



CARPRICE ANALYSIS

Submitted by:
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ACKNOWLEDGMENT

With the covid 19 impact in the market, we have seen lot of changes in the car market. Now some cars are in demand hence making them costly and some are not in demand hence cheaper. One of our clients works with small traders, who sell used cars. With the change in market due to covid 19 impact, our client is facing problems with their previous car price valuation machine learning models. So, they are looking for new machine learning models from new data. We have to make car price valuation model.

INTRODUCTION

- Business Problem Framing

With the covid 19 impact in the market, we have seen lot of changes in the car market. Now some cars are in demand hence making them costly and some are not in demand hence cheaper. One of our clients works with small traders, who sell used cars. With the change in market due to covid 19 impact, our client is facing problems with their previous car price valuation machine learning models. So, they are looking for new machine learning models from new data. We have to make car price valuation model.

- Conceptual Background of the Domain Problem

The project is sub-divided following section. These are:

1. Loading necessary libraries
2. Loading Dataset from a CSV file
3. Summarization of Data to understand Dataset (Descriptive Statistics)
4. Visualization of Data to understand Dataset (Plots, Graphs etc.)
5. Processing the data for modeling
6. skewness and outliers detection for better accuracy
7. Build the model and select the right model and save it

- Review of Literature

There are 14 columns including Name, Location, kilometer driven, fuel-type, engine and main feature is price.

Most of them are object variable and some int variable and main features is float type.

- Motivation for the Problem Undertaken

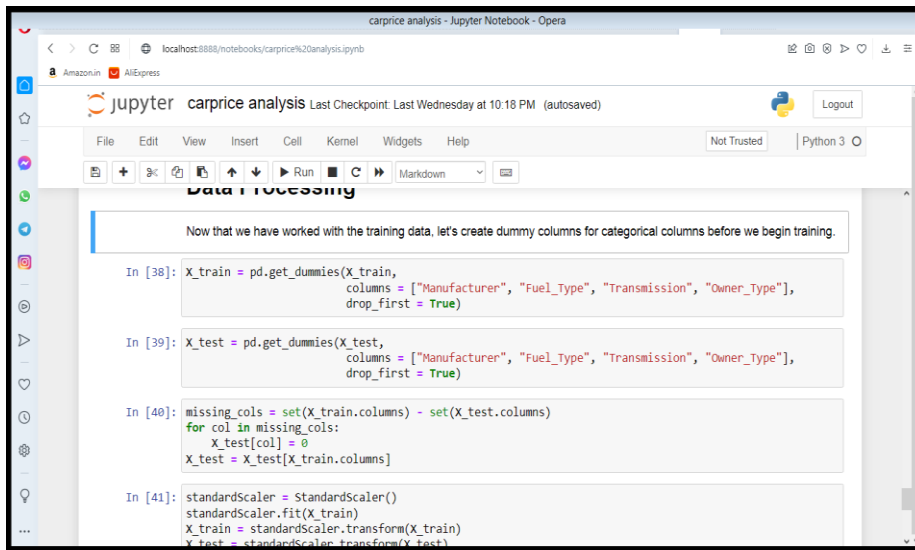
After collecting the data, you need to build a machine learning model. Before model building do all data pre-processing steps. Try different models with different hyper parameters and select the best model.

Follow the complete life cycle of data science. Include all the steps like.

1. Data Cleaning
2. Exploratory Data Analysis
3. Data Pre-processing
4. Model Building
5. Model Evaluation
6. Selecting the best model

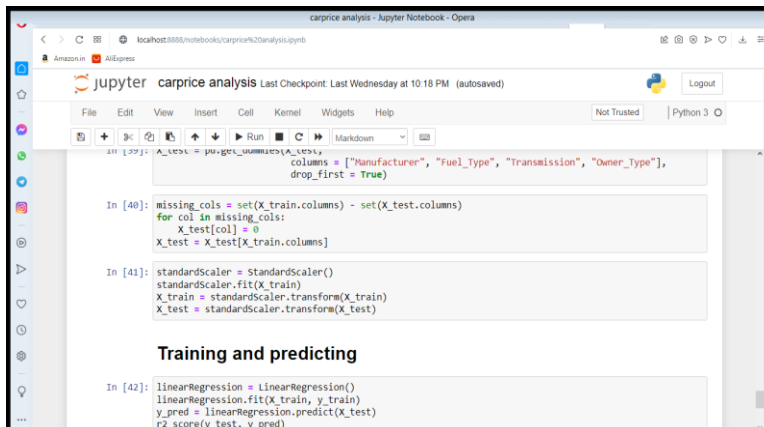
Analytical Problem Framing

- Mathematical/ Analytical Modeling of the Problem
Data Processing: We have worked with the training data, let's create dummy columns for categorical columns before we begin training.



