

Vehicle Price Prediction

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Introduction

- Vehicle price prediction is a task that has become increasingly relevant in recent years, as the automotive industry has seen a surge in demand for personalized vehicles.
- One of the main challenges in this field is predicting the price of a vehicle accurately, which involves analyzing numerous factors such as the make and model, mileage, year, and condition of the vehicle.
- It is possible to leverage historical data to predict the price of a vehicle with a high degree of accuracy.
- Supervised learning algorithms can be trained on past data to identify patterns and relationships between the variables, allowing for the creation of a predictive model that can estimate the price of a vehicle based on its features.
- This project explores the use of various supervised learning algorithms for vehicle price prediction.



Dataset

- DVM Car Dataset: A large-scale dataset for Automotive Applications. DVM
 Car Dataset: A Large-Scale Dataset for Automotive Applications.
 https://deepvisualmarketing.github.io/
- Historical inflation rates: 1914-2023. US Inflation Calculator | Easily calculate how the buying power of the U.S. dollar has changed from 1913 to 2023. Get inflation rates and U.S. inflation news. (2023, April 12).

https://www.usinflationcalculator.com/inflation/historical-inflation-rates/



Column Definitions

Unnamed: 0 -	1.00	0.01	0.05	0.01	-0.01	0.06	-0.01	-0.04	-0.01	0.01	0.11	-0.04	0.04	0.04	0.01	0.07	0.01	-0.03	0.07	-0.02	-0.02	0.09	0.06		1.00
Adv_year -	0.01	1.00	-0.36	-0.01	0.06	0.03	-0.08	-0.06	-0.07	-0.06	-0.02	0.01	-0.04	-0.05	0.03	-0.07	0.03	0.03	0.55	0.04	-0.08	-0.05	0.38		
Adv_month -	0.05	-0.36	1.00	-0.04	0.02	0.05	-0.01	-0.03	-0.01	0.00	0.00	-0.02	-0.01	-0.01	-0.02	-0.00	-0.01	-0.03	0.31	0.07	-0.01	0.02	0.14		0.75
Reg_year -	0.01	-0.01	-0.04	1.00	-1.00	-0.71	-0.09	0.40	0.07	-0.41	0.06	0.12	0.27	0.08	0.48	0.05	0.06	0.18	-0.04	-0.04	0.03	-0.24	0.06		- 0.75
Years -	-0.01	0.06	0.02	-1.00	1.00	0.71	0.08	-0.41	-0.07	0.41	-0.06	-0.12	-0.28	-0.09	-0.48	-0.06	-0.06	-0.18	0.07	0.04	-0.04	0.23	-0.04		
Runned_miles -	0.06	0.03	0.05	-0.71	0.71	1.00	0.11	-0.36	-0.04	0.27	0.01	-0.03	-0.14	0.09	-0.28	0.00	0.07	-0.03	0.06	0.05	-0.01	0.20	0.13		- 0.50
Engin_size -	-0.01	-0.08	-0.01	-0.09	0.08	0.11	1.00	0.54	0.90	0.64	0.26	0.12	0.37	0.65	-0.52	0.75	-0.04	-0.12	-0.07	-0.18	0.81	0.79	0.15		5.000
