

Vehicle Price Prediction

Presented by:
Praveen Kumar Gonuguntla
Pavan Kumar Boddu
Hemanth Varma Dantuluri
Koushik Mannaru
Sreeja Komma Reddy

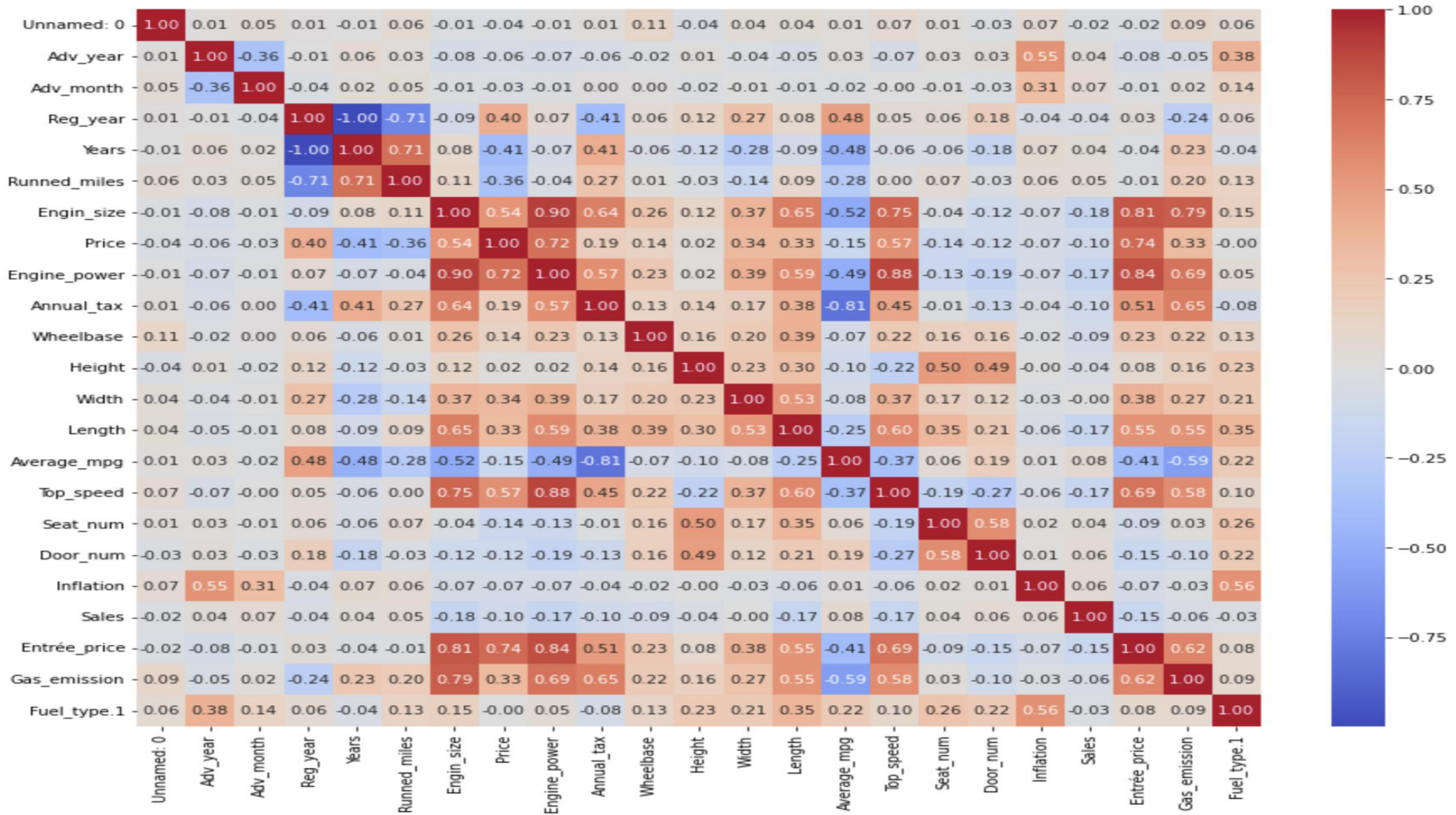
Introduction

- 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



0_df.dtypes

```

Unnamed: 0      int64
Maker           object
Genmodel        object
Genmodel_id     object
Adv_id          object
Adv_year        int64
Adv_month       int64
Reg_year        int64
Years           int64
Bodytype        object
Runned_miles    int64
Engin_size      float64
Gearbox         object
Fuel_type       object
Price           int64
Engine_power    float64
Annual_tax      float64
Wheelbase       int64
Height          int64
Width           int64
Length          int64
Average_mpg     int64
Top_speed       int64
Seat_num        int64
Door_num        int64
Inflation       float64
Sales           int64
Entrée_price    int64
Gas_emission    int64
Fuel_type.1     float64
dtype: object

