

**Minor Project
Report Submitted
For**

ARTIFICIAL INTELLIGENCE-(UCS411)

Submitted by:

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B.E Second Year

Submitted To

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ABSTRACT

Deciding whether a used car is worth the posted price when you see listings online can be difficult. Several factors, including mileage, make, model, year, etc. can influence the actual worth of a car. From the perspective of a seller, it is also a dilemma to price a used car appropriately[2-3]. Based on existing data, the aim is to use machine learning algorithms to develop models for predicting used car prices.

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DECLARATION

I, the undersigned solemnly declare that the project report CAR PRICE PREDICTION USING ML is based on my own work.

I assert the statements made and conclusions drawn are an outcome of my research work. I further certify that

- I. The work contained in the report is original and has been done by me under the general supervision of my supervisor.
- II. The work has not been submitted to any other Institution for any other degree/diploma/certificate in this university or any other University of India or abroad.
- III. We have followed the guidelines provided by the university in writing the report.

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INTRODUCTION

This project focuses on improving the accuracy of existing Machine Learning Data Models using the Linear Regression Model.

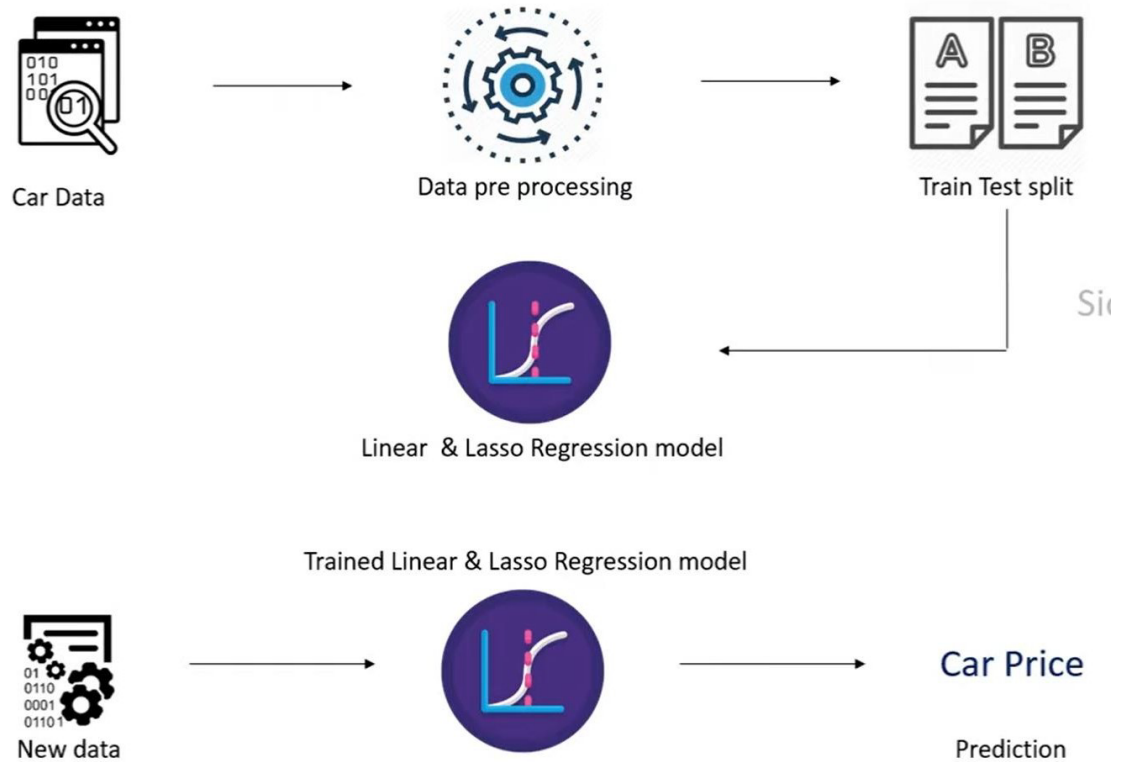
A linear regression model describes the relationship between a dependent variable, y , and one or more independent variables, X . The dependent variable also called as response variable. Independent variables also known as predictor or explanatory variables.

The biggest advantage of linear regression models is linearity: It makes the estimation procedure simple and, most importantly, these linear equations have an easy to understand interpretation on a modular level.

The price of a new car in the industry is fixed by the manufacturer with some additional costs incurred by the Government in the form of taxes. So, customers buying a new car can be assured of the money they invest to be worthy. But, due to the increased prices of new cars and the financial incapability of the customers to buy them, Used Car sales are on a global increase. Therefore, there is an urgent need for a Car Price Prediction system which effectively determines the worthiness of the car using a variety of features. Existing System includes a process where a seller decides a price randomly and buyer has no idea about the car and its value in the

present day scenario. In fact, seller also has no idea about the car's existing value or the price he should be selling the car at. To overcome this problem we have developed a model which will be highly effective. Regression Algorithms are used because they provide us with continuous value as an output and not a categorized value. Because of which it will be possible to predict the actual price a car rather than the price range of a car.

Work Flow



Problem Statement

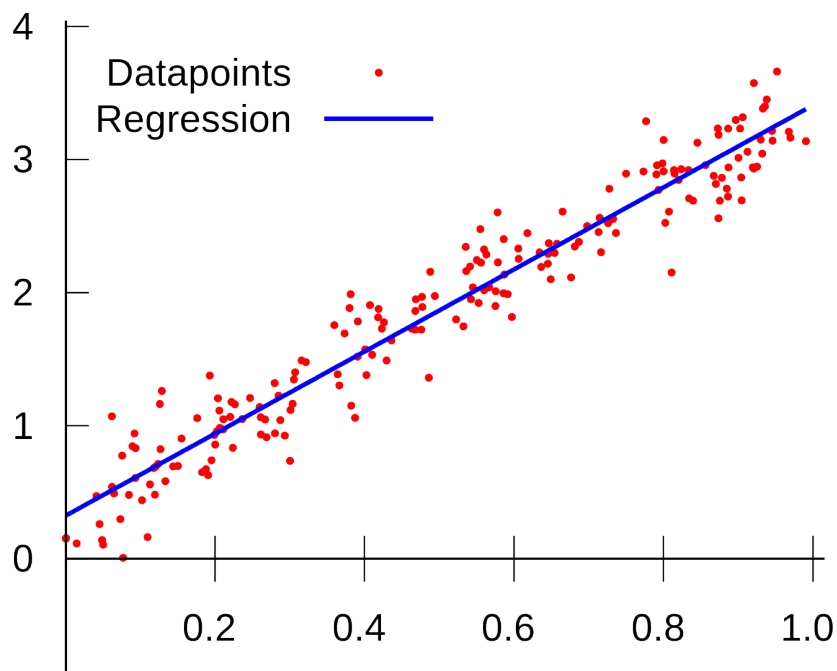
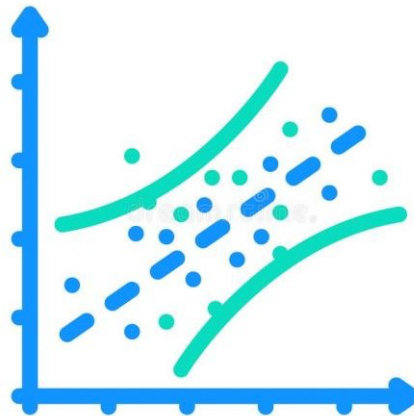
The problem statement is that we have to predict the prices of used cars based on the provided information that includes various fields such as Car brand , year of purchase , fuel type etc .

. Here we are using Linear regression simpler model to predict the price of the used cars.

We need to build a machine learning system that can learn from data and predict what this price can be. For that we will be dividing the data into test and train mode.

Linear Regression

Linear Regression is a machine learning algorithm based on **supervised learning**. It performs a **regression task**. Regression models a target prediction value based on independent variables. It is mostly used for finding out the relationship between variables and forecasting. Different regression models differ based on – the kind of relationship between dependent and independent variables they are considering, and the number of independent variables getting used.