Price -	-0.04	-0.06	-0.03	0.40	-0.41	-0.36	0.54	1.00	0.72	0.19	0.14	0.02	0.34	0.33	-0.15		-0.14	-0.12	-0.07	-0.10	0.74	0.33	-0.00		
Engine_power -	-0.01	-0.07	-0.01	0.07	-0.07	-0.04	0.90	0.72	1.00		0.23	0.02	0.39		-0.49	0.88	-0.13	-0.19	-0.07	-0.17	0.84	0.69	0.05		- 0.25
Annual_tax -	0.01	-0.06	0.00	-0.41	0.41	0.27	0.64	0.19		1.00	0.13	0.14	0.17	0.38	-0.81	0.45	-0.01	-0.13	-0.04	-0.10	0.51		-0.08		
Wheelbase -	0.11	-0.02	0.00	0.06	-0.06	0.01	0.26	0.14	0.23	0.13	1.00	0.16	0.20	0.39	-0.07	0.22	0.16	0.16	-0.02	-0.09	0.23	0.22	0.13		
Height -	-0.04	0.01	-0.02	0.12	-0.12	-0.03	0.12	0.02	0.02	0.14	0.16	1.00	0.23	0.30	-0.10	-0.22	0.50	0.49	-0.00	-0.04	0.08	0.16	0.23		- 0.00
Width -	0.04	-0.04	-0.01	0.27	-0.28	-0.14	0.37	0.34	0.39	0.17	0.20	0.23	1.00	0.53	-0.08	0.37	0.17	0.12	-0.03	-0.00	0.38	0.27	0.21		
Length -	0.04	-0.05	-0.01	0.08	-0.09	0.09		0.33		0.38	0.39	0.30		1.00	-0.25		0.35	0.21	-0.06	-0.17			0.35		
Average_mpg -	0.01	0.03	-0.02	0.48	-0.48	-0.28	-0.52	-0.15	-0.49	-0.81	-0.07	-0.10	-0.08	-0.25	1.00	-0.37	0.06	0.19	0.01	0.08	-0.41		0.22		0.25
Top_speed -	0.07	-0.07	-0.00	0.05	-0.06	0.00	0.75	0.57	0.88	0.45	0.22	-0.22	0.37		-0.37	1.00	-0.19	-0.27	-0.06	-0.17	0.69	0.58	0.10		
Seat_num -	0.01	0.03	-0.01	0.06	-0.06	0.07	-0.04	-0.14	-0.13	-0.01	0.16	0.50	0.17	0.35	0.06	-0.19	1.00	0.58	0.02	0.04	-0.09	0.03	0.26		
Door_num -	-0.03	0.03	-0.03	0.18	-0.18	-0.03	-0.12	-0.12	-0.19	-0.13	0.16	0.49	0.12	0.21	0.19	-0.27	0.58	1.00	0.01	0.06	-0.15	-0.10	0.22		0.50
Inflation -	0.07	0.55	0.31	-0.04	0.07	0.06	-0.07	-0.07	-0.07	-0.04	-0.02	-0.00	-0.03	-0.06	0.01	-0.06	0.02	0.01	1.00	0.06	-0.07	-0.03	0.56		
Sales -	-0.02	0.04	0.07	-0.04	0.04	0.05	-0.18	-0.10	-0.17	-0.10	-0.09	-0.04	-0.00	-0.17	0.08	-0.17	0.04	0.06	0.06	1.00	-0.15	-0.06	-0.03		0.75
Entrée_price -	-0.02	-0.08	-0.01	0.03	-0.04	-0.01	0.81	0.74	0.84	0.51	0.23	0.08	0.38		-0.41	0.69	-0.09	-0.15	-0.07	-0.15	1.00	0.62	0.08		0.75
Gas_emission -	0.09	-0.05	0.02	-0.24	0.23	0.20	0.79	0.33	0.69	0.65	0.22	0.16	0.27			0.58	0.03	-0.10	-0.03	-0.06	0.62	1.00	0.09		
Fuel_type.1 -	0.06	0.38	0.14	0.06	-0.04	0.13	0.15	-0.00	0.05	-0.08	0.13	0.23	0.21	0.35	0.22	0.10	0.26	0.22	0.56	-0.03	0.08	0.09	1.00		
	Unnamed: 0 -	Adv_year -	Adv_month -	Reg_year -	Years -	Runned_miles -	Engin_size -	Price -	Engine_power -	Annual tax -	Wheelbase -	Height -	Width	Length -	Average_mpg -	- pads dg	Seat_num -	- Door num	Inflation -	Sales -	Entrée_price -	Gas_emission -	Fuel_type.1 -		