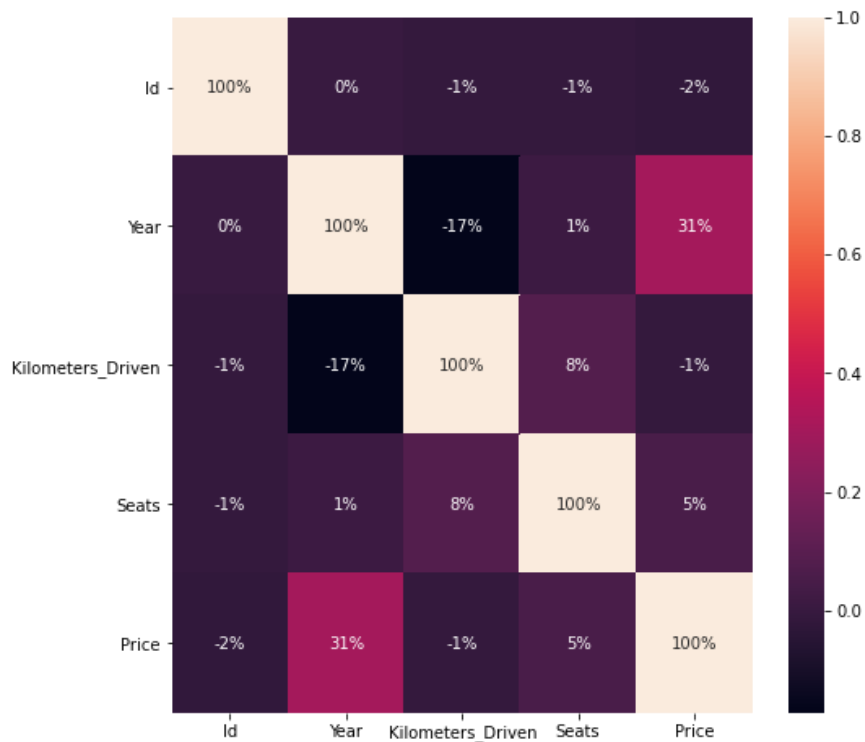
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Amazon.in AliExpress
jupyter carprice analysis Last Checkpoint: Last Wednesday at 10:18 PM (autosaved)
File Edit View Insert Cell Kernel Widgets Help Not Trusted Python 3
Data Processing
Now that we have worked with the training data, let's create dummy columns for categorical columns before we begin training.
In [38]: X_train = pd.get_dummies(X_train,
columns = ["Manufacturer", "Fuel_Type", "Transmission", "Owner_Type"],
drop_first = True)
In [39]: X_test = pd.get_dummies(X_test,
columns = ["Manufacturer", "Fuel_Type", "Transmission", "Owner_Type"],
drop_first = True)
In [40]: missing_cols = set(X_train.columns) - set(X_test.columns)
for col in missing_cols:
X_test[col] = 0
X_test = X_test[X_train.columns]
In [41]: standardScaler = StandardScaler()
standardScaler.fit(X_train)
X_train = standardScaler.transform(X_train)
X_test = standardScaler.transform(X_test)
```

And then scaling the dataset for model building

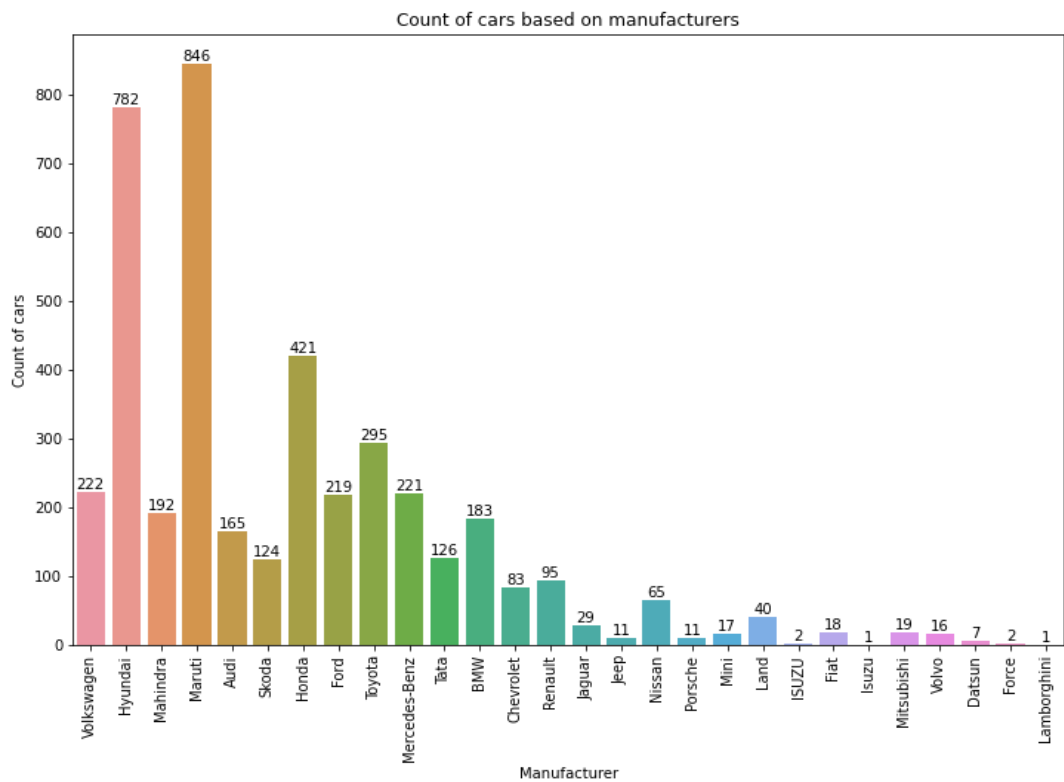


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In [41]: standardScaler = StandardScaler()
standardScaler.fit(X_train)
X_train = standardScaler.transform(X_train)
X_test = standardScaler.transform(X_test)
Training and predicting
In [42]: linearRegression = LinearRegression()
linearRegression.fit(X_train, y_train)
y_pred = linearRegression.predict(X_test)
r2_score(y_test, y_pred)
```

