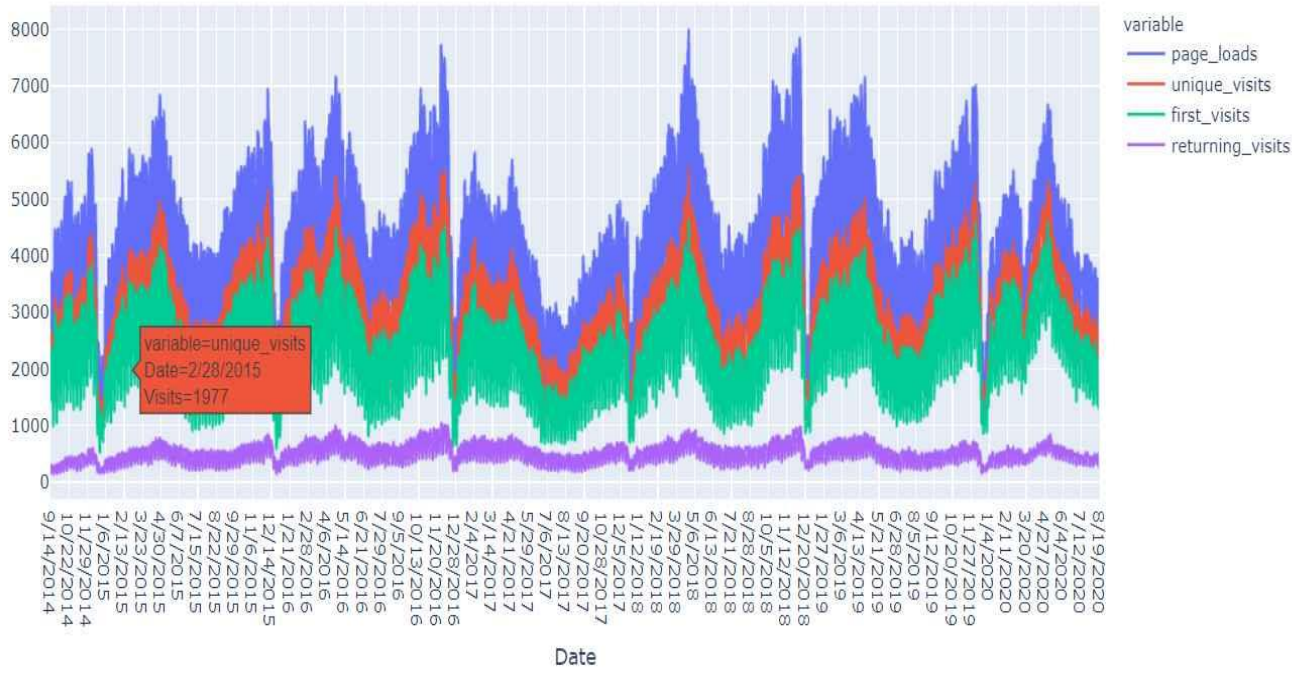
dtypes: int64(6), object(2)

memory usage: 135.6+ KB

In 5:

```
px.line(df,x='Date',y=['page_loads' ,'unique_visits' ,'first_visits'
,'returning_visits'],
      labels={'value':'Visits'}
      ,title='Page Loads & visitors over Time')
```

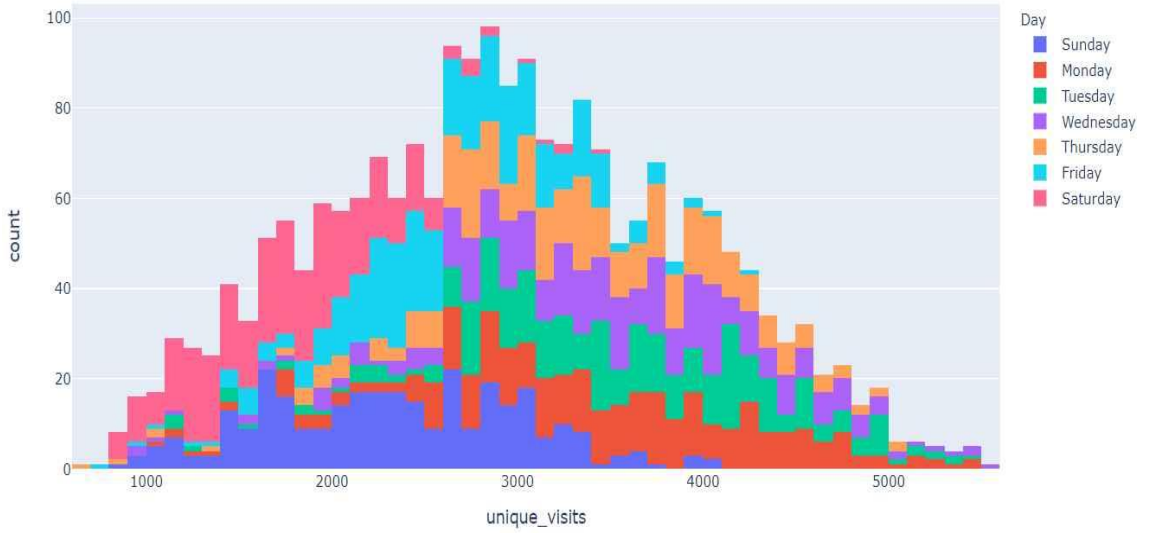
Out 5:



In 6:

```
px.histogram(df,x='unique_visits',color='Day',title='unique visits
for each day')
```

out 6:

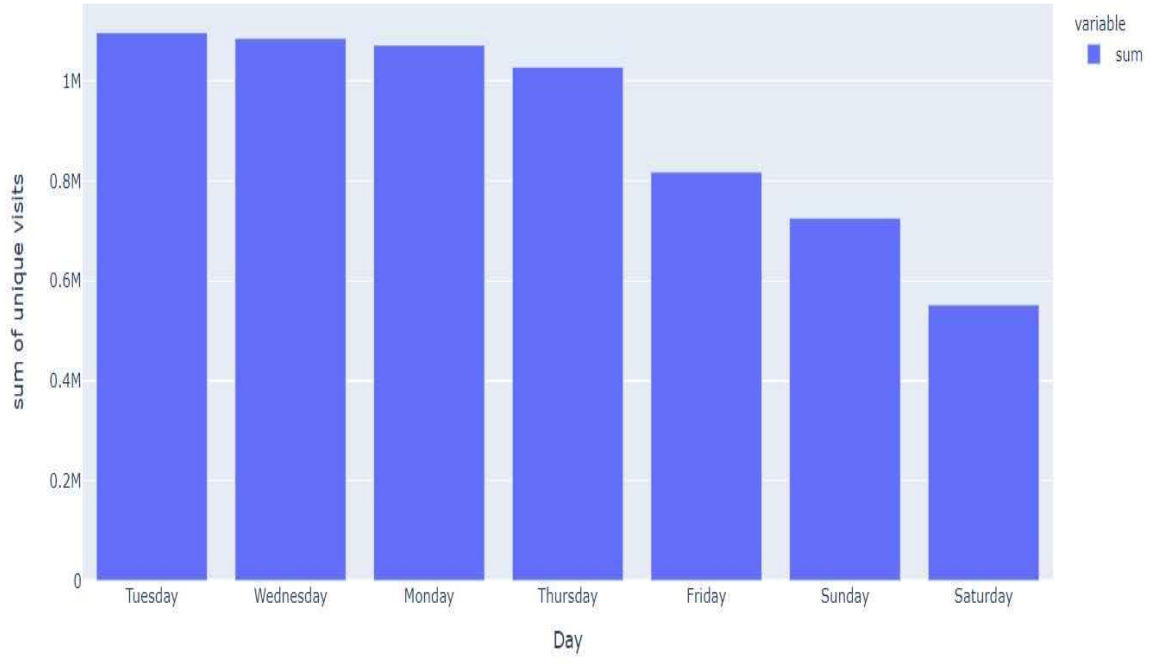


In 7:

```
day_imp=df.groupby(['Day'])['unique_visits'].agg(['sum']).sort_values(by='sum'
,ascending=False)

px.bar(day_imp,labels={'value':'sum of unique visits'},title='Sum of Unique
visits for each day')
```

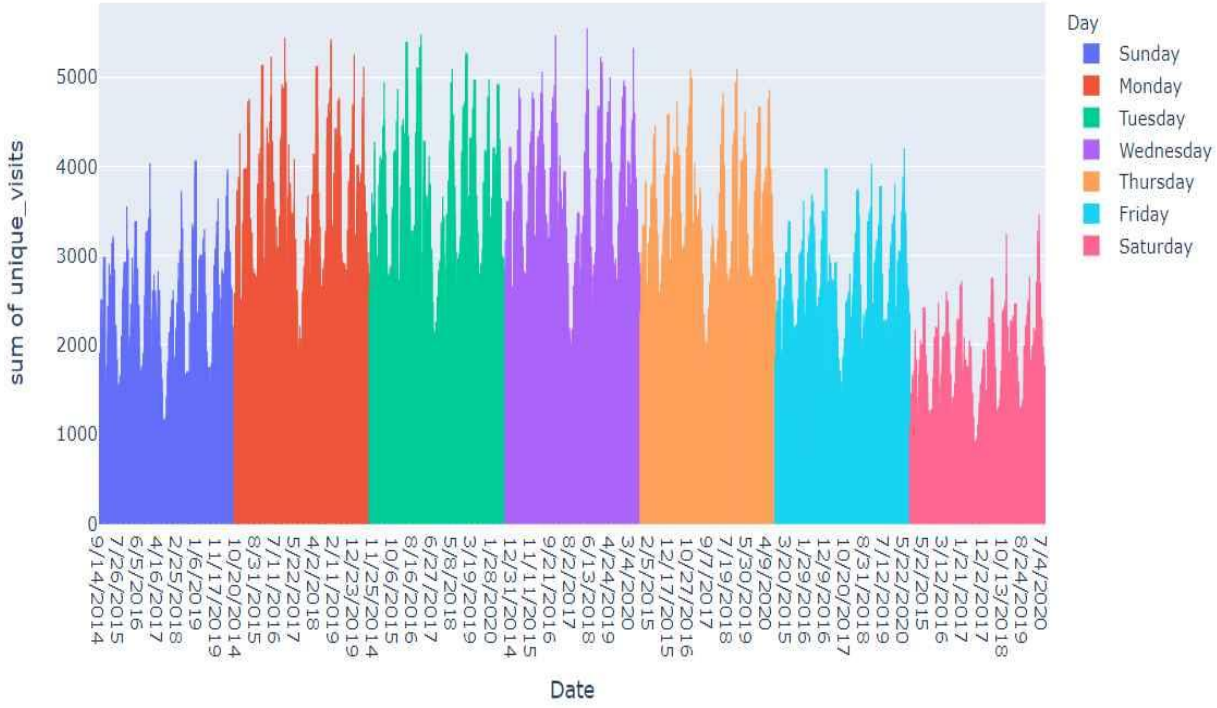
out 7:



In 8:

```
px.histogram(df,x='Date',y='unique_visits',color='Day',title='Sum of unique visits for e
ach day over Time')
```

out 8:



In 9:

```
sums=df.groupby(['Day'])[['page_loads' ,'unique_visits' ,'first_visits' ,'returning_visits']].sum().sort_values(
    by='unique_visits',ascending=False)
sums
```

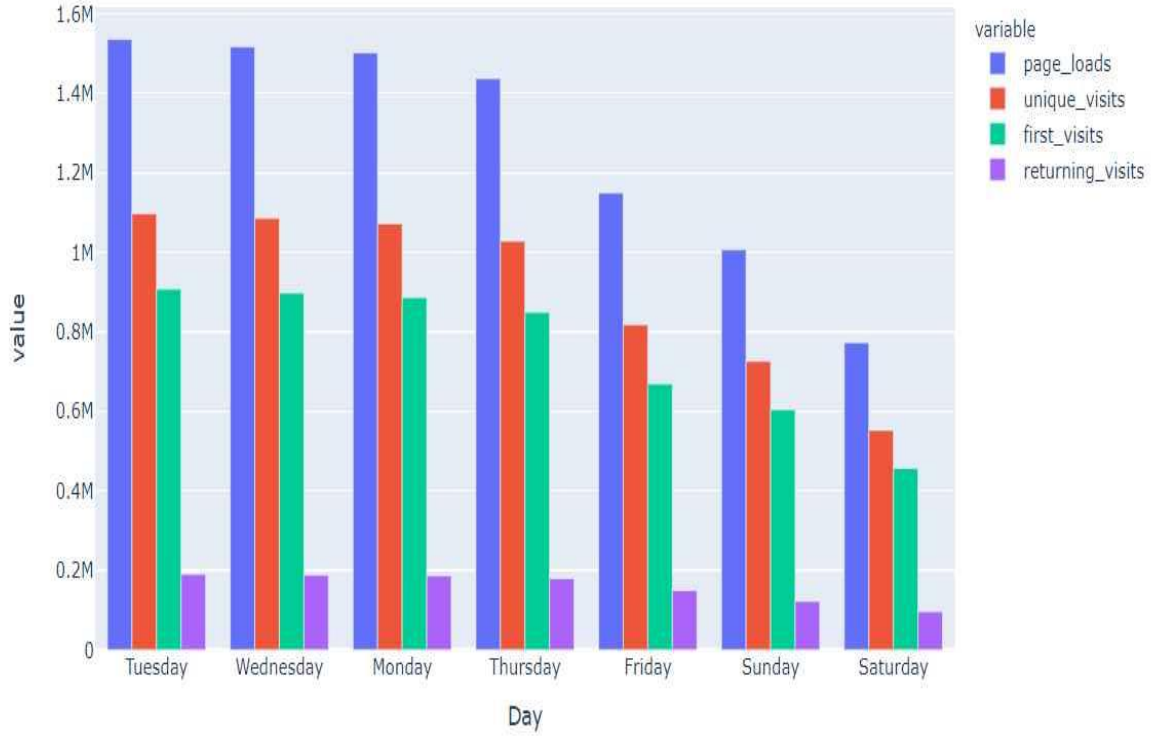
out 9:

page_loads	unique_visits	first_visits	returning_visits
Day			
Tuesday	1536154	1097181	907752
Wednesday	1517114	1085624	897602

In 10:

```
px.bar(sums,barmode='group',title='Sum of page loads and visits for each of their days')
```

