



# Cars24 Used Car Price Analysis

From web scraping to actionable market insights

# Meet the Team

Our dedicated team of experts collaborated to deliver this comprehensive analysis.



Web Scraping

**Anand Raj (Team Lead)**

**Atharv Kakade**



Data Analysing & Cleaning

**Siddhesh Mali (Co Lead)**

**Abhishek Mishra**



Data Visualisation

**Linda Shalini T (Co Lead 2)**



Project Presentation

**Siddhesh Mali (Co Lead)**

**Himanshu Shakya**



Project Report

**Anand Raj (Team Lead)**



# Project Aim and Objectives



## Project Aim

**To extract, clean, and analyze used car data from Cars24.com, focusing on Mumbai-based Mahindra listings. This will develop practical skills in web scraping, data processing, and Python analysis.**



## Key Objectives

- **Scrape key car details (Year, Kilometers, Fuel, Transmission, Price).**
- **Organize and clean data into a structured CSV format.**
- **Analyze and visualize data to derive market insights.**
- **Present findings via Python Notebook, Report, and Presentation.**



## Scope & Coverage

**Target Website: Cars24.com. Brand: Hyundai. Primary Location: Mumbai.**

# Technology Stack



Web Automation

**Selenium WebDriver with ChromeDriver for dynamic page scraping**



Data Processing

**Pandas for manipulation, Regex for text parsing, CSV for storage**



Visualisation

