

Used Car Price Prediction

Alan T Akkara
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Introduction

- The used car market is vast and dynamic.
- Prices vary by mileage, brand, year, fuel type, and other attributes.
- Buyers and sellers struggle with transparency in pricing.
- **Goal:** Build a predictive model to estimate car prices more accurately.

Problem Statement

- Used car buyers risk **overpaying**.
- Seller's risk **undervaluing** their vehicles.
- Businesses need **data-driven tools** for fair and competitive pricing.

Dataset

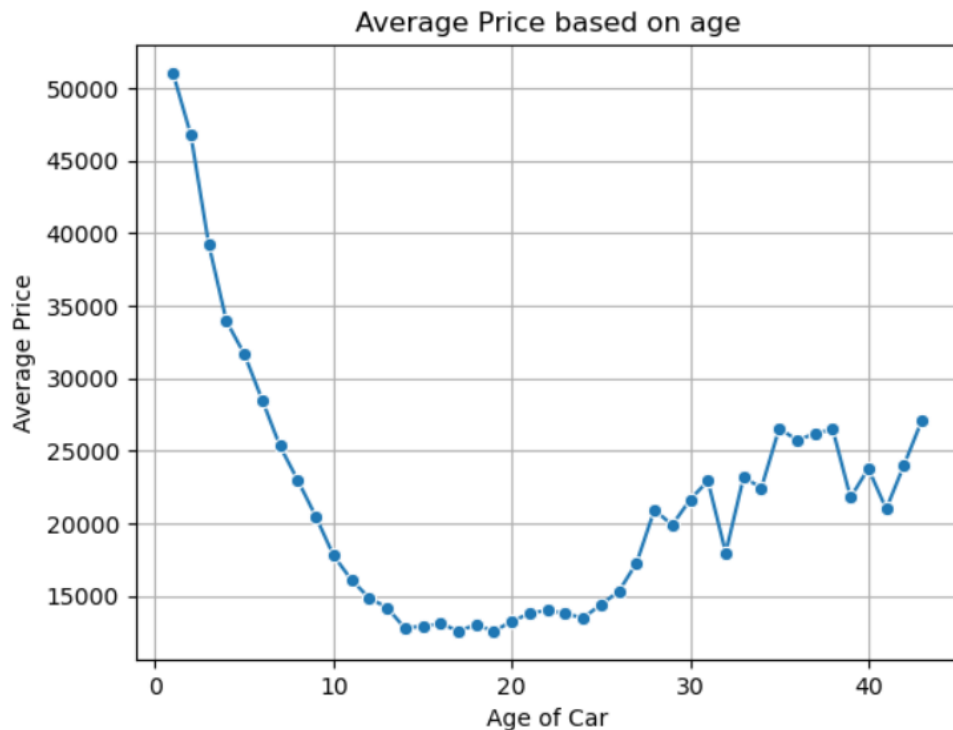
	manufacturer	model	year	mileage	engine	transmission	drivetrain	fuel_type	mpg	exterior_color	interior_color	accidents_or_damage	one_owner	personal_use
0	Acura	ILX Hybrid 1.5L	2013	92945.0	1.5L I-4 i-VTEC variable valve control, engine...	Automatic	Front-wheel Drive	Gasoline	39-38	Black	Parchment	0.0	0.0	
1	Acura	ILX Hybrid 1.5L	2013	47645.0	1.5L I4 8V MPFI SOHC Hybrid	Automatic CVT	Front-wheel Drive	Hybrid	39-38	Gray	Ebony	1.0	1.0	
2	Acura	ILX Hybrid 1.5L	2013	53422.0	1.5L I4 8V MPFI SOHC Hybrid	Automatic CVT	Front-wheel Drive	Hybrid	39-38	Bellanova White Pearl	Ebony	0.0	1.0	
3	Acura	ILX Hybrid 1.5L	2013	117598.0	1.5L I4 8V MPFI SOHC Hybrid	Automatic CVT	Front-wheel Drive	Hybrid	39-38	Polished Metal Metallic	NaN	0.0	1.0	
4	Acura	ILX Hybrid 1.5L	2013	114865.0	1.5L I4 8V MPFI SOHC Hybrid	Automatic CVT	Front-wheel Drive	Hybrid	39-38	NaN	Ebony	1.0	0.0	

