

In [31]:

```
import numpy as np
import pandas as pd
from sklearn import preprocessing
from sklearn import linear_model
```

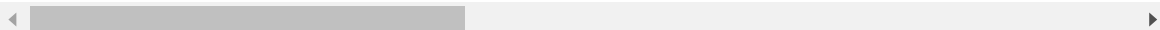
In [4]:

```
df = pd.read_csv("CarPrice_Assignment.csv")
df
```

Out[4]:

	car_ID	symboling	CarName	fueltype	aspiration	doornumber	carbody	drivewheel
0	1	3	alfa-romero giulia	gas	std	two	convertible	rwd
1	2	3	alfa-romero stelvio	gas	std	two	convertible	rwd
2	3	1	alfa-romero Quadrifoglio	gas	std	two	hatchback	rwd
3	4	2	audi 100 ls	gas	std	four	sedan	fwd
4	5	2	audi 100ls	gas	std	four	sedan	4wd
...
200	201	-1	volvo 145e (sw)	gas	std	four	sedan	rwd
201	202	-1	volvo 144ea	gas	turbo	four	sedan	rwd
202	203	-1	volvo 244dl	gas	std	four	sedan	rwd
203	204	-1	volvo 246	diesel	turbo	four	sedan	rwd
204	205	-1	volvo 264gl	gas	turbo	four	sedan	rwd

205 rows × 26 columns



In [7]:

```
df.info
```

Out[7]:

```

<bound method DataFrame.info of          car_ID  symboling          C
arName fueltype aspiration \
0         1         3      alfa-romero giulia      gas      std
1         2         3      alfa-romero stelvio      gas      std
2         3         1  alfa-romero Quadrifoglio      gas      std
3         4         2          audi 100 ls      gas      std
4         5         2          audi 100ls      gas      std
..      ...      ...      ...      ...      ...
200      201        -1        volvo 145e (sw)      gas      std
201      202        -1          volvo 144ea      gas    turbo
202      203        -1        volvo 244dl      gas      std
203      204        -1          volvo 246    diesel    turbo
204      205        -1        volvo 264gl      gas    turbo

      doornumber      carbody drivewheel enginelocation  wheelbase  ... \
0         two  convertible      rwd      front      88.6  ...
1         two  convertible      rwd      front      88.6  ...
2         two    hatchback      rwd      front      94.5  ...
3         four      sedan      fwd      front      99.8  ...
4         four      sedan      4wd      front      99.4  ...
..      ...      ...      ...      ...      ...
200      four      sedan      rwd      front     109.1  ...
201      four      sedan      rwd      front     109.1  ...
202      four      sedan      rwd      front     109.1  ...
203      four      sedan      rwd      front     109.1  ...
204      four      sedan      rwd      front     109.1  ...

      enginesize  fuelsystem  boreratio  stroke  compressionratio  horsepower
\
0         130      mpfi      3.47    2.68      9.0      111
1         130      mpfi      3.47    2.68      9.0      111
2         152      mpfi      2.68    3.47      9.0      154
3         109      mpfi      3.19    3.40     10.0      102
4         136      mpfi      3.19    3.40      8.0      115
..      ...      ...      ...      ...      ...
200      141      mpfi      3.78    3.15      9.5      114
201      141      mpfi      3.78    3.15      8.7      160
202      173      mpfi      3.58    2.87      8.8      134
203      145      idi      3.01    3.40     23.0      106
204      141      mpfi      3.78    3.15      9.5      114

      peakrpm  citympg  highwaympg  price
0         5000      21      27  13495.0
1         5000      21      27  16500.0
2         5000      19      26  16500.0
3         5500      24      30  13950.0
4         5500      18      22  17450.0
..      ...      ...      ...
200      5400      23      28  16845.0
201      5300      19      25  19045.0
202      5500      18      23  21485.0
203      4800      26      27  22470.0
204      5400      19      25  22625.0

