

CAR PRICE PREDICTION

การทำนายราคารถยนต์ในอนาคต

01076032

ELEMENTARY DIFFERENTIAL EQUATIONS
AND LINEAR ALGEBRA

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ขั้นตอนการทำงาน



1

นำข้อมูลเข้าสู่ระบบ

DATA

```
data = pd.read_csv(url)
data.head()
```

	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

```
[ ] data.count()
```

```
Car_Name      301
Year          301
Selling_Price 301
Present_Price 301
Kms_Driven    301
Fuel_Type     301
Seller_Type   301
Transmission  301
Owner         301
dtype: int64
```

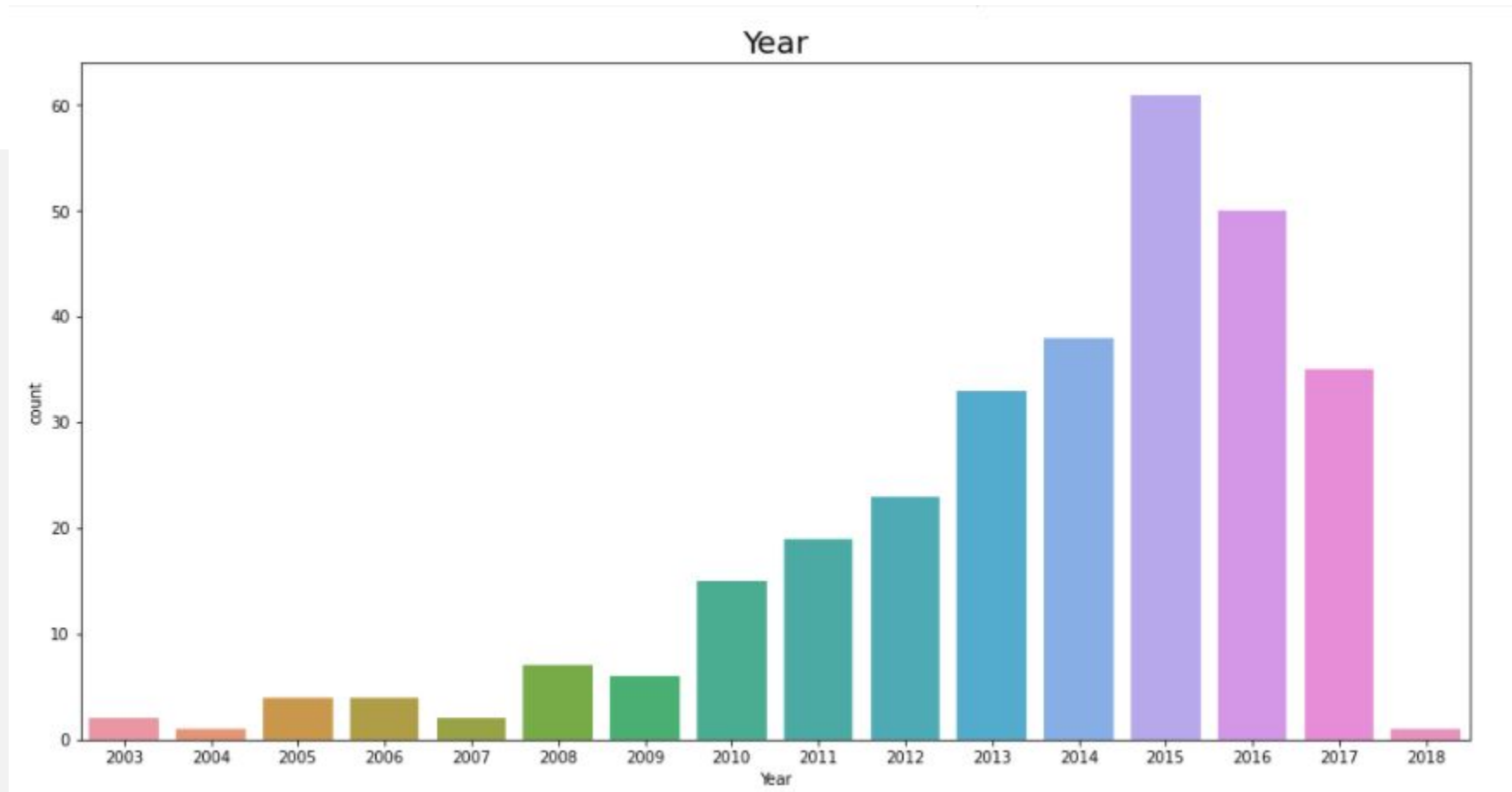
NAME

```
data["Car_Name"].value_counts()
```

city	26
corolla altis	16
verna	14
fortuner	11
brio	10
..	..
Bajaj Pulsar RS200	1
TVS Jupyter	1
omni	1
Bajaj Dominar 400	1
Bajaj Avenger 150 street	1