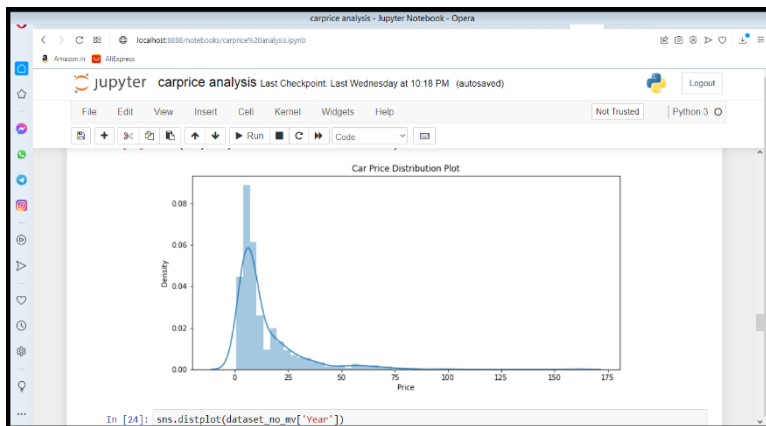
EDA:



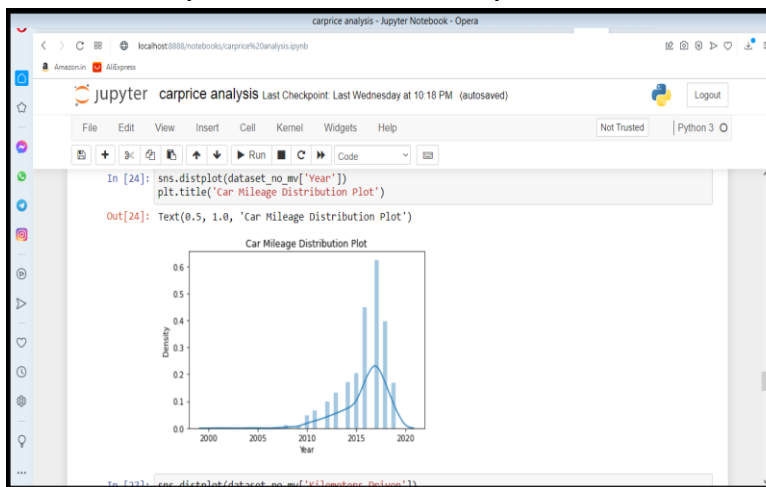
This is the heatmap of the dataset



This is the relation between manufacturer and the count of cars



This is car price distribution plot



Car Mileage distribution plot

- Data Sources and their formats

The data is collected from various site like cardekho,olx and scrapped the data and make a dataset. There are 14 columns and 6019 rows.

