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**Tools Used:** Excel, SQL (SQLite), Python (Pandas, Matplotlib, Seaborn)

## 1. Introduction

The purpose of this project is to analyze Uber request data and identify the major causes of the supply–demand gap.

Using **Excel, SQL, and Python**, data was cleaned, visualized, and analyzed to uncover trends in trip completions, cancellations, and car unavailability.

The dataset contains **6,745 Uber ride requests**, including details such as pickup point, request timestamp, drop timestamp, driver ID, and status of each trip.

## 2. Objectives

- To understand the **supply–demand gap** in Uber ride requests.
- To identify **time periods** and **pickup points** with the largest number of unfulfilled requests.
- To provide **data-driven recommendations** for improving driver availability and reducing cancellations.

## 3. Dataset Overview

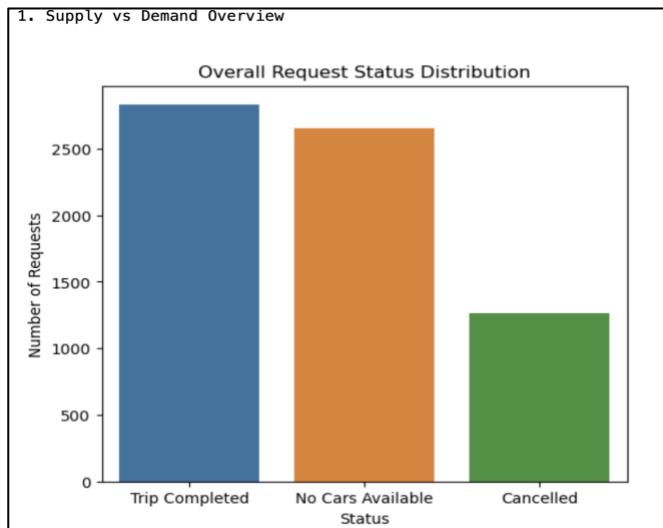
<u>Attribute</u>	<u>Description</u>
<b>Request id</b>	Unique identifier for each trip request.
<b>Pickup point</b>	Either "City" or "Airport".
<b>Driver id</b>	Unique identifier for assigned driver (may be null).
<b>Status</b>	Trip Completed, Cancelled, or No Cars Available.
<b>Request timestamp</b>	Time the customer requested a ride.
<b>Drop timestamp</b>	Time the trip was completed (if applicable).

## 4. Data Cleaning and Preparation

- **Excel** was used for initial cleaning:
  - Removed blank cells and trimmed extra spaces.
  - Standardized text (capitalization and spelling).
  - Converted timestamps into date/time format.
- **SQL (SQLite)** was used to summarize trip statuses and compute supply–demand ratios.
- **Python (Pandas)** was used for deeper exploratory data analysis (EDA) and visualization.

## 5. Visual Analysis and Findings

## 5.1 Overall Request Status Distribution



**Fig. 1: Overall Requests by Status**

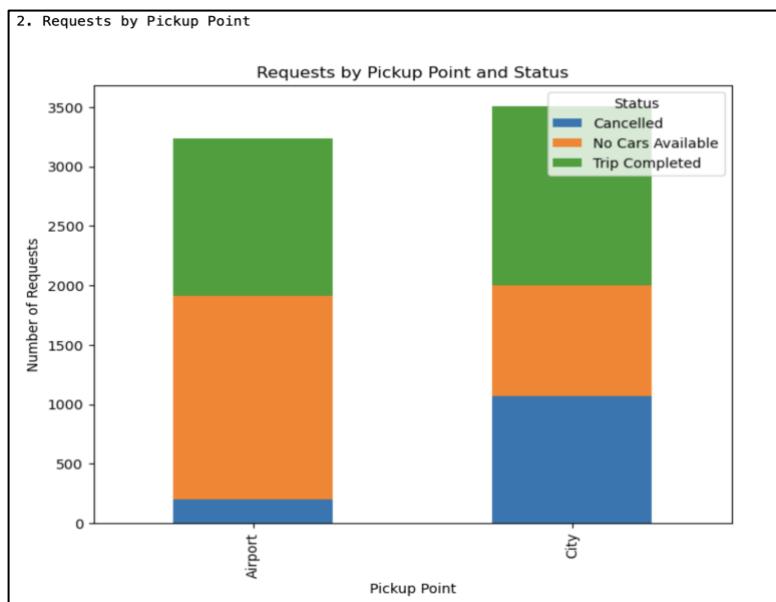
### Insight:

Out of 6,745 total requests:

- **2,831** were completed
- **1,264** were cancelled
- **2,650** had no cars available

This indicates that **over 58% of total requests were not fulfilled**, showing a clear supply–demand imbalance.