**Matplotlib and Seaborn for statistical graphics**



Environment

**Google Colab and Jupyter Notebooks for development**



# Web Scraping Methodology

01

Automated Navigation

**Selenium controlled browser instances to traverse Cars24 listing pages systematically**

03

Pagination Handling

**Implemented robust logic to navigate through multiple pages whilst managing dynamic content loading**

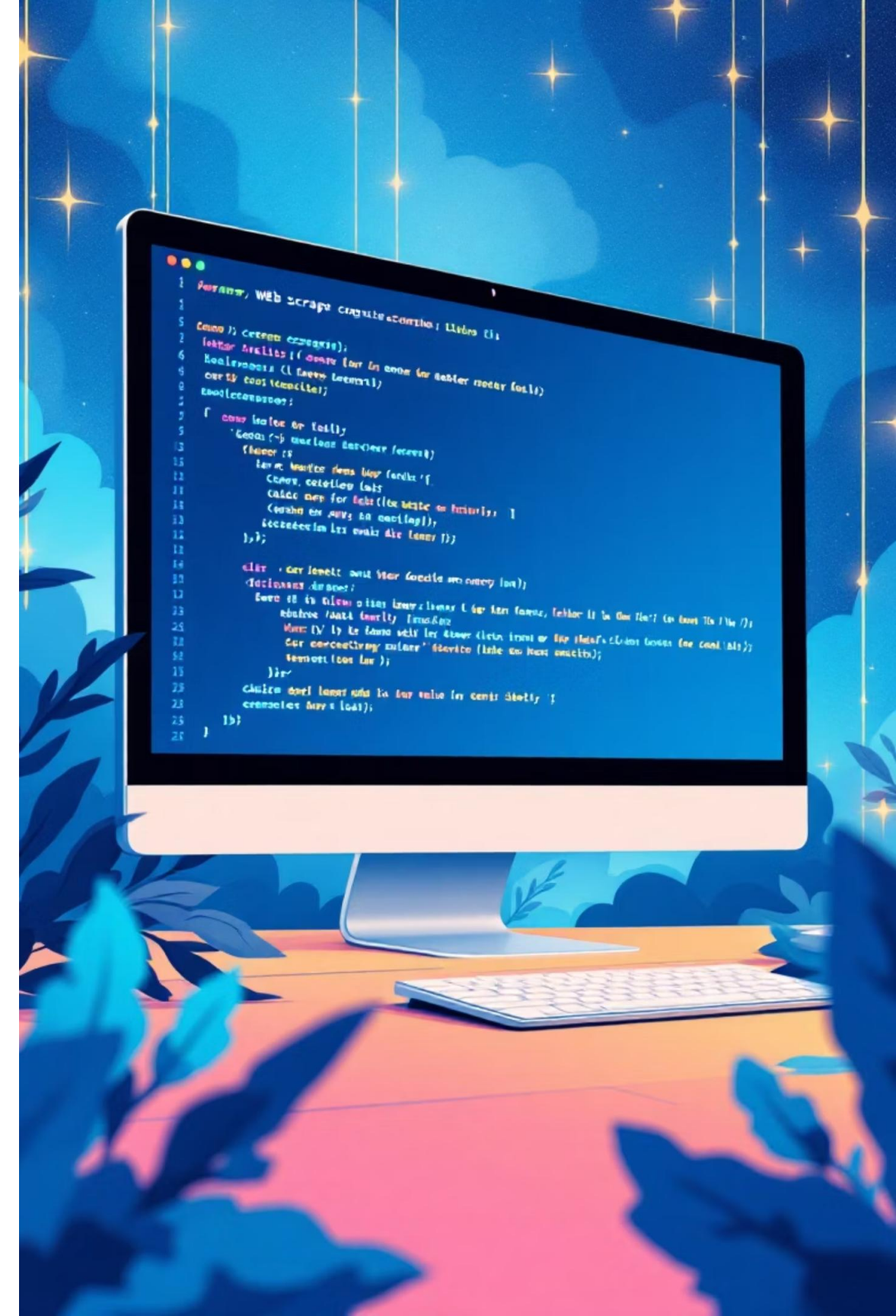
Technical Challenges

**Overcame issues with dynamically loaded content, fragile CSS selectors, and inconsistent page structures across city-specific listings.**

02

Data Extraction

**Captured key attributes: Car Name, Year, Kilometres, Fuel Type, Transmission, Price,**



# Data Cleaning & Preparation

## Cleaning Operations

- Duplicate Removal

**Identified and eliminated redundant entries**

- Type Conversion

**Standardised numeric and categorical fields**

- Missing Values

**Applied appropriate imputation strategies**

- Categorical Consistency

**Normalised fuel types and transmission values**

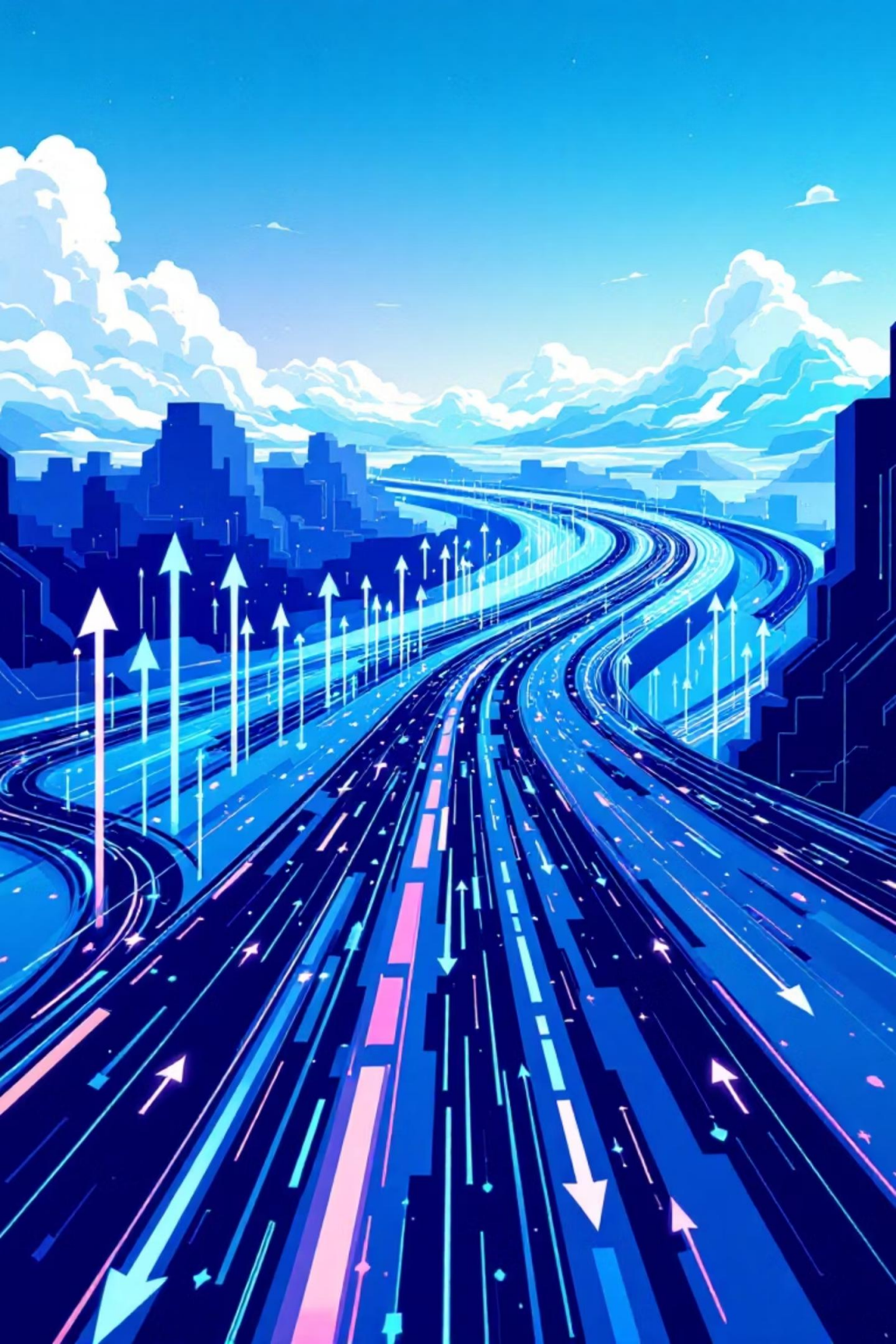
## Derived Features

📄  **$\text{Car\_Age} = 2025 - \text{Year}$**

**$\text{Price\_per\_Km} = \text{Price} \div \text{Kilometres}$**

These engineered features enabled more sophisticated analysis of depreciation patterns and value metrics across the dataset.





