



# Data Analyst Project

By Parth Shethia

## OLA DATA ANALYST PROJECT REPORT

**Prepared By:** Parth Shethia

**Role:** Data Analyst / Business Analyst

**Tools Used:** Python (Pandas), SQL, Power BI

### ChatGPT Prompt to Create Data

Please create a spreadsheet with 1 lac rows, for Bengaluru city. Give the following columns.

The data will be for 1 month. use the following column -

1. Date
2. Time
3. Booking ID
4. Booking Status
5. Customer ID
6. Vehicle Type
  - Auto
  - Prime Plus
  - Prime Sedan
  - Mini
  - Bike
  - eBike
  - Prime SUV
7. Pickup Location (Create dummy location points Take any 50 areas from Bangalore)
8. Drop Location (Take from dummy pickup locations)
9. Avg VTAT (Time taken to arrive at the vehicle)
10. Avg CTAT (Time taken to arrive the Customer)
11. Cancelled Rides by Customer
12. Reason for cancelling by Customer
  - Driver is not moving towards pickup location
  - Driver asked to cancel
  - AC is not working (Only for 4-wheelers)
  - Change of plans

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- Wrong Address
- 13. Cancelled Rides by Driver
  - Personal & Car related issues
  - Customer related issue
  - The customer was coughing/sick
  - More than permitted people in there
- 14. Incomplete Rides
- 15. Incomplete Rides Reason
  - Customer Demand
  - Vehicle Breakdown
  - Other Issue
- 16. Booking Value
- 17. Ride Distance
- 18. Driver Ratings
- 19. Customer Rating

Keep the overall booking status success for this data at 62%. If the booking status is successful, then only fare charge ratings, average VTAT, average CTAT, and other data will be there.

Make sure orders cancelled by customers should not be more than 7%

Make sure orders cancelled drivers should not be more than 18%

Also, increase the number of orders on weekends and match days. Keep match day by using the following dates.

keep incomplete rides less than 6%

Keep order value high on weekends

in Food Category keep around 67 Indian keep order ID with  
10 digits starting with CNR and then digits keep orders  
under 500 value 70% keep orders above 500 value 28%  
keep remaining orders above 1000

## SQL Questions:

1. Retrieve all successful bookings:
2. Find the average ride distance for each vehicle type:
3. Get the total number of cancelled rides by customers:
4. List the top 5 customers who booked the highest number of rides:
5. Get the number of rides cancelled by drivers due to personal and car-related issues:
6. Find the maximum and minimum driver ratings for Prime Sedan bookings:
7. Retrieve all rides where payment was made using UPI:
8. Find the average customer rating per vehicle type:
9. Calculate the total booking value of rides completed successfully:
10. List all incomplete rides along with the reason:

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## Power BI Questions:

1. Ride Volume Over Time
2. Booking Status Breakdown
3. Top 5 Vehicle Types by Ride Distance
4. Average Customer Ratings by Vehicle Type
5. cancelled Rides Reasons
6. Revenue by Payment Method
7. Top 5 Customers by Total Booking Value
8. Ride Distance Distribution Per Day
9. Driver Ratings Distribution
10. Customer vs. Driver Ratings

## Data Columns

- |                    |                                 |
|--------------------|---------------------------------|
| 1. Date            | 10. C_TAT                       |
| 2. Time            | 11. cancelled_Rides_by_Customer |
| 3. Booking_ID      | 12. cancelled_Rides_by_Driver   |
| 4. Booking_Status  | 13. Incomplete_Rides            |
| 5. Customer_ID     | 14. Incomplete_Rides_Reason     |
| 6. Vehicle_Type    | 15. Booking_Value               |
| 7. Pickup_Location | 16. Payment_Method              |
| 8. Drop_Location   | 17. Ride_Distance               |
| 9. V_TAT           | 18. Driver_Ratings              |
|                    | 19. Customer_Rating             |

## Data Characteristics

- **Total Rows:** 100,000 (1 lakh)
- **City:** Bengaluru
- **Duration:** 1 Month
- **Data Nature:** Synthetic (realistic business constraints applied)