```

[205 rows x 26 columns]>

In [9]:

```
df.isna().sum()
```

Out[9]:

```
car_ID          0
symboling       0
CarName         0
fueltype        0
aspiration      0
doornumber      0
carbody         0
drivewheel      0
enginelocation  0
wheelbase       0
carlength       0
carwidth        0
carheight       0
curbweight      0
enginetype      0
cylindernumber  0
enginesize      0
fuelsystem      0
boreratio       0
stroke          0
compressionratio 0
horsepower      0
peakrpm         0
citympg         0
highwaympg      0
price           0
dtype: int64
```

In []:

Data Preprocessing

In [13]:

```
lab_en = preprocessing.LabelEncoder()
lab_en.fit(df.fueltype)
df.fueltype = lab_en.transform(df.fueltype)
```

In [14]:

```
lab_en.fit(df.doornumber)
df.doornumber = lab_en.transform(df.doornumber)
```

In [15]:

```
lab_en.fit(df.aspiration)
df.aspiration = lab_en.transform(df.aspiration)
```

In [16]:

```
lab_en.fit(df.drivewheel)
df.drivewheel = lab_en.transform(df.drivewheel)
```

In [17]:

```
lab_en.fit(df.carbody)
df.carbody = lab_en.transform(df.carbody)
```

In [18]:

```
lab_en.fit(df.enginelocation)
df.enginelocation = lab_en.transform(df.enginelocation)
```

In [19]:

```
lab_en.fit(df.enginetype)
df.enginetype = lab_en.transform(df.enginetype)
```

In [20]:

```
lab_en.fit(df.cylindernumber)
df.cylindernumber = lab_en.transform(df.cylindernumber)
```

In [21]:

```
lab_en.fit(df.fuelsystem)
df.fuelsystem = lab_en.transform(df.fuelsystem)
```

In [39]:

df

Out[39]:

	fueltype	aspiration	doornumber	carbody	drivewheel	enginelocation	wheelbase	...	engine
0	1	0	1	0	2	0	88.6	...	
1	1	0	1	0	2	0	88.6	...	
2	1	0	1	2	2	0	94.5	...	
3	1	0	0	3	1	0	99.8	...	
4	1	0	0	3	0	0	99.4	...	
...	
8	1	0	0	3	2	0	109.1	...	
9	1	1	0	3	2	0	109.1	...	
10	1	0	0	3	2	0	109.1	...	
11	0	1	0	3	2	0	109.1	...	
12	1	1	0	3	2	0	109.1	...	

In [34]:

```
lreg = linear_model.LinearRegression()

independent_var = df[[ 'fueltype', 'aspiration',
                        'doornumber', 'carbody', 'drivewheel', 'enginelocation', 'wheelbase',
                        'carlength', 'carwidth', 'carheight', 'curbweight', 'enginetype',
                        'cylindernumber', 'engine size', 'fuelsystem', 'boreratio', 'stroke',
                        'compressionratio', 'horsepower', 'peakrpm', 'citympg', 'highwaympg']]
dependent_var = df.price

lreg.fit(independent_var,dependent_var )
```

Out[34]:

LinearRegression(copy_X=True, fit_intercept=True, n_jobs=None, normalize=False)

In [35]:

```
lreg.coef_
```

Out[35]:

```
array([-9.21677014e+03,  2.42352286e+02, -5.13930613e+02, -8.19139857e+02,
        1.16616708e+03,  1.03974427e+04,  1.10593867e+02, -2.91102195e+01,
        7.12602036e+02,  1.84232716e+02,  2.35662037e+00,  1.90948576e+02,
        1.28166071e+02,  1.01622195e+02, -1.48294905e+02, -2.74450799e+03,
       -2.82460445e+03, -5.48092812e+02,  2.06295041e+01,  2.10440991e+00,
       -1.45461967e+02,  1.35183217e+02])
```

In [37]:

```
lreg.intercept_
```

Out[37]:

```
-49313.25002503426
```

In [44]:

```
lreg.predict([[1,0,1,0,2,0,88.6,168.8,64.1,48.8,2548,0,0,130,0,3.47,2.68,9,111,5000,21,25]])
```

Out[44]:

```
array([13438.04815291])
```

In [38]:

```
df['price'][0]
```

Out[38]:

```
13495.0
```

In []:

```
predicted price is similar to the actual value
```