# Kabhi Khushi Kabhie Cab

#### **Presented By:**

Ankit Kumar Noyonica Chatterjee Surabhi Tannu

### Problem Statement

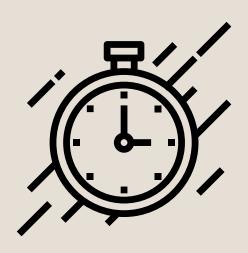
Addressing the uncertainties in cab pricing, availability, and traffic-related delays to enhance users' overall transportation experience

# Solutions



### Predictive Pricing Model

Predicting minimum cab ride prices based on routes and times for cost-effective traveler choices



### Cab Availability Forecasting

Estimating cab waiting times to reduce uncertainty and enhance user convenience



### Traffic Condition Prediction

Forecasting traffic congestion to aid efficient travel planning and reduce delays

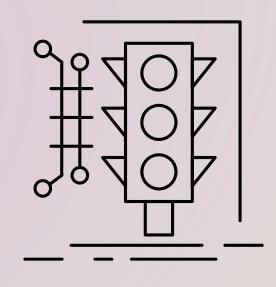
# Potential Applications and Impact



### **Enhanced User Experience**

Would enhance convenience and cost-efficiency for cab users, leading to increased customer satisfaction

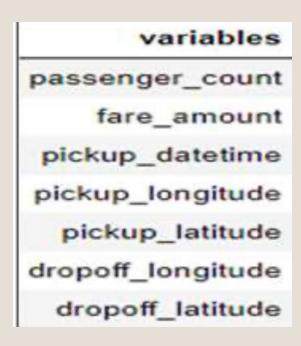


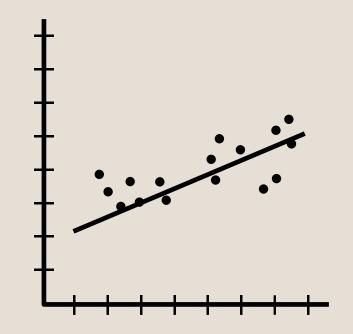


### Smart Traffic Prediction

Traffic prediction data can aid in time efficient time experience.

### Literature Review





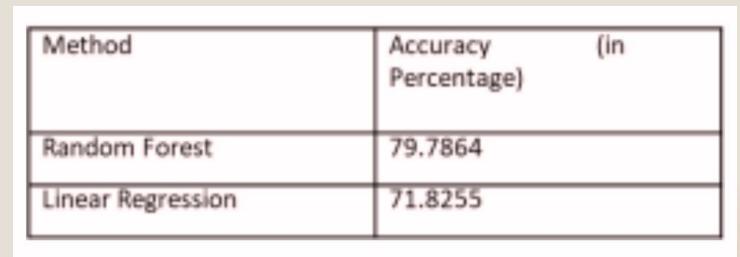


Fig 9. Accuracy

Real\_time prediction of Cab Fair using Machine Learning (Dr. A. Pravin, et al.)

The company uses historical data and machine-learning models combined to predict ETA.

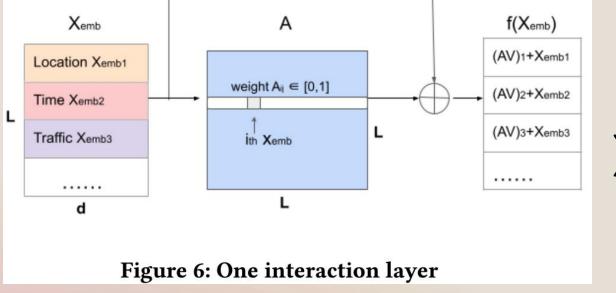
It uses two layered neural network to predict ETA.

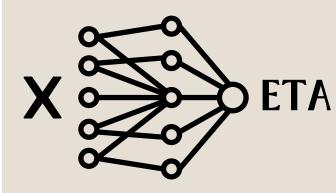
$$A_{ij} = \frac{\exp(QK^T/\sqrt{d})_{i,j}}{\sum_{j=1}^{L} \exp(QK^T/\sqrt{d})_{i,j}} \quad Q = X_{emb}W_q$$
$$K = X_{emb}W_k$$

$$Q = X_{emb}W_{q}$$

$$K = X_{emb}W_{k}$$

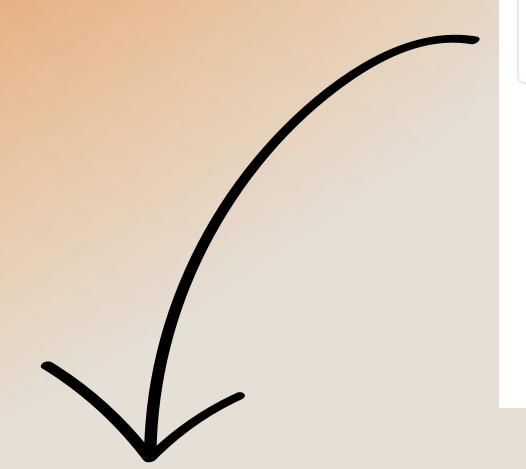
$$V = X_{emb}W_{v}$$

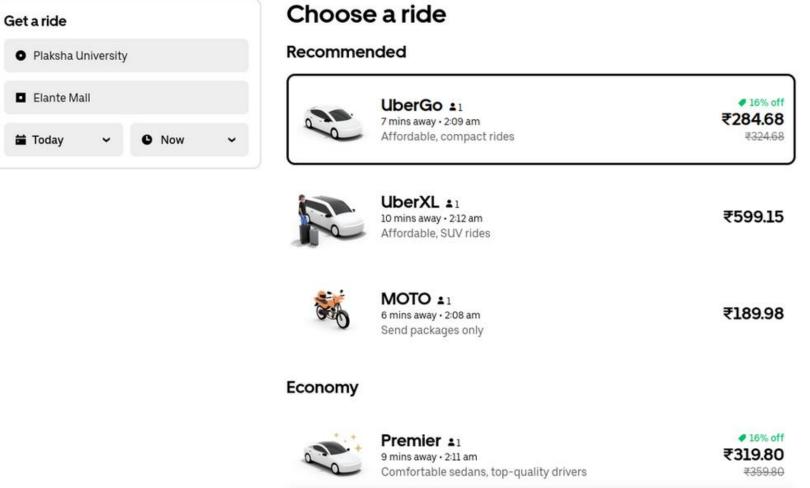




Deepreta: An ETA Post-processing System at Scale (Uber Technologies Inc.)