## Business Constraints Applied

- 62% successful bookings
- Customer cancellations < 7%
- Driver cancellations < 18%
- Incomplete rides < 6%
- Higher booking volume & value on weekends and match days
- 70% of orders under ₹500

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- 28% orders between ₹500–₹1000
- Remaining orders above ₹1000
- Booking IDs start with **CNR** followed by digits

## Data Cleaning & Preparation (Python – Pandas)

BEFORE CLEANING:-

```
[4]: df.head()
df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 103024 entries, 0 to 103023
Data columns (total 20 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   Date              103024 non-null   datetime64[ns]
 1   Time              103024 non-null   object 
 2   Booking_ID        103024 non-null   object 
 3   Booking_Status    103024 non-null   object 
 4   Customer_ID       103024 non-null   object 
 5   Vehicle_Type      103024 non-null   object 
 6   Pickup_Location   103024 non-null   object 
 7   Drop_Location     103024 non-null   object 
 8   V_TAT             63967 non-null   float64
 9   C_TAT             63967 non-null   float64
 10  Canceled_Rides_by_Customer 10499 non-null   object 
 11  Canceled_Rides_by_Driver   18434 non-null   object 
 12  Incomplete_Rides      63967 non-null   object 
 13  Incomplete_Rides_Reason 3926 non-null   object 
 14  Booking_Value        103024 non-null   int64 
 15  Payment_Method       63967 non-null   object 
 16  Ride_Distance        103024 non-null   int64 
 17  Driver_Ratings       63967 non-null   float64
 18  Customer_Rating      63967 non-null   float64
 19  Vehicle_Images       103024 non-null   object 
dtypes: datetime64[ns](1), float64(4), int64(2), object(13)
memory usage: 15.7+ MB
```

AFTER CLEANING:-

```
[12]: df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 103024 entries, 0 to 103023
Data columns (total 20 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   Booking_ID        103024 non-null   object 
 1   Booking_Status    103024 non-null   object 
 2   Customer_ID       103024 non-null   object 
 3   Vehicle_Type      103024 non-null   object 
 4   Pickup_Location   103024 non-null   object 
 5   Drop_Location     103024 non-null   object 
 6   V_TAT             63967 non-null   float64
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 11  Incomplete_Rides_Reason 3926 non-null   object 
 12  Booking_Value        103024 non-null   int64 
 13  Payment_Method       63967 non-null   object 
 14  Ride_Distance        103024 non-null   int64 
 15  Driver_Ratings       63967 non-null   float64
 16  Customer_Rating      63967 non-null   float64
 17  booking_datetime    103024 non-null   datetime64[ns]
 18  ride_outcome        103024 non-null   object 
 19  cancelled_by         28933 non-null   object 
dtypes: datetime64[ns](1), float64(4), int64(2), object(13)
memory usage: 15.7+ MB
```

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## Cleaning Steps Performed

- **Loaded Excel data** and inspected structure, datatypes, and missing values.
- **Removed irrelevant columns** (Vehicle\_Images).
- **Converted text nulls** ("null", #NAME?) into real missing values.
- **Merged Date and Time** into a single booking\_datetime column.
- **Standardized booking outcomes** (Success, Canceled, Failed) and identified who canceled.
- **Unified cancellation reasons** into one column and dropped redundant fields.
- **Converted numeric columns** to proper numeric types.
- **Applied business rules:** ratings kept only for successful rides.
- **Standardized categorical fields** (case, spacing).
- **Renamed columns** to MySQL-friendly format.
- **Exported cleaned data** as CSV and imported into MySQL.
- **Removed hidden carriage-return characters (CHAR(13))** and whitespace from text fields.
-  **Outcome:**

A clean, structured dataset ready for SQL analysis and Power BI visualization.

