CAR PRICE PREDICTION WITH MACHINE LEARNING: (Task-3) In [1]: #importing basic libraries import numpy as np import pandas as pd import matplotlib.pyplot as plt import seaborn as sns from sklearn import linear\_model data\_set=pd.read\_csv(r"C:\Users\HP\OneDrive\Documents\oasis infobytes\car data.csv") data\_set.head(50) Out[2]: Car\_Name Year Selling\_Price Present\_Price Driven\_kms Fuel\_Type Selling\_type Transmission Owner 0 0 ritz 2014 3.35 27000 5.59 Petrol Dealer Manual 1 sx4 2013 4.75 9.54 43000 Diesel Dealer Manual 0 2 ciaz 2017 9.85 Petrol 0 7.25 6900 Dealer Manual wagon r 2011 2.85 4.15 5200 Petrol Dealer Manual 6.87 42450 0 swift 2014 4.60 Diesel Dealer Manual 5 vitara brezza 2018 9.83 2071 Diesel Dealer Manual 18796 Petrol 0 ciaz 2015 6.75 8.12 Dealer Manual s cross 2015 6.50 8.61 33429 Diesel Dealer Manual 8 8.75 8.89 20273 0 ciaz 2016 Diesel Dealer Manual ciaz 2015 7.45 8.92 42367 Diesel Dealer Manual 0 10 alto 800 2017 2.85 3.60 2135 0 Petrol Dealer Manual ciaz 2015 11 6.85 10.38 51000 Diesel Dealer Manual Petrol 12 15000 0 ciaz 2015 7.50 9.94 Dealer Automatic 13 ertiga 2015 6.10 7.71 26000 Petrol Dealer Manual 0 2.25 7.21 77427 0 dzire 2009 Petrol Dealer 14 Manual ertiga 2016 15 7.75 10.79 43000 Diesel Dealer Manual 0 7.25 10.79 41678 16 ertiga 2015 Diesel Dealer Manual 17 ertiga 2016 7.75 10.79 43000 Diesel Dealer Manual 0 18 wagon r 2015 3.25 5.09 35500 CNG Dealer 0 Manual 19 sx4 2010 2.65 7.98 41442 Petrol Dealer Manual 25000 Petrol 0 20 alto k10 2016 2.85 3.95 Dealer Manual 21 ignis 2017 4.90 2400 Petrol Dealer Manual 22 4.40 8.01 50000 Petrol 0 sx4 2011 Dealer Automatic 23 alto k10 2014 2.50 3.46 45280 Petrol Dealer Manual 0 24 2.90 4.41 56879 wagon r 2013 Petrol Dealer Manual 25 swift 2011 3.00 4.99 20000 Petrol Dealer Manual 0 26 5.87 Petrol 0 swift 2013 4.15 55138 Dealer Manual 27 swift 2017 6.00 6.49 16200 Petrol Individual Manual 28 alto k10 2010 1.95 3.95 44542 Petrol 0 Dealer Manual 29 ciaz 2015 7.45 10.38 45000 Diesel Dealer Manual 0 0 30 ritz 2012 3.10 5.98 51439 Dealer Diesel Manual ritz 2011 31 2.35 4.89 54200 Petrol Dealer Manual 39000 32 0 swift 2014 4.95 7.49 Diesel Dealer Manual 34 45000 0 dzire 2014 5.50 8.06 Dealer Diesel Manual 35 sx4 2011 2.95 7.74 49998 CNG Dealer Manual Petrol 36 4.65 7.20 48767 Dealer 0 dzire 2015 Manual 800 2003 37 0.35 2.28 127000 Petrol Individual Manual 0 38 alto k10 2016 3.00 3.76 10079 Petrol Dealer Manual 0 39 sx4 2003 2.25 7.98 62000 Petrol Dealer Manual 0 40 5.85 7.87 24524 0 baleno 2016 Petrol Dealer Automatic alto k10 2014 41 2.55 3.98 46706 Petrol Dealer Manual 0 Petrol 42 sx4 2008 1.95 7.15 58000 Dealer 0 Manual 43 dzire 2014 5.50 8.06 45780 Diesel Dealer Manual 0 44 omni 2012 1.25 2.69 50000 Petrol Dealer 0 Manual 45 ciaz 2014 7.50 12.04 15000 Petrol Dealer Automatic Petrol 46 ritz 