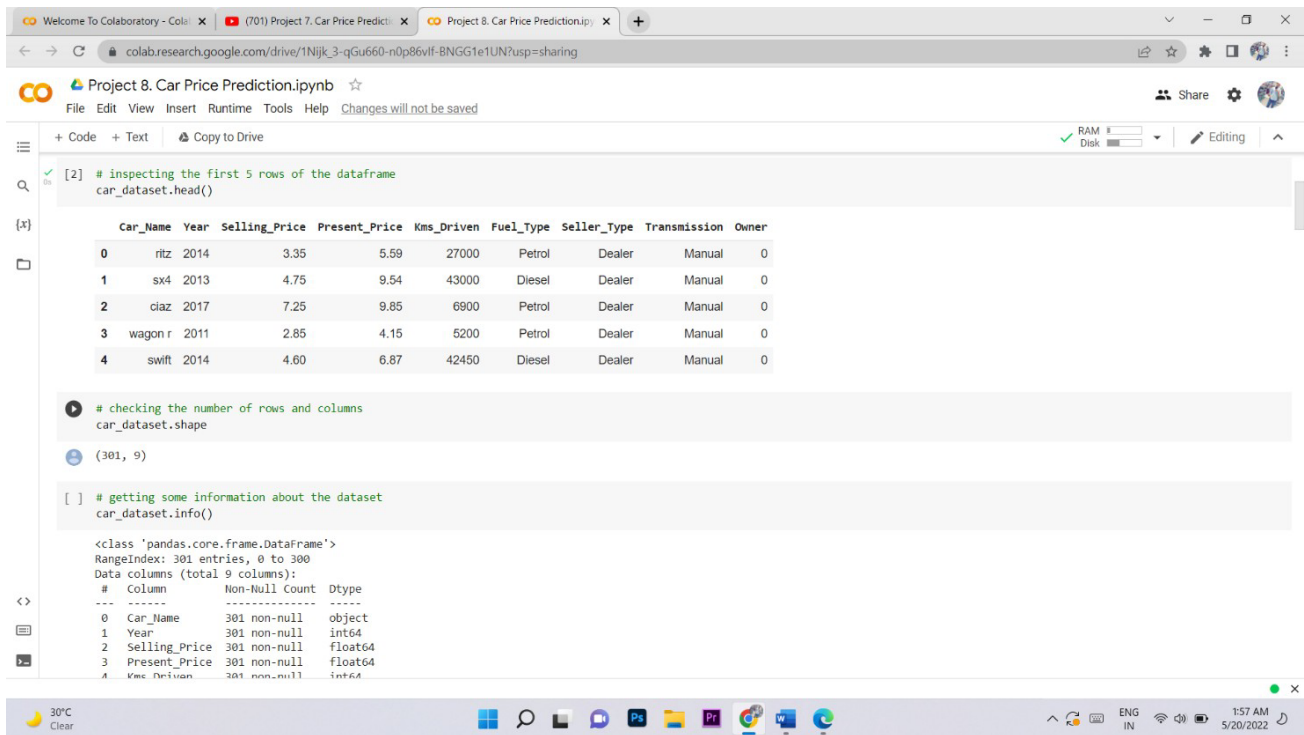


RESULT ANALYSIS

We analysed learning algorithms on our chosen data set. By using the linear regression model we aimed to maintain the linearity of the database so as to predict the data very closely. The biggest advantage of linear regression models is linearity.

Through this project, we found that most suited and, therefore, most common algorithm used to work on big data with linearity, this is very important classifier. It makes the estimation procedure simple and, most importantly, these linear equations have an easy to understand interpretation on a modular level.

Working Screenshots



The screenshot shows a Google Colab notebook titled "Project 8. Car Price Prediction.ipynb". The code cell [2] contains the following code:

```
[2] # inspecting the first 5 rows of the dataframe
car_dataset.head()
```

The output is a table with 9 columns: Car_Name, Year, Selling_Price, Present_Price, Kms_Driven, Fuel_Type, Seller_Type, Transmission, and Owner. The first 5 rows are displayed:

	Car_Name	Year	Selling_Price	Present_Price	Kms_Driven	Fuel_Type	Seller_Type	Transmission	Owner
0	ritz	2014	3.35	5.59	27000	Petrol	Dealer	Manual	0
1	sx4	2013	4.75	9.54	43000	Diesel	Dealer	Manual	0
2	ciaz	2017	7.25	9.85	6900	Petrol	Dealer	Manual	0
3	wagon r	2011	2.85	4.15	5200	Petrol	Dealer	Manual	0
4	swift	2014	4.60	6.87	42450	Diesel	Dealer	Manual	0

The next code cell contains the following code:

```
# checking the number of rows and columns
car_dataset.shape
```

The output is:

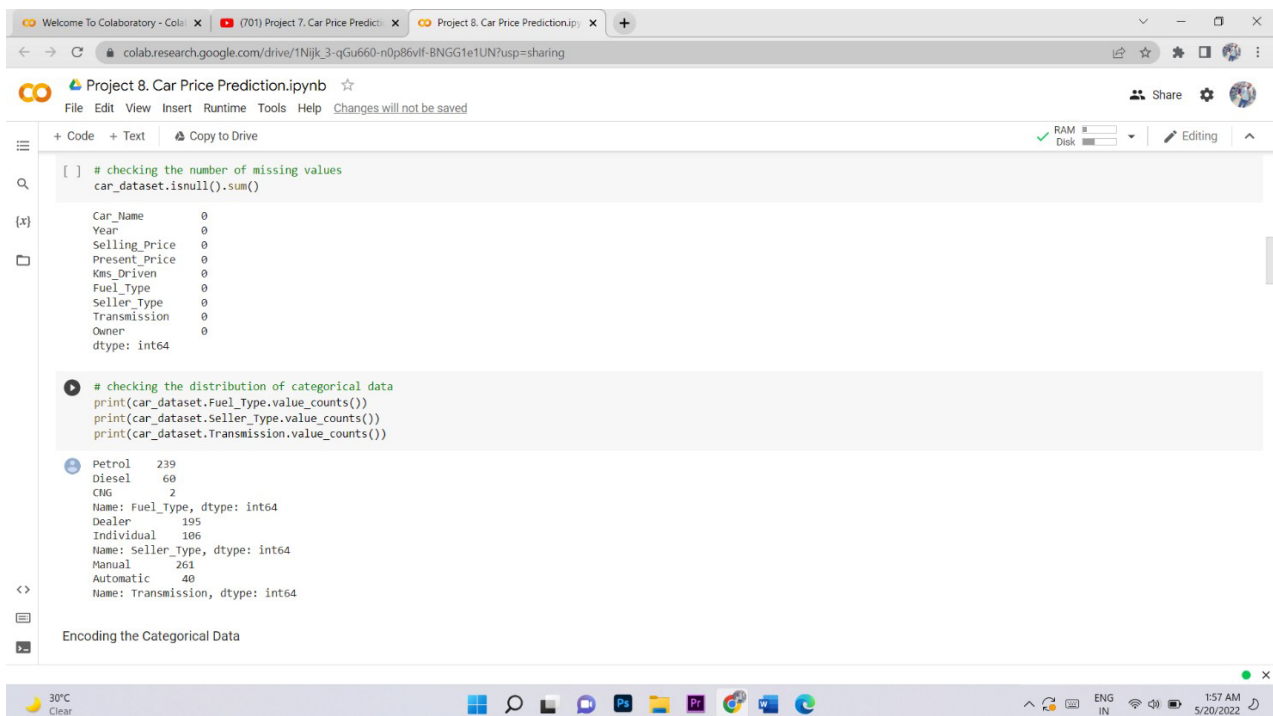
```
(301, 9)
```

The final code cell contains the following code:

```
[ ] # getting some information about the dataset
car_dataset.info()
```

The output is:

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 301 entries, 0 to 300
Data columns (total 9 columns):
#   Column          Non-Null Count  Dtype
---  ---
0   Car_Name        301 non-null   object
1   Year            301 non-null   int64
2   Selling_Price    301 non-null   float64
3   Present_Price    301 non-null   float64
4   Kms_Driven       301 non-null   int64
```



The screenshot shows a Google Colab notebook titled "Project 8. Car Price Prediction.ipynb". The code cell [] contains the following code:

```
[ ] # checking the number of missing values
car_dataset.isnull().sum()
```

The output is:

```
Car_Name      0
Year          0
Selling_Price  0
Present_Price  0
Kms_Driven    0
Fuel_Type     0
Seller_Type   0
Transmission  0
Owner         0
dtype: int64
```

The next code cell contains the following code:

```
# checking the distribution of categorical data
print(car_dataset.Fuel_Type.value_counts())
print(car_dataset.Seller_Type.value_counts())
print(car_dataset.Transmission.value_counts())
```

The output is:

```
Petrol    239
Diesel    60
CNG       2
Name: Fuel_Type, dtype: int64
Dealer    195
Individual 106
Name: Seller_Type, dtype: int64
Manual    261
Automatic  40
Name: Transmission, dtype: int64
```

The final code cell contains the following code:

```
Encoding the Categorical Data
```

Project 8. Car Price Prediction.ipynb

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Splitting the data and Target

```
[ ] X = car_dataset.drop(['Car_Name', 'Selling_Price'], axis=1)
Y = car_dataset['Selling_Price']

[ ] print(X)
```

	Year	Present_Price	Kms_Driven	...	Seller_Type	Transmission	Owner
0	2014	5.59	27000	...	0	0	0
1	2013	9.54	43000	...	0	0	0
2	2017	9.85	6900	...	0	0	0
3	2011	4.15	5200	...	0	0	0
4	2014	6.87	42450	...	0	0	0
...
296	2016	11.60	33988	...	0	0	0
297	2015	5.90	60000	...	0	0	0
298	2009	11.00	87934	...	0	0	0
299	2017	12.50	9000	...	0	0	0
300	2016	5.90	5464	...	0	0	0

[301 rows x 7 columns]

```
[ ] print(Y)
```

0	3.35
1	4.75
2	7.25
3	2.85
4	4.60
...	...
296	9.50
297	4.00
...	...

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Project 8. Car Price Prediction.ipynb

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```
plt.scatter(Y_train, training_data_prediction)
plt.xlabel("Actual Price")
plt.ylabel("Predicted Price")
plt.title("Actual Prices vs Predicted Prices")
plt.show()
```



```
[ ] # prediction on Training data
test_data_prediction = lin_reg_model.predict(X_test)

[ ] # R squared Error
error_score = metrics.r2_score(Y_test, test_data_prediction)
print("R squared Error : ", error_score)
```

R squared Error : 0.8365766715027051

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