## SQL Analysis & Data Modeling

### Database Setup

```
CREATE DATABASE Ola;
USE Ola;
```

### SQL Views Created

- Successful\_Bookings
- ride\_distance\_for\_each\_vehicle
- cancelled\_rides\_by\_customers
- Top\_5\_Customers
- Rides\_cancelled\_by\_Drivers\_P\_C\_Issues

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- Max\_Min\_Driver\_Rating
- UPI\_Payment
- AVG\_Cust\_Rating
- total\_successful\_ride\_value
- Incomplete\_Rides\_Reason

## 📌 Business Benefit:

SQL views improve query performance and enable seamless integration with Power BI.

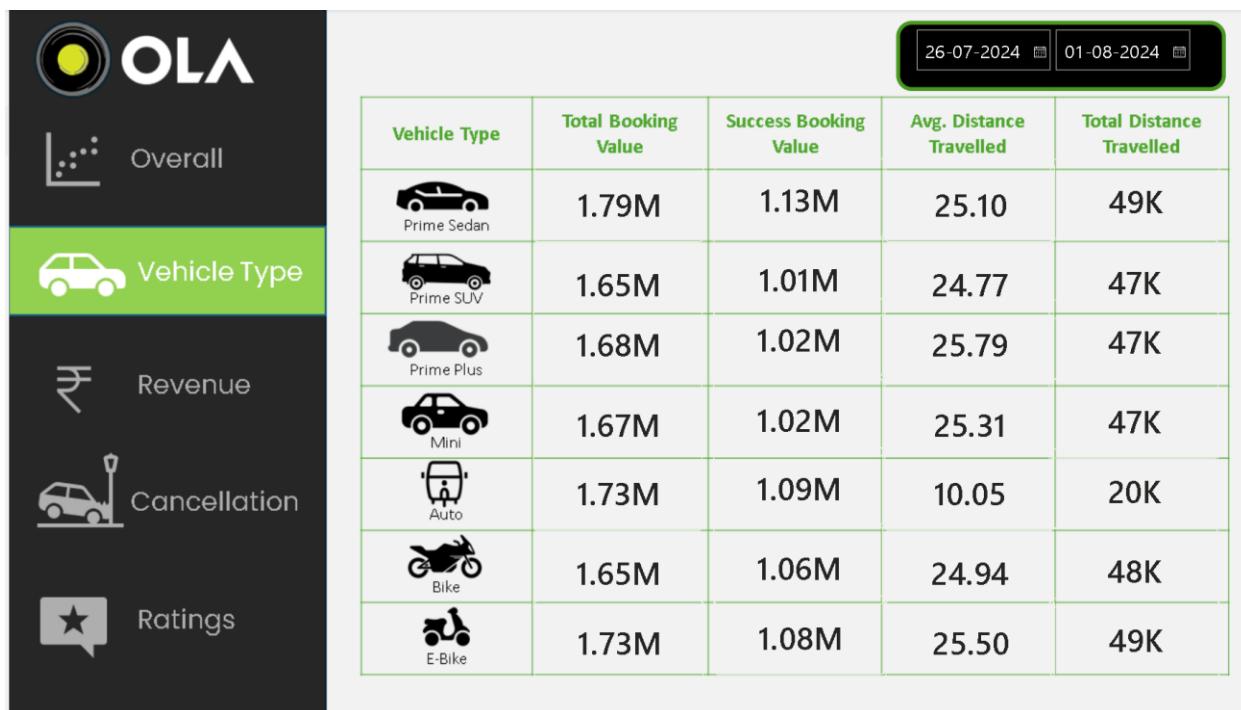
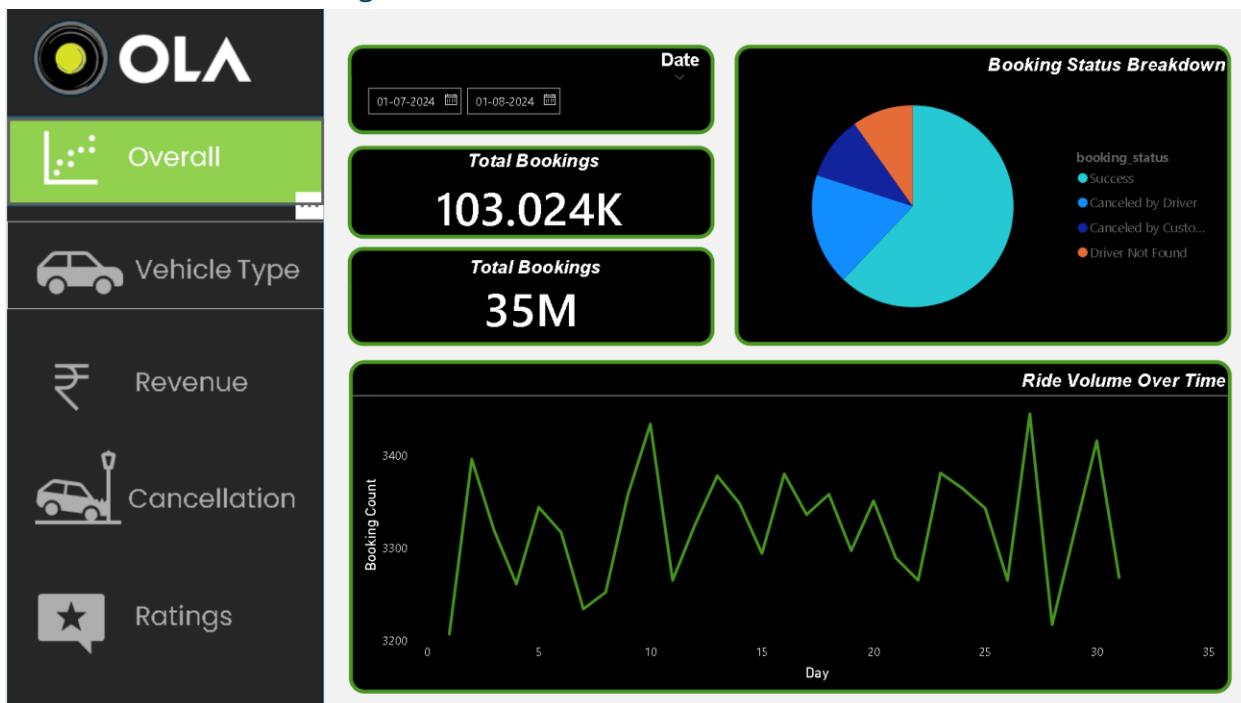
## SQL Questions & Insights