# Feature Transformation

Transformed raw scraped data into analysis-ready features for deeper insight into pricing and usage patterns.

1

Cleaned Car Name

**Removed year prefixes (e.g., 2016 Honda City → Honda City) for consistent brand and model grouping.**

2

Standardised Kilometers Driven

**Converted all formats (e.g., “25k km”, “2.5L km”) into integer kilometers for accurate numeric comparison.**

3

Normalized Price (INR)

**Removed currency symbols and standardised ₹, lakh, crore values into a uniform INR scale.**

4

Derived Car Age

**Calculated as `2025 - Year` to capture depreciation effects and support price-age relationship analysis.**

5

Created Price per Kilometer

**Derived from `Price(INR) / Kilometers_Driven` to identify value efficiency and pricing anomalies.**

Outcome

- Ensured uniform numeric types for robust modeling and visualization.
- Enabled comparative insights between car brands, fuel types, and usage patterns.
- Prepared data effectively for advanced statistical analysis and machine learning models.

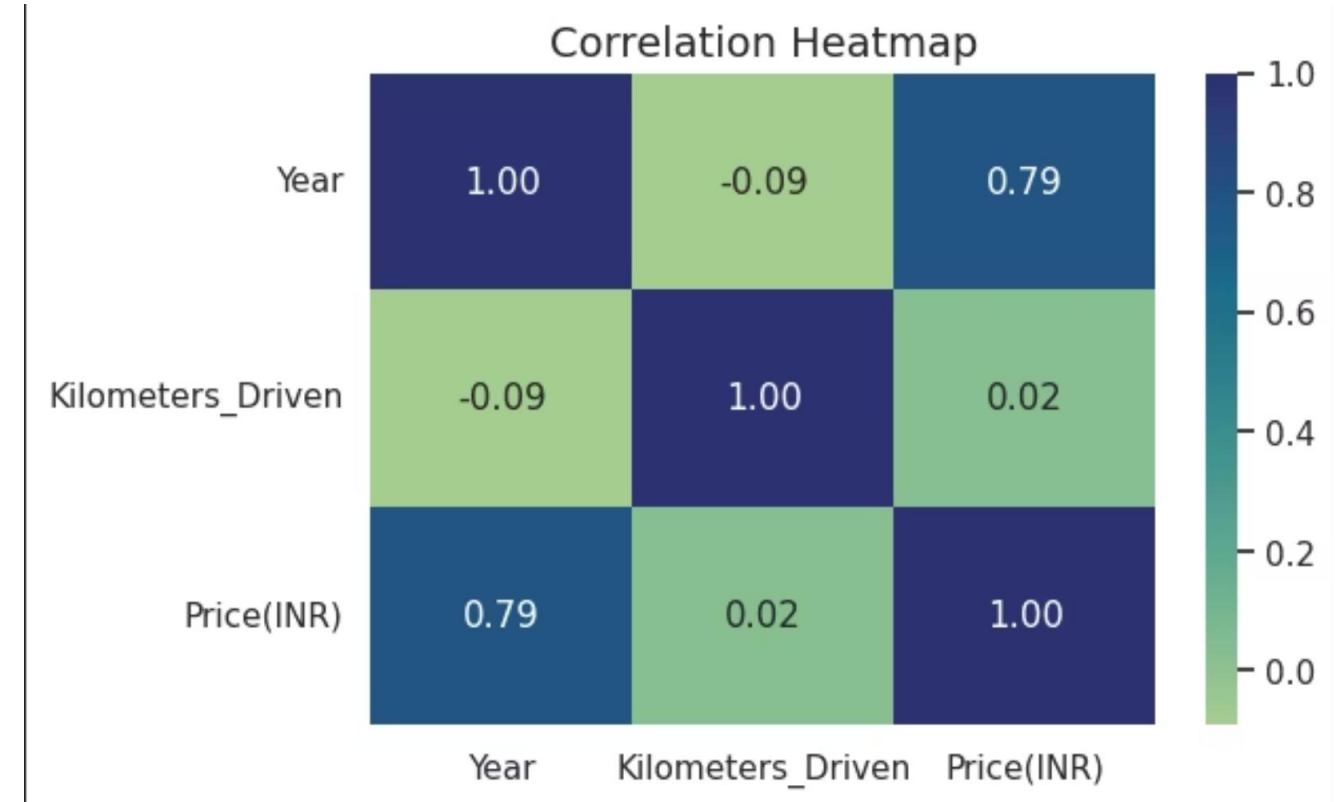
# Exploratory Data Analysis

## Univariate Analysis

- Price distribution showed right-skewed pattern
- Car age concentrated in 3–7 year range
- Kilometre readings varied widely across listings