Name: Car_Name, Length: 98, dtype: int64

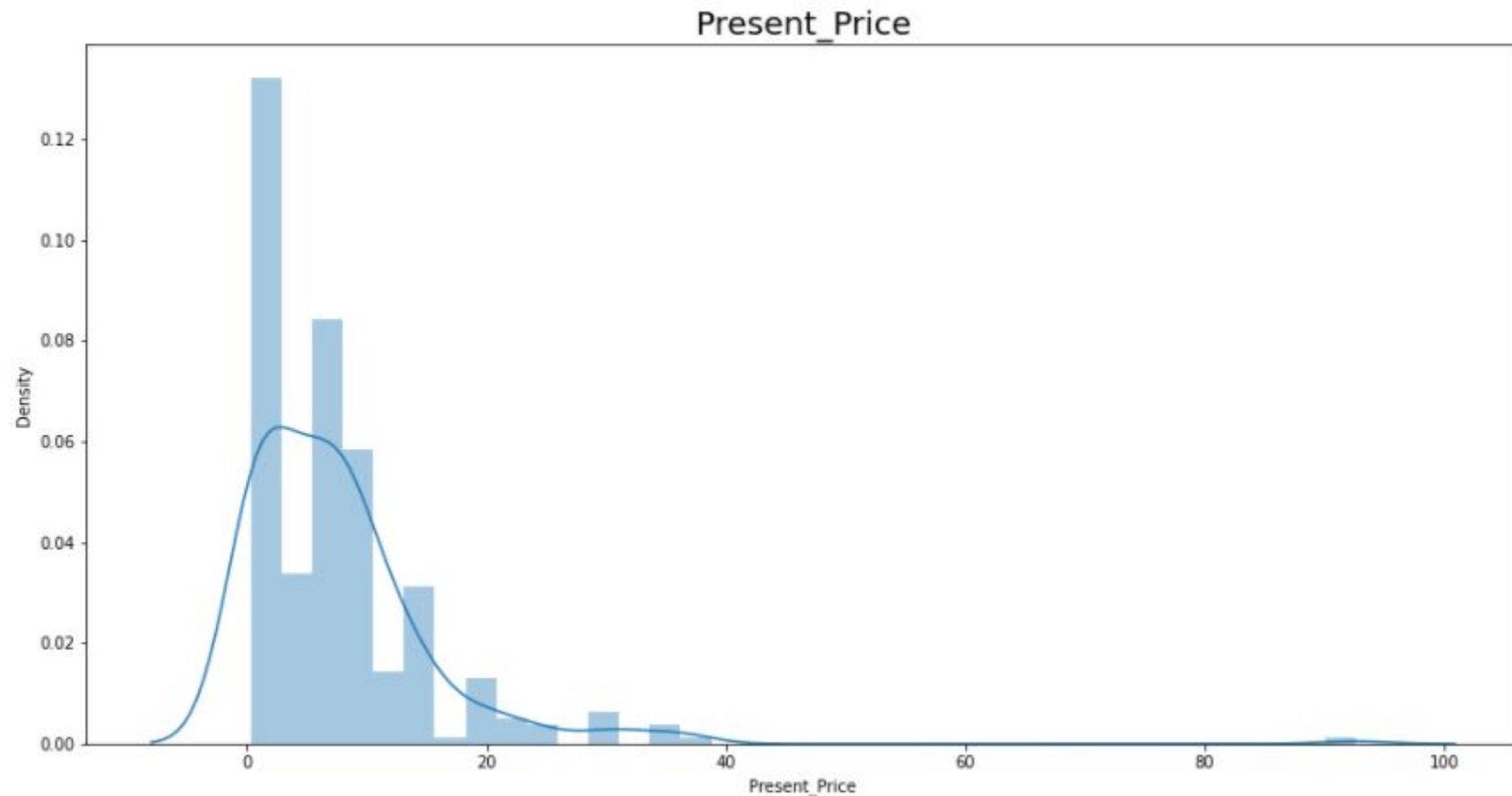
YEAR



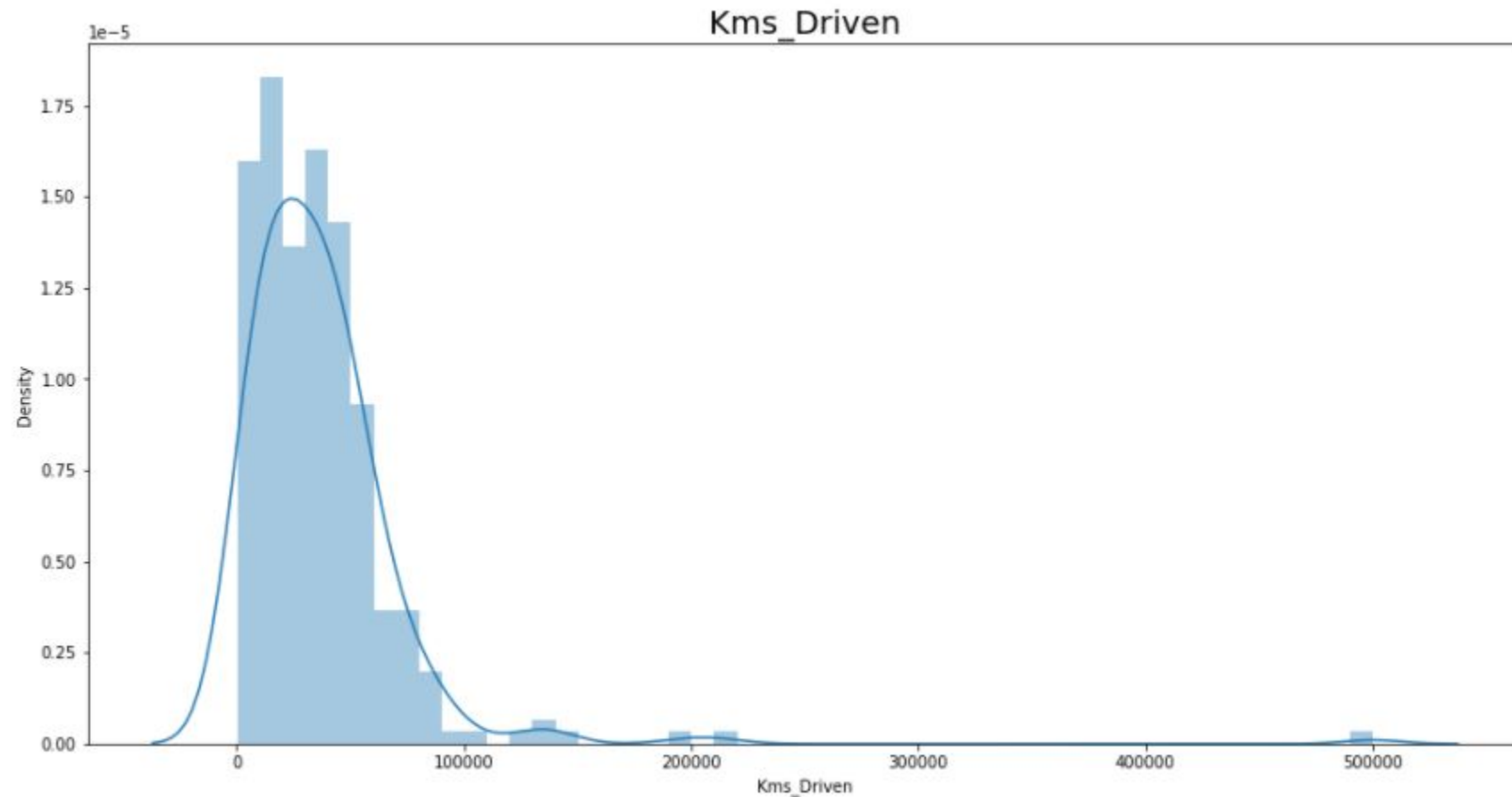
SELLING PRICE



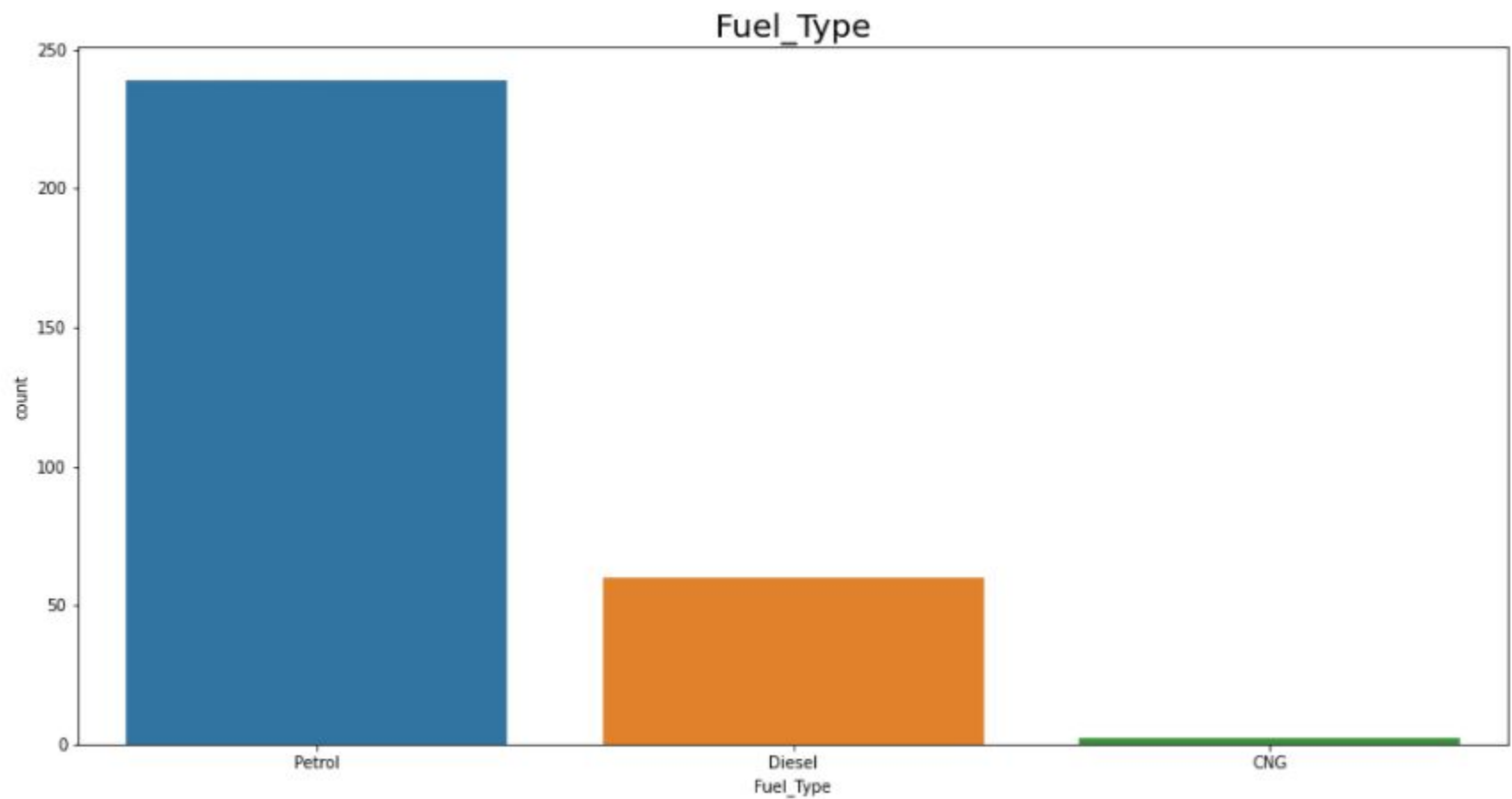
PRESENT PRICE



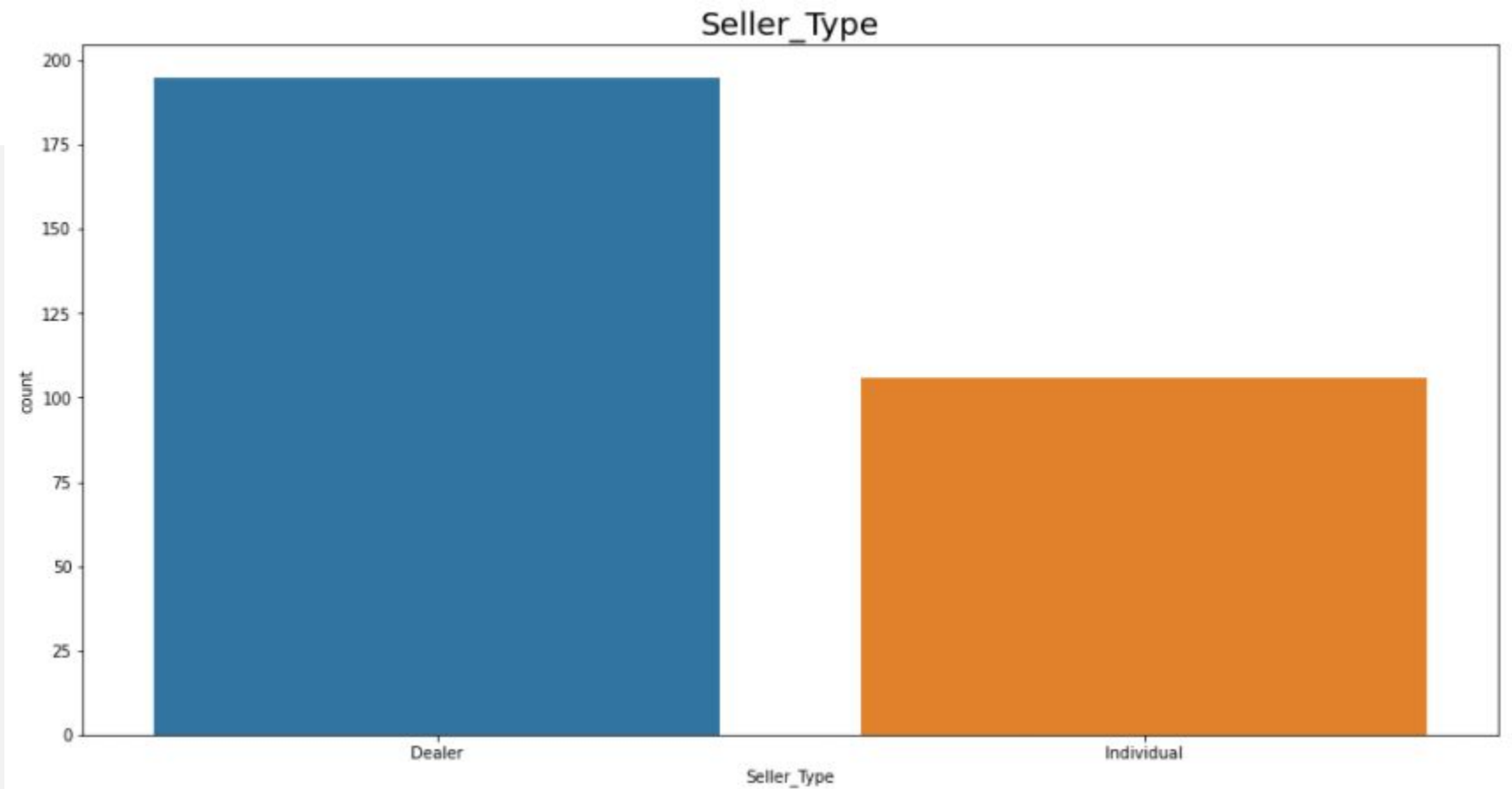
KMS DRIVEN



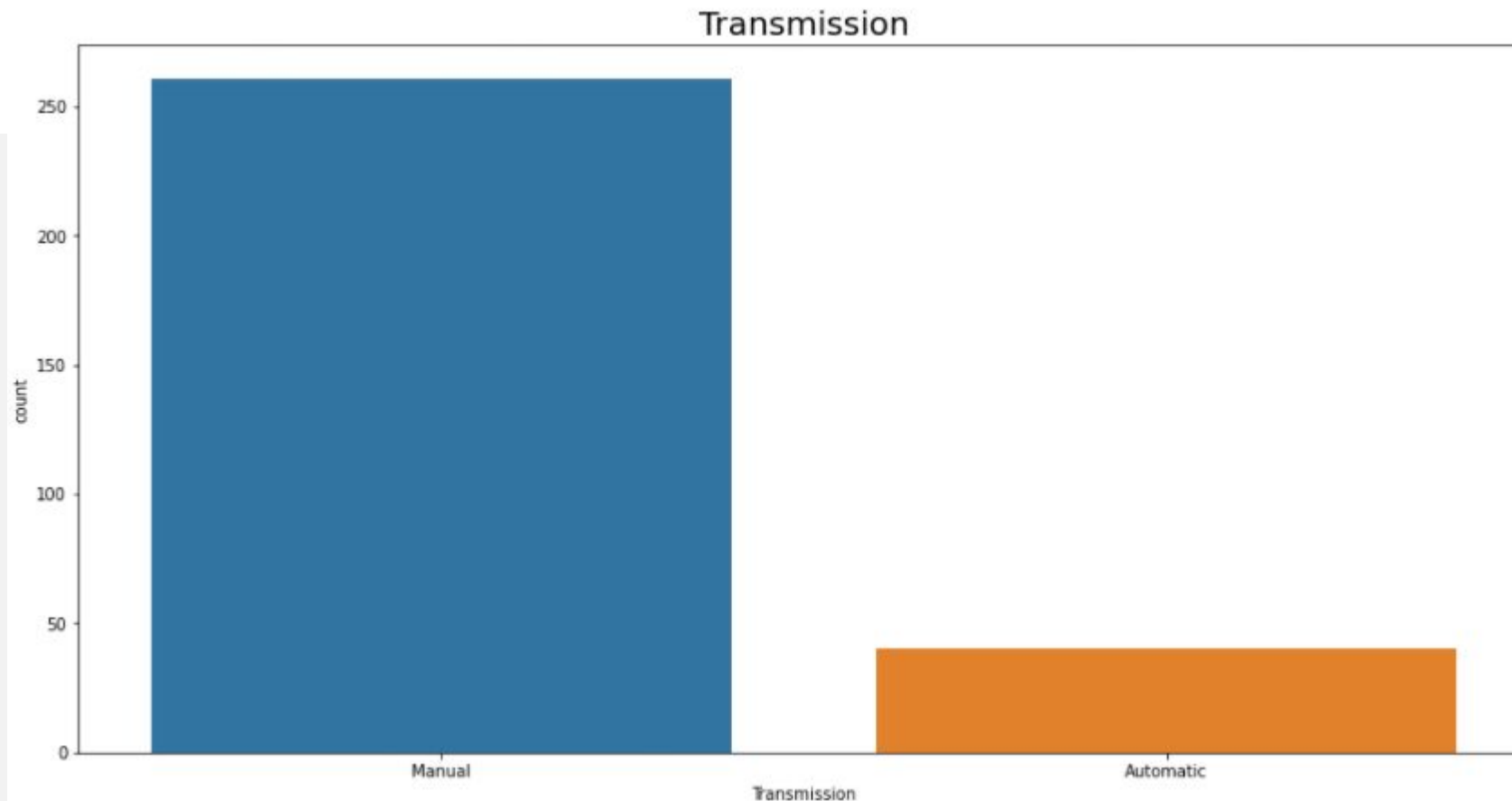
FUEL TYPE



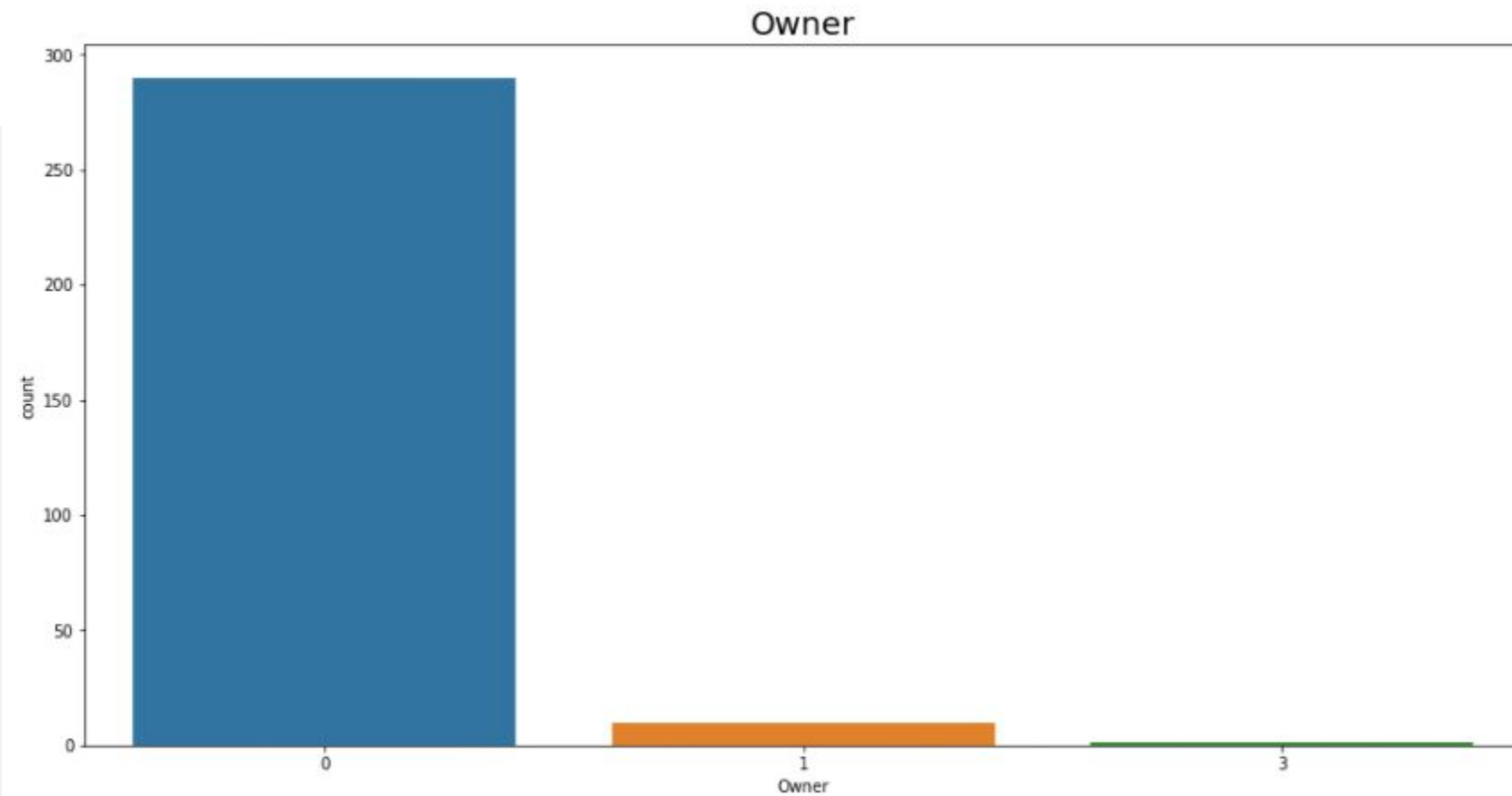
SELLER TYPE



TRANSMISSION



OWNER





2

ปรับแต่งข้อมูล

NAME

```
data.drop('Car_Name', axis=1, inplace=True)  
data.head()
```

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

YEAR

```
[ ] data['Year'].replace({2010 : 1, 2011 : 2, 2012 : 3, 2013 : 4, 2014 : 5, 2015 : 6, 2016 : 7, 2017 : 8, 2018 : 9}, inplace=True)
data.loc[(data['Year'] >= 2003) & (data['Year'] <= 2009), 'Year'] = 0
data['Year'].value_counts()
```

```
6    61
7    50
5    38
8    35
4    33
0    26
3    23
2    19
1    15
9     1
Name: Year, dtype: int64
```

KMS DRIVEN

```
[ ] data["Kms_Driven"].describe()
```

```
count      301.000000