### Key Queries Answered

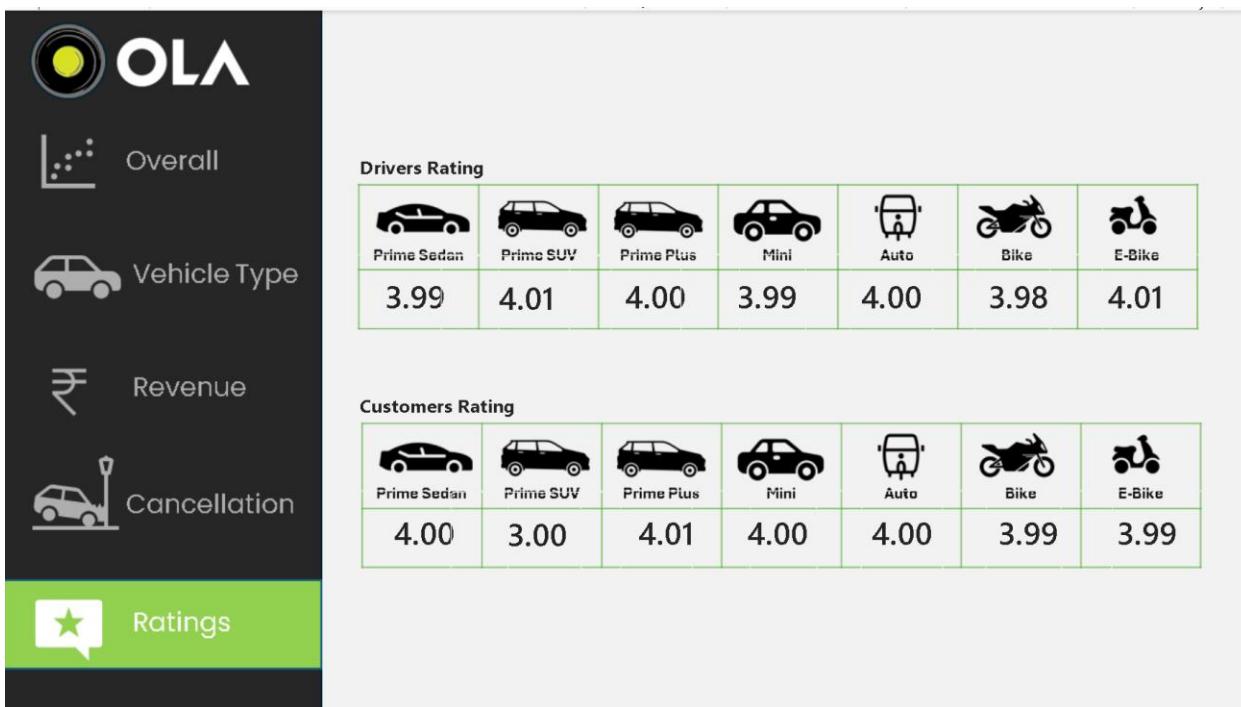
1. Successful bookings list
2. Average ride distance per vehicle type
3. Total customer cancellations
4. Top 5 customers by number of rides
5. Driver cancellations due to personal/car issues
6. Max & min driver ratings for Prime Sedan
7. UPI-based payments
8. Average customer ratings by vehicle
9. Total revenue from successful rides
10. Incomplete rides with reasons