## Bivariate Relationships

- Strong negative correlation between price and vehicle age
- Inverse relationship between kilometres driven and valuation
- Transmission type significantly influenced pricing



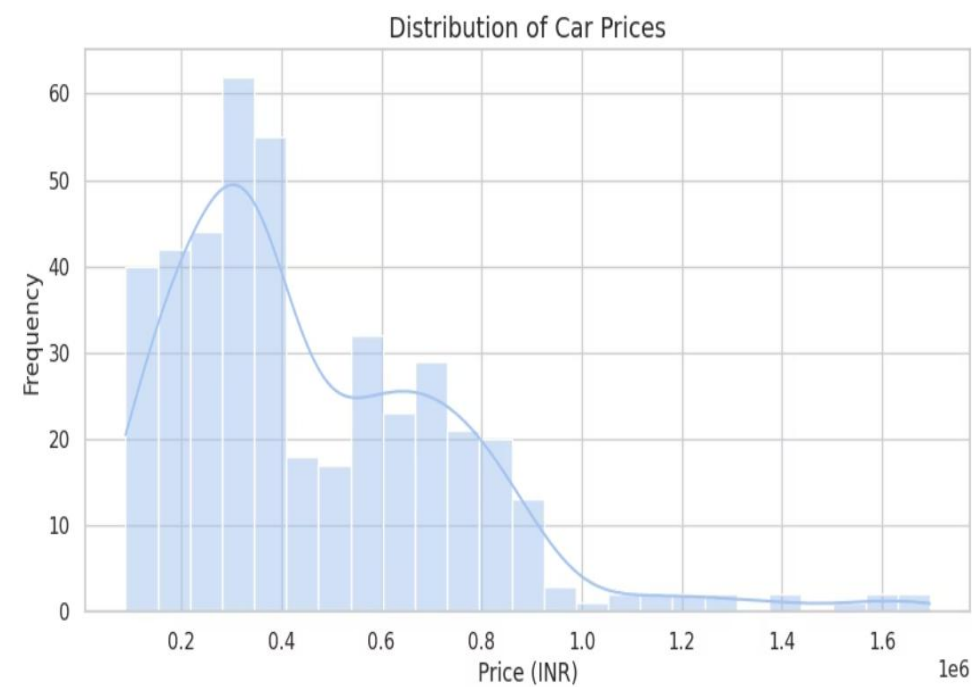
Correlation matrix revealed strongest predictive relationships, with age and mileage showing the highest negative correlation with price.