Scraping the Web





Chandigarh

From Plaksha University >

Mahendra Chaudhary Zoological Park Chhat. To Elante Mall >

Kishanpura

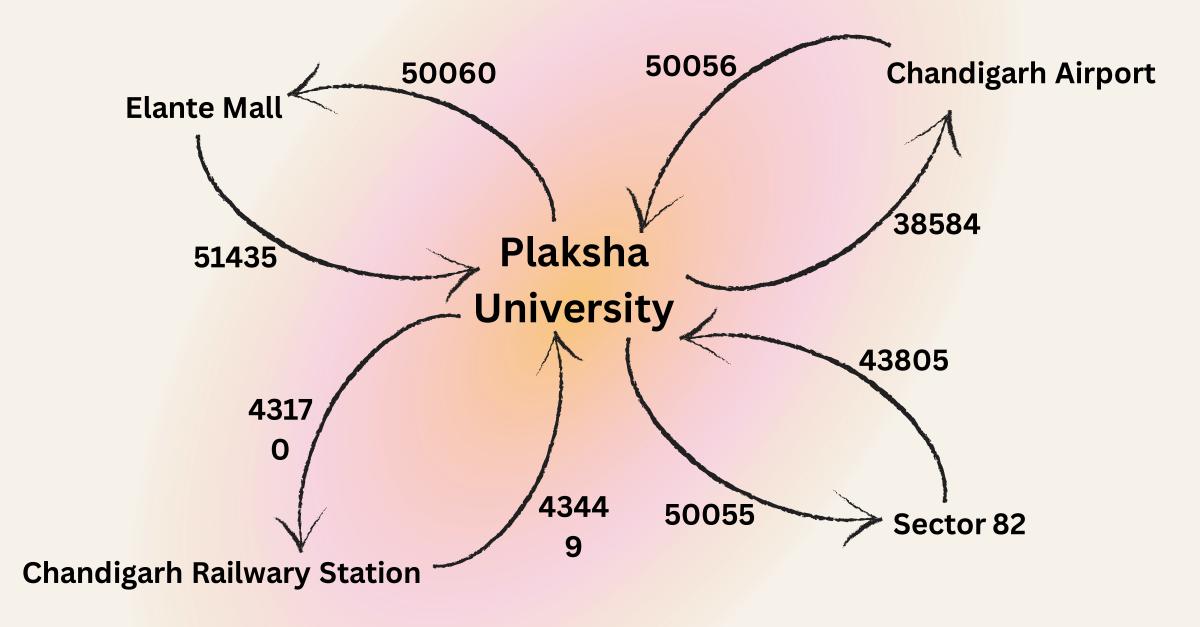
|        | Cab_Type       | Cab_Price | Arrival_Time      | Pick_Up            | Destination                                 | <b>Current Time</b>  | <b>Current Day</b> |
|--------|----------------|-----------|-------------------|--------------------|---|----------------------|--------------------|
| 330818 | UberXLPerson1  | ₹232.09   | 7 mins away•20:18 | Plaksha University | Sector 82                                   | 03/10/2023, 20:01:04 | Tuesday            |
| 188085 | UberXLPerson1  | ₹609.93   | 9 mins away•21:02 | Plaksha University | Railway Station Chandigarh (Panchkula Side) | 20/09/2023, 20:13:49 | Wednesday          |
| 44997  | PremierPerson1 | ₹202.01   | Unavailable       | Plaksha University | Chandigarh Airport Departure Terminal (IXC) | 12/09/2023, 0:17:05  | Tuesday            |
| 218293 | UberGoPerson1  | ₹174.22   | 5 mins away•15:07 | Sector 82          | Plaksha University                          | 22/09/2023, 14:52:05 | Friday             |
| 231396 | PremierPerson1 | ₹316.27   | 8 mins away•7:06  | Plaksha University | Railway Station Chandigarh (Panchkula Side) | 23/09/2023, 06:28:19 | Saturday           |

### Let's talk Data

We have collected a total of 3,75,348 datapoints across 7 attributes

#### **Attributes**

| Cab_Type      | object |
|---------------|--------|
| Cab_Price     | object |
| Arrival_Time  | object |
| Pick_Up       | object |
| Destination   | object |
| Current Time  | object |
| Current Day   | object |
| dtype: object |        |



