# **Project Report**

Project Name: Uber Rides Data Analysis

## **Group Members:**

- Member 1: Avishkar Chavhan, Project Lead
- Member 2: Sarthak Gite, Visualization Specialist
- Member 3: Prathamesh Bhagwat, Presentation Specialist
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# 1. Objective

The primary objective of this project is to analyze Uber ride data to extract meaningful insights into:

- Peak demand hours and days.
- Popular pickup and drop-off locations.
- Trends in ride volumes over time (hourly, daily, or monthly).
- Customer payment preferences.
- Average trip durations and anomalies in trip durations.

The analysis aims to improve operational efficiency, customer experience, and strategic planning.

#### 2. Sources

# References, Tools, and Resources Used:

- **Programming Language:** Python
- Libraries:
  - o pandas and numpy for data preprocessing and manipulation.
  - o matplotlib and seaborn for data visualizations.
  - o folium for location-based visualizations (maps).
- Dataset: A CSV file containing Uber ride details for India.
- **Development Environment:** Python Environment(venv), Streamlit, Colab.

# 3. Steps Taken

## **Step-by-Step Explanation:**

# 1. Data Collection and Preparation:

- Loaded the provided CSV dataset containing details of Uber rides across India.
- o Verified the integrity of the dataset by checking for missing or incorrect values.
- Standardized column names and converted timestamp fields to a datetime format.

# 2. Data Preprocessing:

- Extracted relevant features such as hour, day of the week, and month from pickup timestamps.
- Calculated trip durations in minutes from pickup and drop-off timestamps.
- o Filtered invalid or negative trip durations to ensure data quality.

#### 3. Analysis:

- Peak Demand Analysis: Identified the busiest hours of the day and busiest days of the week for rides.
- High-Demand Zones: Determined the most popular pickup and drop-off locations using latitude and longitude data.
- **Payment Preferences:** Analyzed payment methods (cash, card, digital wallet) to identify trends.
- o **Trip Duration Insights:** Calculated average trip durations for different times of the day and flagged anomalies.

#### 4. Visualization:

- Created bar charts for peak hours and payment distribution.
- o Generated heatmaps to visualize high-demand zones using folium.

#### 5. Output Generation:

- o Saved visualizations in the output folder as PNG files.
- o Exported the heatmap as an HTML file for interactive exploration.

## 4. Code

#### **Key Portions of Code:**

## 1. Data Preprocessing:

```
import pandas as pd

def preprocess_data(file_path):
    df = pd.read_csv(file_path)
    df['pickup_time'] = pd.to_datetime(df['Pickup Timestamp'],
    errors='coerce')
    df['dropoff_time'] = pd.to_datetime(df['Dropoff Timestamp'],
    errors='coerce')
    df['trip_duration'] = (df['dropoff_time'] -
    df['pickup_time']).dt.total_seconds() / 60
    df['hour'] = df['pickup_time'].dt.hour
    df['day_of_week'] = df['pickup_time'].dt.day_name()
    return df
```

## 2. Peak Hours Analysis:

```
def peak_hours_analysis(df):
    return df.groupby('hour').size()
```

## 3. High-Demand Zones:

```
def high_demand_zones_analysis(df):
    return df.groupby(['Pickup Latitude', 'Pickup
Longitude']).size().sort_values(ascending=False).head(10)
```

#### 4. Visualization:

```
import matplotlib.pyplot as plt
import seaborn as sns

def plot_peak_hours(peak_hours, output_path):
    plt.figure(figsize=(10, 6))
    sns.barplot(x=peak_hours.index, y=peak_hours.values, palette="viridis")
    plt.title("Peak Hours for Uber Rides")
    plt.xlabel("Hour of Day")
    plt.ylabel("Number of Rides")
    plt.savefig(output_path)
```

## 5. Results

# **Summary of Outcomes:**

#### 1. Peak Hours:

o The busiest hours for Uber rides were between [Insert Hours].

#### 2. High-Demand Zones:

• The most popular pickup location was latitude [Insert Latitude] and longitude [Insert Longitude].

## 3. Payment Preferences:

 Digital wallets accounted for [Insert Percentage] of all rides, indicating a trend towards cashless transactions.

#### 4. Trip Duration:

- o The average trip duration was [Insert Duration] minutes.
- o Trips with unusually long durations were flagged for further investigation.

## 6. Conclusion

The analysis provided valuable insights into ride patterns, high-demand zones, and payment trends for Uber rides in India. These findings can help Uber optimize operations, improve customer satisfaction, and strategize for future growth.