Source : <https://www.kaggle.com/datasets/andreinovikov/used-cars-dataset>

Data Wrangling

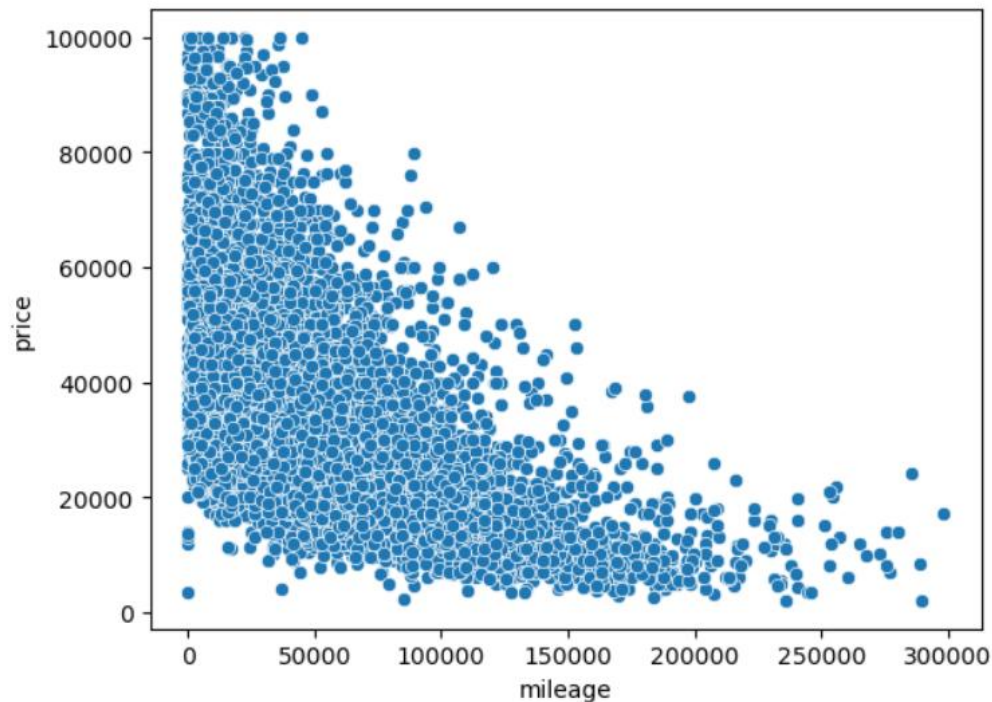
- Removed duplicate rows to ensure data integrity.
- Limited extreme values in **year**, **mileage**, and **price**.
- Restricted categorical features to top 10 values.
- Converted **mpg** ranges into their mean values for consistency.
- After preprocessing: **690,370 rows × 17 columns**

Exploratory Data Analysis



- Price decreases as car age increases.
- Older classic cars show a slight price rise