# Data Preprocessing

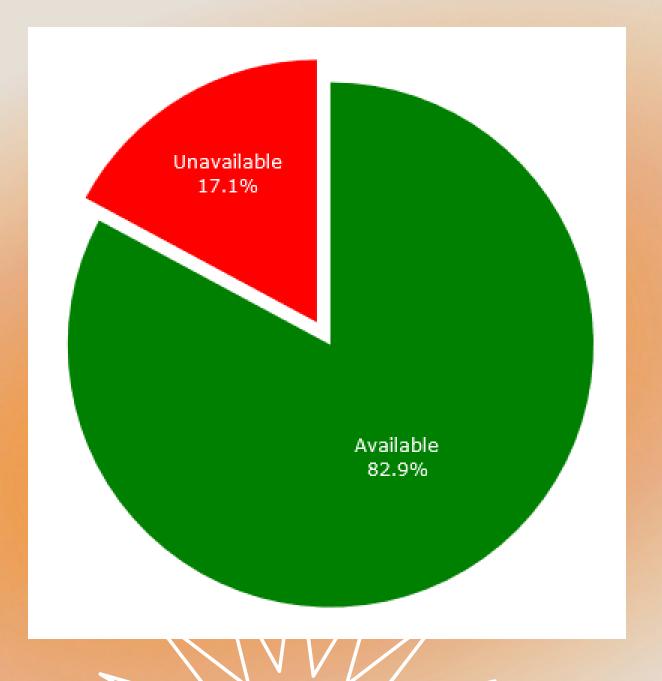
|   | Cab_Type   | Cab_Price | Arrival_Time               | Pick_Up     | Destination        | Current Time         | <b>Current Day</b> |
|---|------------|-----------|----------------------------|-------------|--------------------|----------------------|--------------------|
| 0 | Hatchbacks | ₹280.96   | 2 mins away•21:16 drop-off | Elante Mall | Plaksha University | 09/09/2023, 20:39:04 | Saturday           |
| 1 | SUV        | ₹473.86   | 3 mins away•21:16 drop-off | Elante Mall | Plaksha University | 09/09/2023, 20:39:04 | Saturday           |
| 2 | Auto       | ₹225.62   | 4 mins away•21:16 drop-off | Elante Mall | Plaksha University | 09/09/2023, 20:39:04 | Saturday           |
| 3 | Bike       | ₹156.81   | 2 mins away•21:15 drop-off | Elante Mall | Plaksha University | 09/09/2023, 20:39:04 | Saturday           |
| 4 | Sedan      | ₹319.02   | 7 mins away•21:17 drop-off | Elante Mall | Plaksha University | 09/09/2023, 20:39:04 | Saturday           |

|   | Cab_Type   | Cab_Price | Pick_Up        | Destination           | Current<br>Day | Cab_Arrival_Time | Cab_Destination_Time | Current_Date | Current_Time | Availability | Route_Time |
|---|------------|-----------|----------------|-----------------------|----------------|------------------|----------------------|--------------|--------------|--------------|------------|
| 0 | Hatchbacks | 280.96    | Elante<br>Mall | Plaksha<br>University | Saturday       | 2 mins away      | 21:16                | 09/09/2023   | 20:39        | 1            | 37.0       |
| 1 | SUV        | 473.86    | Elante<br>Mall | Plaksha<br>University | Saturday       | 3 mins away      | 21:16                | 09/09/2023   | 20:39        | 1            | 37.0       |
| 2 | Auto       | 225.62    | Elante<br>Mall | Plaksha<br>University | Saturday       | 4 mins away      | 21:16                | 09/09/2023   | 20:39        | 1            | 37.0       |
| 3 | Bike       | 156.81    | Elante<br>Mall | Plaksha<br>University | Saturday       | 2 mins away      | 21:15                | 09/09/2023   | 20:39        | 1            | 36.0       |
| 4 | Sedan      | 319.02    | Elante<br>Mall | Plaksha<br>University | Saturday       | 7 mins away      | 21:17                | 09/09/2023   | 20:39        | 1            | 38.0       |