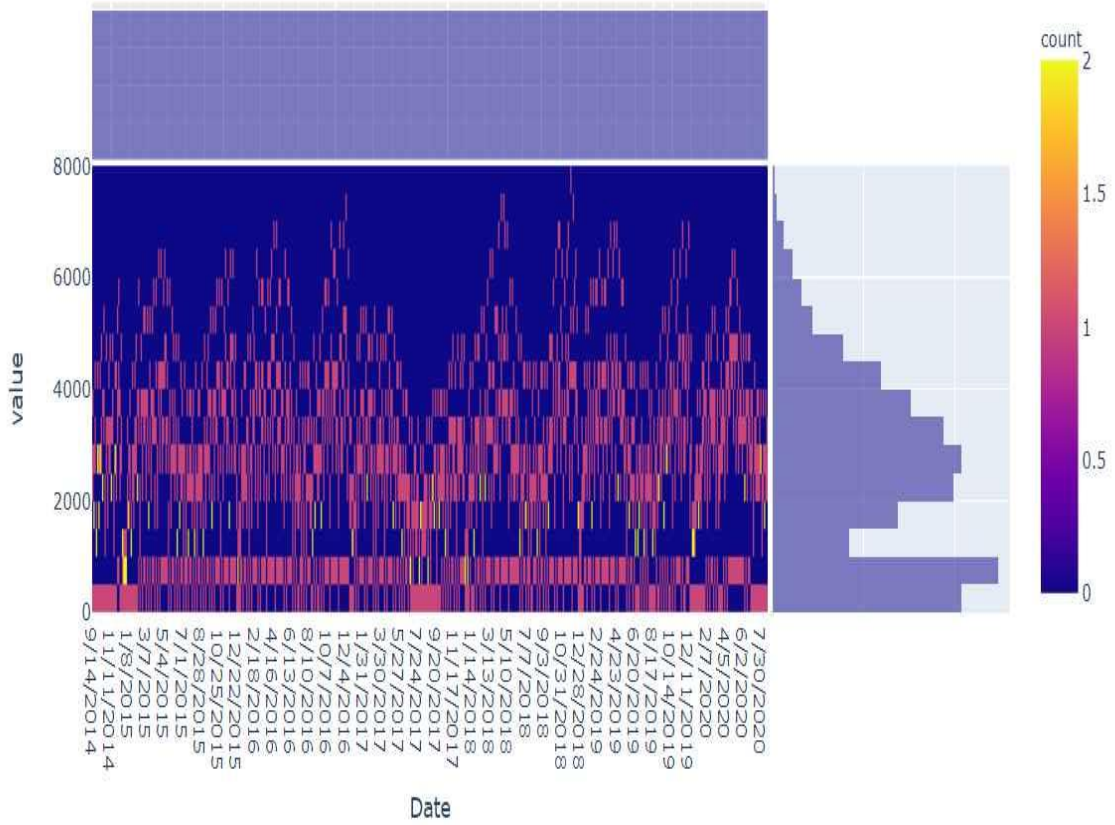
out 10:



In 11:

```
px.density_heatmap(df, x='Date',y=['page_loads' ,'unique_visits' ,'first_visits' ,'returning_visits']
                   color_continuous_scale="Viridis"
                   ,marginal_x="histogram", marginal_y="histogram",title='Correlation for each data point')
```

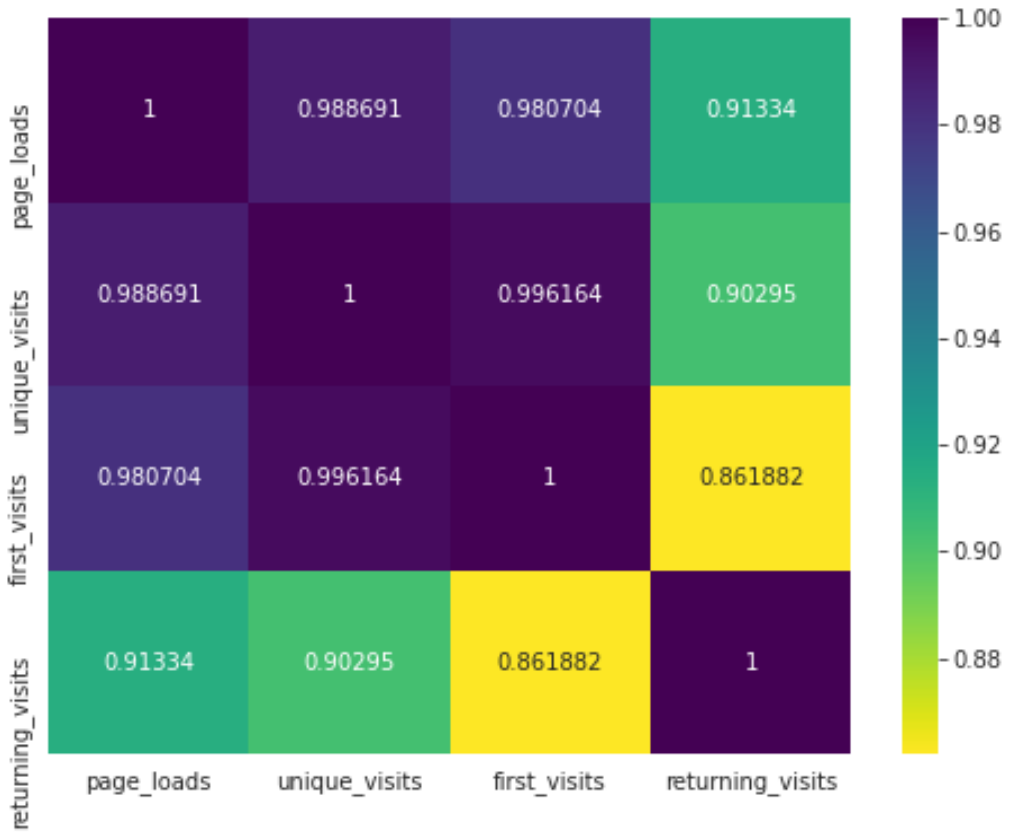
Out 11:



In 12:

```
fig, ax = plt.subplots()
fig.set_size_inches(8, 6)
sns.heatmap(df[['page_loads' ,'unique_visits' ,'first_visits' ,'returning_visits']].corr(),
            annot=True,
            cmap='viridis_r',
            fmt='g')
```

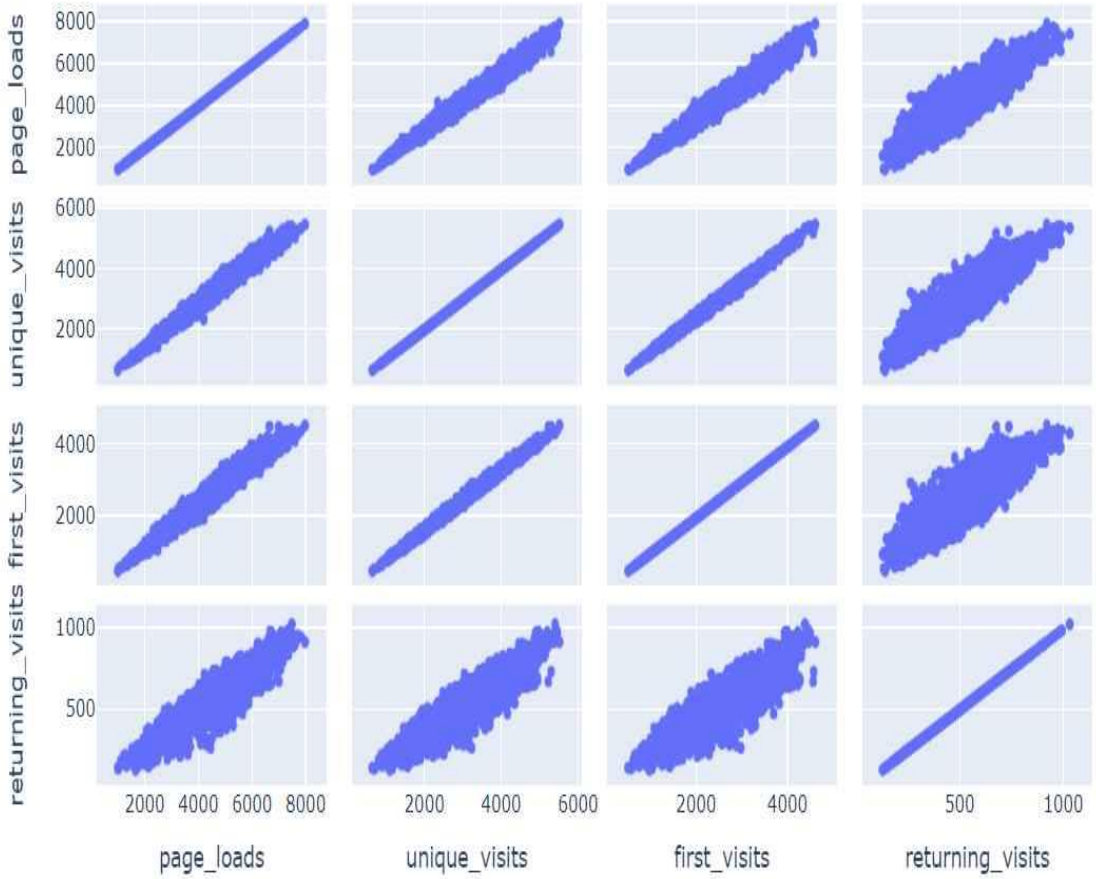
out 12:



In 13:

```
px.scatter_matrix(df[['page_loads' ,'unique_visits' ,'first_visits' ,'returning_visits']])
```

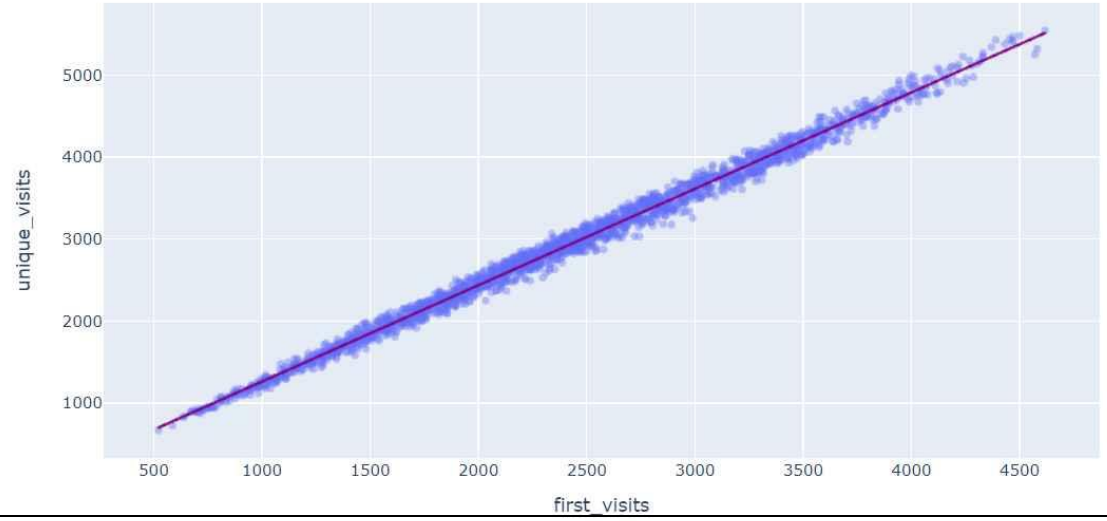
out 13:



