

# TOOL USED:

- ❖ EXCEL for cleaning the data and creating the dashboard.
- SQL insights for Visualization.
- EDA-PYHTON (Pandas, NumPy, Matplotlib)

#### PROJECT OVERVIEW:

- ❖ ANALYZE UBER RIDE REQUEST DATA TO DETECT SUPPLY-DEMAND ISSUES.
- ❖ DATASET: RIDE LOGS WITH TIMESTAMPS, PICKUP POINT, AND TRIP STATUS.
- ❖ TOOLS: EXCEL (DASHBOARDS AND PIVOT TABLES), SQL (INSIGHTS), PYTHON (PANDAS, NUMPY, MATPLOTLIB, SEABORN).

#### DATA EXTRACTION:

- Standard python libraries used are:
  - NumPy 1.19.2,
  - Pandas 1.2.3,
  - Matplotlib 3.3.4
  - Seaborn 0.11.1
- Loading the dataset from csv to pandas data frame.

```
uberdata = pd.read_csv("Uber Request Data.csv")
```

Checking different columns data types.

```
uberdata.info()
```

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

Request id	Pickup point	Driver id	Status	Request timestamp	Drop timestamp
619	Airport	1.0	Trip Completed	11/7/2016 11:51	11/7/2016 13:00
867	Airport	1.0	Trip Completed	11/7/2016 17:57	11/7/2016 18:47
1807	City	1.0	Trip Completed	12/7/2016 9:17	12/7/2016 9:58
2532	Airport	1.0	Trip Completed	12/7/2016 21:08	12/7/2016 22:03
3112	City	1.0	Trip Completed	13-07-2016 08:33:16	13-07-2016 09:25:47

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 6745 entries, 0 to 6744
Data columns (total 6 columns):
    Column
                        Non-Null Count Dtype
8 Request id 6745 non-null
1 Pickup point 6745 non-null
                                         int64
                                        object
2 Driver id
                        4095 non-null
                                        float64
                        6745 non-null
                                         object
4 Request timestamp 6745 non-null
                                         object
    Drop timestamp
                        2831 non-null
                                        object
dtypes: float64(1), int64(1), object(4)
```

memory usage: 316.3+ KB

#### DATA CLEANING:

- > Checking Null column values:
  - 'Drop timestamp' has 58 percent of null values.
  - 'Driver ID' has 39 percent of null values.
  - These entries are the rides where trip was never assigned to a driver.

```
pd.DataFrame(round((100*(uberdata.isnull().sum()/len(uberdata.index))),2))
```

Request id	0.00
Pickup point	0.00
Driver id	39.29
Status	0.00
Request timestamp	0.00
Drop timestamp	58.03

- Correcting data types of datetime columns:
  - Converting format of 'Request timestamp' and 'Drop timestamp' columns to datetime object.

```
uberdata['Request timestamp'] = pd.to_datetime(uberdata['Request timestamp'])
uberdata['Drop timestamp'] = pd.to_datetime(uberdata['Drop timestamp'])
uberdata.head(10)
```

Request id	Request timestamp	Drop timestamp
619	2016-11-07 11:51:00	2016-11-07 13:00:00
867	2016-11-07 17:57:00	2016-11-07 18:47:00
1807	2016-12-07 09:17:00	2016-12-07 09:58:00

#### FEATURE ENGINEERING:

#### > Addition of new columns:

- 'Request Hours': By extracting hours from 'Request Timestamp' object column.
- 'Drop Hours': By extracting hours from 'Drop Timestamp' object column.

```
uberdata['Request Hours'] = uberdata['Request timestamp'].apply(lambda x:x.hour)
uberdata['Drop Hours'] = uberdata['Drop timestamp'].apply(lambda x: x.hour)
uberdata.head(5)
```

Request id	Request timestamp	Drop timestamp	Request Hours	Drop Hours
619	2016-11-07 11:51:00	2016-11-07 13:00:00	31	13.0
867	2016-11-07 17:57:00	2016-11-07 18:47:00	17	18.0
1807	2016-12-07 09:17:00	2016-12-07 09:58:00	9	9.0
2532	2016-12-07 21:08:00	2016-12-07 22:03:00	21	22.0
3112	2016-07-13 08:33:16	2016-07-13 09:25:47	8	9.0

Dividing all requests into different time slots:

```
def determine_time_slot(x):
    if (x >=0 and x < 8):
        return "Early morning hours"
                                         #12am-7am
    elif (x >= 8 \text{ and } x < 12):
        return "Peak morning hours"
                                          #Sam-11am
    elif (x >= 12 and x < 17):
        return "Noon hours"
                                         #12pm-4pm
    elif (x >= 17 \text{ and } x < 21):
        return "Evening hours"
                                         #5pm-8pm
    elif (x >= 21):
        return "Night hours"
                                          #9pm onwards
uberdata['Request Time Slot'] = uberdata['Request Hours'].apply(determine time slot)
uberdata[['Request id', 'Pickup point', 'Request Time Slot']].head(5)
```

Request Time Slot	Pickup point	Request id	
Peak morning hours	Airport	619	
Evening hours	Airport	867	
Peak morning hours	City	1807	
Night hours	Airport	2532	
Peak morning hours	City	3112	

## PLOTS AND OBSERVATIONS:

Most problematic pickup point status wise:

```
sns.countplot(x="Status",hue="Pickup point",data = uberdata)
plt.title('Status vs Airport/City trips count')
plt.ylabel('Airport/City trip count')
```

Table 1. Distribution of airport/City trips status wise

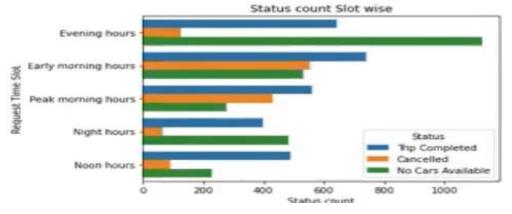
	Pickup Point		
Status	Airport	City	
No Cars Available	1600-1700	900	
Cancelled	175-200	1400-1500	
Trip Completed	1300-1350	1500	

Most problematic time slots where rides were unsuccessful:

Table 1. Problematic time slots

	Request Time Slot		
Status	Evening	Early Morning	
No Cars Available	1100-1200	500-600	
Cancelled	100	550	
Trip Completed	700	750-775	





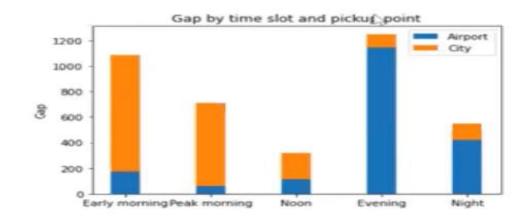
## SUMMARY OF EDA INSIGHTS:

 Gap for airport pickup point is maximum in evening.

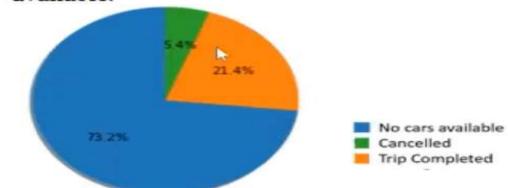
Request Time Slot	Demand_From_Airport	Supply_To_Airport	Gap_From_Airport
Evening	1457	312	1145

 Gap for city pickup point is maximum in early morning.

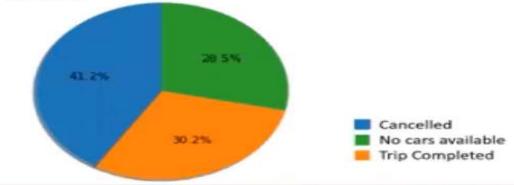
Request Time Slot	Demand_From_City	Supply_To_City	Gap_From_City	
Early morning	1310	396	914	



 For 73 % of total airport requests cars were not available.



 41% of city requests were cancelled by the drivers.



# SUMMARY OF SQL INSIGHTS:

Uber\_SQL\_Insights.sql × C: > Users > hp > Desktop > Uber\_SQL\_Insights.sql -- Total Requests by Pickup Point and Status SELECT pickup\_point, status, COUNT(\*) AS total\_requests FROM uber\_requests GROUP By pickup point, status ORDER BY pickup point, total requests DESC; -- Hourly Demand Trend SELECT 9 HOUR(request\_timestamp) AS hour of day, 1.63 COUNT(\*) AS total\_requests FROM uber requests 11 12 GROUP BY hour of day 13 ORDER BY hour\_of\_day; 14 2.55 -- Requests with No Cars Available SELECT COUNT(\*) As no\_car\_requests 11.05 17 FROM uber requests WHERE status - 'No Cars Available'; 19 -- Supply vs Demand Per Hour 20 21 SELECT 22 HOUR(request\_timestamp) AS hour, 23 COUNT(\*) As total requests, SUM(CASE WHEN status - 'Trip Completed' THEN 1 ELSE 0 END) AS completed\_trips. 24 SUM(CASE WHEN status I= 'Trip Completed' THEN 1 ELSE 0 END) AS unfulfilled\_requests 25 FROM uber requests 20 GROUP BY hour 27 28 ORDER BY hour; 29 -- Driver-wise Trip Completion Count 30 SELECT driver\_id, COUNT(\*) AS completed\_trips 31 32 FROM uber\_requests WHERE status - 'Trip Completed' 34 GROUP BY driver id 34 ORDER BY completed\_trips DESC; 36 -- Cancellation Rates by Pickup Point 37 SELECT 38 39 pickup point. 40 COUNT(\*) AS total requests, SUM(CASE WHEN status = 'Cancelled' THEN 1 ELSE 0 END) AS cancelled. 41 ROUND(SUM(CASE WHEN status - 'Cancelled' THEN 1 ELSE Ø END) \* 100.0 / COUNT(\*), 2) AS cancel rate\_percent 42 43 FROM uber requests GROUP BY pickup point; 44 -- Peak Hours for No Cars Available 46 47 48 HOUR(request\_timestamp) AS hour, COUNT(\*) As no\_car\_count 49 50 FROM uber\_requests 51 WHERE status = 'No Cars Available' GROUP BY hour ORDER BY no car count DESC LIMIT 5:

# SUMMARY OF EXCEL INSIGHTS:



# SOLUTION FOR THE SUPPLY DEMAND GAP:

- For the trips in the early morning, drivers can be incentivized to make those trips.
  - 1. Uber can pay for the gas mileage of drivers to come back to the city without a rider.
  - 2. Uber can increase the demand at the airport to reduce idle time by increased marketing and price cuts for the passengers.
  - 3. Uber can request a feedback from drivers to understand reasons behind ride cancellation.
- For the trips in evening, some of the ways are:
  - 1. Uber can also pay drivers to come without a passenger to the airport.
  - 2. Another innovative way can be to encourage passengers to pool the ride with others so that lesser number of cars can serve more passengers.

