Bike Price Prediction Project - Internship Report

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Tools: Python, Pandas, NumPy, Scikit-learn, Jupyter Notebook, Power BI

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1. Introduction

The goal of this project is to predict the prices of used bikes in India based on various features such as model year, kilometers driven, engine power, mileage, owner type, and location. This helps buyers and sellers make informed decisions.

2. Problem Statement

Predicting accurate used bike prices is challenging due to varying factors like bike age, usage, engine power, and ownership history. The project aims to build a regression model to estimate the price and visualize insights through a dashboard.

3. Dataset Description

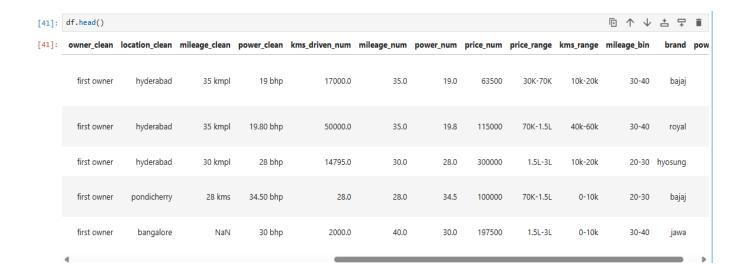


Dataset Size: 7857, 8

4. Data Cleaning & Preprocessing

Steps performed:

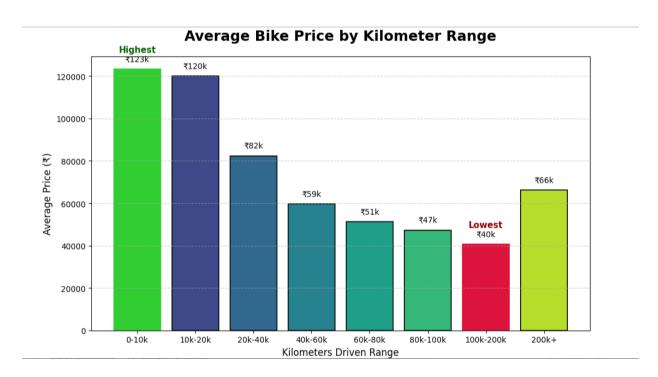
- Removed duplicates and inconsistent formatting
- Converted km, mileage, and power columns to numeric
- Handled missing values
- Standardized categorical columns (owner_clean, location_clean)
- Created numerical columns for modeling (kms_driven_num, mileage_num, power_num, price_num)
- Created bins for price, mileage, and kms driven

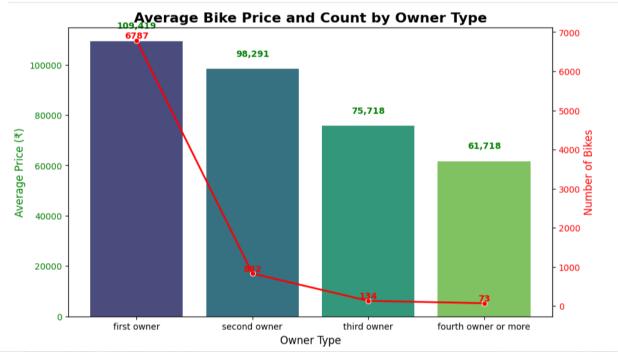


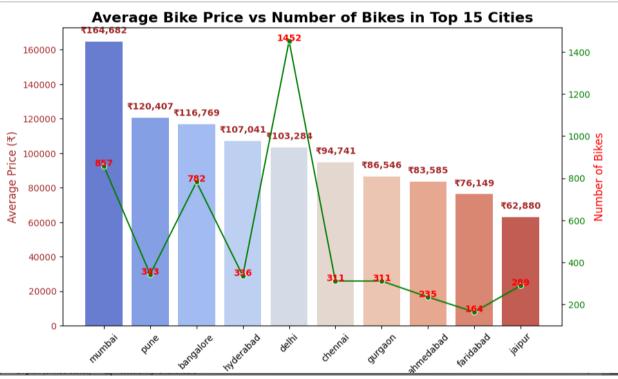
5. Exploratory Data Analysis (EDA)