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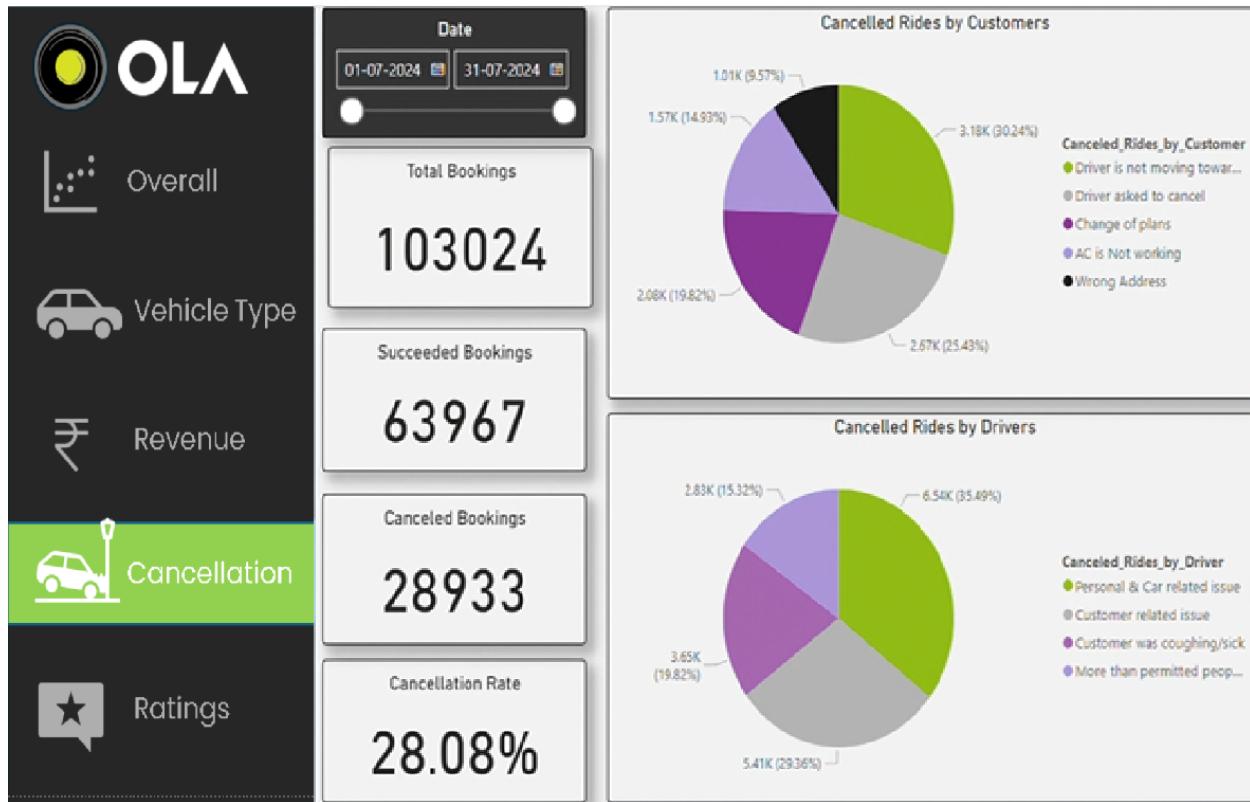
## Power BI Dashboard Design