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Before >>	Unnamed: 0 Maker Genmodel Genmodel_id Adv_id Adv_year Adv_month Reg_year Years Bodytype Runned_miles Engin_size Gearbox Fuel_type Price Engine_power Annual_tax Wheelbase Height Width Length Average_mpg Top_speed Seat_num Door_num Inflation Sales Entrée_price Gas_emission Fuel_type.1 dtype: object	int64 object object object int64 int64 int64 int64 object int64 float64 float64 float64 int64 float64	After >>	Unnamed: 0 Adv_year Adv_month Reg_year Years Bodytype Runned_miles Engin_size Gearbox Fuel_type Price Engine_power Annual_tax Wheelbase Height Width Length Average_mpg Top_speed Seat_num Door_num Inflation Sales Entrée_price Gas_emission Fuel_type.1 dtype: object	int64 int64 int64 int64 int64 object int64 float64 float64 int64 int64 int64 int64 int64 int64 int64 int64 int64 int64 int64 int64 float64	Final >>	Unnamed: 0 Adv_year Adv_month Reg_year Years Runned_miles Engin_size Price Engine_power Annual_tax Wheelbase Height Width Length Average_mpg Top_speed Seat_num Door_num Inflation Sales Entrée_price Gas_emission Fuel_type.1 Bodytype_Convertible Bodytype_Saloon Bodytype_Limousine Bodytype_Limousine Bodytype_Window Van Gearbox_Manual Fuel_type_Diesel Fuel_type_Hybrid Fuel_type_Hybrid Fuel_type_Hybrid Fuel_type_Hybrid Fuel_type_Hybrid Fuel_type_Hybrid Fuel_type_Petrol Fuel_type_Pet

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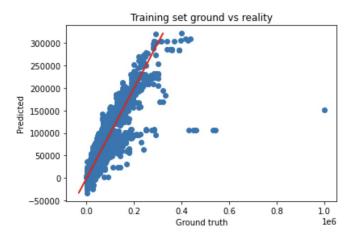
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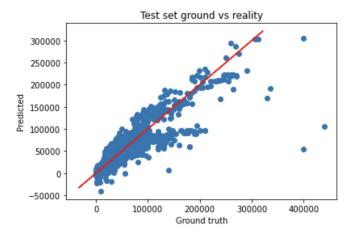


Polynomial Regression

Training set Mean Squared Error (MSE): 39898541.767 Training set R^2 score: 0.871

Training set Mean Squared Error (MSE): 39898541.767







OLS Regression Results

Dep. Variable:	Price x	R-squared:	0.719
Model:	0LS	Adj. R-squared:	0.719
Method:	Least Squares	F-statistic:	2.564e+04
Date:	Tue, 09 May 2023	<pre>Prob (F-statistic):</pre>	0.00
Time:	23:25:35	Log-Likelihood:	-1.5915e+06
No. Observations:	150686	AIC:	3.183e+06
Df Residuals:	150670	BIC:	3.183e+06
Df Model:	15		
Covariance Type:	nonrobust		