# Visualisation Highlights

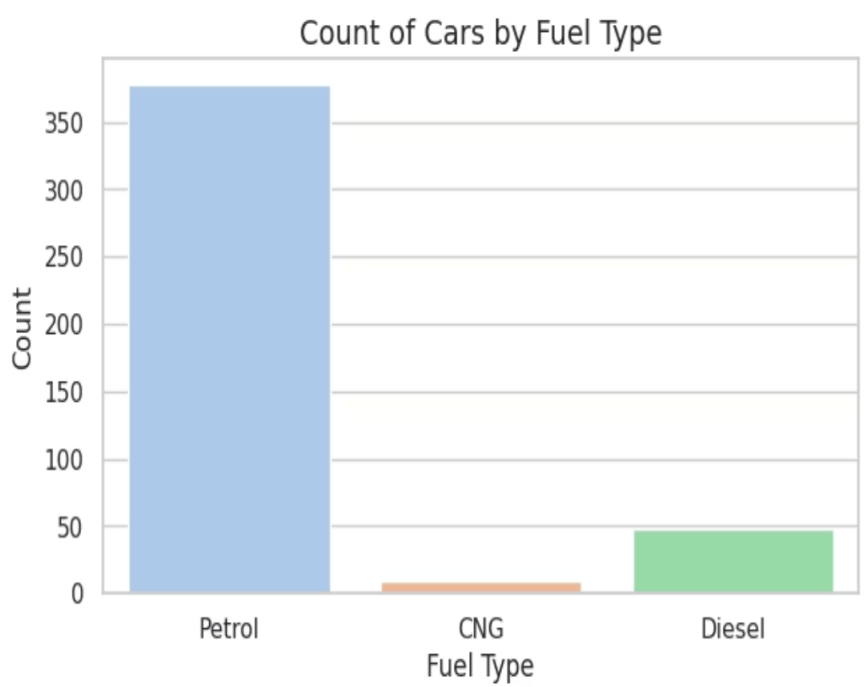
## Price Distribution

Right-skewed histogram indicating concentration in mid-range segment



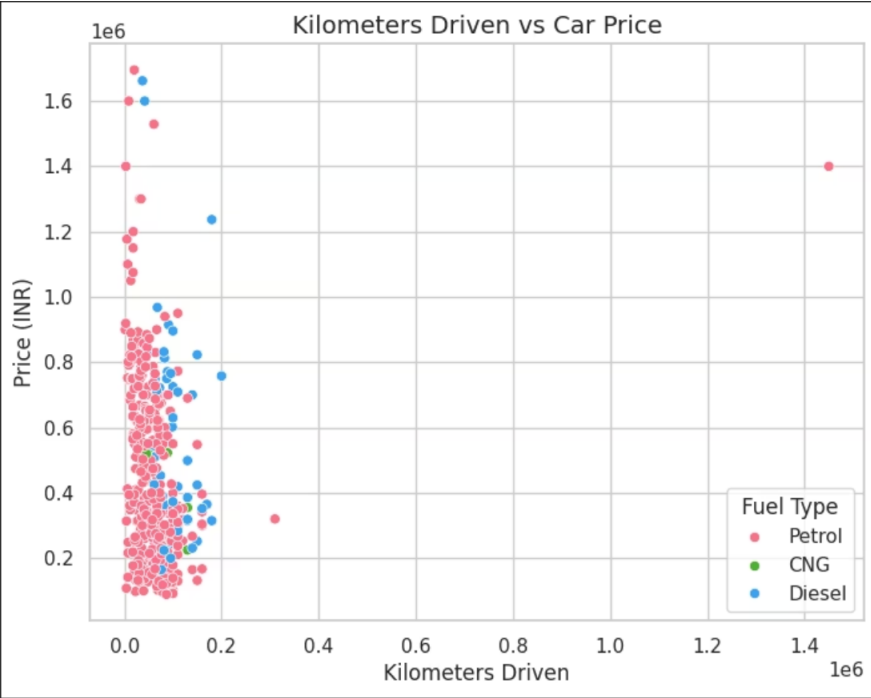
## Fuel Type Analysis

Countplots revealed petrol dominance in the used car market



## Price vs Kilometres

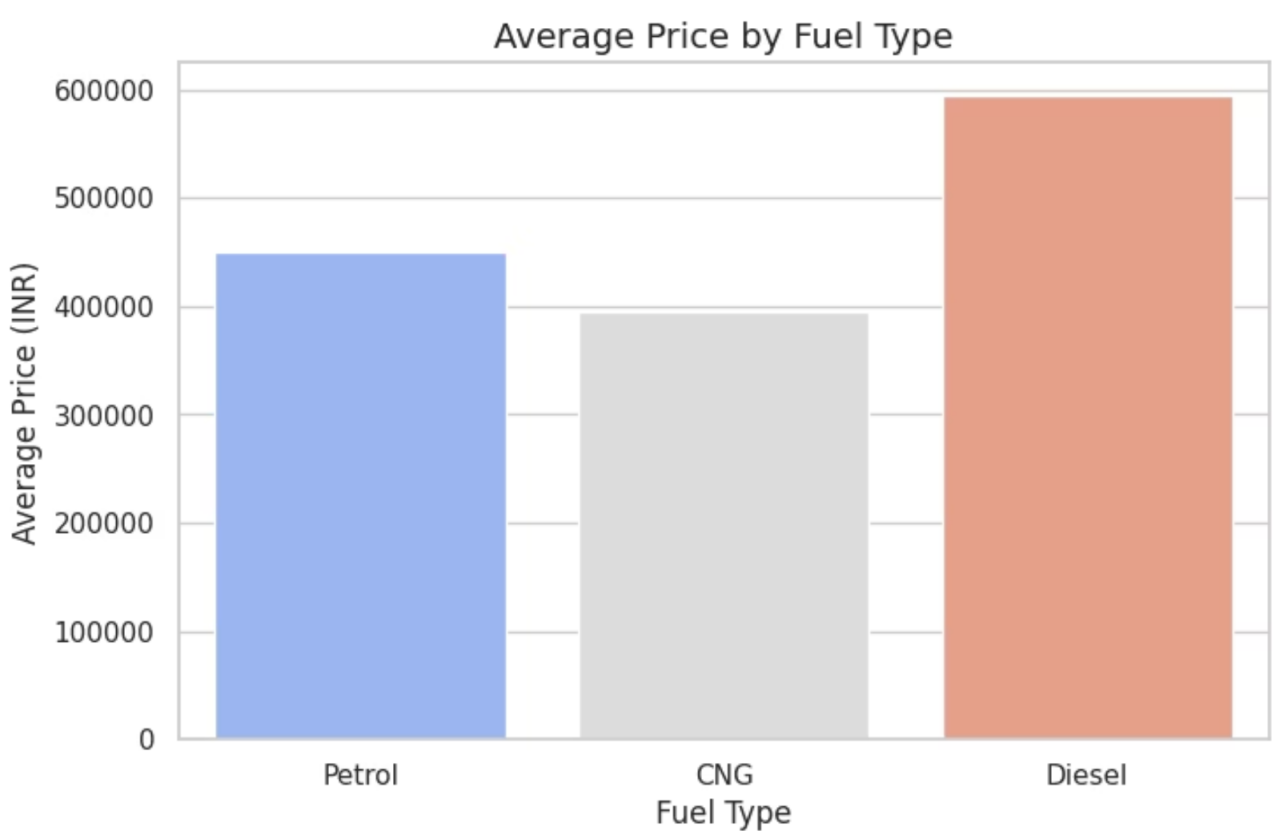
Scatterplot demonstrated depreciation with increased mileage by fuel type



