# **Understanding Factors Affecting Used Car Prices**

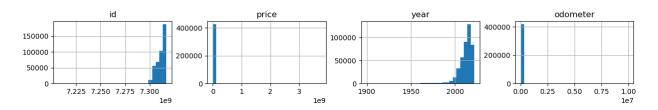
### **Executive Summary**

This report outlines findings from an analysis of a dataset including 426K used cars. Features that significantly influence the prices of used cars are identified. These features include odometer, year, region, model, condition, cylinders, fuel, drive, size, type, paint color, and state. These factors can be considered in determining a price of a car.

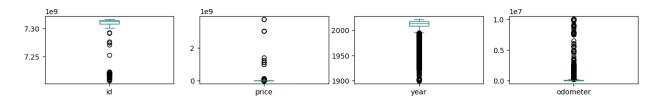
## **Data Analysis**

The dataset includes some numerical features and categorical features. Numerical features include ID, year, odometer, and price. Their distributions are visualized below.

Histograms of Numerical Features



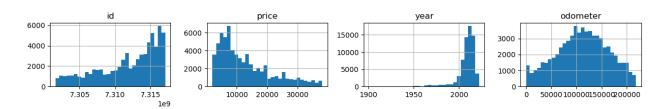
**Boxplots of Numerical Features** 



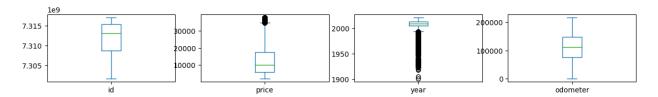
### **Data Preparation**

Outlier prices and odometer readings are filtered out.

Histograms of Numerical Features



#### **Boxplots of Numerical Features**



Pearson correlation for numerical features are calculated. Ranked Numerical Features by Correlation Strength indicating a degree of association between the variable and price:

- odometer 0.414961
- year 0.240999

ANOVA for each categorical feature is calculated. Ranked Categorical Features by ANOVA p-value, indicating whether observed differences among group means could be due to specific variable being studied or if they are just due to chance:

- region: F-Statistic = 17.11, p-value = 0.00
- model: F-Statistic = 8.37, p-value = 0.00
- condition: F-Statistic = 1305.45, p-value = 0.00
- cylinders: F-Statistic = 1198.19, p-value = 0.00
- fuel: F-Statistic = 2747.30, p-value = 0.00
- drive: F-Statistic = 5651.68, p-value = 0.00
- size: F-Statistic = 1999.38, p-value = 0.00
- type: F-Statistic = 1719.88, p-value = 0.00
- paint\_color: F-Statistic = 304.11, p-value = 0.00
- state: F-Statistic = 56.00, p-value = 0.00

#### **Modeling**

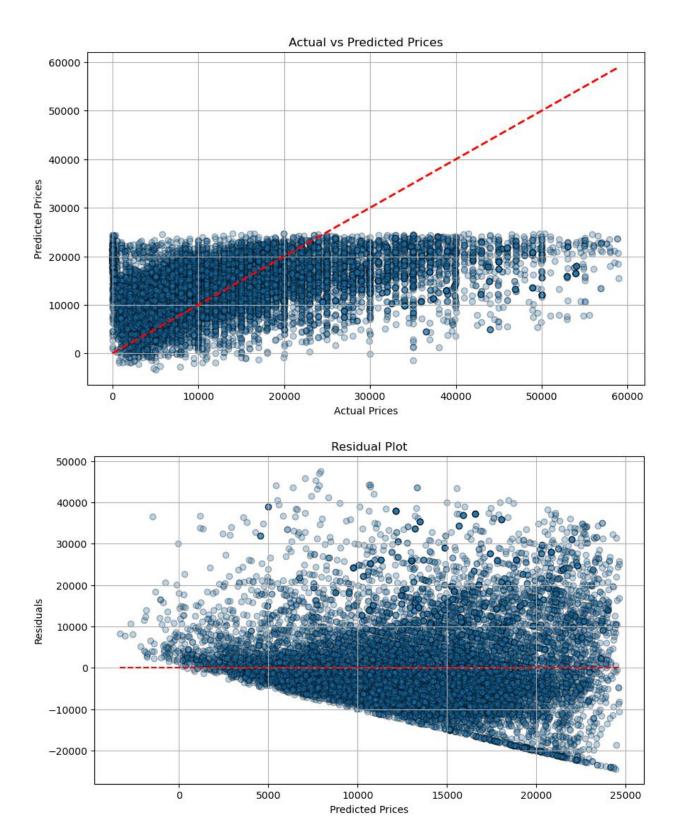
Three models are built using different training data or algorithms.

1. Model 1 – linear regression model trained using only two numerical features, odometer and year.

Model Evaluation Metrics:

Mean Squared Error: 98024874.99423555

R^2 Score: 0.21155398282392646

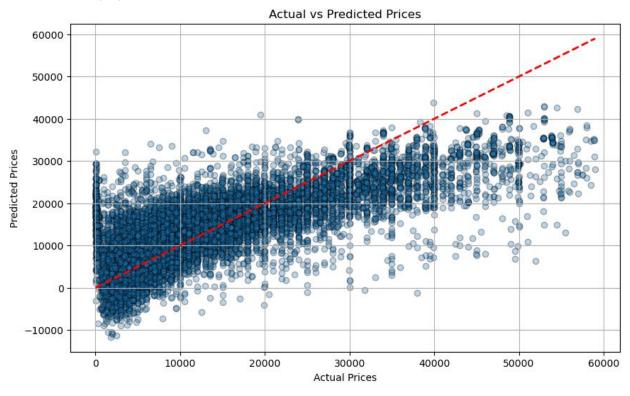


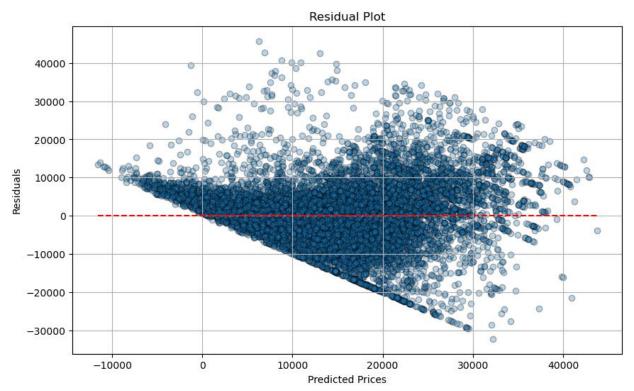
2. Model 2 – linear regression model trained using the two numerical features and the above listed categorical features.

Model Evaluation Metrics:

Mean Squared Error (MSE): 17041510.93282232

R-Squared (R2): 0.8629295735145145





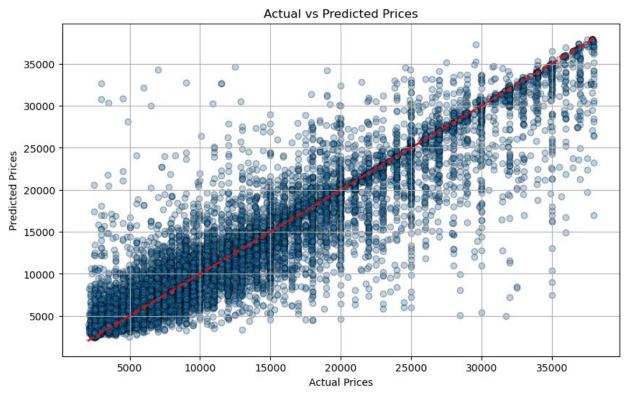
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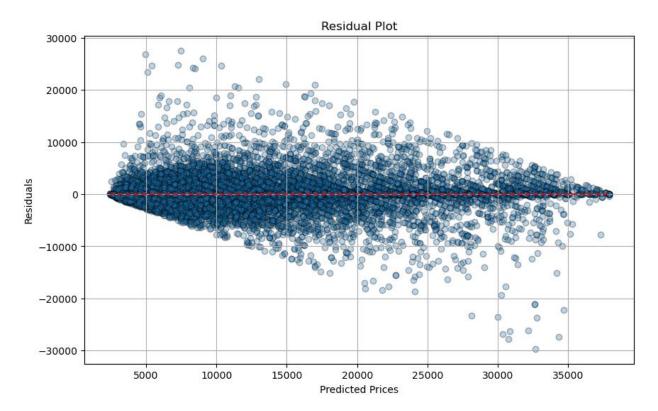
3. Model 3 – random forest regressor model trained using the two numerical features and the above listed categorical features.

Model Evaluation Metrics:

Mean Squared Error (MSE): 17041510.93282232

R-Squared (R2): 0.8629295735145145



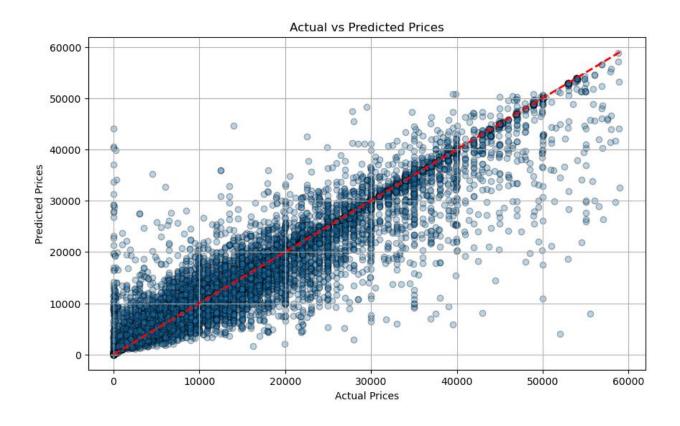


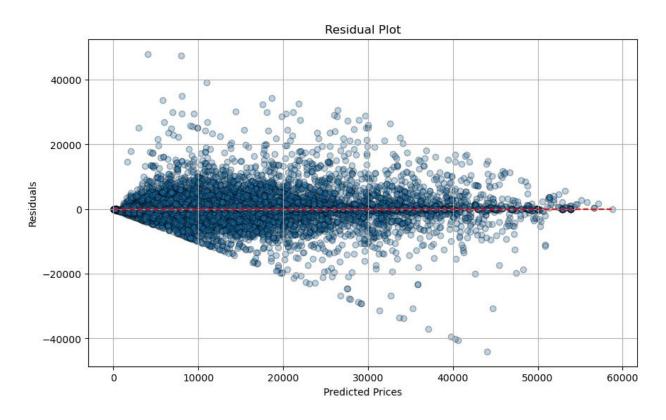
4. Model 4 – grid search for the random forest regressor model.

Model Evaluation Metrics:

Mean Squared Error: 17039247.069625437

R^2 Score: 0.8629477825040298





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### **Evaluation**

The second model surpasses the first model significantly, and the third model surpasses the second one significantly. The grid search model closely resembles the third model with a slight improvement.