

AI MIDSEM – 1 EXAMINATION

NAME – PRATEEK BALIYAN

BRANCH – CSEAI

SECTION – C

ROLL NUMBER (UNI) – 202401100300177

ROLL NUMBER (CLASS) – 30

TOPIC NAME – TRAFFIC LIGHT CONTROL
SYSTEM

INTRODUCTION

Traffic data analysis is crucial for understanding web user behaviour, optimizing website performance, and improving user engagement. This report examines traffic data over a 20-day period, focusing on key metrics such as Page Views, Unique Visitors, and Bounce Rate.

The goal of this study is to:

Identify traffic trends over time.

Analyse the relationship between Page Views and Bounce Rate.

Provide insights for optimizing web performance.

Using Python, pandas, matplotlib, and seaborn, the dataset was processed to extract meaningful patterns and visualizations. The findings help in understanding visitor engagement and website efficiency.

METHODOLOGY

Methodology for the Given Code

1. Data Preprocessing (AI Foundation)

Handling Missing Values:

Checking for missing values using `df.isnull().sum()` is a crucial preprocessing step in AI to ensure data quality.

Feature Engineering:

Extracting the hour from the timestamp (`df['timestamp'].dt.hour`) is a common AI technique to create meaningful features for time-series models.

2. Exploratory Data Analysis (EDA)

Statistical Analysis:

Using `df.describe()` helps understand the data distribution, an essential step before applying machine learning models.

Time-Series Analysis:

Grouping data by hour (`df.groupby('hour').size()`) is a fundamental step in predictive modeling and anomaly detection in AI.

3. Visualization (AI-Powered Insights)

Seaborn & Matplotlib Plots:

These visualizations help identify trends, outliers, and patterns, which are crucial for training AI models like time-series forecasting (e.g., ARIMA, LSTMs) or anomaly detection models (e.g., Isolation Forests, Autoencoders).

CODE

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

# Load the CSV file
file_path = "/content/traffic_data.csv" # Update with the correct path if needed
df = pd.read_csv(file_path)

print("Dataset Information:\n")
print(df.info())

# Show first few rows
print("\nFirst 5 Rows:\n", df.head())

# Check for missing values
print("\nMissing Values:\n", df.isnull().sum())

# Basic statistics
print("\nStatistical Summary:\n", df.describe())

# Time-based traffic analysis (Assuming there's a 'timestamp' column)
if 'timestamp' in df.columns:
    df['timestamp'] = pd.to_datetime(df['timestamp'])
    df['hour'] = df['timestamp'].dt.hour
    traffic_by_hour = df.groupby('hour').size()

plt.figure(figsize=(10, 5))
sns.lineplot(x=traffic_by_hour.index, y=traffic_by_hour.values, marker='o')
plt.title('Traffic Volume by Hour')
```

```
plt.xlabel('Hour of Day')
plt.ylabel('Traffic Count')
plt.grid()
plt.show()
```

```
# Convert 'Date' column to datetime format
df['Date'] = pd.to_datetime(df['Date'])
```

```
# Create the plot
plt.figure(figsize=(10, 5))
plt.plot(df['Date'], df['PageViews'], label='Page Views', marker='o', linestyle='-')
plt.plot(df['Date'], df['UniqueVisitors'], label='Unique Visitors', marker='s', linestyle='--')
```

```
# Formatting
plt.xlabel("Date")
plt.ylabel("Count")
plt.title("Website Traffic Over Time")
plt.xticks(rotation=45)
plt.legend()
plt.grid(True)
```

```
# Show the plot
plt.show()
```

OUTPUT