Exploratory Data Analysis



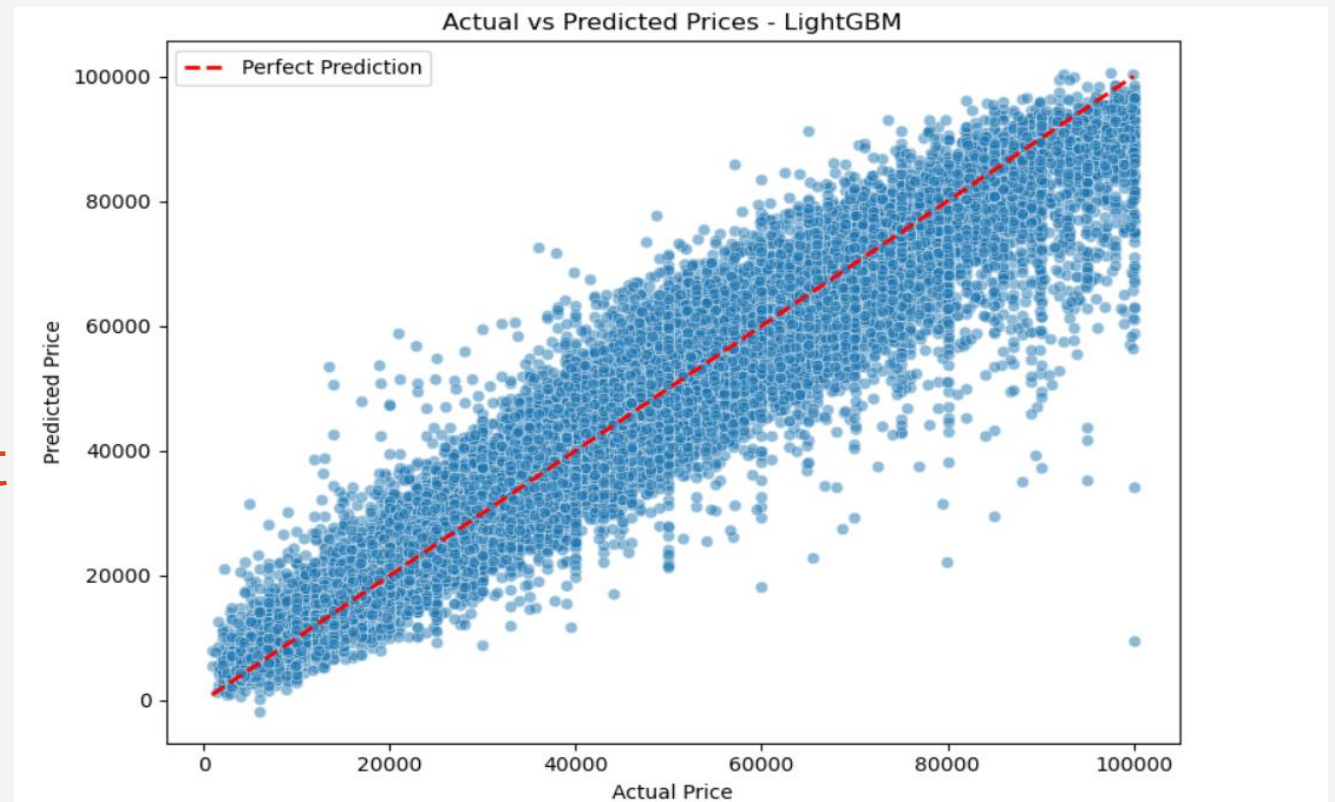
- Cars with higher mileage generally have lower prices

Modelling Approach

- Models tested:
 - Linear Regression
 - Random Forest
 - LightGBM
 - XGBoost
- Evaluation metric: **RMSE & R^2** .
- Final model: **LightGBM** due to best performance.