mean      36947.205980
std       38886.883882
min         500.000000
25%      15000.000000
50%      32000.000000
75%      48767.000000
max      500000.000000
Name: Kms_Driven, dtype: float64
```

```
[ ] data.loc[(data['Kms_Driven'] >= 500) & (data['Kms_Driven'] <= 15000), 'Kms_Driven'] = 0
data.loc[(data['Kms_Driven'] > 15000) & (data['Kms_Driven'] <= 32000), 'Kms_Driven'] = 1
data.loc[(data['Kms_Driven'] > 32000) & (data['Kms_Driven'] <= 49000), 'Kms_Driven'] = 2
data.loc[(data['Kms_Driven'] > 49000), 'Kms_Driven'] = 3
data["Kms_Driven"].value_counts()
```

```
0      77
2      76
1      75
3      73
Name: Kms_Driven, dtype: int64
```

FUEL TYPE

```
[ ] data["Fuel_Type"].value_counts()
```

```
Petrol    239  
Diesel    60  
CNG        2  
Name: Fuel_Type, dtype: int64
```

```
[ ] numFuel = {"Petrol":0,"Diesel":1,"CNG":2}  
data["Fuel_Type"].replace(numFuel, inplace = True)  
data["Fuel_Type"].value_counts()
```

```
0    239  
1     60  
2      2  