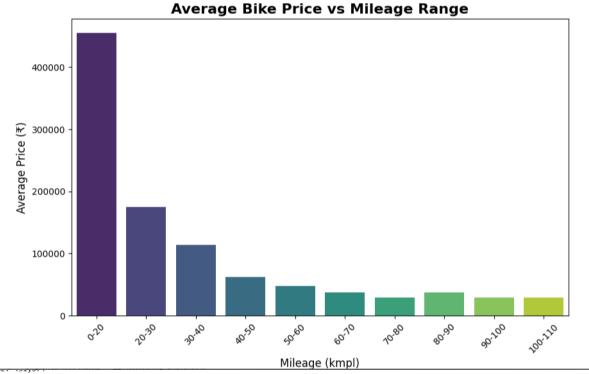
- Distribution of bikes by price, mileage, and owner
- Trends observed:
 - Older bikes cost less
 - o 1st owner bikes sell at higher prices

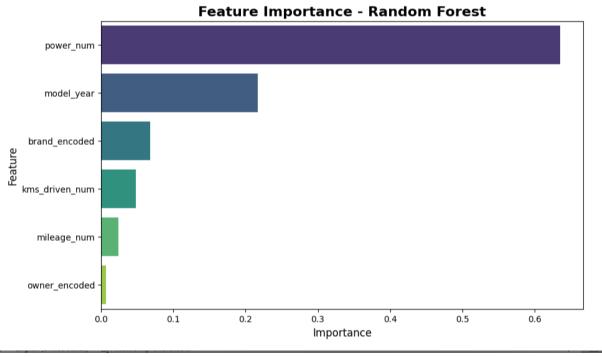
Higher mileage correlates slightly with higher price











6. Feature Engineering

- Created derived columns: price_range, kms_range, mileage_bin
- Reason: To group values for better visualization and model performance

7. Machine Learning Model

Model Used: Random Forest Regressor

Input Features: model_year, kms_driven_num, mileage_num, power_num, owner_clean,

location_clean

Target Variable: price_num

Why Random Forest: Handles non-linear relationships and performs well on tabular data

8. Model Evaluation

Metric	Log Scale	Original Scale
R ² Score	0.91	-
MAE	0.14	₹11,758
RMSE	0.19	₹17,236

• R² Score: Measures model fit

• MAE: Average absolute error

• **RMSE:** Root mean squared error

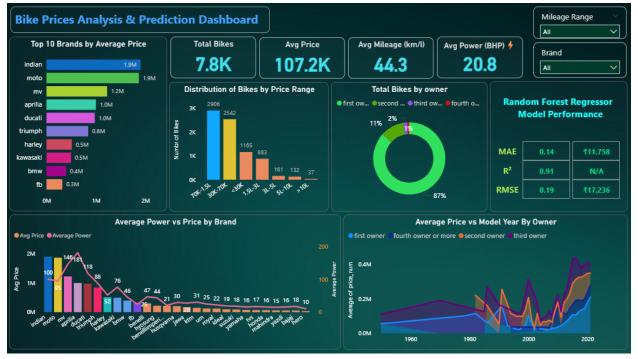
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Random Forest Performance (log scale):
R² Score: 0.90
MAE: 0.17
RMSE: 0.25

Random Forest Performance (original scale):
MAE: ₹14,397
RMSE: ₹31,074
```

9. Dashboard & Visualization

- KPIs: Total bikes, Average price, Average mileage, KM driven
- Charts:
 - Price vs Model Year (Scatter)
 - Brand distribution (Bar)

- Mileage vs Price (Scatter)
- Cards showing model performance metrics



10. Insights & Conclusion

- 1st owner bikes cost more than 2nd or 3rd owner bikes
- Newer bikes have higher prices
- Mileage has a slight correlation with price
- Dashboard helps buyers and sellers make data-driven decisions

11. Tools & Technologies

- Python
- Pandas
- NumPy
- Scikit-learn

- Jupyter Notebook
- Power BI

12. References

- Google General research on used bike pricing factors and model evaluation techniques
- YouTube Tutorials for Power BI dashboard design and Python data analysis
- Scikit-learn Documentation *Implementation details for Random Forest Regressor and performance metrics*