```

df.dtypes

```

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df_encoded.dtypes

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Top_speed       int64
Seat_num        int64
Door_num        int64
Inflation       float64
Sales           int64
Entrée_price    int64
Gas_emission    int64
Fuel_type.1     float64
Bodytype_Convertible  uint8
Bodytype_Coupe      uint8
Bodytype_Estate     uint8
Bodytype_Hatchback  uint8
Bodytype_Limousine  uint8
Bodytype_MPV        uint8
Bodytype_SUV        uint8
Bodytype_Saloon     uint8
Bodytype_Window Van uint8
Gearbox_Manual      uint8
Fuel_type_Diesel    uint8
Fuel_type_Hybrid    uint8
Fuel_type_Hybrid    Diesel/Electric Plug-in  uint8
Fuel_type_Hybrid    Petrol/Electric          uint8
Fuel_type_Hybrid    Petrol/Electric Plug-in  uint8
Fuel_type_Petrol    uint8
Fuel_type_Petrol    Ethanol                  uint8
dtype: object

```

Before >>

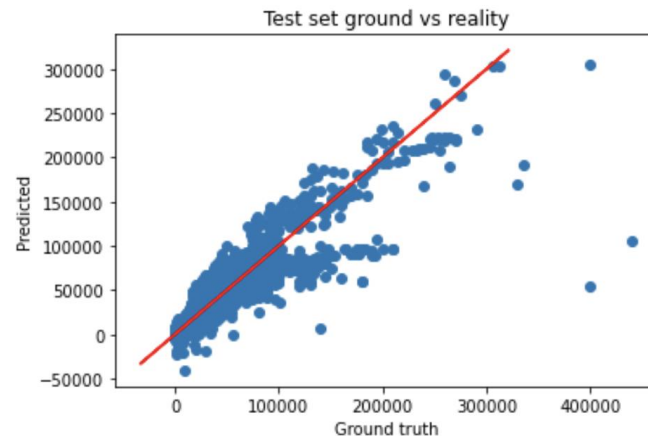
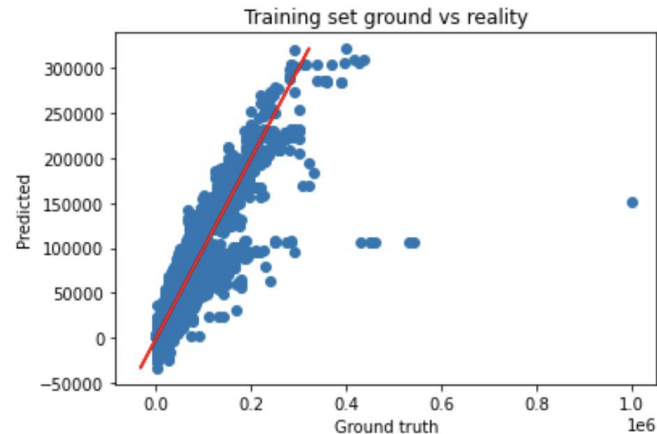
After >>

Final >>

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
Test set R^2 score: 0.876



OLS Regression Results

```
=====
Dep. Variable:      Price_x      R-squared:      0.719
Model:              OLS          Adj. R-squared:    0.719
Method:             Least Squares  F-statistic:    2.564e+04
Date:               Tue, 09 May 2023  Prob (F-statistic): 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
=====
```