In 14:

```
px.scatter(
    df, x='first_visits', y='unique_visits',opacity=0.4,
    trendline='ols',
trendline_color_override='purple',title="Regression line for unique visits and first visits"
)
```

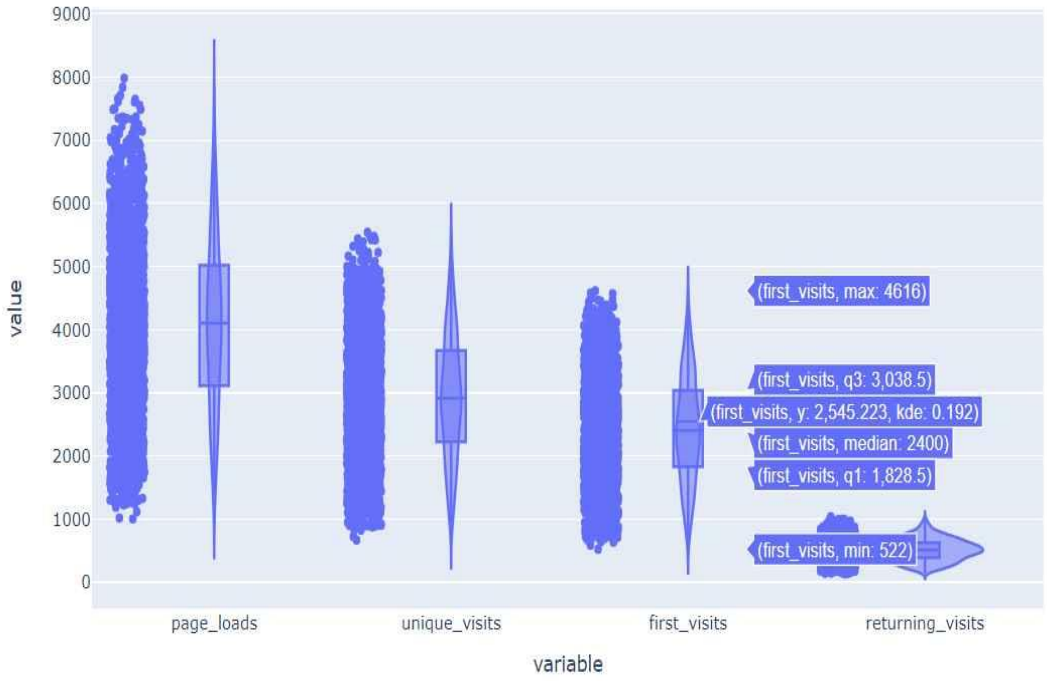
Out 14:



In 15:

```
px.violin(df,y=['page_loads' ,'unique_visits' ,'first_visits' ,'returning_visits'],box=True,points='all')
```

out 15:



In 16:

```
regressor2=LinearRegression(fit_intercept=False,normalize=True)

regressor2.fit(X_train, y_train)
```

out 16:

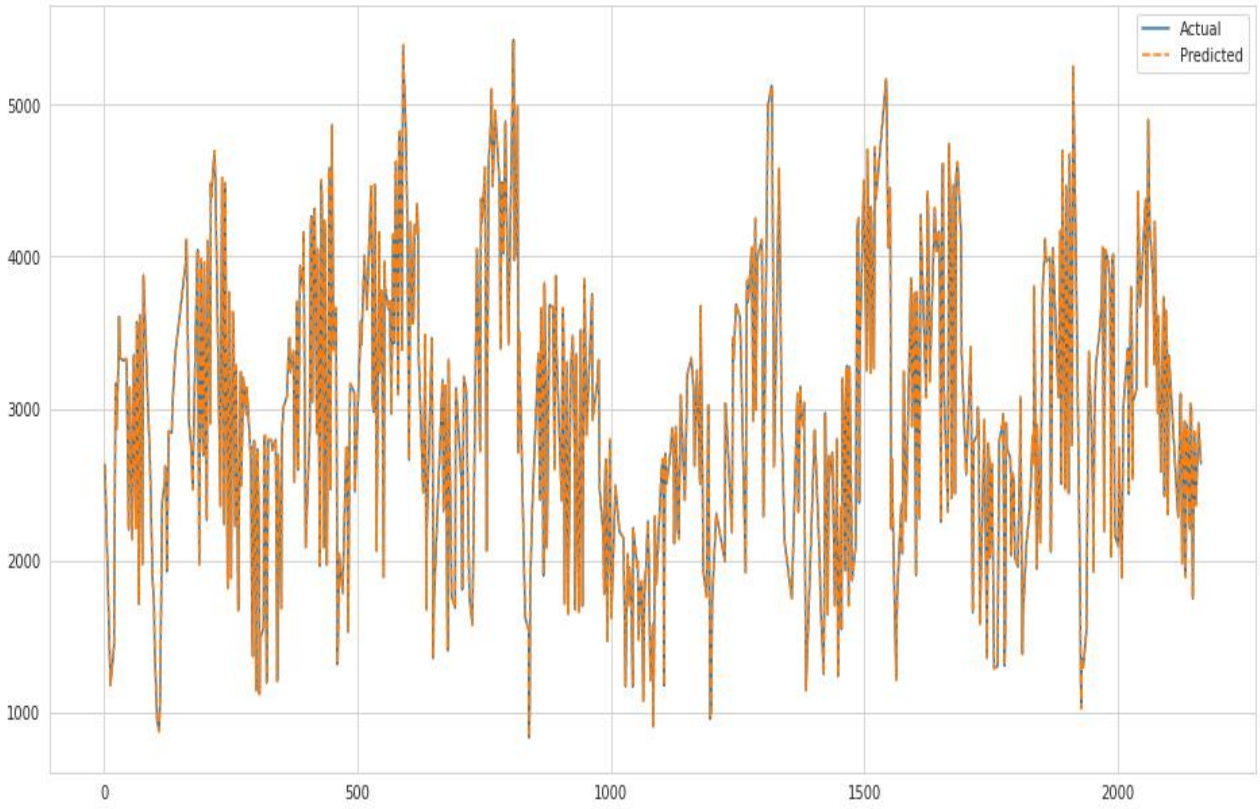
```
LinearRegression(fit_intercept=False, normalize=True)
```