Target variable is price of the cars.

	Id	Name	Location	Year	Kilometers_Driven	Fuel_Type	Transmission	Owner_Type	Mileage	Engine	Power	Seats
Out[2]:	0	Maruti Wagon R LXI CNG	Mumbai	2010	72000	CNG	Manual	First	26.6 km/kg	998 CC	58.16 bhp	5.0
	1	Hyundai Creta 1.6 CRDi SX Option	Pune	2015	41000	Diesel	Manual	First	19.67 kmpl	1582 CC	126.2 bhp	5.0
	2	Honda Jazz V	Chennai	2011	46000	Petrol	Manual	First	18.2 kmpl	1199 CC	88.7 bhp	5.0
	3	Maruti Ertiga VDI	Chennai	2012	87000	Diesel	Manual	First	20.77 kmpl	1248 CC	88.76 bhp	7.0
	4	Audi A4 New 2.0 TDI Multitronic	Coimbatore	2013	40670	Diesel	Automatic	Second	15.2 kmpl	1968 CC	140.8 bhp	5.0

- Data Inputs- Logic- Output Relationships

Describe the relationship behind the data input, its format, the logic in between and the output. Describe how the input affects the output.

- State the set of assumptions (if any) related to the problem under consideration

Here, you can describe any presumptions taken by you.

- Hardware and Software Requirements and Tools Used

Here we use lots of liaberries like pandas,numpy,matplot,seaborn, and we use python language for the coding purpose and import some other metrics liaberries also for model building like sklearn metrics ,Linear Regression,Random Forest Regressor ,accuracy report etc.

```

In [26]: import datetime
import numpy as np
import pandas as pd

import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline

from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn.ensemble import RandomForestRegressor
from sklearn.preprocessing import StandardScaler
from sklearn.metrics import r2_score
import warnings
warnings.filterwarnings('ignore')

In [2]: dataset = pd.read_csv("carprice.csv")
dataset.head(5)

Out[2]:
```

	Id	Name	Location	Year	Kilometers_Driven	Fuel_Type	Transmission	Owner_Type	Mileage	Engine	Power	Seats
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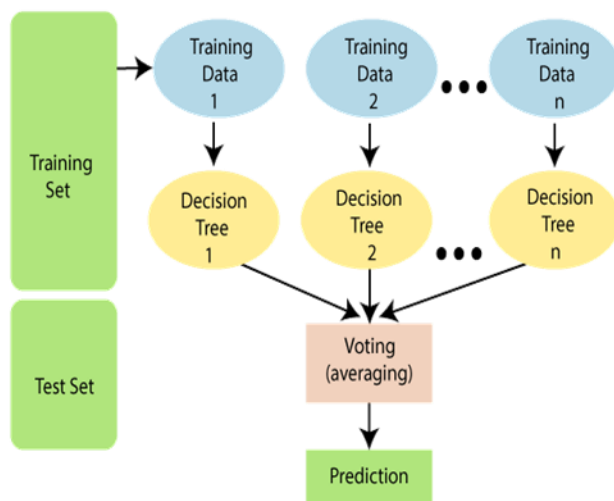
Model/s Development and Evaluation

- Identification of possible problem-solving approaches (methods)

We use randomforest classifier and Linear Regression for the model building as it is regression problem and finally we save the RandomForest Classifier model for gd accuracy.

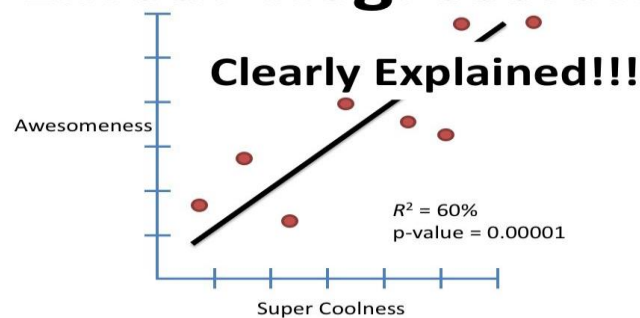
- Testing of Identified Approaches (Algorithms)

Random Forest Classifier:

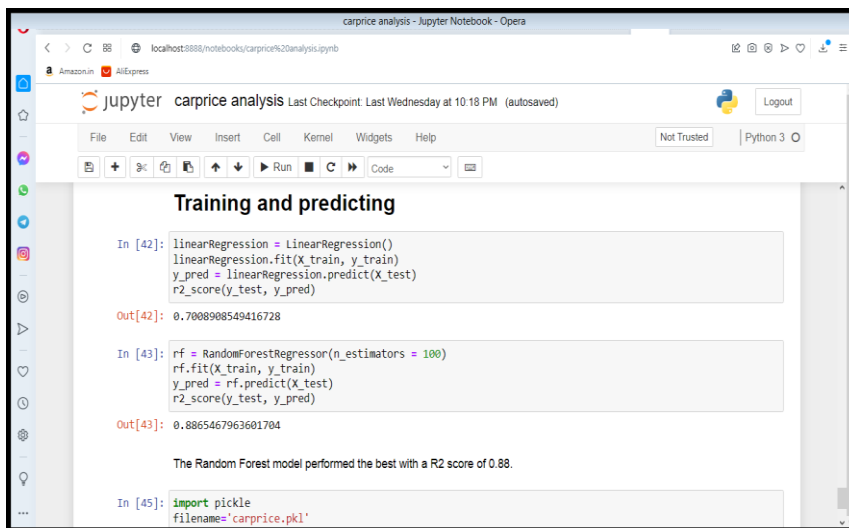


Linear Regression:

Linear Regression



- Run and Evaluate selected models



The screenshot shows a Jupyter Notebook titled 'carprice analysis' with the following code and output:

```
In [42]: linearRegression = LinearRegression()
linearRegression.fit(X_train, y_train)
y_pred = linearRegression.predict(X_test)
r2_score(y_test, y_pred)

Out[42]: 0.7008908549416728
```

```
In [43]: rf = RandomForestRegressor(n_estimators = 100)
rf.fit(X_train, y_train)
y_pred = rf.predict(X_test)
r2_score(y_test, y_pred)

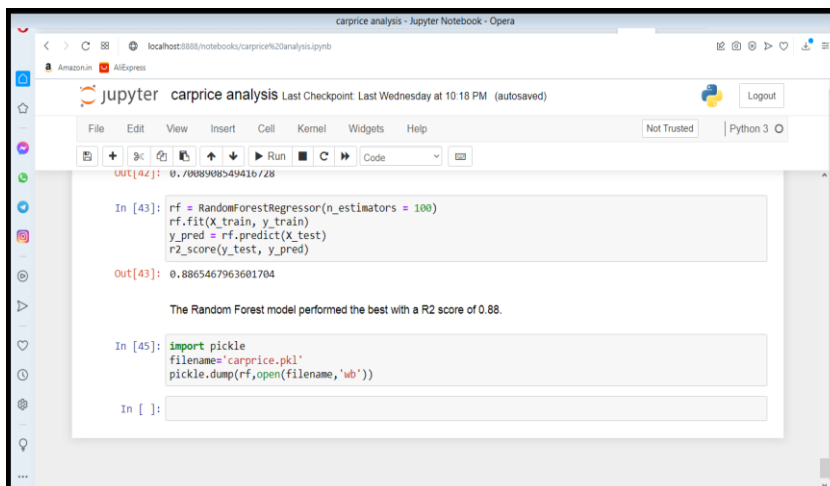
Out[43]: 0.8865467963601704
```

The Random Forest model performed the best with a R2 score of 0.88.

```
In [45]: import pickle
filename='carprice.pkl'
```

We save the best model and this is RandomForest Classifier model.

- Interpretation of the Results



The screenshot shows the continuation of the Jupyter Notebook with the following code and output:

```
Out[42]: 0.7008908549416728
```

```
In [43]: rf = RandomForestRegressor(n_estimators = 100)
rf.fit(X_train, y_train)
y_pred = rf.predict(X_test)
r2_score(y_test, y_pred)

Out[43]: 0.8865467963601704
```

The Random Forest model performed the best with a R2 score of 0.88.

```
In [45]: import pickle
filename='carprice.pkl'
pickle.dump(rf,open(filename,'wb'))

In [ ]:
```

And we save Random Forest Classifier model and accuracy is so good.

CONCLUSION

- Key Findings and Conclusions of the Study

After the Final Submission of test data, my accuracy score was 88%

Feature engineering helped me increase my accuracy.

Amazingly Random Forest Classifier worked better than all other Ensemble models.

- Learning Outcomes of the Study in respect of Data Science

Data science comes as a very important tool to solve problems in the domain to help the companies increase their overall revenue, profits, improving their marketing strategies and focusing on changing trends in car sales and purchases. Predictive modelling, Market mix modelling, recommendation systems are some of the machine learning techniques used for achieving the business goals for car companies.