# Deeper Dive into Market Trends

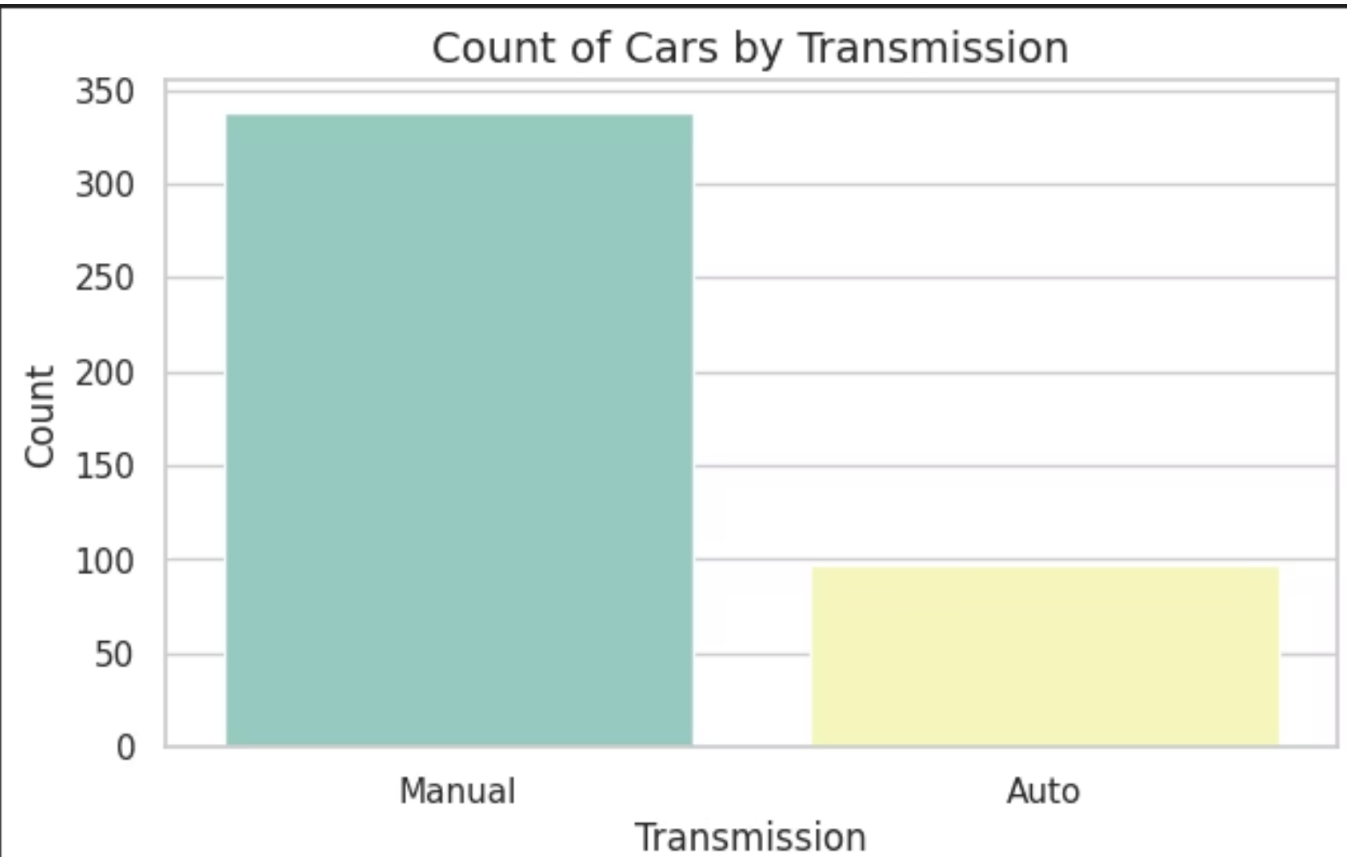
## Average Price by Fuel Type

Diesel cars command the highest average prices (~₹6 lakh), reflecting their long-distance efficiency and perceived market value. Petrol models follow at ₹4–4.5 lakh, while CNG cars are the most affordable, catering to budget-conscious and eco-friendly buyers. This indicates fuel type is a strong predictor in pricing models.