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The Power BI report is divided into **five analytical sections** for clear business storytelling.

## Key Business Insights

### 1. Overall Booking Performance

- Total bookings during the period: **103,024**
- Successful bookings: **63,967**
- Cancelled bookings: **28,933**
- Overall cancellation rate: **28.08%**
- Nearly **1 out of every 3 bookings** does not convert into a completed ride, directly impacting revenue and customer experience.
- Ride volume across days remains relatively stable with minor peaks, indicating predictable demand patterns.

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## 2. Cancellation Analysis

### Customer-Side Cancellations

- Major reasons include:
  - Driver not moving towards pickup location
  - Driver requesting customer to cancel
  - Change of travel plans
  - Wrong pickup address
- These reasons indicate **operational execution issues** rather than lack of customer intent.

### Driver-Side Cancellations

- Primary reasons include:
  - Personal and vehicle-related issues
  - Customer-related issues
  - Passenger policy violations (exceeding permitted passengers)
- Driver reliability and availability remain critical contributors to cancellations.

## 3. Revenue Insights

- Total booking value generated: **₹35 million**
- Cash is the dominant payment method, followed by UPI.
- Credit and debit card usage is minimal.
- Heavy cash reliance increases payment handling risk and reduces operational efficiency.

## 4. Vehicle Type Performance

- Total booking value by vehicle type:
  - Prime Sedan: **₹1.79M**
  - Prime SUV: **₹1.65M**
  - Prime Plus: **₹1.68M**
  - Mini: **₹1.67M**
  - Auto: **₹1.73M**
  - Bike: **₹1.65M**
  - E-Bike: **₹1.73M**
- Average distance travelled:
  - Auto: **~10 km** (short-distance trips)
  - Other vehicle types: **24–26 km** (medium to long-distance trips)
- Different vehicle types cater to distinct travel needs and pricing strategies should reflect this difference.

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## 5. Customer and Driver Ratings

- Average driver ratings across vehicle categories: **3.98 – 4.01**
- Average customer ratings across vehicle categories: **3.99 – 4.01**
- Ratings indicate acceptable service quality but highlight scope for improvement to achieve a premium experience.

## 6. Customer Behavior and Ride Distance

- A small group of customers contributes a disproportionately high share of total revenue.
- Revenue concentration is higher in short and mid-distance rides, with fewer long-distance bookings.
- Retention of high-value customers is critical for long-term growth.

## 7. Key Business Recommendations

- Reduce cancellation rate by:
  - Tracking driver movement immediately after ride acceptance
  - Penalizing drivers who request customers to cancel
  - Automatically reallocating drivers when delays are detected
- Promote digital payments by incentivizing UPI and reducing cash dependency.
- Optimize vehicle-type strategy by:
  - Promoting autos for short urban trips
  - Offering premium pricing and bundles for longer rides in Prime categories
- Improve service quality by focusing on driver training and pickup punctuality to raise average ratings beyond **4.5**.
- Introduce loyalty programs to retain high-value repeat customers and stabilize revenue.

## 8. Summary

- The platform demonstrates **strong demand and revenue potential**, but operational inefficiencies—especially cancellations—limit overall performance.
- Addressing driver reliability, payment digitization, and customer retention can significantly improve profitability and customer satisfaction without increasing acquisition costs.

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