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const	1.887e+04	4690.171	4.023	0.000	9676.551	2.81e+04	
Years_x	-1205.5911	8.570	-140.678	0.000	-1222.388	-1188.794	
Entrée_price_x	0.7699	0.002	380.510	0.000	0.766	0.774	
Runned_miles_x	-0.0497	0.001	-61.877	0.000	-0.051	-0.048	
Length_x	-3.0603	0.098	-31.234	0.000	-3.252	-2.868	
Seat_num_x	-666.6300	44.947	-14.831	0.000	-754.726	-578.534	
Bodytype_Convertible	1843.8668	4676.919	0.394	0.693	-7322.799	1.1e+04	
Bodytype_Coupe	4485.2271	4675.823	0.959	0.337	-4679.290	1.36e+04	
Bodytype_Estate	2944.2580	4674.976	0.630	0.529	-6218.601	1.21e+04	
Bodytype_Hatchback	2160.1706	4674.156	0.462	0.644	-7001.080	1.13e+04	
Bodytype_Limousine	-5.747e+04	4828.462	-11.902	0.000	-6.69e+04	-4.8e+04	
Bodytype_MPV	1446.2631	4674.483	0.309	0.757	-7715.629	1.06e+04	
Bodytype_SUV	1621.9588	4674.067	0.347	0.729	-7539.118	1.08e+04	
Bodytype_Saloon	-933.1512	4674.604	-0.200	0.842	-1.01e+04	8228.978	
Bodytype_Window Van	1.42e+04	5167.320	2.748	0.006	4073.899	2.43e+04	
<pre>Gas_emission_x</pre>	-3.4861	0.676	-5.159	0.000	-4.810	-2.162	
Omnibuci	20277	5 067 Durl	ain Watcon.		2 000	λ	

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Omnibus:	282//6.06/	Durbin-watson:	2.000					
Prob(Omnibus):	0.000	Jarque-Bera (JB):	4080192985.611					
Skew:	13.447	Prob(JB):	0.00					
Kurtosis:	808.689	Cond. No.	4.28e+07					

Notes:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

[2] The condition number is large, 4.28e+07. This might indicate that there are strong multicollinearity or other numerical problems.

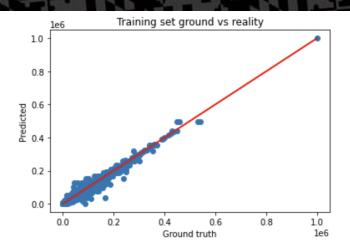
Training set R^2 score: 0.719

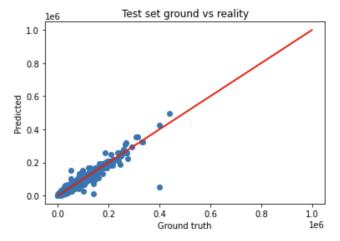


Decision Tree

Training set Mean Squared Error (MSE): 8099126.783 Training set R^2 score: 0.974

Test set Mean Squared Error (MSE): 13731375.131



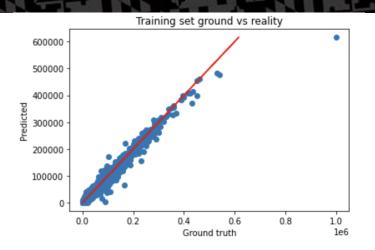


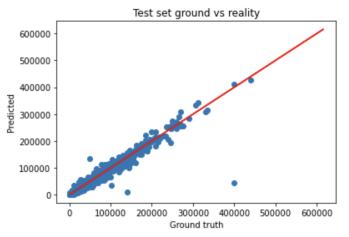


Random Forest

Training set Mean Squared Error (MSE): 6754562.675 Training set R^2 score: 0.978

Test set Mean Squared Error (MSE): 10515721.002 Test set R^2 score: 0.964



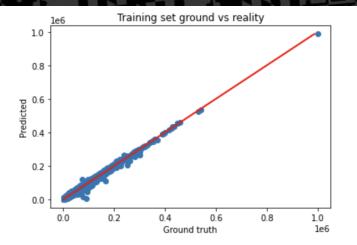


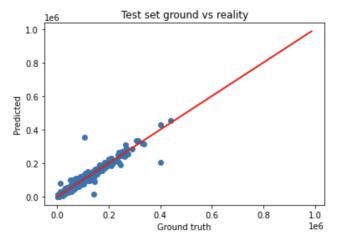


Gradient Boosting

Training set Mean Squared Error (MSE): 2643583.247 Training set R^2 score: 0.991

Test set Mean Squared Error (MSE): 7291046.840







Bagging Regressor

R^2 score for train dataset = 0.9914

R^2 score for test dataset = 0.9738