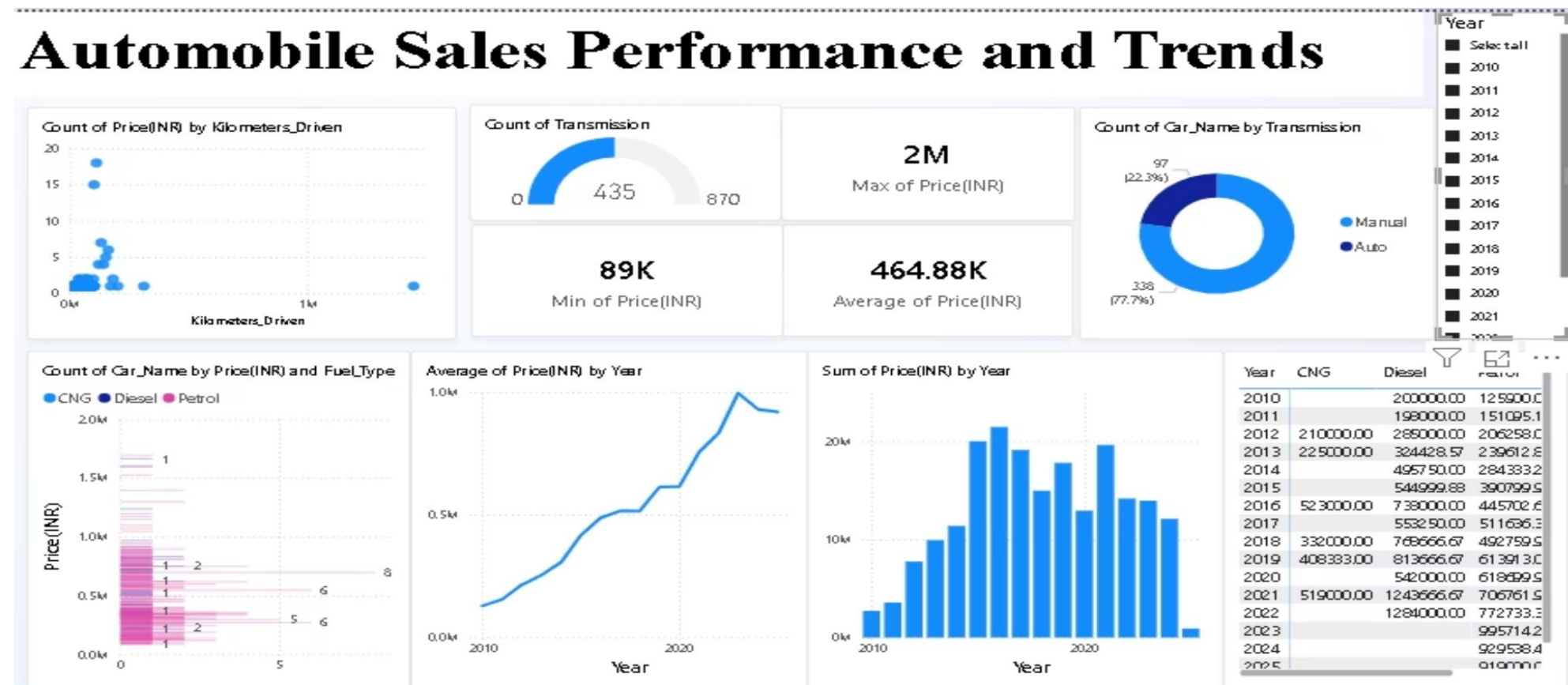


## Count of Cars by Transmission

Manual transmission vehicles dominate the market with listings three times higher than automatics, suggesting a strong preference among budget-oriented consumers. Automatic cars form a smaller, premium segment, appealing to urban users prioritising comfort. Transmission type significantly influences both market availability and demand.



# Automobile Sales Trends & Insights (Power BI Dashboard)



## Key Observations from Dashboard

- Price vs Kilometers Driven:** As kilometers increase, car prices decrease, confirming mileage as a strong negative predictor of resale value.
- Transmission Trends:** Automatic cars dominate (870 units) compared to Manual (435), indicating a shift toward convenience-focused buyers. This suggests a changing market preference.
- Fuel Type & Price Patterns:** Diesel cars command the highest prices, followed by Petrol, while CNG remains the most budget-friendly. This confirms fuel type as a key factor in price modeling.
- Yearly Price Trends:** Average and total prices show an upward trend from 2010–2020, reflecting market growth and rising consumer spending power.