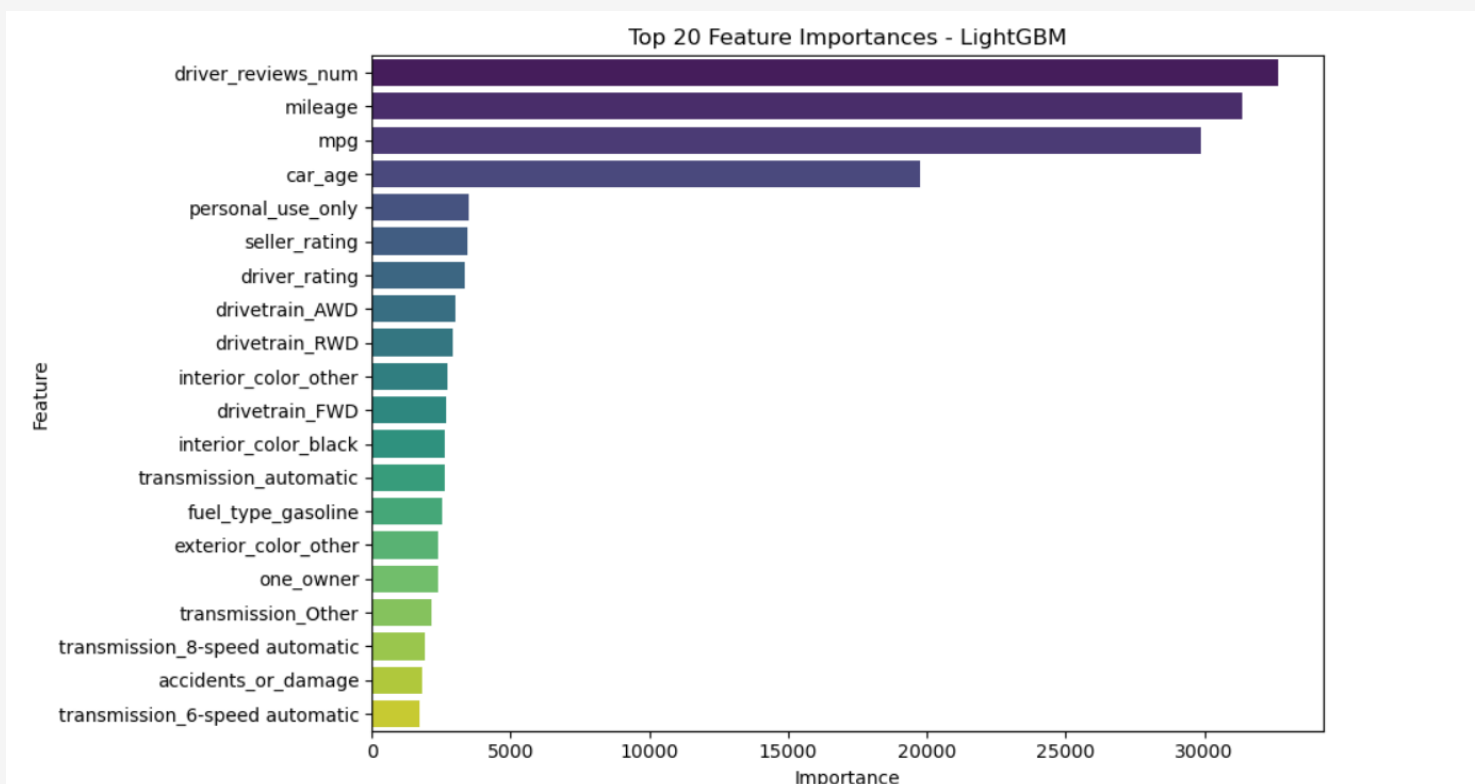
Model Performance

- LightGBM Results:
 - RMSE: 3857.48
 - R^2 : 0.94
- Strong predictive power on test set.



Feature Importance

- Top features :
 - Number of Driver Reviews
 - Mileage
 - Miles per Gallon
 - Age of the Car



Recommendations

- Businesses can use model outputs to:
 - **Set competitive prices** for listings.
 - **Provide transparency** to customers.
 - **Improve trust** in used car platforms.
- Buyers and sellers benefit from **fair pricing benchmarks**.

Future Research

- Develop a real-time car price prediction app.
- Add **features** like accident history and service records for better accuracy
- Analyze **time trends** to predict future price depreciation

Conclusion

- Built and evaluated multiple ML models.
- LightGBM achieved **94% accuracy**.
- Strong negative correlation between **car age, mileage, and price**
- The model can **enhance decision-making** in used car pricing.

THANK YOU