Name: Fuel_Type, dtype: int64
```

SELLER TYPE

```
[ ] data["Seller_Type"].value_counts()
```

```
Dealer      195  
Individual   106  
Name: Seller_Type, dtype: int64
```

```
[ ] numSeller = {"Individual":0,"Dealer":1}  
data["Seller_Type"].replace(numSeller, inplace = True)  
data["Seller_Type"].value_counts()
```

```
1    195  
0    106  
Name: Seller_Type, dtype: int64
```

TRANSMISSION

```
[ ] data["Transmission"].value_counts()
```

```
Manual      261  
Automatic   40  
Name: Transmission, dtype: int64
```

```
[ ] numTrans = {"Manual":0,"Automatic":1}  
data["Transmission"].replace(numTrans, inplace = True)  
data["Transmission"].value_counts()
```

```
0      261  
1       40  
Name: Transmission, dtype: int64
```

DATA

```
[ ] data.head()
```

	Year	Selling_Price	Present_Price	Kms_Driven	Fuel_Type	Seller_Type	Transmission	Owner
0	5	3.35	5.59	1	0	1	0	0
1	4	4.75	9.54	2	1	1	0	0
2	8	7.25	9.85	0	0	1	0	0
3	2	2.85	4.15	0	0	1	0	0
4	5	4.60	6.87	2	1	1	0	0