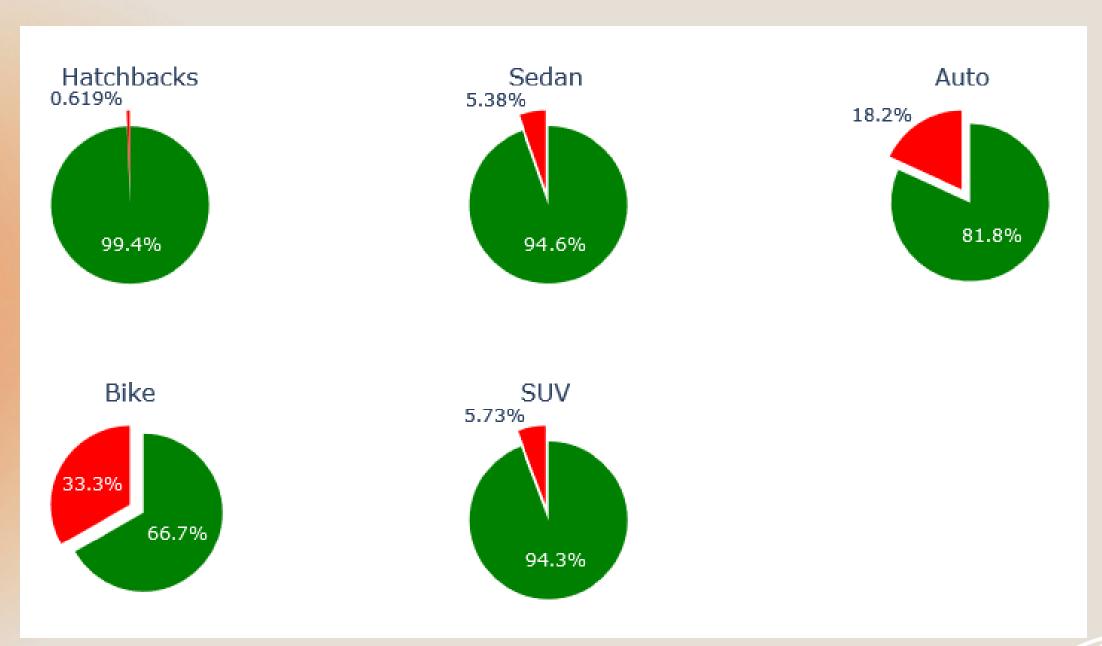
# Cab Availability



#### Availability of All Types of Cabs



#### Availability of Cabs from Plaksha University to Elante Mall



# Cab Price Distribution

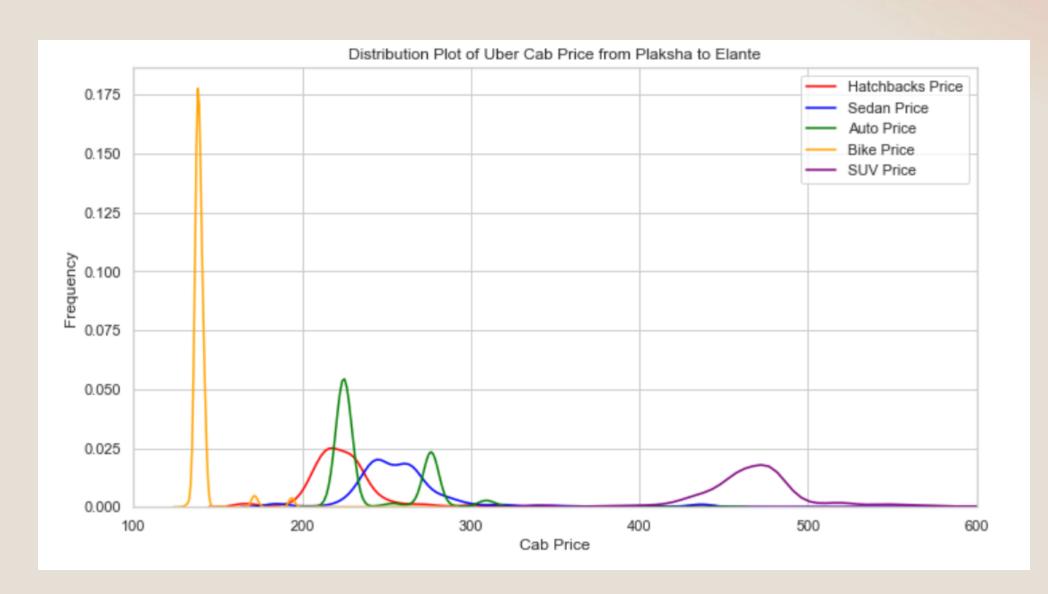
Cab\_Type

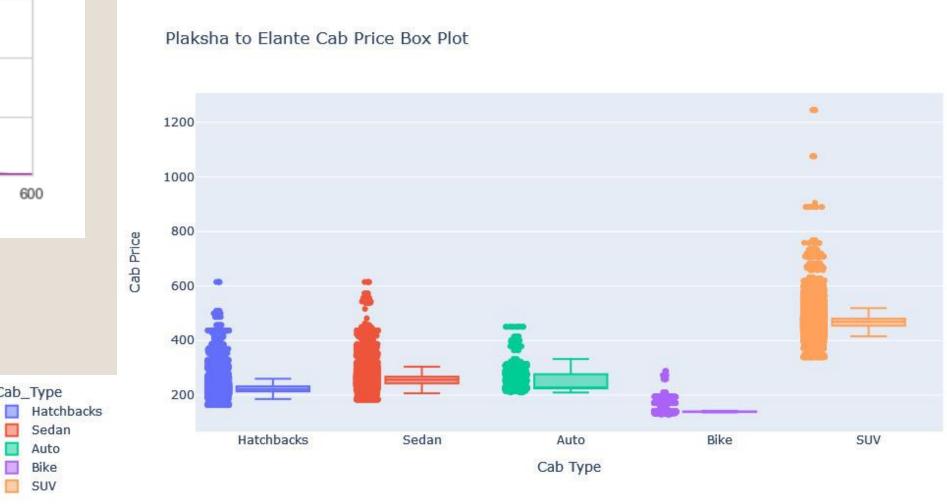
Sedan

Bike

Auto

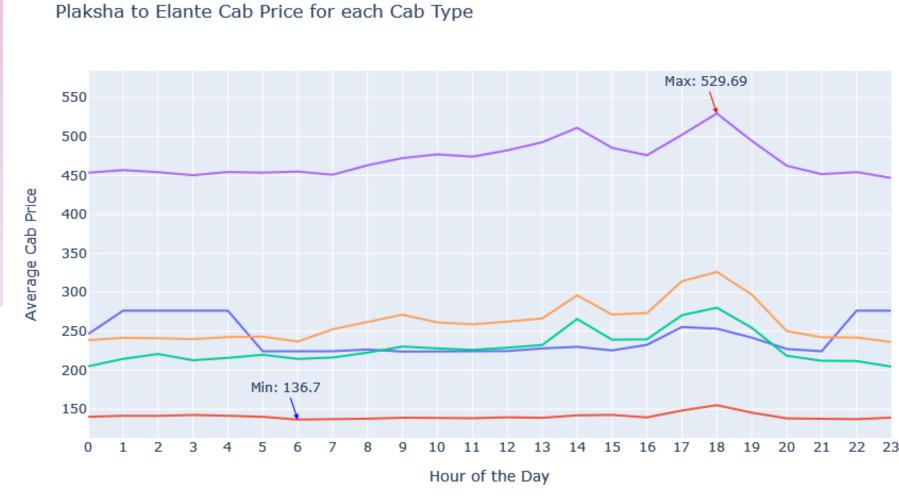
SUV





# Hourly Cab Price





Tuesday

Cab Type

---- Auto

---- Bike

- SUV

---- Sedan

Hatchbacks

# ML Methodology

Cab Price Prediction

