

Exploring the Influencing Factors of Automobile Prices Based on the Dataset for Vehicle Price Prediction

Boyu Wang ^a

School of Data Science and Artificial Intelligence, Dongbei University of Finance and Economics, Dalian, Liaoning, 116025, China

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Abstract: With the continuous development of the automotive market, the research on the influencing factors of car prices plays a crucial role in industry decision-making and consumer guidance. However, previous studies on the factors affecting car prices have only remained at the level of single variable analysis, lacking a systematic analysis of the combined influence of multiple variables, making it difficult to provide a scientific basis for market behavior. Therefore, this paper aims to study the influence mechanism of key variables such as production year, engine displacement, and mileage on car prices. This study uses descriptive statistics, correlation analysis, and multiple linear regression models to quantify the impact of each variable on car prices. To address the issue of insufficient variable correlation analysis, correlation analysis is used to explore the initial connections between variables. Further, through multiple linear regression models, the specific impact of each factor on price is clarified. Based on 10,000 car data, the model was constructed, and it was found that the production year has the most significant impact on price, followed by engine displacement and mileage, with the model's explanatory power reaching 87.6%. The number of doors and the number of owners have no statistically significant impact on price. These conclusions provide consumers with a "year + mileage" dual-core evaluation framework.


1 INTRODUCTION

In the automotive market, the influencing factors of car prices have always been the focus of attention for consumers, automakers, and dealers. Understanding these factors not only helps consumers make wiser decisions when purchasing a car but also enables automakers to formulate more competitive pricing strategies and provides a basis for dealers to optimize inventory management and cost control. Therefore, exploring the influencing factors of car prices has significant practical significance.

Previous studies have identified multiple key factors that affect car prices. Firstly, brand is one of the important factors influencing car prices. Luxury brands such as BMW and Mercedes-Benz, with their high quality and good market reputation, can usually sell at higher prices (Wu, 2025). Secondly, different models within the same brand also show significant price differences. For instance, Toyota's Corolla, as a family car for the general public, is relatively

affordable, while the Prado, as a mid-to-large SUV with higher configurations, has a correspondingly higher price (Luo, 2025). Additionally, the production year also has a significant impact on car prices. With technological advancements, newly produced cars typically adopt more advanced technologies and manufacturing processes, offering better performance and safety, and thus command higher prices (Zhao, 2023).

Engine size is another crucial factor. Cars equipped with larger engines usually have stronger power, but due to their higher manufacturing difficulty and cost, they tend to be more expensive (Deng et al., 2022). Meanwhile, with the increasing awareness of environmental protection and the innovation of energy technologies, the type of fuel has become an increasingly significant factor affecting car prices. Traditional gasoline and diesel vehicles have obvious price differences due to fuel costs and technological differences. However, hybrid and electric vehicles, due to their high research and

^a <https://orcid.org/0009-0004-4875-0967>

development costs and rapid updates in battery technology, experience greater price fluctuations (Zhao, 2023). Moreover, the type of transmission also affects car prices. Cars with automatic transmissions are generally more expensive than those with manual transmissions because automatic transmissions better suit the driving habits of most consumers (Zhang, 2022).

Although previous studies have achieved certain results in identifying the factors influencing car prices, they generally suffer from the limitations of single data sources and limited sample sizes, leading to an insufficiently comprehensive coverage of scenarios (Chen, 2023). Moreover, the existing analytical methods are mostly simple correlation analyses, failing to deeply explore the complex interactions among various factors, leaving room for further research.

To break through the limitations of previous studies, this paper mainly uses correlation analysis and multiple linear regression (MLR) analysis methods to explore the influence of various factors on car prices, reveal their intrinsic relationships, and provide scientific decision-making bases for all parties in the automotive market, promoting its healthy development.

2 METHODS

2.1 Data Sources and Description

This paper is based on a car price prediction dataset containing 10,000 entries from the Kaggle website, which covers various aspects of information such as brand, model, production year, engine size, fuel type, transmission type, mileage, number of doors, and the number of previous owners (Mustafa, 2025; Wang, 2023). Before the analysis, this study cleaned and preprocessed the data to ensure its reliability. All the data will be used in the subsequent research.