3

วิเคราะห์ข้อมูล

Correlations

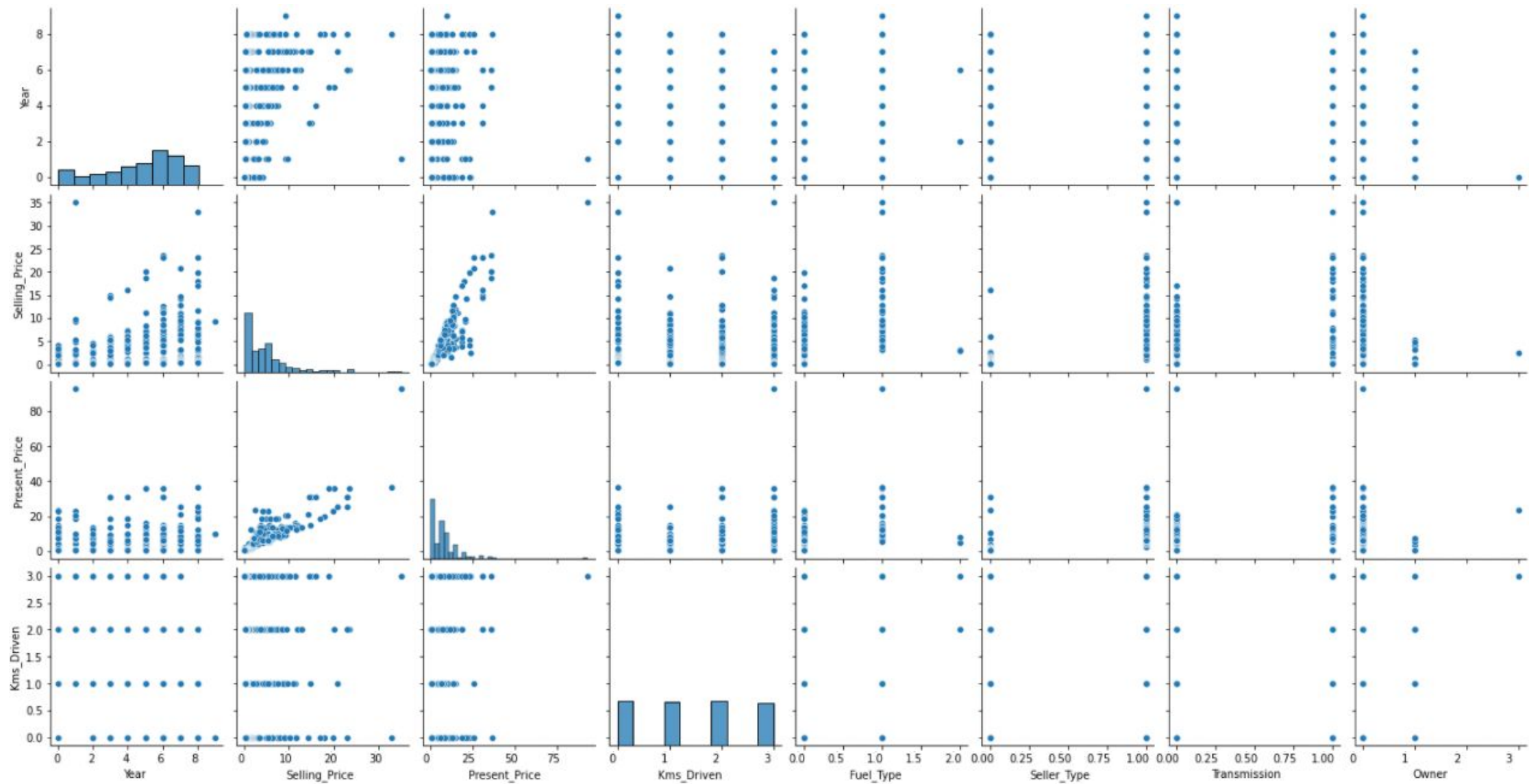
Linear regression

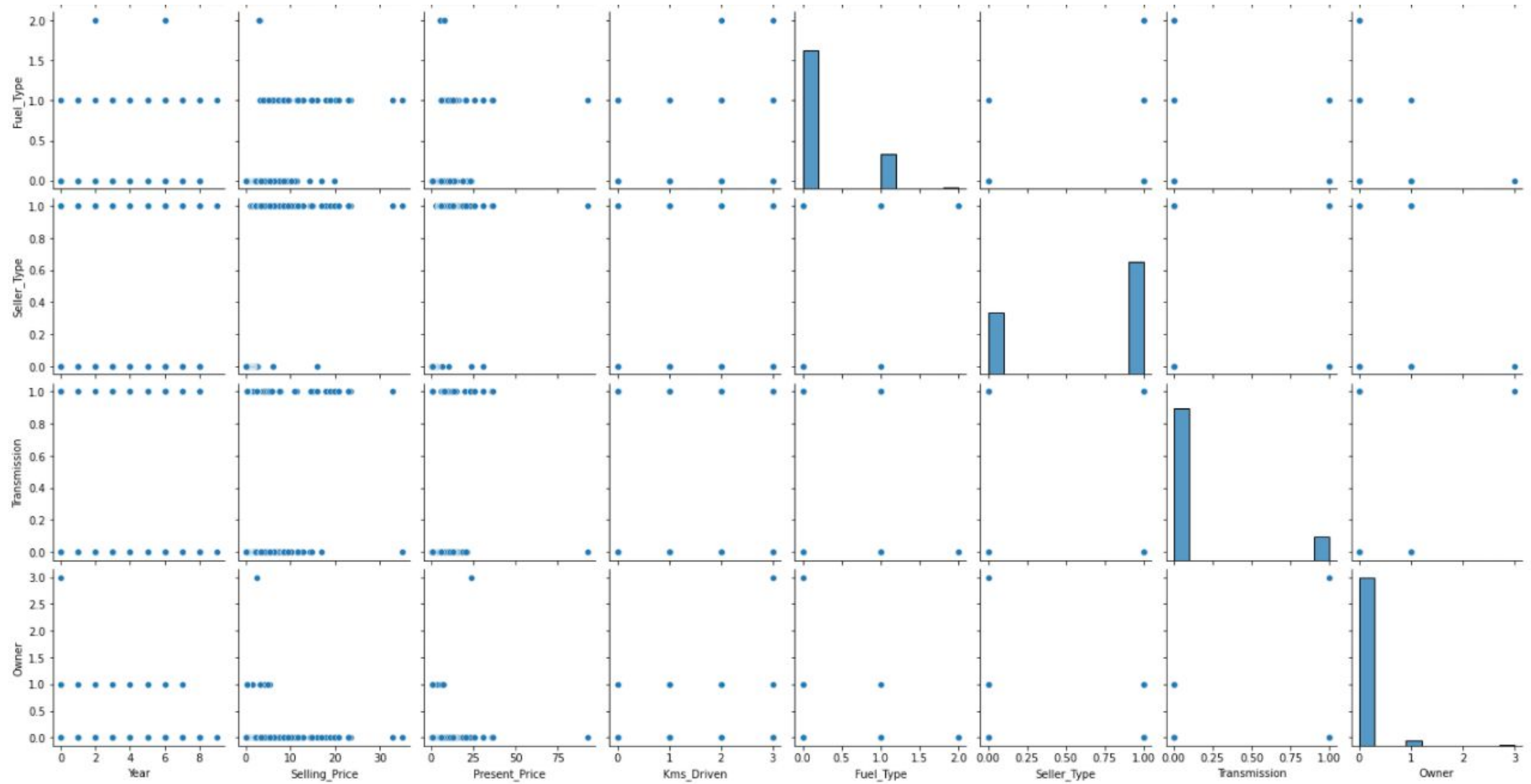
4

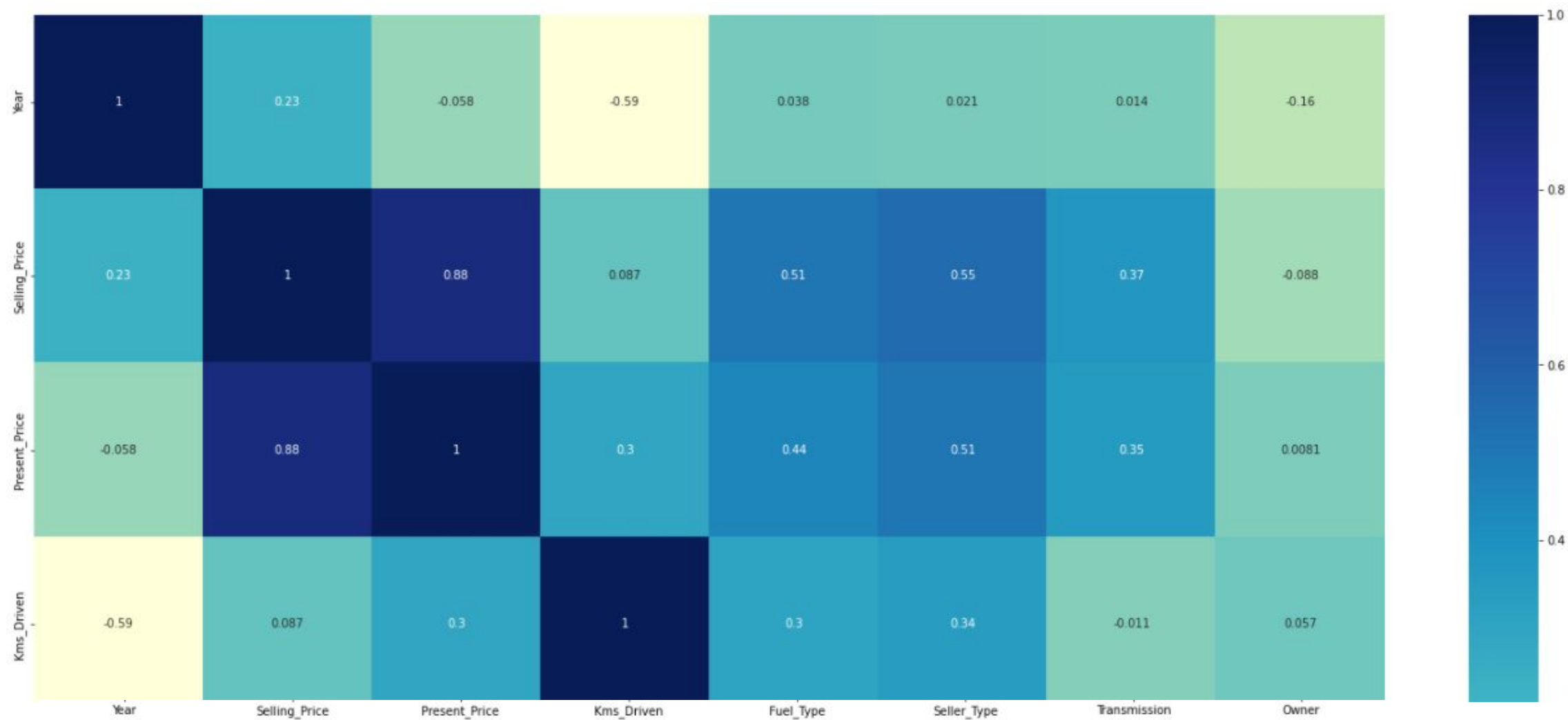
M O D E L

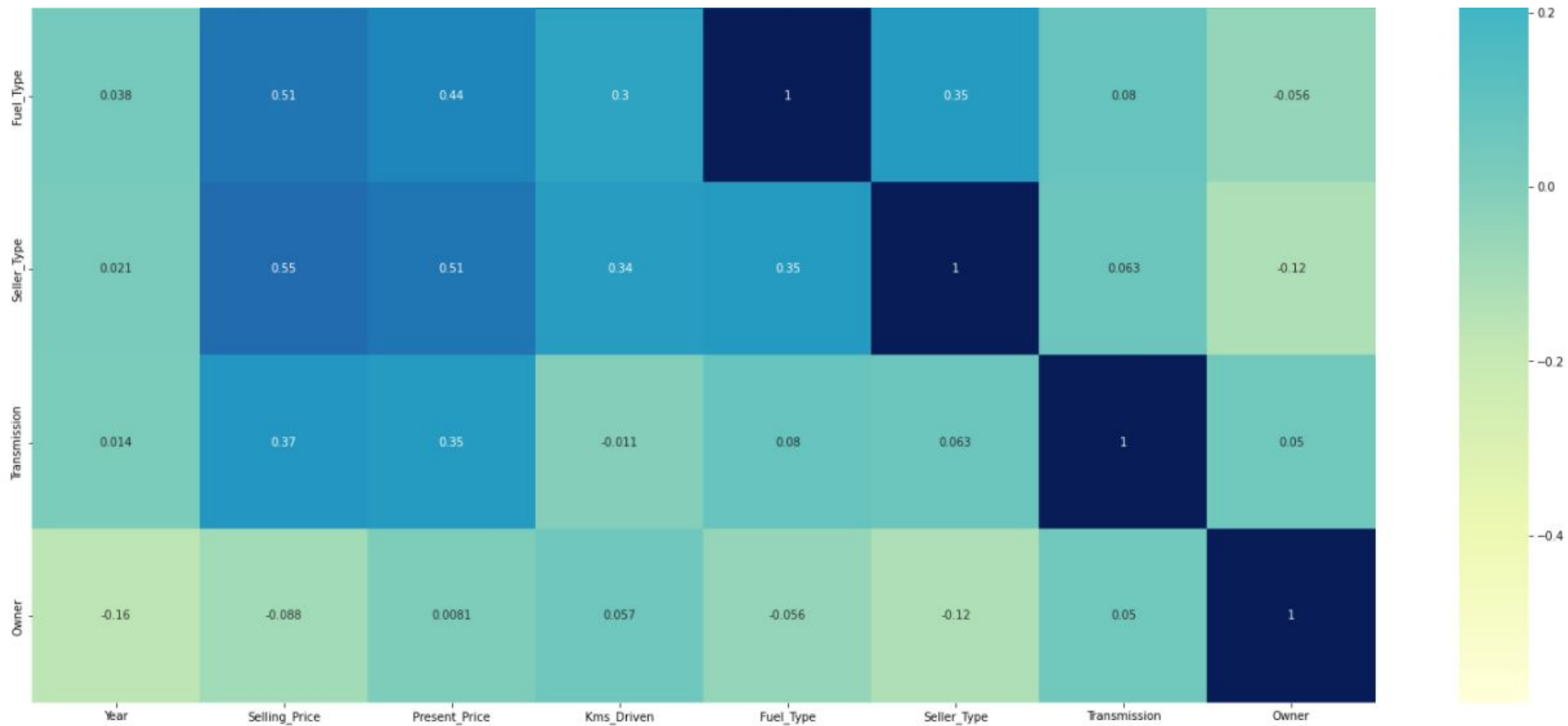
```
*** Car Selling Price Prediction ***
Input :
Year : 2010
Present Price (THB) : 459000
Kms Driven (km) : 0
Fuel type (Select one: Petrol = 0 / Diesel = 1 / CNG = 2): 1
Seller Type (Select one: Individual = 0 / Dealer = 1): 1
Car Transmission (Select one: Manual = 0 / Automatic = 1): 0
Number of Car Owner (Select one: First hand = 0 / Second hand = 1 / Third hand = 2 / Fourth hand or more = 3): 0
Output :
Predict selling price : 281664.60 THB
```

การประมวลผล









```
[ ] #Training (Selling_Price)
x1 = data.drop(['Selling_Price'], axis=1)      #Drop both Price makes Linear Regression score lower
y1 = data['Selling_Price']

x1 = RobustScaler().fit_transform(x1)

x1Train, x1Test, y1Train, y1Test = train_test_split(x1, y1, test_size = 0.2, random_state = 42)
x1Train.shape, x1Test.shape, y1Train.shape, y1Test.shape

((240, 7), (61, 7), (240,), (61,))
```

นำชุดข้อมูลที่ผ่านการสุ่มมาทำการ Train โมเดล

```
[ ] #Linear Regression (Selling_Price)
aLR = LinearRegression()
aLR.fit(x1Train, y1Train)
yPredict = aLR.predict(x1Test)
print("Linear regression score : ", aLR.score(x1Test, y1Test))
print("Mean squared error : ", mean_squared_error(y1Test, yPredict))

Linear regression score : 0.8592697783382605
Mean squared error : 3.2418029156481825
```