#### Linear Regression

Linear Regression is ideal for predicting cab prices when the connection between factors like distance, time, and traffic conditions and the fare is roughly linear. Its interpretability provides insights into how individual features impact the predicted fare.

#### Random Forest Regressor

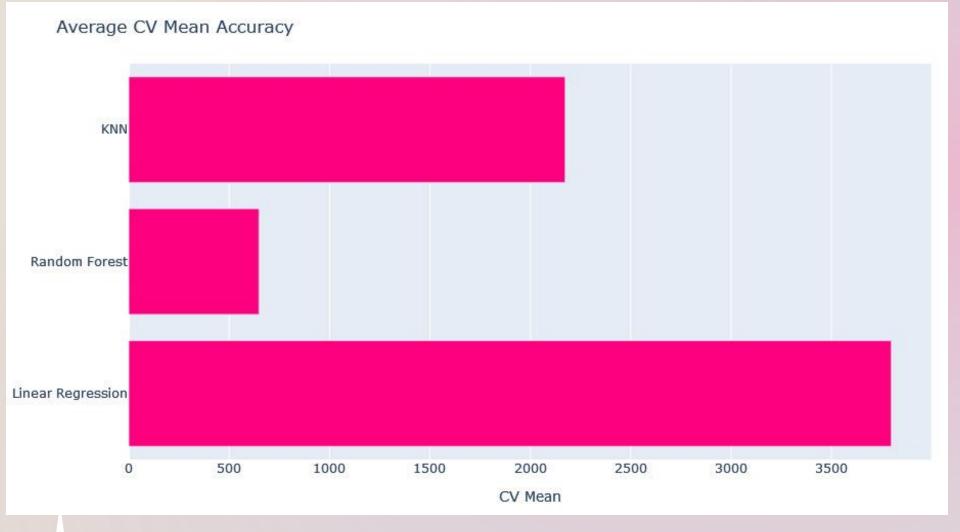
For cab fare prediction with intricate, potentially non-linear relationships among factors,
Random Forest Regressor excels at capturing complex dependencies.