	coef	std err	t	P> t	[0.025	0.975]
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
Gas_emission_x	-3.4861	0.676	-5.159	0.000	-4.810	-2.162

```
=====
Omnibus:            282776.067  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 R² score: 0.719

Test set R² score: 0.722

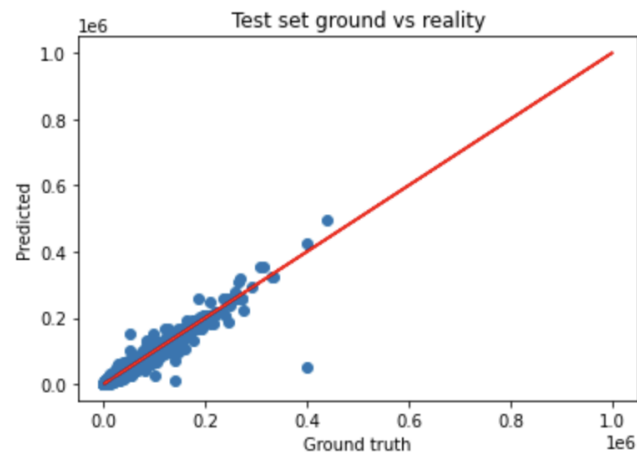
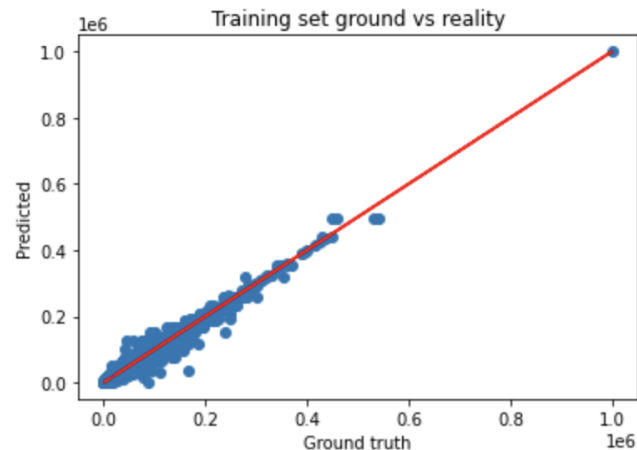
Decision Tree

Training set Mean Squared Error (MSE): 8099126.783

Training set R^2 score: 0.974

Test set Mean Squared Error (MSE): 13731375.131

Test set R^2 score: 0.953



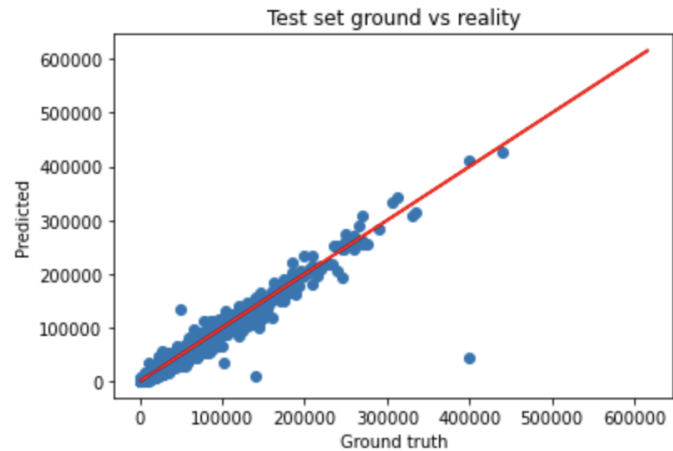
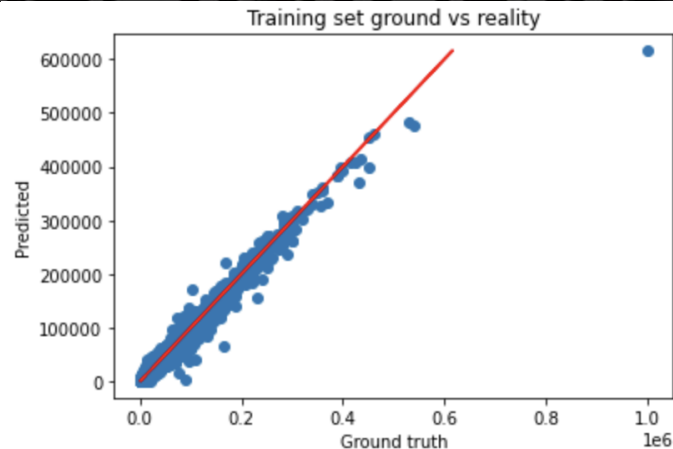
Random Forest

Training set Mean Squared Error (MSE): 6754562.675

Training set R^2 score: 0.978

Test set Mean Squared Error (MSE): 10515721.002

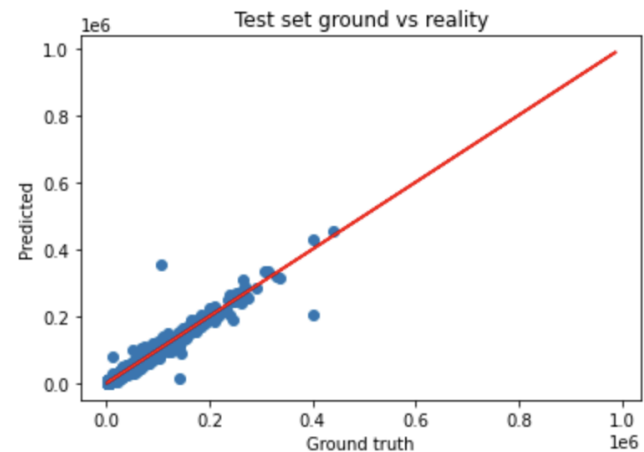
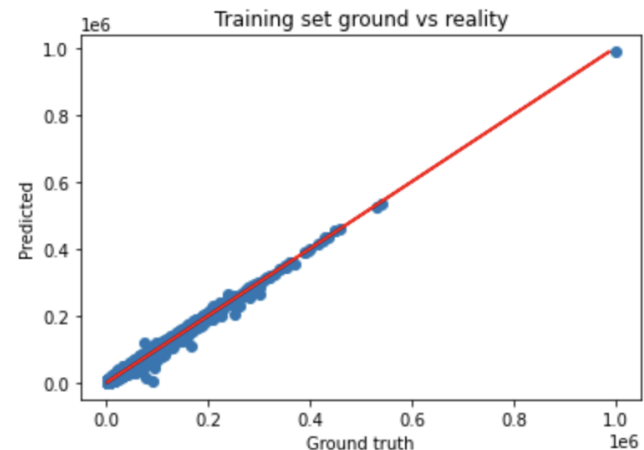
Test set R^2 score: 0.964



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
Test set R^2 score: 0.975



Bagging Regressor

R^2 score for train dataset = 0.9914

R^2 score for test dataset = 0.9738