2.2 Selection and Explanation of Indicators

Table 1 presents the selection and explanation table of key indicators affecting car prices, listing seven variable names such as year and engine size, and briefly explaining the principles by which each variable affects car prices.

Table 1: Key Indicators Selection and Explanation for Factors Affecting Automobile Prices

Number	Variable Name	Brief Description
x_1	Year	New cars are usually more expensive
x_2	Engine Size	Large engines are costly, making the car price higher
x_3	Mileage	The higher the mileage, the lower the price usually is
x_4	Doors	Different door numbers represent different models, and the prices vary
x_5	Owner Count	Frequent changes of ownership lead to a lower price

2.3 Method Introduction

This study employs descriptive statistics, correlation analysis, and MLR models for analysis. Descriptive statistics are used to summarize the characteristics of the data, revealing the central tendency and dispersion of variables through the calculation of means, standard deviations, and other indicators, providing a foundational framework for subsequent analysis. Correlation analysis, based on Pearson's correlation coefficient, quantifies the strength and direction of linear relationships between variables, helping to identify those significantly related to car prices and select core independent variables for model construction. The MLR model is constructed with car price as the dependent variable and x_1 , x_2 ,

x_3 , x_4 , and x_5 as independent variables, with the formula as follows:

$$\text{Price} = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + \beta_3 x_3 + \beta_4 x_4 + \beta_5 x_5 + \varepsilon \quad (1)$$

This method can control the influence of other variables and independently assess the marginal effect of each independent variable on the price. The advantage of the model lies in its strong interpretability, allowing for a direct comparison of the degree of variable influence through standardized coefficients. This method is particularly suitable for the quantitative analysis of the combined influence of multiple variables, breaking through the limitations of single-variable analysis.

3 RESULTS AND DISCUSSION

Through descriptive statistics of the variables, the key results in Table 2 were obtained:

3.1 Descriptive Statistical Analysis

Table 2: Descriptive Statistics of Variables

Variable	Case Number		Average Value	Standard Deviation
	Effective	Missing		
Year	10000	0	2011.540	6.898
Engine Size	10000	0	3.001	1.149
Mileage	10000	0	149239.110	86322.349
Doors	10000	0	3.500	1.110
Owner Count	10000	0	2.990	1.423
Price	10000	0	8852.96	3112.597

Table 2 indicates that the production years of the vehicles are mainly concentrated around 2011, with a relatively concentrated time distribution. The average engine displacement is 3.001, mainly medium displacement. The standard deviation of mileage is as high as 86,322.349, indicating significant differences in vehicle mileage; the number of doors is mainly concentrated between 3 and 4. The number of vehicle owners fluctuates, but is mainly distributed around 3

people. The price of cars has a high degree of dispersion and significant individual differences.

3.2 Correlation Analysis of Variables and Prices

Through the correlation analysis of variables, the key results in Table 3 were obtained:

Table 3: Correlation Coefficient Table of Key Variables Affecting Automobile Prices

Relevance						
	Year	Engine Size	Mileage	Doors	Owner Count	Price
Year	1					
Engine Size	-0.001	1				
Mileage	-0.002	0.015	1			
Doors	0.015	-0.01	0.008	1		
Owner Count	0	0.007	0.006	-0.005	1	
Price	0.663**	0.357**	-0.551**	0.001	0.003	1
** At the 0.01 level (two-tailed), the correlation is significant.						

Table 3 indicates that the production year is significantly positively correlated with the car price; the newer the year, the higher the price. Engine displacement is significantly positively correlated with the price; the larger the displacement, the higher the price. Mileage is significantly negatively correlated with the price; the more mileage, the lower the price. However, the correlation coefficients of the number of doors and the number of previous owners with the car price are close to 0 and do not show a significant correlation, indicating that these two variables have little impact on the car price.