In 17:

```
plt.figure(figsize=(16,8))

sns.lineplot(data=lr2)
```

out 17:



In 18:

regressor2.score(X_test,y_test)*100

out 18:

100.0

In 19:

svr_rbf = SVR(kernel='rbf', C=1e3, gamma=0.00001)

svr_rbf.fit(X_train, y_train)

out 19:

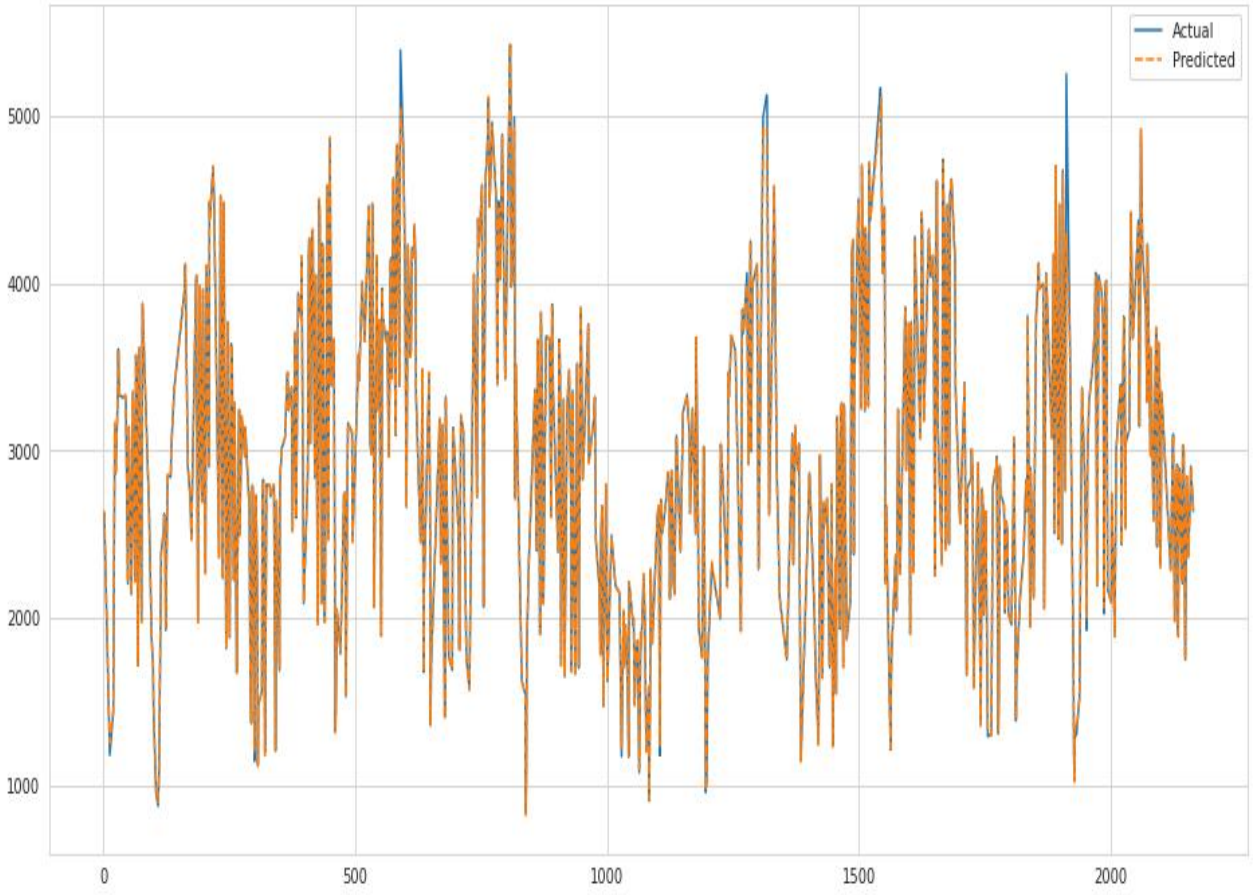
SVR(C=1000.0, gamma=1e-05)