2013 2.65 4.89 64532 Dealer 0 Manual 47 wagon r 2006 1.05 4.15 65000 Petrol Dealer Manual 48 ertiga 2015 5.80 7.71 25870 Petrol Dealer Manual 0 ciaz 2017 7.75 9.29 37000 Petrol Dealer Automatic In [3]: data\_set.shape (301, 9) Out[3]: data\_set.isnull().sum() #checking the null value Car\_Name Out[4]: Year 0 Selling\_Price 0 Present\_Price 0 Driven\_kms 0 Fuel\_Type 0 Selling\_type 0 Transmission 0 0wner 0 dtype: int64 In [5]: data\_set.info() <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 301 non-null int64 1 Year Selling\_Price 301 non-null 2 float64 3 Present\_Price 301 non-null float64 4 Driven\_kms 301 non-null int64 Fuel\_Type 5 301 non-null object 6 Selling\_type 301 non-null object 7 Transmission 301 non-null object 301 non-null int64 dtypes: float64(2), int64(3), object(4) memory usage: 21.3+ KB data\_set.describe() Year Selling\_Price Present\_Price Driven\_kms Owner Out[6]: 301.000000 301.000000 **count** 301.000000 301.000000 301.000000 36947.205980 mean 2013.627907 4.661296 7.628472 0.043189 2.891554 5.082812 8.642584 38886.883882 0.247915 min 2003.000000 0.100000 0.320000 500.000000 0.000000 0.000000 **25**% 2012.000000 0.900000 1.200000 15000.000000 **50%** 2014.000000 3.600000 6.400000 0.000000 32000.000000 **75%** 2016.000000 6.000000 9.900000 48767.000000 0.000000 max 2018.000000 35.000000 92.600000 500000.000000 3.000000 In [7]: data\_set.columns Index(['Car\_Name', 'Year', 'Selling\_Price', 'Present\_Price', 'Driven\_kms', 'Fuel\_Type', 'Selling\_type', 'Transmission', 'Owner'], dtype='object') **Data Modifications** In [8]: inputs=data\_set.drop(['Car\_Name', 'Owner', 'Selling\_type'], axis='columns') inputs Year Selling\_Price Present\_Price Driven\_kms Fuel\_Type Transmission Out[8]: **0** 2014 3.35 5.59 27000 Petrol Manual **1** 2013 4.75 9.54 43000 Manual Diesel **2** 2017 7.25 9.85 6900 Petrol Manual **3** 2011 2.85 4.15 5200 Petrol Manual **4** 2014 4.60 6.87 42450 Diesel Manual **296** 2016 9.50 11.60 33988 Diesel Manual **297** 2015 4.00 5.90 60000 Petrol Manual **298** 2009 3.35 11.00 87934 Petrol Manual **299** 2017 11.50 12.50 9000 Manual Diesel **300** 2016 5.30 5.90 5464 Petrol Manual 301 rows  $\times$  6 columns target=data\_set.Selling\_Price In [9]: 3.35 Out[9]: 4.75 7.25 3 2.85 4 4.60 296 9.50 297 4.00 298 3.35 299 11.50 300 5.30 Name: Selling\_Price, Length: 301, dtype: float64 Data visualization plt.figure(figsize=(5,4)) sns.countplot(x='Fuel\_Type',data=data\_set) <Axes: xlabel='Fuel\_Type', ylabel='count'> Out[10]: 250 200 150 100 50 Diesel CNG Petrol Fuel\_Type In [11]: plt.figure(figsize=(5,4)) sns.countplot(x='Transmission', data=data\_set) <Axes: xlabel='Transmission', ylabel='count'> 250 200 th 150 100 50 Automatic Manual Transmission In [12]: plt.figure(figsize=(5,4)) sns.countplot(x='Selling\_type', data=data\_set) <Axes: xlabel='Selling\_type', ylabel='count'> Out[12]: 200 175 150 125 100 75 50 25 Dealer Individual Selling\_type In [13]: plt.figure(figsize=(5,5)) sns.kdeplot(data\_set['Selling\_Price']) <Axes: xlabel='Selling\_Price', ylabel='Density'> Out[13]: 0.10 0.08 Density 90.0 0.04 0.02 0.00 10 20 30 Selling\_Price In [14]: plt.figure(figsize=(5,5)) sns.kdeplot(data\_set['Present\_Price']) <Axes: xlabel='Present\_Price', ylabel='Density'> Out[14]: 0.06 0.05 0.04 Density oo w 0.02 0.01 20 40 60 80 100 Present\_Price In [15]: z=data\_set.drop(['Car\_Name', 'Year', 'Driven\_kms', 'Fuel\_Type', 'Selling\_type', 'Transmission', 'Owner'], axis=1) Out[15]: Selling\_Price Present\_Price 3.35 5.59 1 4.75 9.54 2 7.25 9.85 4.15 4 4.60 6.87 296 9.50 11.60 297 5.90 298 3.35 11.00 299 11.50 12.50 300 5.30 5.90 301 rows × 2 columns In [16]: sns.kdeplot(z) <Axes: ylabel='Density'> Selling\_Price Present\_Price 0.05 0.04 Density o o o 0.02 0.01 0.00 20 60 80 100 In [17]: sns.heatmap(data\_set.corr(), cmap='Blues') C:\Users\HP\AppData\Local\Temp\ipykernel\_20408\2110432753.py:1: FutureWarning: The default value of numeric\_only in DataFrame.corr is deprecated. In a future version, it will defau It to False. Select only valid columns or specify the value of numeric\_only to silence this warning. sns.heatmap(data\_set.corr(),cmap='Blues') <Axes: > Out[17]: Year 0.6 Selling\_Price -- 0.4 Present\_Price -- 0.2 - 0.0 Driven\_kms -- -0.2 Owner -- -0.4 Year Selling\_Price Present\_Price Driven\_kms In [ ]: Training the model In [18]: #Encoding from sklearn.preprocessing import LabelEncoder Numerics=LabelEncoder() In [22]: inputs['Fuel\_Type']=Numerics.fit\_transform(inputs['Fuel\_Type']) #encoded fuel type CNG-0 , Diesel-1, Petrol-2 inputs['Transmission']=Numerics.fit\_transform(inputs['Transmission']) #encoded transmission type Automatic-0 , Manual-1 inputs Year Selling\_Price Present\_Price Driven\_kms Fuel\_Type Transmission Out[22]: **0** 2014 3.35 5.59 27000 1 **1** 2013 4.75 9.54 43000 **2** 2017 7.25 9.85 6900 2 1 **3** 2011 2.85 4.15 5200 42450 **4** 2014 4.60 6.87 1 1 **296** 2016 9.50 11.60 33988 1 1 **297** 2015 4.00 5.90 60000 2 1 **298** 2009 3.35 11.00 87934 **299** 2017 11.50 12.50 9000 **300** 2016 2 1 5.30 5.90 5464 301 rows × 6 columns model=linear\_model.LinearRegression() inputs=inputs.values In [24]: model.fit(inputs, target) ▼ LinearRegression LinearRegression() In [ ] In [ ]: Final result of prediction

prediction=model.predict( [[2020,1500,100,1000,2,0]]) # (Year, Selling\_Price, Present\_Price, Driven\_kms, Fuel\_Type, Transission)

#Thanking You...

#encoded fuel type CNG-0 , Diesel-1, Petrol-2
#encoded transmission type Automatic-0 ,Manual-1

- Thiruvalluvan G

print("Car price predicted value:", prediction)

Car price predicted value: [1500.]