## 5.2 Requests by Pickup Point

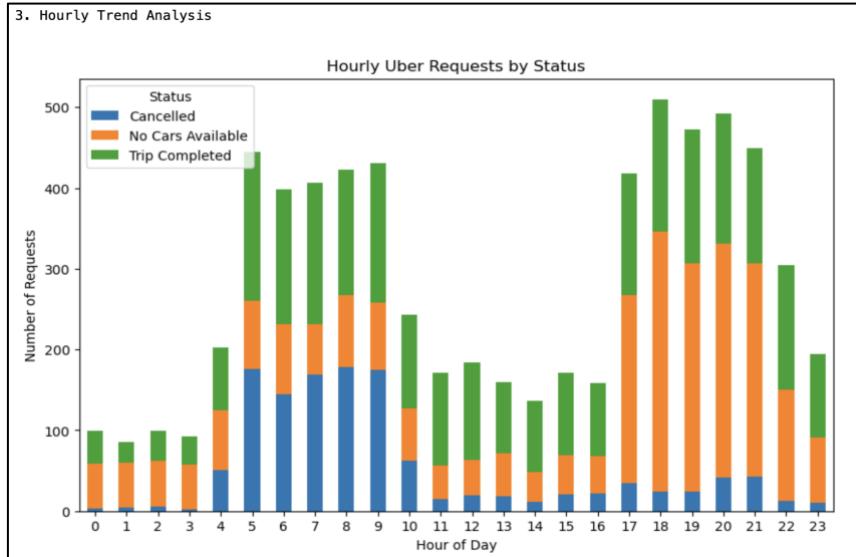


**Fig. 2: Requests by Pickup Point**

### Insight:

- At the **Airport**, more requests failed due to *No Cars Available* (~59%).
  - In the **City**, most unfulfilled requests were *Cancelled* by drivers (~57%).
- This suggests that **car availability is a bigger issue at the Airport**, while **driver cancellations dominate in the City**.

### 5.3 Hourly Trend Analysis



**Fig. 3 : Hourly Supply–Demand Pattern**

#### Insight:

- Demand spikes between **5 AM–9 AM** (Airport pickups to City).
- Another rise is seen around **10 PM–12 AM** (City to Airport).
- During these times, the number of completed trips drops significantly, showing **severe driver shortage during early morning and night hours.**

## 6. SQL Analysis Summary

#### Queries Run:

1. Total Requests by Status
2. Requests by Pickup Point
3. Hourly Supply–Demand Summary
4. Supply–Demand Gap Percentage.

#### Key SQL Results:

Pickup Point	Total Requests	Completed	Gap (%)
Airport	3,238	1,327	59.02%
City	3,507	1,504	57.11%

Part 1: SQL Analysis				
<b>1. Total Requests by Status:</b>				
	Status	Total_Requests		
0	Cancelled	1264		
1	No Cars Available	2650		
2	Trip Completed	2831		
<b>2. Hourly Supply vs Demand Gap:</b>				
	Hour	Supply	Demand_Gap	
0	None	2831	3914	
<b>3. Requests by Pickup Point:</b>				
	Pickup point	Completed	Cancelled	
0	Airport	1327	198	
1	City	1504	1066	
<b>4. Supply-Demand Gap Percentage:</b>				
	Pickup point	Total_Requests	Completed	Gap_Percentage
0	Airport	3238	1327	59.02
1	City	3507	1504	57.11

**Fig. 4: Results of SQL Analysis**

### Insight:

The Airport shows a slightly higher supply–demand gap compared to the City, mainly due to lower driver availability during peak hours.

## 7. Python EDA Summary

### Techniques Used:

- pandas - data manipulation.
- matplotlib & seaborn – visualization.
- datetime features - analyze hourly trends.

### Core Insights:

- Supply–demand gap  $\approx 58\text{--}60\%$
- Early morning and late-night requests show highest unavailability
- Airport requests face driver shortage; City requests face higher cancellations

## 8. Recommendations

1. **Increase driver incentives** for early morning (4–8 AM) and late-night (10 PM–1 AM) shifts.
2. **Improve coordination** for Airport pickups, possibly with pre-scheduled driver allocation.
3. **Introduce penalty or warning system** for frequent driver cancellations.
4. **Monitor demand patterns** weekly to adjust supply dynamically.

## 9. Conclusion

The analysis clearly highlights a **major supply–demand gap** in Uber operations, especially during off-peak hours and Airport pickups.

By optimizing driver allocation and encouraging participation during low-supply periods, Uber can significantly reduce unfulfilled requests and improve customer satisfaction.