In 20:

plt.figure(figsize=(16,8))

sns.lineplot(data=svr)

out 20:



In 21:

svr_rbf.score(X_test,y_test)*100

out 21:

99.80054455767926

In 22:

xgb_r = xg.XGBRegressor(objective ='reg:squarederror',n_estimators = 10, seed = 123)

xgb_r.fit(X_train, y_train)

out 22:

XGBRegressor(base_score=0.5, booster='gbtree', colsample_bylevel=1,

colsample_bynode=1, colsample_bytree=1, gamma=0, gpu_id=-1,

importance_type='gain', interaction_constraints='',

learning_rate=0.300000012, max_delta_step=0,
max_depth=6,

min_child_weight=1, missing=nan,
monotone_constraints='()',

n_estimators=10, n_jobs=4, num_parallel_tree=1,
random_state=123,

reg_alpha=0, reg_lambda=1, scale_pos_weight=1,
seed=123,

subsample=1, tree_method='exact',
validate_parameters=1,

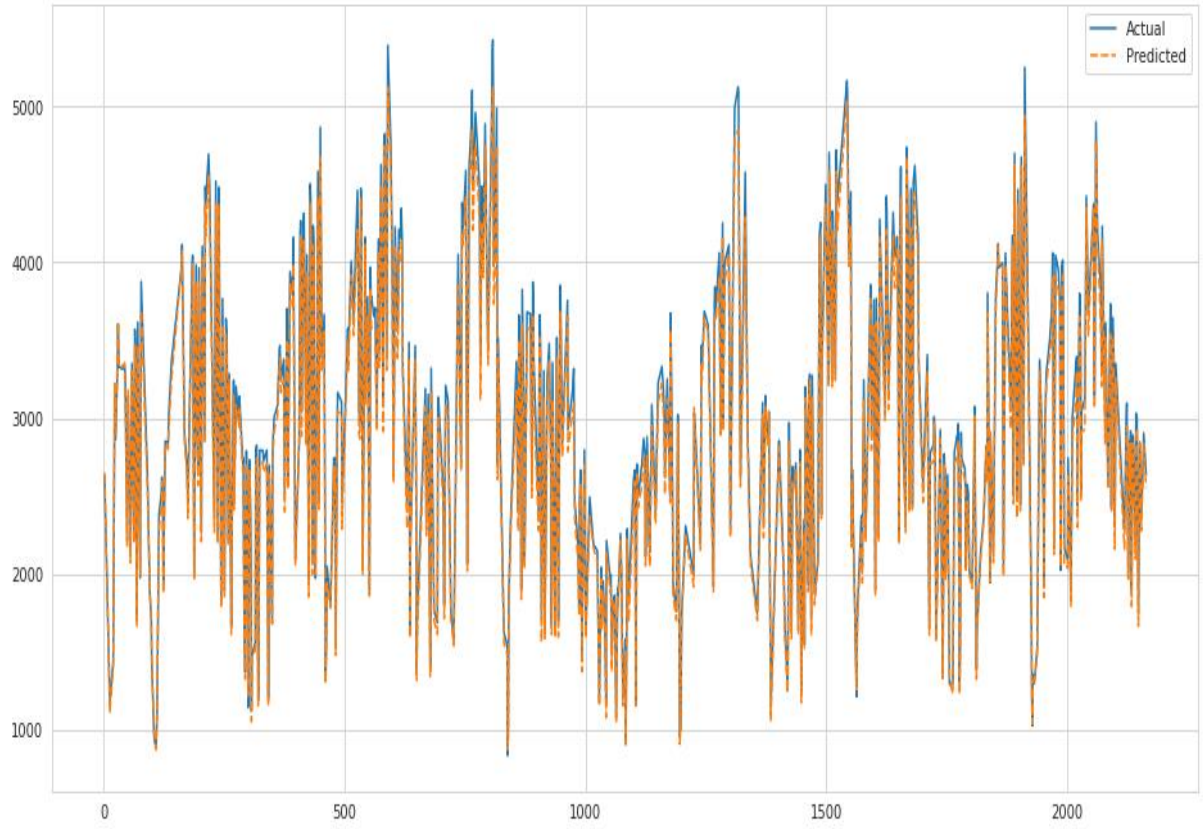
verbosity=None)

In 23:

```
plt.figure(figsize=(16,8))

sns.lineplot(data=xgb_df)
```

out 23:



In 24:

```
xgb_r.score(X_test,y_test)*100
```

out 24:

98.7655882096893

Conclusion:

- In today's Web development, a good page design is essential. A bad design will lead to the loss of visitors and that can lead to a loss of business. In general, a good page layout has to satisfy the basic elements of a good page design. This includes color contrast, text organization, font selection, style of a page, page size, graphics used, and consistency. In order to create a well-designed page for a specific audience. The developer needs to organized and analyze the users' statistics and the background of the users. Although it can be hard to come up with a design that is well suited to all of the users, there will be a design that is appropriate for most of the audience. The better the page design, the more hits a page will get. That implies an increase in accessibility and a possible increase in business.