#### K-Nearest Neighbors

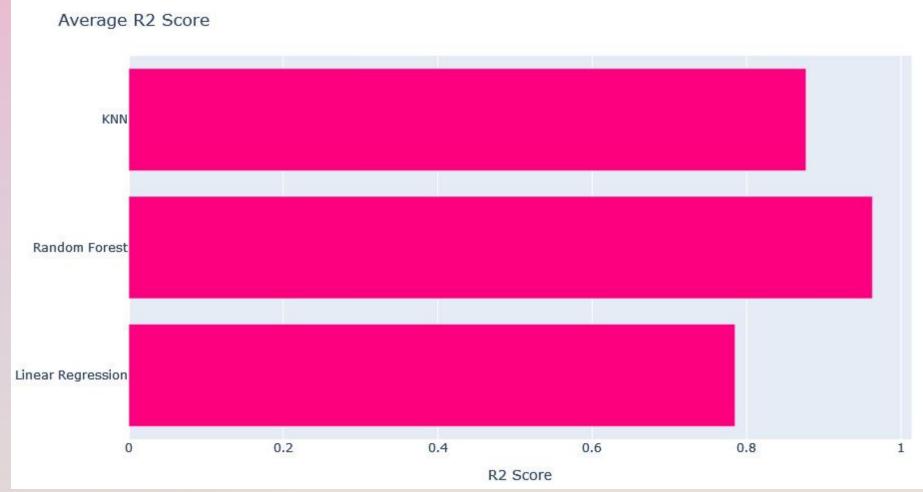
K-Nearest Neighbors is valuable for cab price prediction when there are local pricing patterns, such as neighborhood-specific or time-dependent variations. KNN effectively captures these patterns, adapting to location proximity or temporal factors.

### Performance Metrics

#### Cab Price Prediction



|                   | CV Mean     | Std        | R2 Score |
|-------------------|-------------|------------|----------|
| Linear Regression | 3797.716902 | 162.524465 | 0.784966 |
| Random Forest     | 656.601318  | 33.316359  | 0.962803 |
| KNN               | 2173.335933 | 52.001714  | 0.876900 |



# ML Methodology

#### Route Time Prediction

#### Linear Regression

Linear Regression is a suitable choice for predicting route time when input features, such as distance, traffic conditions, and time of day, exhibit a linear relationship.