3.3 MLR Analysis

This study constructed an MLR model to explore the key factors influencing car prices. The dependent variable of the model is the car price, and the independent variables include: production year, engine size, mileage, number of doors, and number of previous owners. The results are shown in Table 4:

Table 4: Results of MLR Analysis of Factors Affecting Car Prices

Linear regression analysis results (n=10000)					
	Unstandardized coefficient		Standardized coefficient	<i>t</i>	<i>p</i>
	β	standard error	<i>Beta</i>		
Constant	-592198.767	3196.549	-	-185.262	0.000**
Year	298.801	1.589	0.662	188.033	0.000**
Engine Size	992.914	9.538	0.367	104.105	0.000**
Mileage	-0.020	0.000	-0.555	-157.636	0.000**
Doors	-3.732	9.875	-0.001	-0.378	0.705
Owner Count	7.551	7.704	0.003	0.980	0.327
R ²	0.876				
Adjusted R ²	0.876				
<i>F</i>	<i>F</i> (5,9994)=14132.594, <i>p</i> =0.000				
Note: Dependent variable = Price					
* <i>p</i> <0.05 ** <i>p</i> <0.01					

As shown in Table 3, with price as the dependent variable, year and engine size have a significant positive impact on price ($p < 0.01$), with standardized coefficients of 0.662 and 0.367 respectively; mileage has a significant negative impact on price ($p < 0.01$), with a standardized coefficient of -0.555; the number of doors and the number of owners have no significant impact on price (p values are 0.705 and 0.327 respectively), with standardized coefficients close to 0. The model's R^2 is 0.876, and the adjusted R^2 is also 0.876, indicating that the model explains 87.6% of the price variation, validating the model's effectiveness and suggesting that the model accounts for approximately 87.6% of the price variation, with a very good fit. The F -statistic value is 14132.594, with a p -value < 0.01 , indicating that the model as a whole is significant. In summary, year, engine size, and mileage are the main factors affecting price. The prediction formula is:

$$\text{Price} = -592198.767 + 298.801x_1 + 992.914x_2 - 0.020x_3 - 3.732x_4 + 7.551x_5 \quad (2)$$

3.4 Limitations Analysis

This study has certain limitations in analyzing the factors influencing car prices. Although the introduction emphasized the significant role of brand and model in determining prices, in the actual modeling process, due to the high diversity of brands and models in the dataset leading to an excessive number of categorical variables and insufficient sample sizes for some niche brands or unpopular models, to control the complexity of the model, these variables were ultimately not included in the MLR analysis(Hao, 2025). This means the model failed to

fully reflect the impact of brand premiums and the differences between popular and niche models on prices, which may lead to deviations in the interpretation of real market prices.

Furthermore, the study only explored the relationship between variables based on a linear model, while in reality, car prices may be influenced by nonlinear factors or interaction effects. For instance, the impact of engine displacement on price may vary depending on brand positioning, and the depreciation effect of mileage may show nonlinear attenuation over the years. Additionally, fuel type and transmission type were not deeply explored as categorical variables in the regression analysis, which, to some extent, simplified the complex influence mechanisms of their technical costs and market acceptance on prices.

The singularity of the data source also limits the coverage of the sample in terms of the range of car models and regional market environments, resulting in insufficient explanatory power for niche brands or special models (Wu et al., 2025). Future research could attempt to include brand and model as categorical variables, adopt nonlinear models, or combine more diverse data sources to more comprehensively reveal the price influence mechanism under the interaction of multiple variables.

4 CONCLUSION

This study, through descriptive statistics, correlation analysis, and MLR models, found that production year, engine displacement, and mileage are the main factors affecting car prices. The newer the year, the

larger the displacement, and the lower the mileage, the higher the price, with the year having a relatively more significant impact. The number of doors and the number of previous owners have no significant impact on price. The model's explanatory power reached 87.6%, indicating that the selected variables can effectively explain price variations. The research results provide data support for consumers' car purchase decisions, car manufacturers' pricing strategies, and dealers' inventory management.

However, the study has certain limitations. Due to the characteristics of the data and the complexity of the model, variables such as brand and model that may have significant impacts on prices were not included in the analysis, and the differences in brand premiums and model positioning were not fully reflected. Only a linear model was used to examine the relationship between variables, which may overlook nonlinear effects or interaction effects. The sample coverage of a single dataset is limited, and its explanatory power for niche models or specific markets is insufficient. Future research could include brand and model variables, combine nonlinear models or interaction terms, and more finely depict the mechanism of multi-variable interaction, and introduce more diverse data sources to enhance the applicability of the conclusions to complex market scenarios.

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