นำโมเดลที่ผ่านการ Train มาทำนายชุดข้อมูลที่เหลือ


```
[ ] #Linear Regression (Selling_Price)
yPredict = aLR.predict(x1)
print("Linear regression score : ",aLR.score(x1, y1))
print("Mean squared error : ", mean_squared_error(y1, yPredict))
```

```
Linear regression score : 0.8852375928700549
Mean squared error : 2.955033613791859
```

นำชุดข้อมูลทั้งหมดมาทำนายราคารถยนต์

Predict Y(Selling_Price)

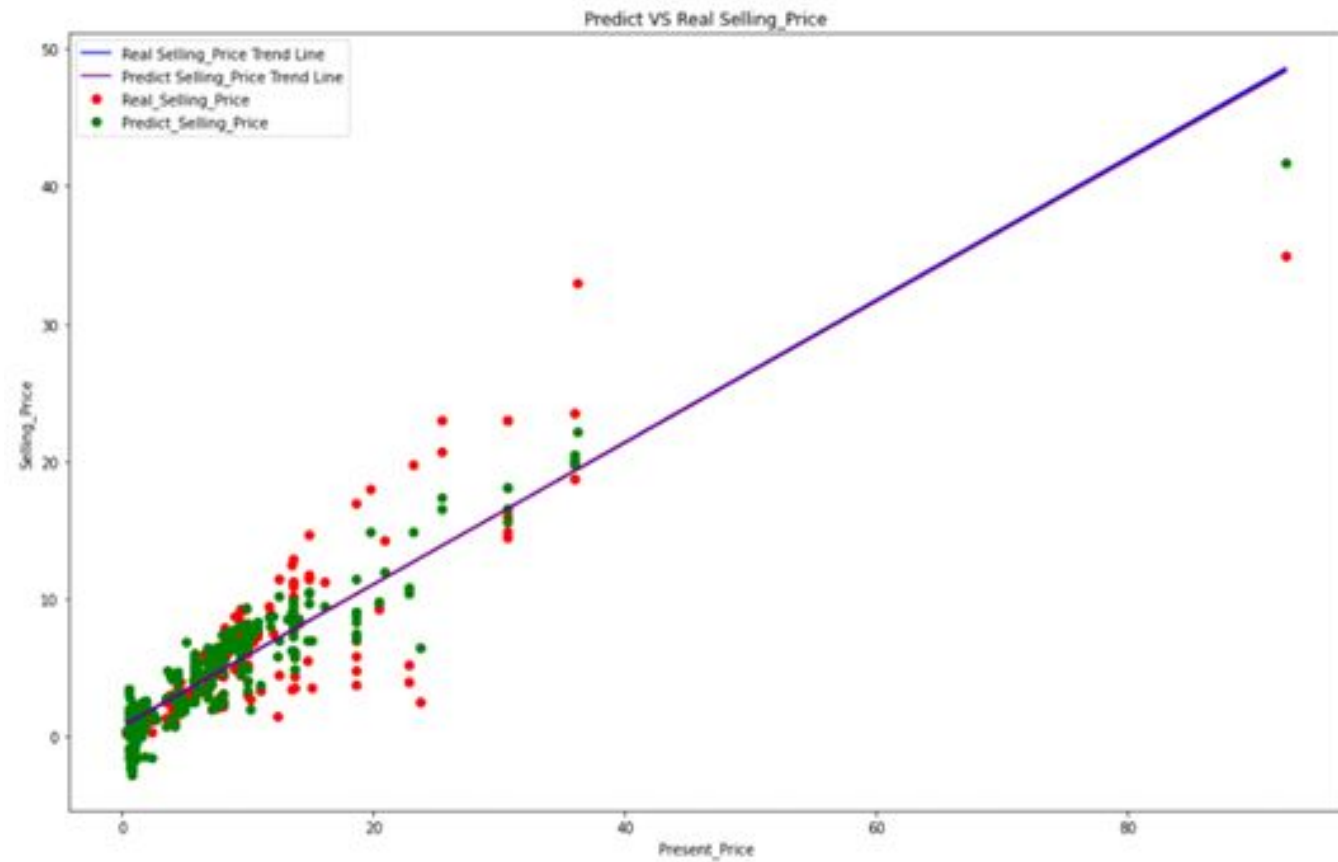
4.08	6.55	7.53	2.57	5.77	9.4	5.61	6.95	7.83	7.08	4.77	7.38
8.22	5.42	2.06	8.32	7.91	8.32	6.86	3.16	4.17	5.7	4.69	2.79
2.45	2.59	3.1	4.31	1.38	7.73	4.21	1.85	6.04	7.13	6.3	6.06
4.85	-1.5	4.43	2.41	7.36	3.02	2.04	6.3	1.28	8.74	2.66	0.71
5.42	8.05	16.58	18.15	14.86	15.6	2.01	6.99	4.88	7.49	3.28	20.11
9.09	5.07	19.76	20.52	22.15	5.8	14.88	9.81	5.77	12.02	5.37	7.03
8.74	2.56	6.33	4.67	7.	5.8	10.85	16.58	10.48	6.61	17.35	9.09
6.31	6.52	41.76	6.18	3.53	4.25	7.11	9.5	4.94	18.15	10.42	8.33
16.59	11.43	9.09	9.81	2.22	2.59	2.58	2.5	2.44	1.67	0.77	0.83
1.7	2.44	2.17	2.05	1.63	1.3	1.28	1.28	0.11	1.82	1.28	1.
1.94	0.01	1.9	0.11	0.53	-1.39	2.21	1.74	2.17	2.16	2.17	0.99
2.21	1.8	2.15	1.3	0.75	0.16	1.77	1.05	0.34	1.39	1.	1.4
0.66	-0.24	0.92	0.3	-1.34	1.45	-0.29	1.28	0.19	-0.1	0.96	3.47
2.02	1.	2.03	3.47	-0.58	0.58	0.19	-0.99	1.62	3.08	1.62	0.23
0.49	1.21	0.12	1.22	0.85	3.47	0.59	-1.35	-0.54	2.74	1.91	-1.64
-1.07	0.55	0.07	0.02	-2.46	0.25	0.28	0.04	0.06	-2.26	-2.18	-1.32
-2.81	-0.88	-2.16	0.42	-0.82	-1.53	-2.08	-2.26	-2.18	2.78	4.53	1.72
3.97	4.31	6.33	5.64	6.76	6.01	2.48	9.68	8.44	3.39	5.35	4.6
4.73	4.73	5.42	4.6	4.43	5.66	5.74	6.95	6.47	2.42	4.53	1.65
5.73	4.57	6.48	6.99	10.37	4.5	4.88	6.99	4.73	8.81	4.66	2.75
6.08	3.84	4.41	6.13	6.48	5.73	3.45	5.82	3.23	5.78	9.56	4.88
6.49	7.11	4.88	2.46	7.74	7.68	8.03	4.35	8.44	5.41	4.23	5.83
4.76	6.68	4.35	8.05	5.39	6.44	4.11	3.3	6.44	2.6	4.61	9.9
8.03	8.03	6.49	7.28	4.97	2.24	8.58	7.99	4.16	7.64	5.91	6.44
7.68	8.79	4.44	4.71	6.49	4.01	4.26	8.58	8.68	3.92	3.74	10.18
5.38]											

ค่าที่ได้จากการทำนาย

Real Y(Selling Price)