## Strategic Insights

Factor	Observation	Impact
Mileage	Higher km → Lower price	Use in depreciation modeling
Transmission	Automatics rising	Market demand shift
Fuel Type	Diesel premium	Pricing & eco-strategy
Year Trend	Rising prices	Time-based forecasting

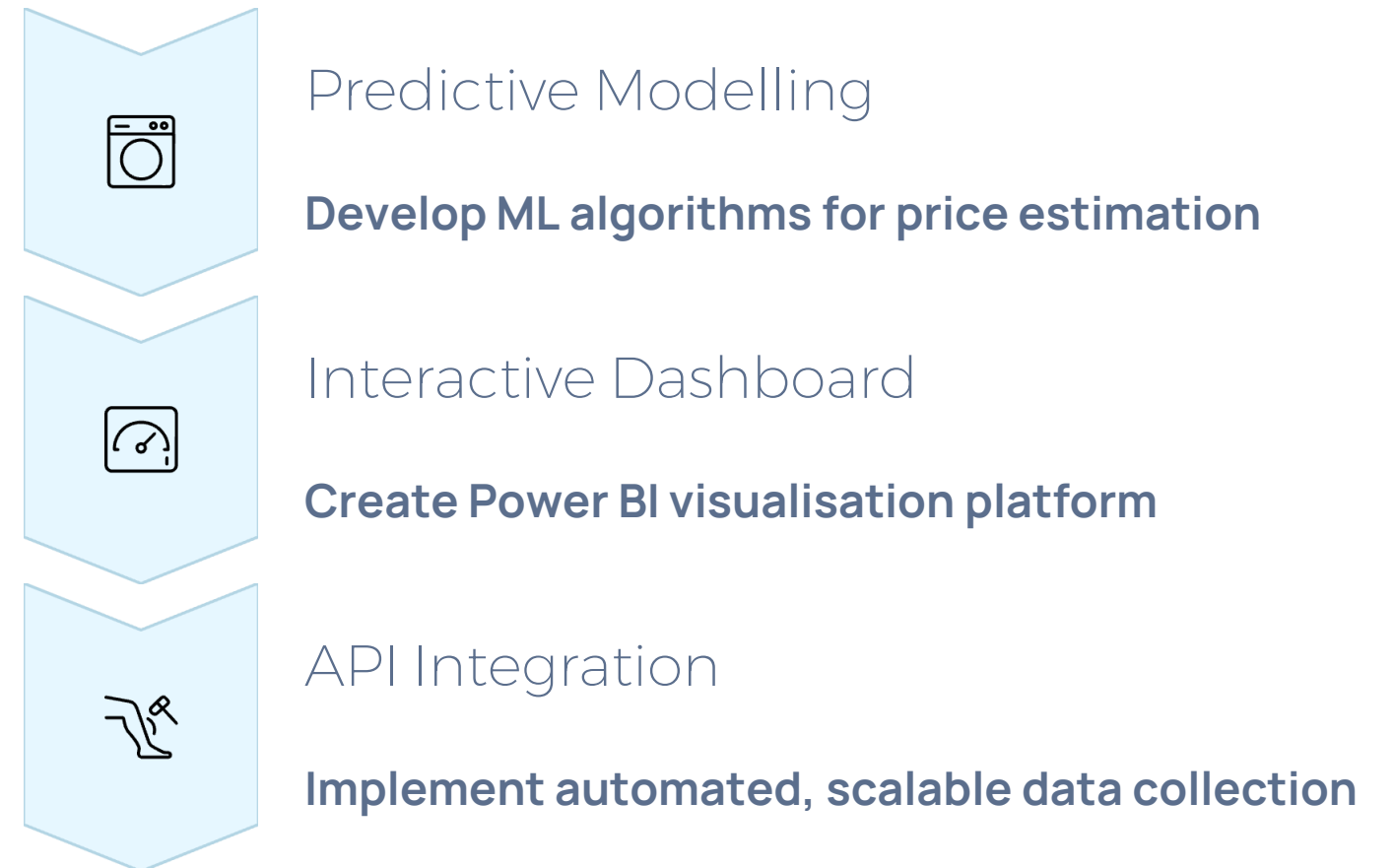


# Conclusion & Future Directions

## Project Accomplishments

- 1 Successfully scraped comprehensive vehicle data
- 2 Cleaned and engineered analytical features
- 3 Generated actionable market insights

## Next Steps



This project demonstrates the complete data science pipeline from raw data acquisition to meaningful business intelligence.

# Thank You!

We appreciate your time and interest in our Cars24 Used Car Price Analysis project. We hope our insights provide valuable understanding of the market dynamics.