#### Random Forest Regressor

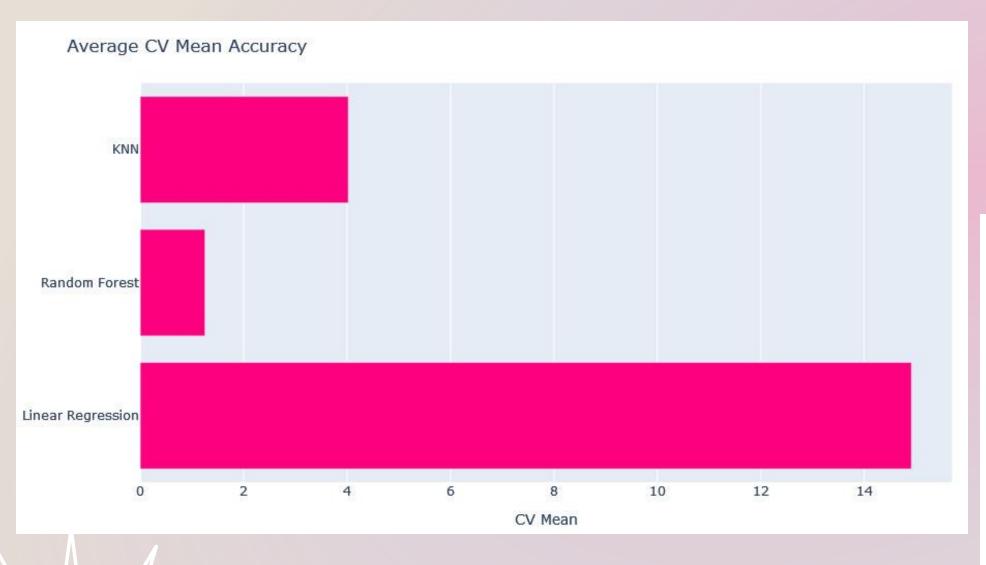
Similar to cab fare prediction, Random Forest Regressor is advantageous for route time prediction when intricate, non-linear relationships exist between input features and route duration.

#### K-Nearest Neighbors

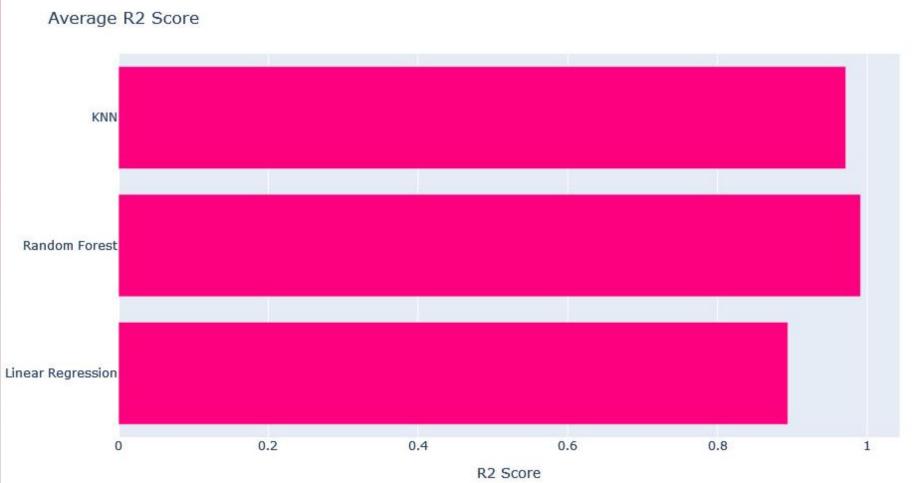
K Nearest Neighbors is useful when predicting route time, especially in scenarios where the travel time varies based on local patterns.

### Performance Metrics

#### Route Time Prediction



|                   | CV Mean   | Std      | R2 Score |
|-------------------|-----------|----------|----------|
| Linear Regression | 14.913452 | 0.113069 | 0.893972 |
| Random Forest     | 1.251514  | 0.026682 | 0.991102 |
| KNN               | 4.022650  | 0.113075 | 0.971400 |



# ML Modeling Obstacles



#### Persistent Web Scraping

A laptop was relentlessly employed in the continuous extraction of data from the internet through web scraping



### Limited Storage Capacity

We continuously created new Google Sheets after reaching the 250,000-row limit in each, to accommodate our extensive data storage needs



### Time Intensive Processing

Training our machine learning model proved to be a time-consuming task, particularly due to the substantial size of our dataset

# Deployment Hurdles

# Debloa

#### Data Volume and Management:

Handling larger volumes of data effectively, including storage and processing, becomes more complex as the scale grows.

#### Real-Time Data Processing:

As your service scales, the need for real-time data processing increases, which can be technically demanding.

#### Changing Data Patterns:

Over time, patterns in cab usage, pricing, and traffic conditions may change, requiring continuous updates to the model to maintain accuracy.