3.35	4.75	7.25	2.85	4.6	9.25	6.75	6.5	8.75	7.45	2.85	6.85	7.5
6.1	2.25	7.75	7.25	7.75	3.25	2.65	2.85	4.9	4.4	2.5	2.9	
3.0	4.15	6.0	1.95	7.45	3.1	2.35	4.95	6.0	5.5	2.95	4.65	
0.35	3.0	2.25	5.85	2.55	1.95	5.5	1.25	7.5	2.65	1.05	5.8	
7.75	14.9	23.0	18.0	16.0	2.75	3.6	4.5	4.75	4.1	19.99	6.95	
4.5	18.75	23.5	33.0	4.75	19.75	9.25	4.35	14.25	3.95	4.5	7.45	
2.65	4.9	3.95	5.5	1.5	5.25	14.5	14.73	4.75	23.0	12.5	3.49	
2.5	35.0	5.9	3.45	4.75	3.8	11.25	3.51	23.0	4.0	5.85	20.75	
17.0	7.05	9.65	1.75	1.7	1.65	1.45	1.35	1.35	1.35	1.25	1.2	
1.2	1.2	1.15	1.15	1.15	1.15	1.11	1.1	1.1	1.1	1.05	1.05	
1.05	1.05	1.0	0.95	0.9	0.9	0.75	0.8	0.78	0.75	0.75	0.75	
0.72	0.65	0.65	0.65	0.65	0.6	0.6	0.6	0.6	0.6	0.6	0.6	
0.6	0.55	0.55	0.52	0.51	0.5	0.5	0.5	0.5	0.5	0.48	0.48	
0.48	0.48	0.45	0.45	0.45	0.45	0.45	0.45	0.45	0.45	0.45	0.42	0.42
0.4	0.4	0.4	0.4	0.4	0.38	0.38	0.35	0.35	0.35	0.31	0.3	
0.3	0.3	0.27	0.25	0.25	0.25	0.25	0.25	0.25	0.2	0.2	0.2	
0.2	0.2	0.18	0.17	0.16	0.15	0.12	0.1	3.25	4.4	2.95	2.75	
5.25	5.75	5.15	7.9	4.85	3.1	11.75	11.25	2.9	5.25	4.5	2.9	
3.15	6.45	4.5	3.5	4.5	6.0	8.25	5.11	2.7	5.25	2.55	4.95	
3.1	6.15	9.25	11.45	3.9	5.5	9.1	3.1	11.25	4.8	2.0	5.35	
4.75	4.4	6.25	5.95	5.2	3.75	5.95	4.0	5.25	12.9	5.0	5.4	
7.2	5.25	3.0	10.25	8.5	8.4	3.9	9.15	5.5	4.0	6.6	4.0	
6.5	3.65	8.35	4.8	6.7	4.1	3.0	7.5	2.25	5.3	10.9	8.65	
9.7	6.0	6.25	5.25	2.1	8.25	8.99	3.5	7.4	5.65	5.75	8.4	
10.11	4.5	5.4	6.4	3.25	3.75	8.55	9.5	4.0	3.35	11.5	5.3	
]												

ค่าจริง



กราฟเปรียบเทียบค่าที่ได้จาก Model การทำนาย และ ค่าจริง

```
[ ] #Cosine Similarity
    real = data['Selling_Price'].values.tolist()           #List of real values
    predict = yPredict.tolist()                           #List of predict values
    dot = np.dot(real,predict)                             #Dot product
    magReal = np.linalg.norm(real)                        #magnitude of real values
    magPredict = np.linalg.norm(predict)                  #magnitude of predict values
    cosine = dot/(magReal*magPredict)                     #cosine value
    degree = float("{0:.3f}".format((math.acos(cosine)*180)/math.pi)) #value in degree

    print("Cosine =", "{0:.4f}".format(cosine))
    print("Degree =", degree, "°")

Cosine = 0.9684
Degree = 14.447 °
```

หาค่าของ Cosine Similarity เพื่อวัดประสิทธิภาพของ Model การทำนาย

ผลการทดสอบ

```
[ ] #Predict Output from Input
#Use : Y = aLR.predict(X)
print(" *** Car Selling Price Prediction ***\nInput :")
cYear = int(input("Year : "))
cPP = int(input("Present Price (THB) : "))
cKM = int(input("Kms Driven (km) : "))
cFuel = int(input("Fuel type (Select one: Petrol = 0 / Diesel = 1 / CNG = 2): "))
cSeller = int(input("Seller Type (Select one: Individual = 0 / Dealer = 1): "))
cTrans = int(input("Car Transmission (Select one: Manual = 0 / Automatic = 1): "))
cOwner = int(input("Number of Car Owner (Select one: First hand = 0 / Second hand = 1 / Third hand = 2 / Fourth hand or more = 3): "))

#Convert input
if cYear < 2009 : cYear = 0
elif cYear == 2010 : cYear = 1
elif cYear == 2011 : cYear = 2
elif cYear == 2012 : cYear = 3
elif cYear == 2013 : cYear = 4
elif cYear == 2014 : cYear = 5
elif cYear == 2015 : cYear = 6
elif cYear == 2016 : cYear = 7
elif cYear == 2017 : cYear = 8
elif cYear == 2018 : cYear = 9
elif cYear >= 2019 : cYear = cYear - 2009

if cKM >= 0 and cKM <= 15000 : cKM = 0
elif cKM > 15000 and cKM <= 32000 : cKM = 1
elif cKM > 32000 and cKM <= 49000 : cKM = 2
elif cKM > 49000 : cKM = 3

cPP = (cPP/0.44)/pow(10,5)
cPP = "{0:.2f}".format(cPP)
#Check var
#print(cYear)
#print(cPP)
#print(cKM)

#Calulte
inpData = {'Year':[cYear],'Present_Price':[cPP],'Kms_Driven':[cKM],'Fuel_Type':[cFuel],'Seller_Type':[cSeller],'Transmission':[cTrans],'Owner':[cOwner]}
newDF = pd.DataFrame(inpData)
newX = data.drop('Selling_Price',axis=1)
newX = newX.append(newDF,ignore_index=True)
inpX = RobustScaler().fit_transform(newX)
outputPredict = aLR.predict(inpX)
opPred = outputPredict[len(outputPredict)-1]
opPred = str(opPred*pow(10,5)*0.44)
opPred = "{0:.2f}".format(float(opPred))
#ans = opPred
```

การรับข้อมูลจากผู้ใช้งานเข้ามาสู่ Model การทำนายและแสดงค่าการทำนาย

*** Car Selling Price Prediction ***

Input :

Year : 2017

Present Price (THB) : 500000

Kms Driven (km) : 100

Fuel type (Select one: Petrol = 0 / Diesel = 1 / CNG = 2): 0

Seller Type (Select one: Individual = 0 / Dealer = 1): 0

Car Transmission (Select one: Manual = 0 / Automatic = 1): 1

Number of Car Owner (Select one: First hand = 0 / Second hand = 1 / Third hand = 2 / Fourth hand or more = 3): 0

Output :

Predict selling price : 363487.06 THB

*** Car Selling Price Prediction ***

Input :

Year : 2017

Present Price (THB) : 500000

Kms Driven (km) : 100

Fuel type (Select one: Petrol = 0 / Diesel = 1 / CNG = 2): 0

Seller Type (Select one: Individual = 0 / Dealer = 1): 0

Car Transmission (Select one: Manual = 0 / Automatic = 1): 1

Number of Car Owner (Select one: First hand = 0 / Second hand = 1 / Third hand = 2 / Fourth hand or more = 3): 1

Output :

Predict selling price : 320370.19 THB

*** Car Selling Price Prediction ***

Input :

Year : 2019

Present Price (THB) : 10000000

Kms Driven (km) : 5000

Fuel type (Select one: Petrol = 0 / Diesel = 1 / CNG = 2): 0

Seller Type (Select one: Individual = 0 / Dealer = 1): 0

Car Transmission (Select one: Manual = 0 / Automatic = 1): 0

Number of Car Owner (Select one: First hand = 0 / Second hand = 1 / Third hand = 2 / Fourth hand or more = 3): 0

Output :

Predict selling price : 4549297.19 THB

*** Car Selling Price Prediction ***

Input :

Year : 2015

Present Price (THB) : 200000

Kms Driven (km) : 300

Fuel type (Select one: Petrol = 0 / Diesel = 1 / CNG = 2): 0

Seller Type (Select one: Individual = 0 / Dealer = 1): 0

Car Transmission (Select one: Manual = 0 / Automatic = 1): 0

Number of Car Owner (Select one: First hand = 0 / Second hand = 1 / Third hand = 2 / Fourth hand or more = 3): 1

Output :

Predict selling price : 91245.07 THB

THANKS !