PROBLEM STATEMENT

To predict and analyze which fuel giving high mileage. By using linear regression

In [1]:

```
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn import preprocessing,svm
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
```

In [2]:

df=pd.read_csv(r"C:\Users\Teju\Downloads\used_cars_data.csv")
df

Out[2]:

	S.No.	Name	Location	Year	Kilometers_Driven	Fuel_Type	Transmission	Owr
0	0	Maruti Wagon R LXI CNG	Mumbai	2010	72000	CNG	Manual	
1	1	Hyundai Creta 1.6 CRDi SX Option	Pune	2015	41000	Diesel	Manual	
2	2	Honda Jazz V	Chennai	2011	46000	Petrol	Manual	
3	3	Maruti Ertiga VDI	Chennai	2012	87000	Diesel	Manual	
4	4	Audi A4 New 2.0 TDI Multitronic	Coimbatore	2013	40670	Diesel	Automatic	
7248	7248	Volkswagen Vento Diesel Trendline	Hyderabad	2011	89411	Diesel	Manual	
7249	7249	Volkswagen Polo GT TSI	Mumbai	2015	59000	Petrol	Automatic	
7250	7250	Nissan Micra Diesel XV	Kolkata	2012	28000	Diesel	Manual	
7251	7251	Volkswagen Polo GT TSI	Pune	2013	52262	Petrol	Automatic	
7252	7252	Mercedes- Benz E- Class 2009- 2013 E 220 CDI Avan	Kochi	2014	72443	Diesel	Automatic	

7253 rows × 14 columns

In [3]:

df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 7253 entries, 0 to 7252
Data columns (total 14 columns):

#	Column	Non-Null Count	Dtype
0	S.No.	7253 non-null	int64
1	Name	7253 non-null	object
2	Location	7253 non-null	object
3	Year	7253 non-null	int64
4	Kilometers_Driven	7253 non-null	int64
5	Fuel_Type	7253 non-null	object
6	Transmission	7253 non-null	object
7	Owner_Type	7253 non-null	object
8	Mileage	7251 non-null	object
9	Engine	7207 non-null	object
10	Power	7207 non-null	object
11	Seats	7200 non-null	float64
12	New_Price	1006 non-null	object
13	Price	6019 non-null	float64
44	C1+C4/2\+	(4/3) - - + /0)	

dtypes: float64(2), int64(3), object(9)

memory usage: 793.4+ KB

In [4]:

df.head()

Out[4]:

	S.No.	Name	Location	Year	Kilometers_Driven	Fuel_Type	Transmission	Owner_T
0	0	Maruti Wagon R LXI CNG	Mumbai	2010	72000	CNG	Manual	F
1	1	Hyundai Creta 1.6 CRDi SX Option	Pune	2015	41000	Diesel	Manual	F
2	2	Honda Jazz V	Chennai	2011	46000	Petrol	Manual	F
3	3	Maruti Ertiga VDI	Chennai	2012	87000	Diesel	Manual	F
4	4	Audi A4 New 2.0 TDI Multitronic	Coimbatore	2013	40670	Diesel	Automatic	Sec
4								•

In [5]:

df.tail()

Out[5]:

	S.No.	Name	Location	Year	Kilometers_Driven	Fuel_Type	Transmission	Own
7248	7248	Volkswagen Vento Diesel Trendline	Hyderabad	2011	89411	Diesel	Manual	
7249	7249	Volkswagen Polo GT TSI	Mumbai	2015	59000	Petrol	Automatic	
7250	7250	Nissan Micra Diesel XV	Kolkata	2012	28000	Diesel	Manual	
7251	7251	Volkswagen Polo GT TSI	Pune	2013	52262	Petrol	Automatic	
7252	7252	Mercedes- Benz E- Class 2009- 2013 E 220 CDI Avan	Kochi	2014	72443	Diesel	Automatic	
4.6	_	_		_				

In [7]:

df.shape

Out[7]:

(7253, 14)

In [8]:

df.describe

Out[8]:

<box< th=""><th></th><th>NDFrame.des</th><th>cribe</th><th>of S.</th><th>No.</th><th></th><th></th></box<>		NDFrame.des	cribe	of S.	No.		
0 1 2 3 4	0 1 2 3 4	Maruti Wagon R LXI CNG Hyundai Creta 1.6 CRDi SX Option Honda Jazz V Maruti Ertiga VDI					
7248 7249 7250 7251 7252	7248 7249 7250 7251 7252	Mercedes-Ben		Ni Vo	lkswagen Po ssan Micra lkswagen Po	olo GT TSI Diesel XV olo GT TSI	Hyderabad Mumbai Kolkata Pune Kochi
٥ ١	Year K	ilometers_Dr	iven F	uel_Type Tr	ansmission	Owner_Type	Mileag
e \ 0	2010	7	2000	CNG	Manual	First	26.6 km/k
g 1	2015	4	1000	Diesel	Manual	First	19.67 kmp
1 2	2011	4	6000	Petrol	Manual	First	18.2 kmp
1 3	2012	8	7000	Diesel	Manual	First	20.77 kmp
1 4 1	2013	4	0670	Diesel	Automatic	Second	15.2 kmp
•••	•••		•••	•••			
7248 1	2011	8	9411	Diesel	Manual	First	20.54 kmp
7249 1	2015	5	9000	Petrol	Automatic	First	17.21 kmp
7250 1	2012	2	8000	Diesel	Manual	First	23.08 kmp
7251 1	2013	5	2262	Petrol	Automatic	Third	17.2 kmp
7252 1	2014	7	2443	Diesel	Automatic	First	10.0 kmp
0 1 2 3 4 7248 7249 7250 7251 7252	Engine 998 CC 1582 CC 1199 CC 1248 CC 1968 CC 1598 CC 1197 CC 1461 CC 1197 CC 2148 CC	58.16 bhp 126.2 bhp 88.7 bhp 88.76 bhp 140.8 bhp 103.6 bhp 103.6 bhp 103.6 bhp	Seats 5.6 5.6 7.6 5.6 5.6 5.6 5.6	NaN	1.75 12.50 4.50 6.00 17.74 NaN NaN		

[7253 rows x 14 columns]>

In [9]:

```
df.isna().any()
```

Out[9]:

S.No. False Name False Location False False Year Kilometers_Driven False Fuel_Type False Transmission False Owner_Type False Mileage True Engine True Power True True Seats New_Price True Price True dtype: bool

In [11]:

```
df=df[['Kilometers_Driven','Price']]
#Changing the column names
df.columns=['kilometers','price']
df
```

Out[11]:

	kilometers	price
0	72000	1.75
1	41000	12.50
2	46000	4.50
3	87000	6.00
4	40670	17.74
7248	89411	NaN
7249	59000	NaN
7250	28000	NaN
7251	52262	NaN
7252	72443	NaN

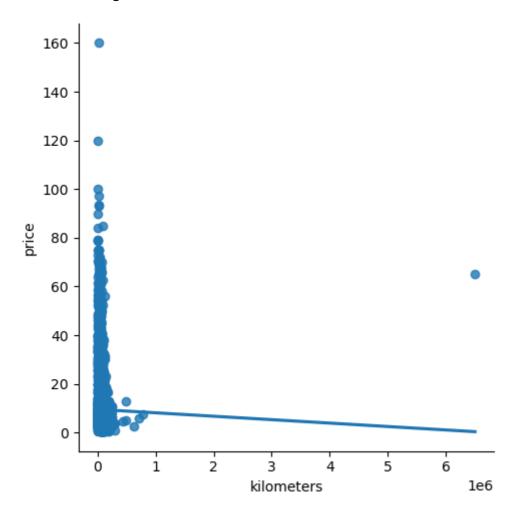
7253 rows × 2 columns

In [12]:

```
sns.lmplot(x="kilometers",y="price",data=df,order=1,ci=None)
```

Out[12]:

<seaborn.axisgrid.FacetGrid at 0x154c6c43fd0>



In [13]:

```
df.fillna(method='ffill',inplace=True)
```

In [14]:

```
x = np.array(df['kilometers']).reshape(-1,1)
y = np.array(df['price']).reshape(-1,1)
```

In [15]:

```
df.dropna(inplace=True)
```

In [16]:

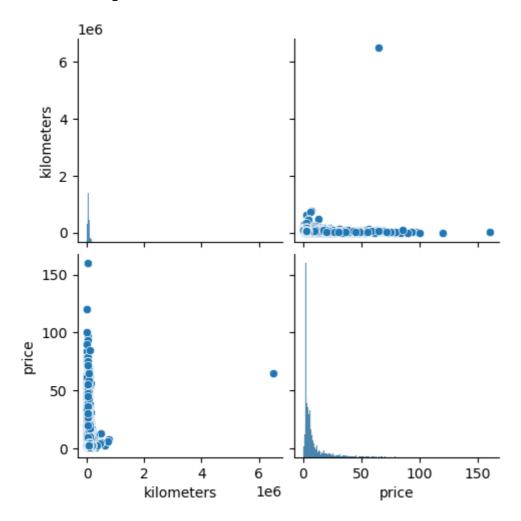
```
x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.25)
regr = LinearRegression()
regr.fit(x_train,y_train)
print(regr.score(x_test,y_test))
```

In [17]:

sns.pairplot(df)

Out[17]:

<seaborn.axisgrid.PairGrid at 0x154c800f880>



In []: