

Importing Libraries

In [46]:

```
import numpy as np
import pandas as pd
from pandas import DataFrame as df
import matplotlib.pyplot as plt
from scipy import stats
import scipy.stats as ss
import seaborn as sns
```

Loading file

In [2]:

```
filename = 'https://s3-api.us-geo.objectstorage.softlayer.net/cf-courses-data/CognitiveClass/DA0101EN/automobileEDA.csv'
```

In [3]:

```
cars = pd.read_csv(filename)
```

In [4]:

```
cars
```

Out[4]:

	symboling	normalized-losses	make	aspiration	num-of-doors	body-style	drive-wheels	engine-location	wheel-base
0	3	122	alfa-romero	std	two	convertible	rwd	front	88
1	3	122	alfa-romero	std	two	convertible	rwd	front	88
2	1	122	alfa-romero	std	two	hatchback	rwd	front	94
3	2	164	audi	std	four	sedan	fwd	front	99
4	2	164	audi	std	four	sedan	4wd	front	99
...
196	-1	95	volvo	std	four	sedan	rwd	front	109
197	-1	95	volvo	turbo	four	sedan	rwd	front	109
198	-1	95	volvo	std	four	sedan	rwd	front	109
199	-1	95	volvo	turbo	four	sedan	rwd	front	109
200	-1	95	volvo	turbo	four	sedan	rwd	front	109

201 rows × 29 columns

Viewing DataTypes

In [5]:

```

print("Dimensions are:")
print(cars.shape)
print("")
print("Datatypes:")
print(cars.dtypes)

```

Dimensions are:
(201, 29)

Datatypes:

symboling	int64
normalized-losses	int64
make	object
aspiration	object
num-of-doors	object
body-style	object
drive-wheels	object
engine-location	object
wheel-base	float64
length	float64
width	float64
height	float64
curb-weight	int64
engine-type	object
num-of-cylinders	object
engine-size	int64
fuel-system	object
bore	float64
stroke	float64
compression-ratio	float64
horsepower	float64
peak-rpm	float64
city-mpg	int64
highway-mpg	int64
price	float64
city-L/100km	float64
horsepower-binned	object
diesel	int64
gas	int64
dtype:	object

Viewing correlation between variables 'price' and 'engine-size'

In [6]:

```
correlation = cars.corr()  
correlation['price']['engine-size']
```

Out[6]:

0.8723351674455199

Checking for negative or positive correlation of variables with respect to price

In [7]:

```
for i in cars.describe().columns:
    a = correlation['price'][i]
    if (a > 0 or a < 0):
        sign = "Positive"
        if (a < 0):
            sign = "Negative"
    else:
        sign = "Positive"
    print("Column name: {gog:s}          Correlation: {fog:s}".format(gog = i,
fog = sign))
```

```
Column name: symboling          Correlation: Negative
Column name: normalized-losses  Correlation: Positive
Column name: wheel-base        Correlation: Positive
Column name: length            Correlation: Positive
Column name: width             Correlation: Positive
Column name: height            Correlation: Positive
Column name: curb-weight       Correlation: Positive
Column name: engine-size       Correlation: Positive
Column name: bore              Correlation: Positive
Column name: stroke            Correlation: Positive
Column name: compression-ratio Correlation: Positive
Column name: horsepower        Correlation: Positive
Column name: peak-rpm          Correlation: Negative
Column name: city-mpg          Correlation: Negative
Column name: highway-mpg       Correlation: Negative
Column name: price             Correlation: Positive
Column name: city-L/100km      Correlation: Positive
Column name: diesel            Correlation: Positive
Column name: gas               Correlation: Negative
```

Checking datatype of variable 'peak-rpm'

In [8]:

```
type(cars['peak-rpm'][1])
```

Out[8]:

```
numpy.float64
```

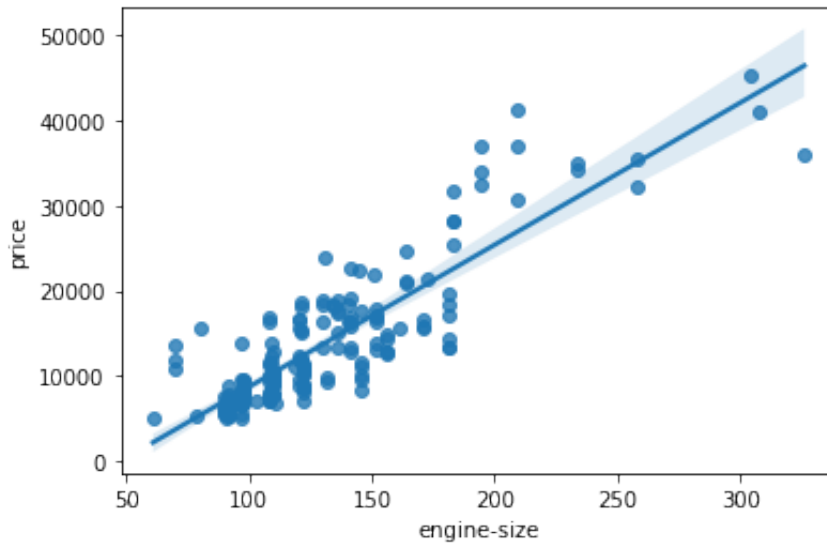
Using seaborn regplot() - plot relation between “engine-size” and “price”.

In [9]:

```
sns.regplot('engine-size', 'price', data=cars)
```

Out[9]:

<matplotlib.axes._subplots.AxesSubplot at 0x7fed7f511c50>



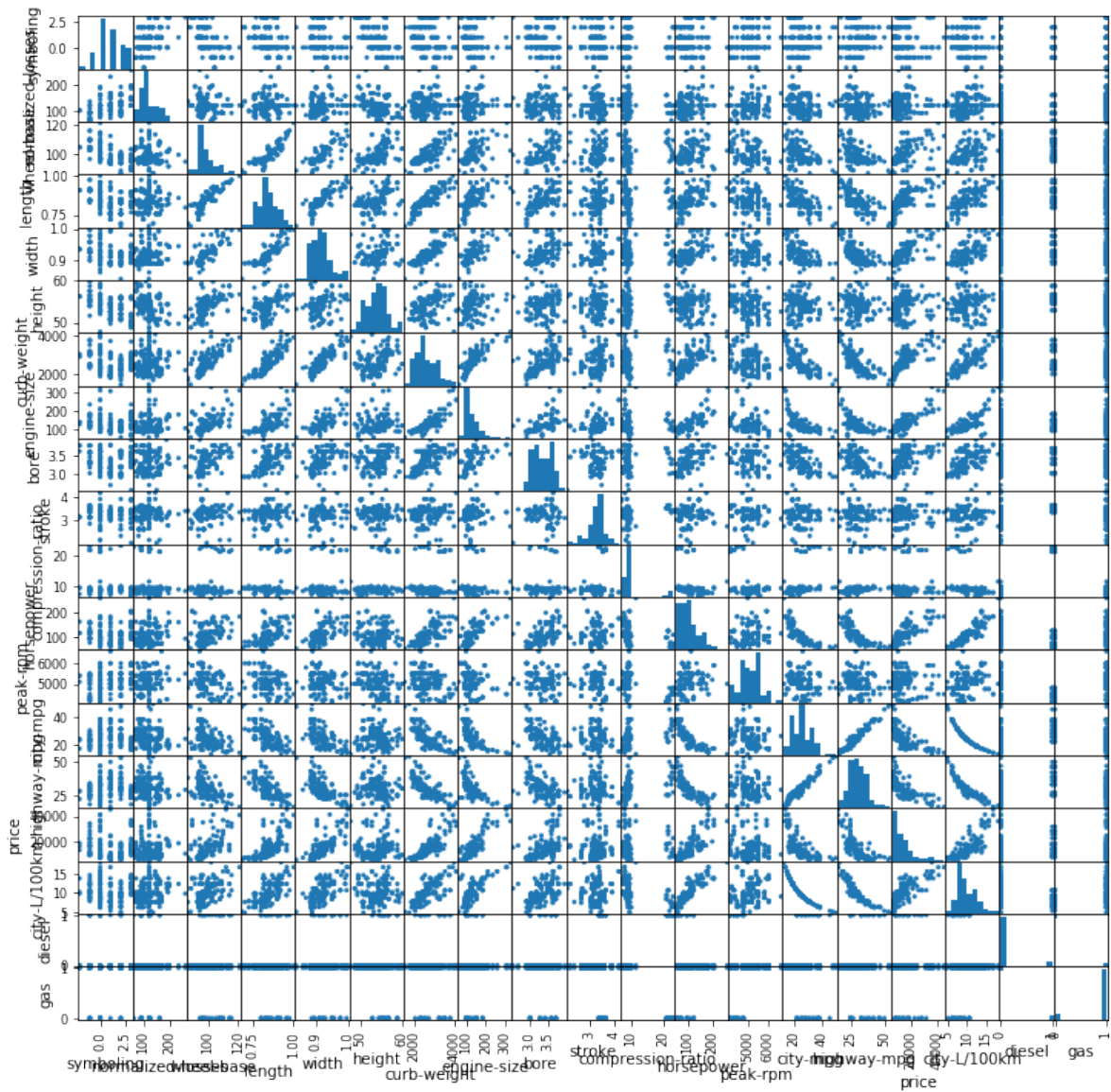
Engine size and price are positively correlated. Checking pandas scatter_matrix

In [12]:

```
from pandas.plotting import scatter_matrix
```

In [18]:

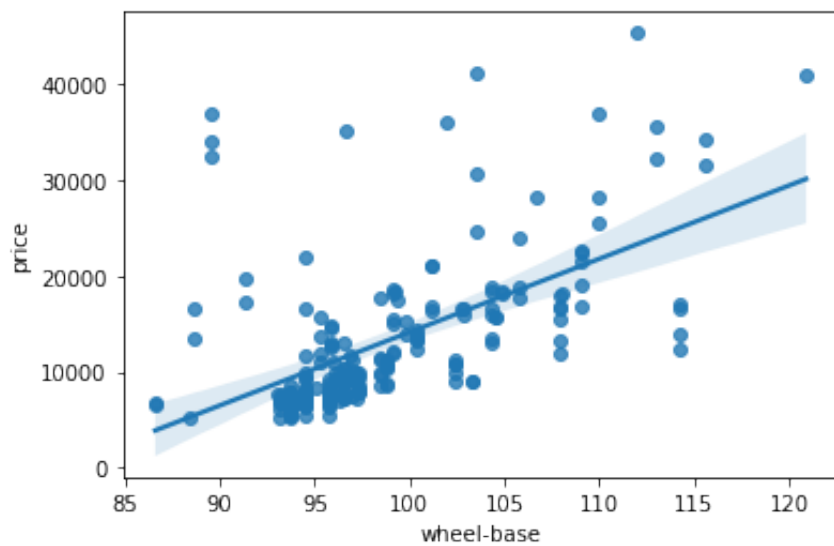
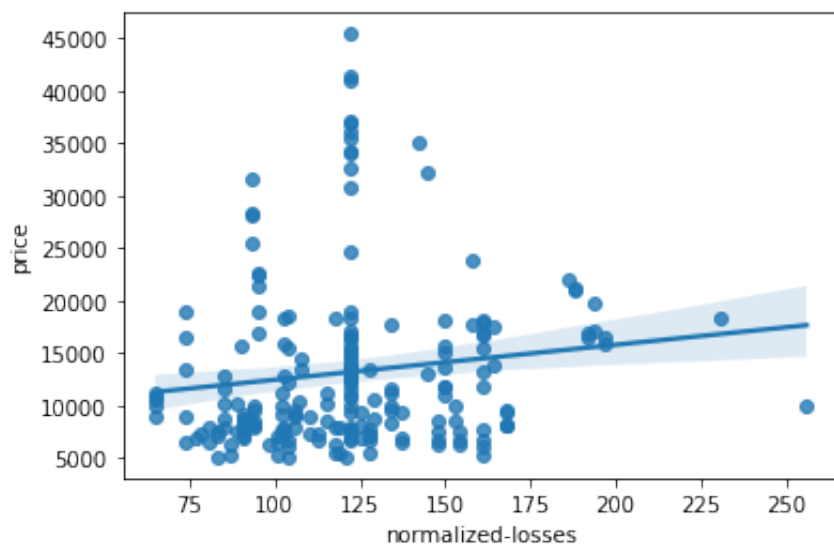
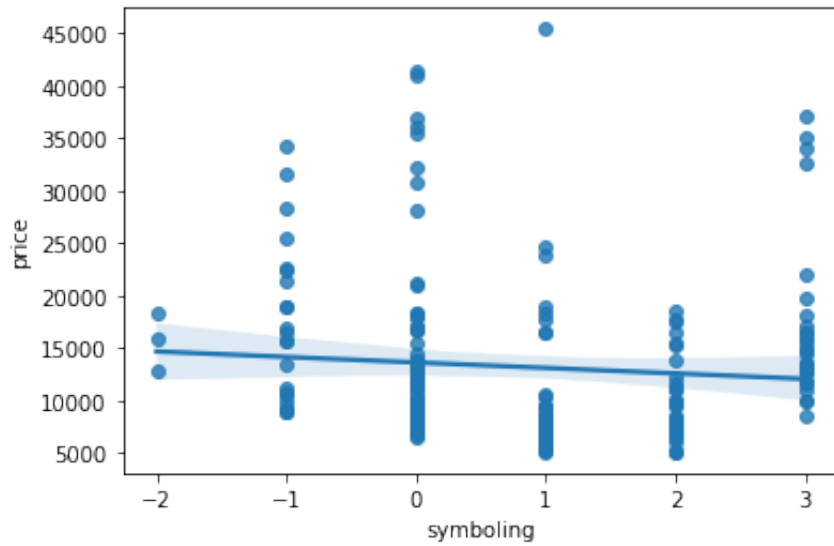
```
scatter_matrix(cars, alpha = 1, figsize = (12,12));
```

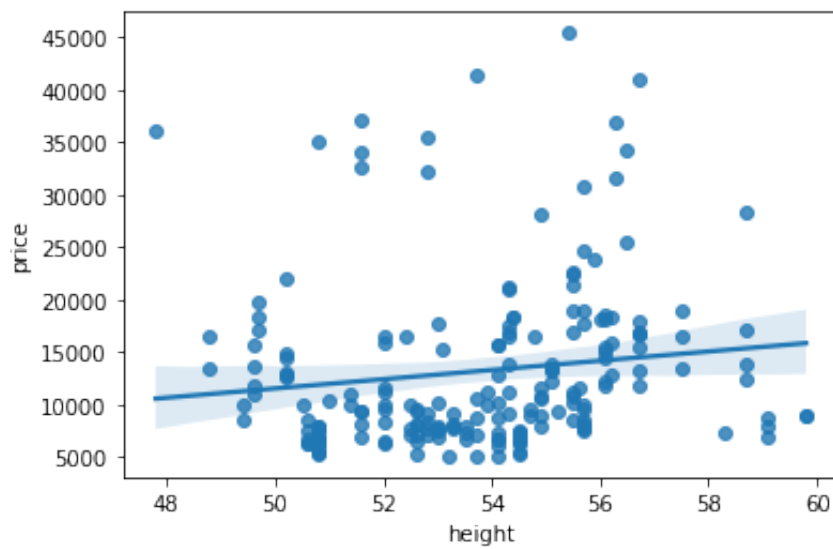
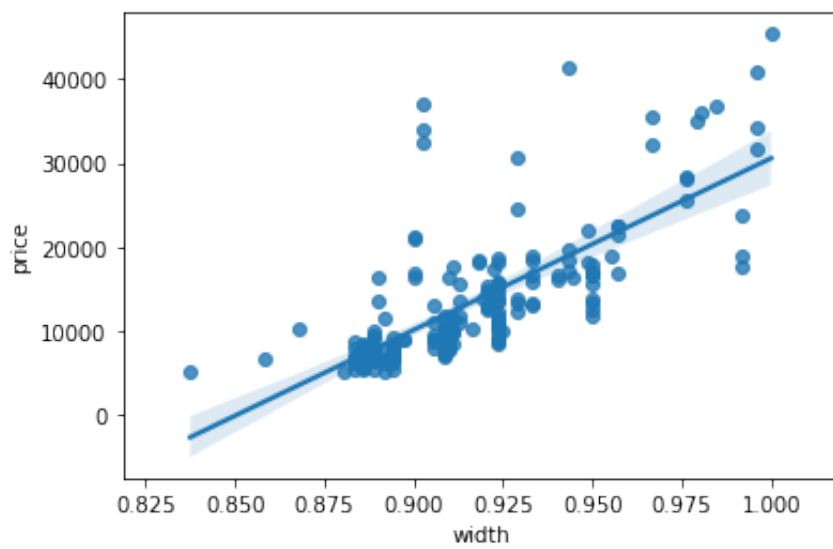
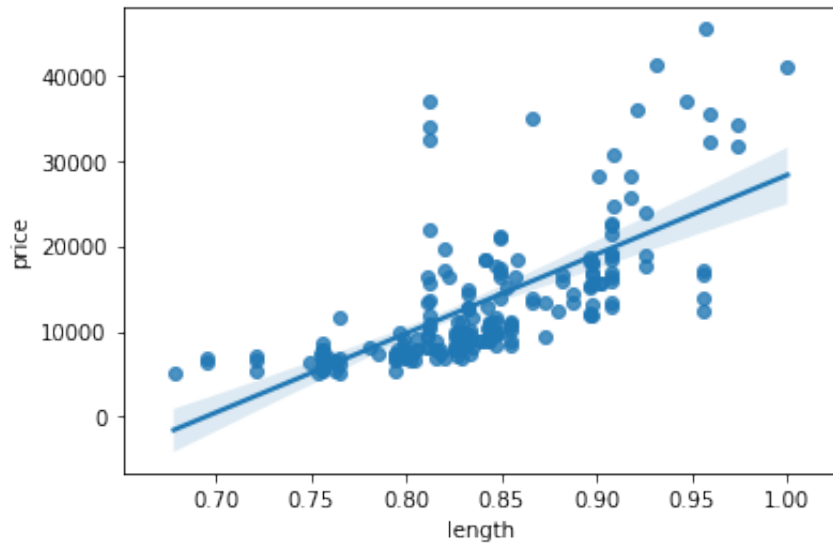


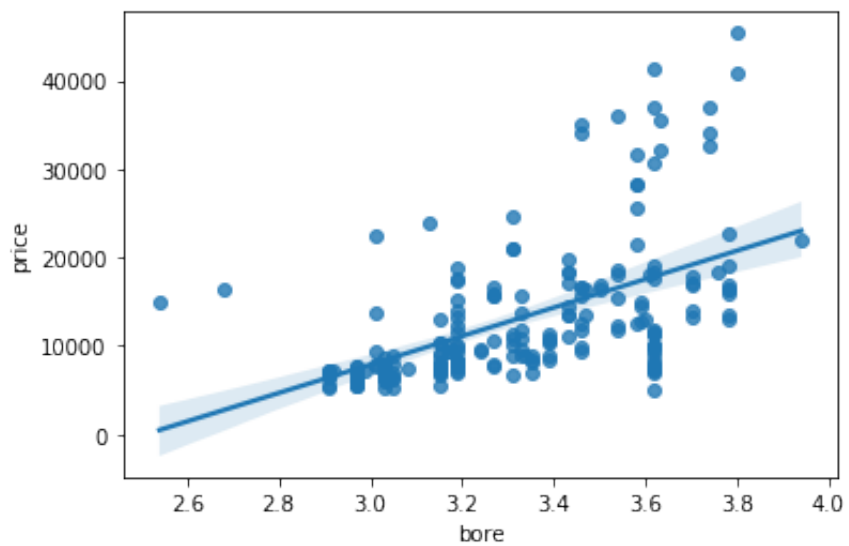
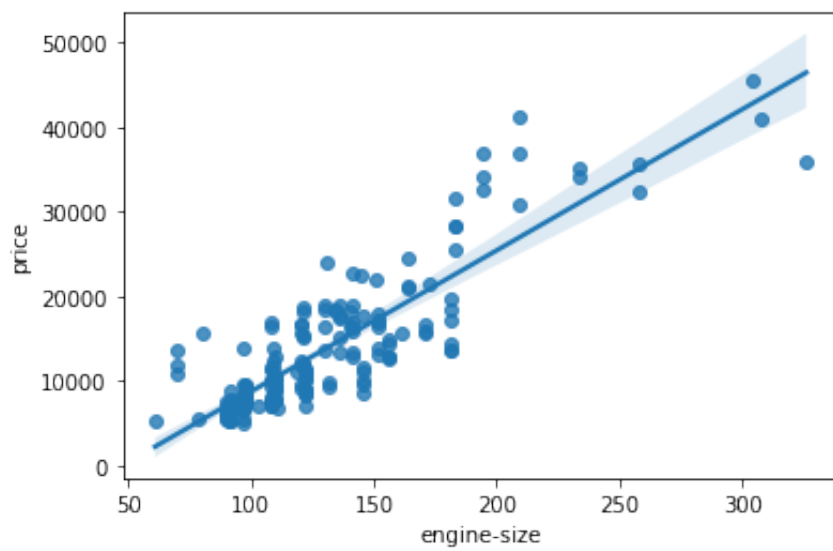
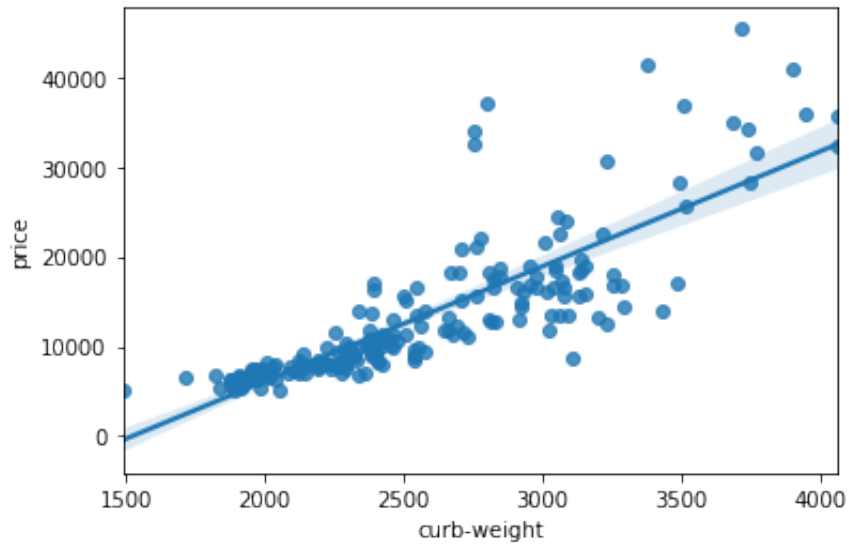
Viewing individual - variable Regplot

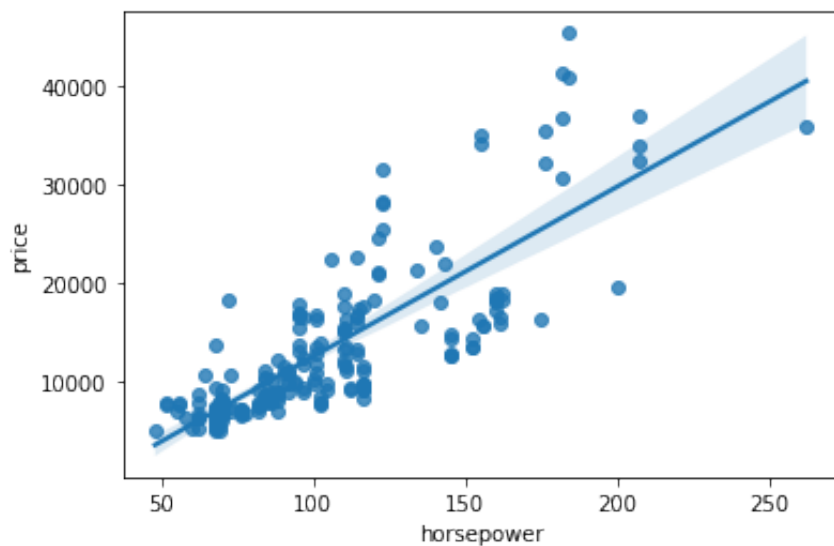
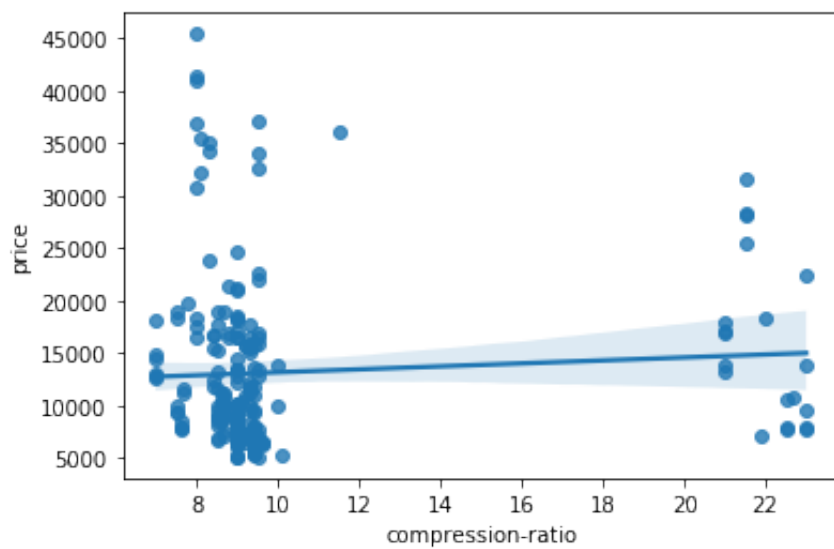
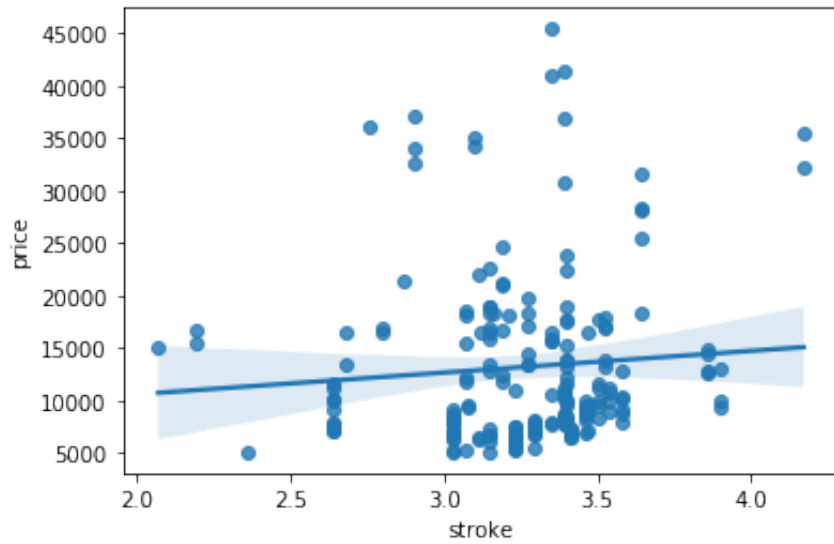
In [19]:

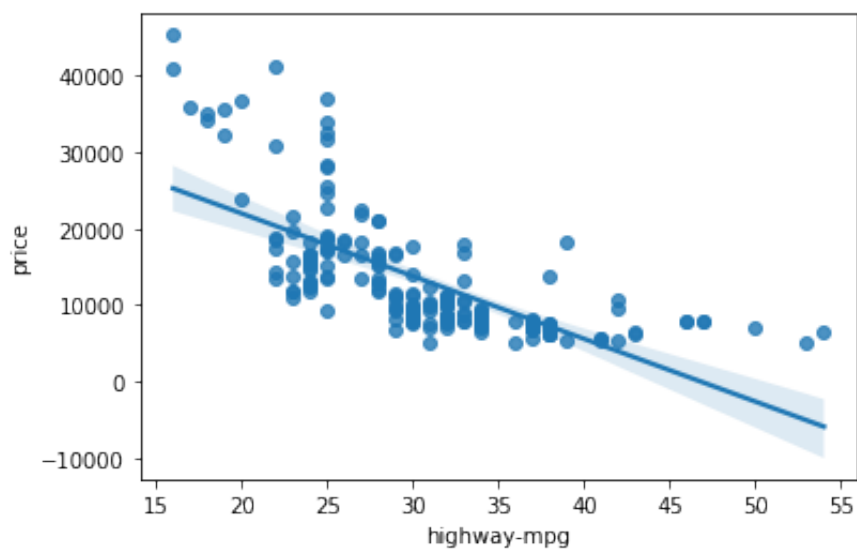
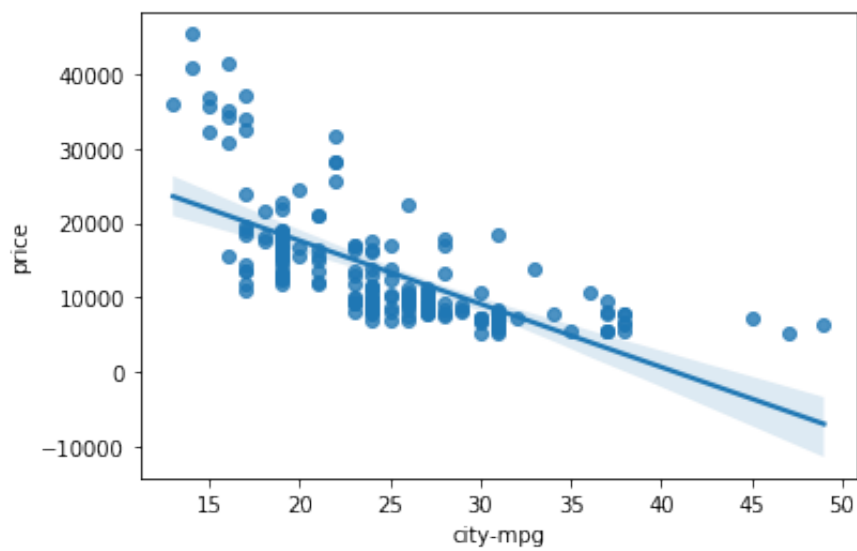
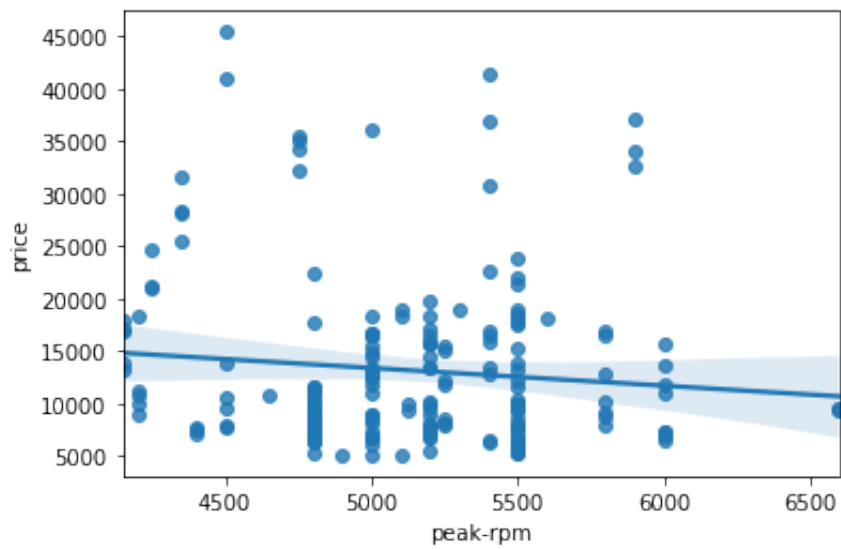
```
for i in cars.describe().columns:
    if i!='price':
        sns.regplot(y='price', x=i, data = cars)
        plt.show()
```

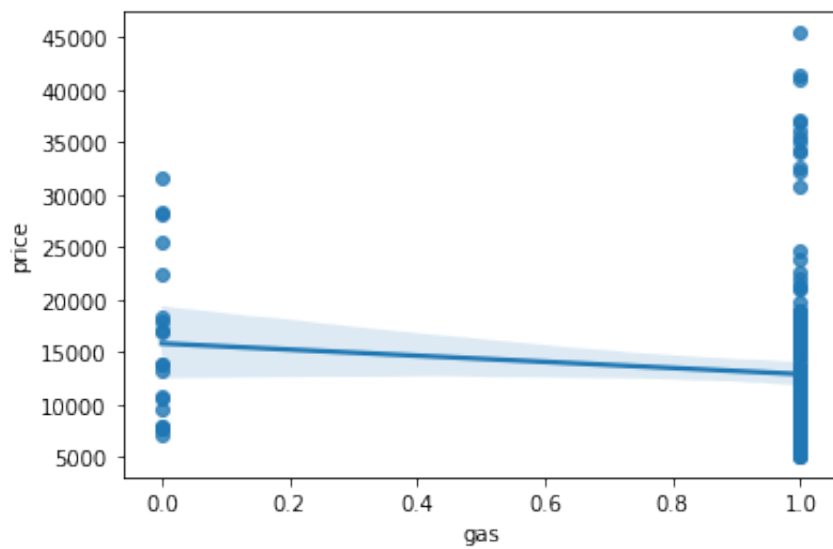
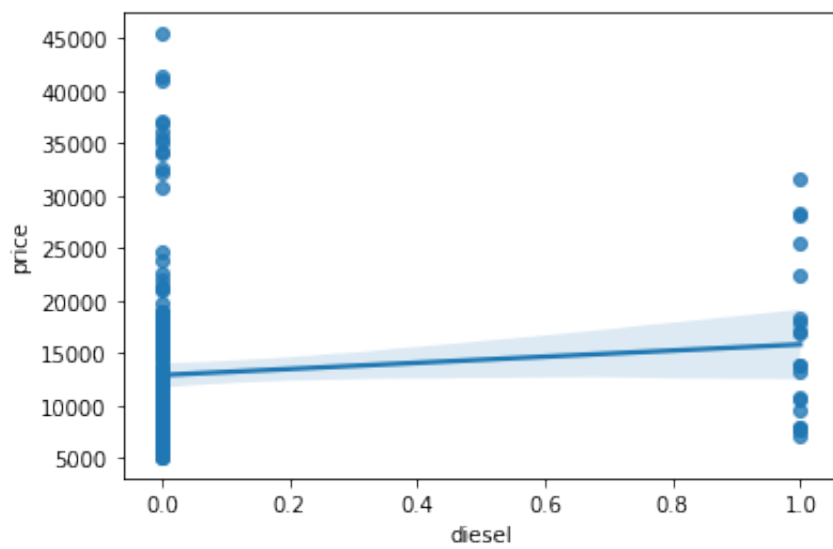
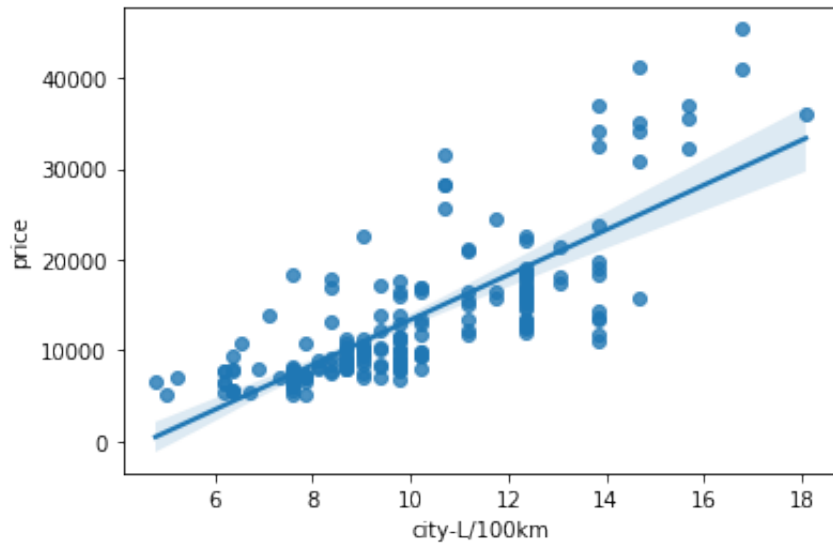








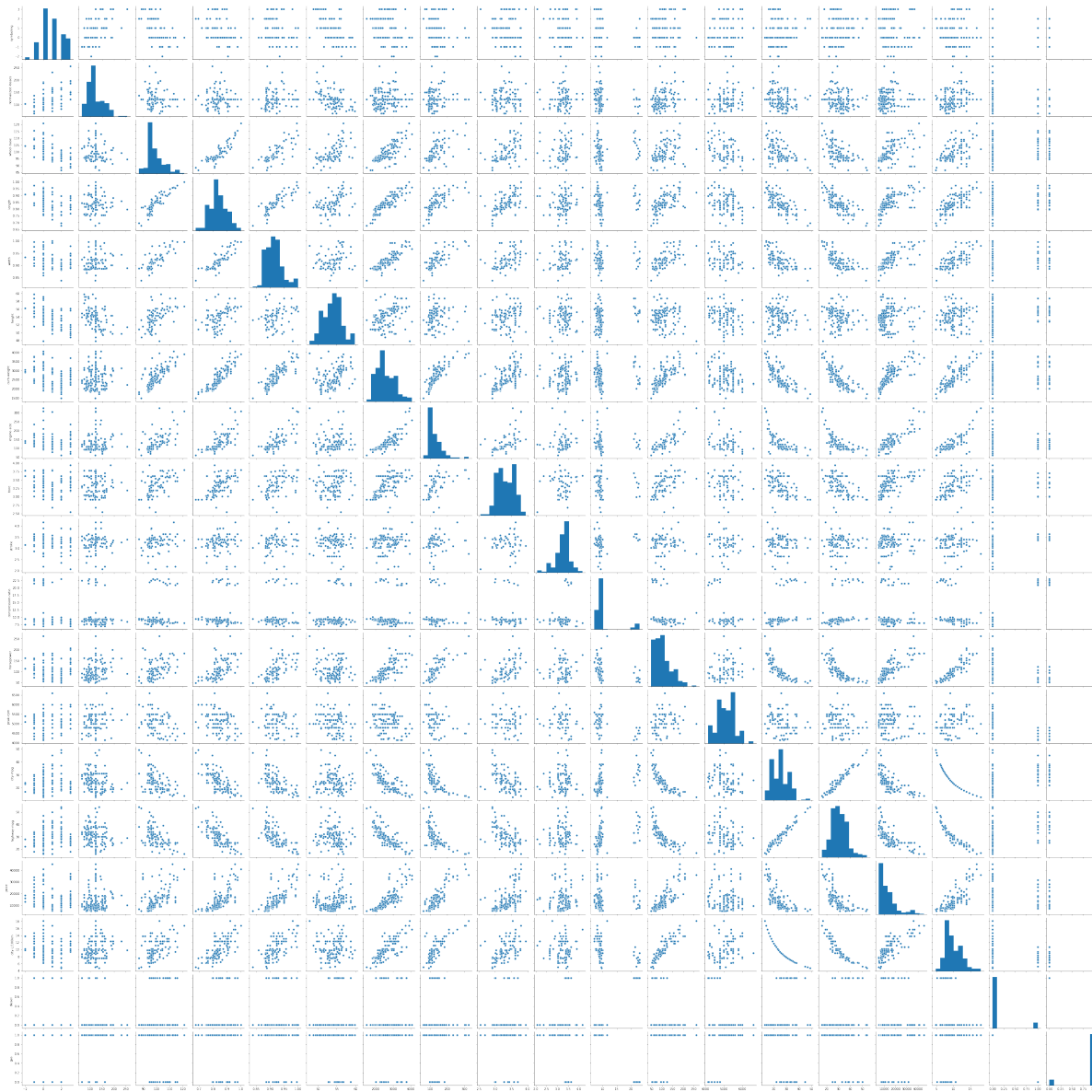




Using the SNS pairplot function

In [20]:

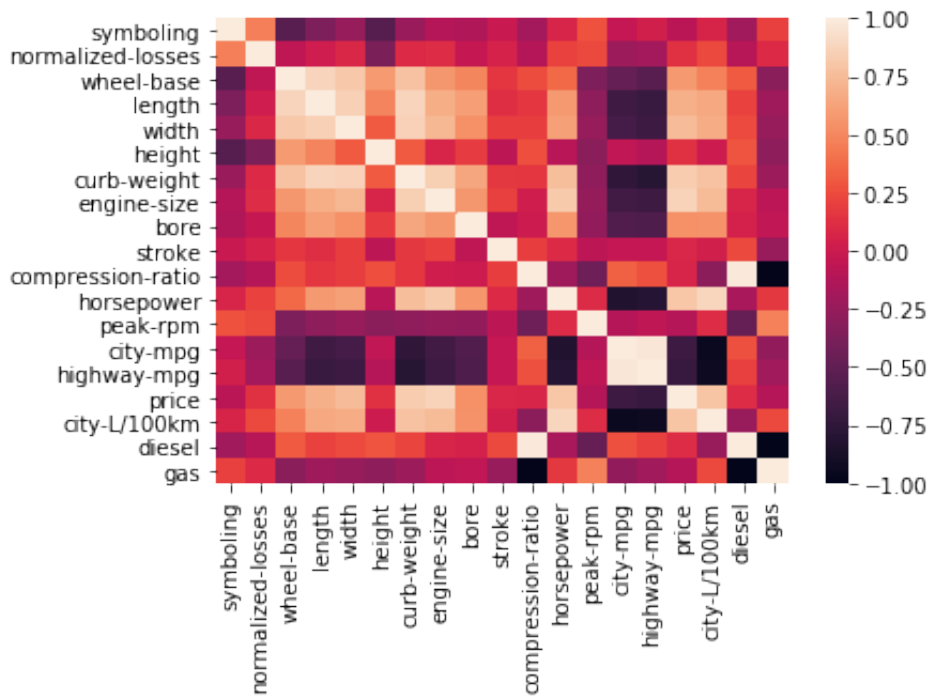
```
sns.pairplot(cars);
```



Viewing Heatmap of the correlation

In [21]:

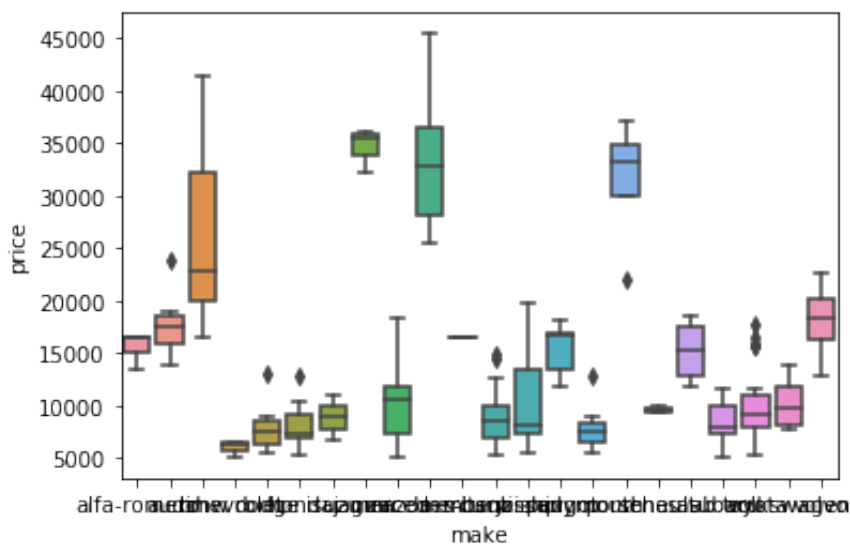
```
sns.heatmap(correlation)
plt.show()
```

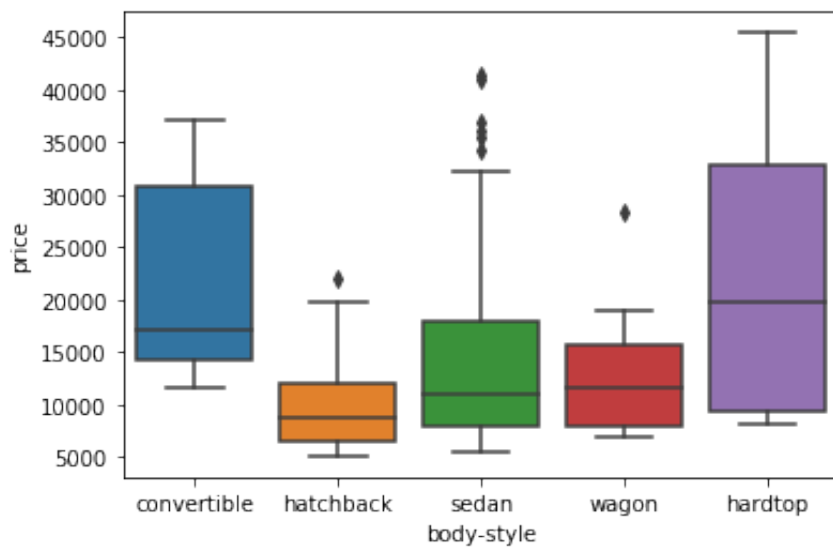
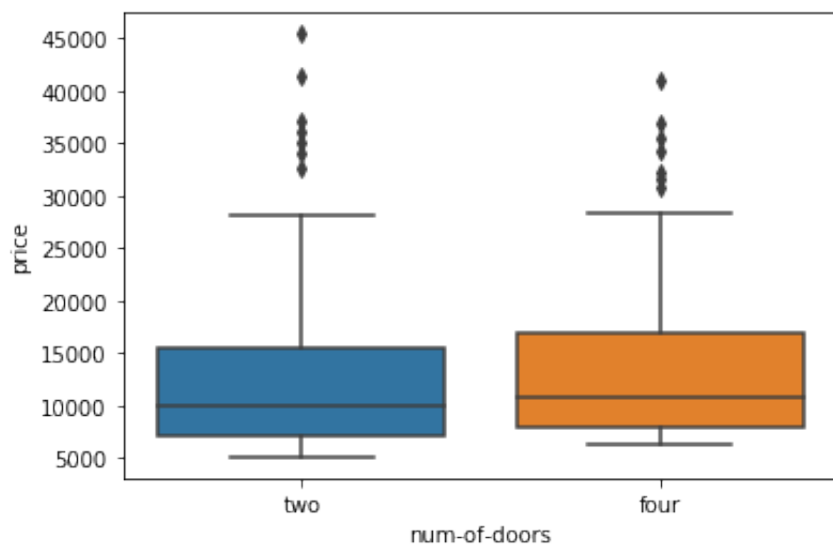
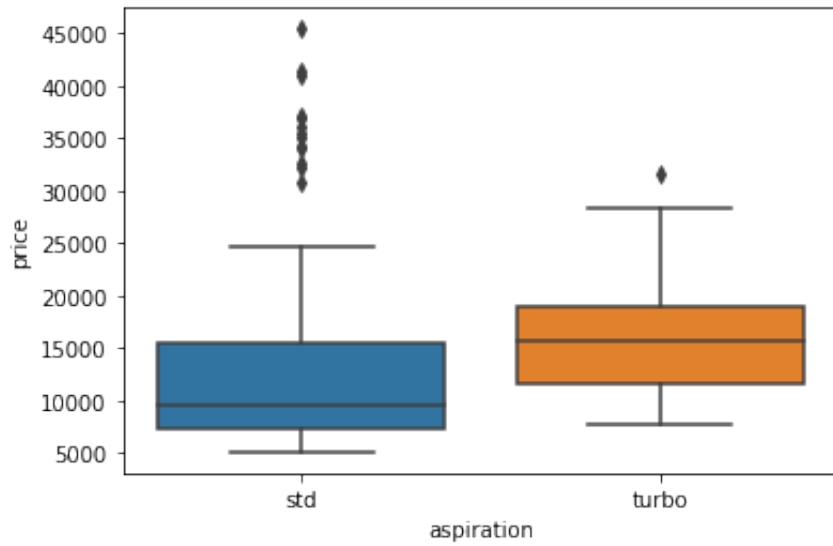


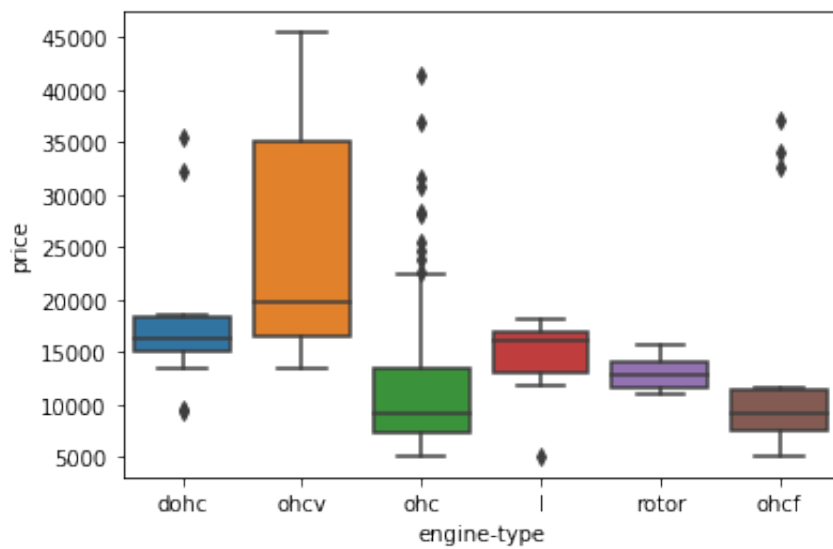
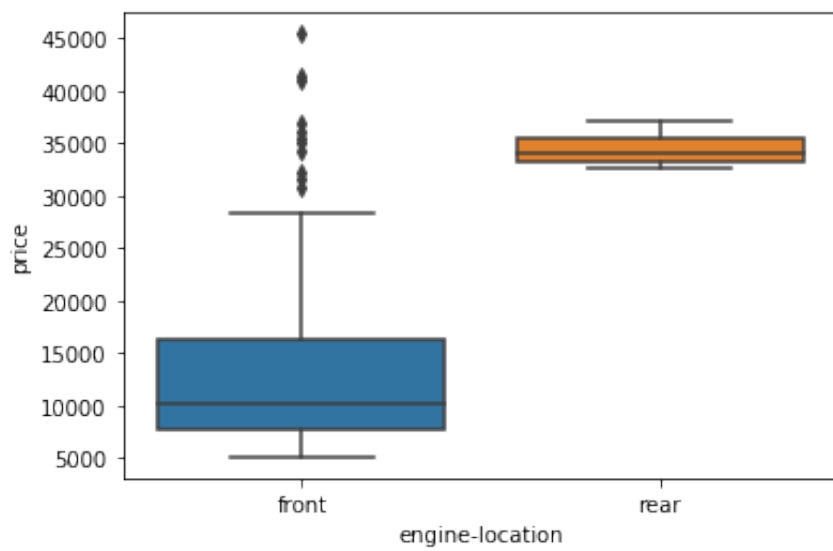
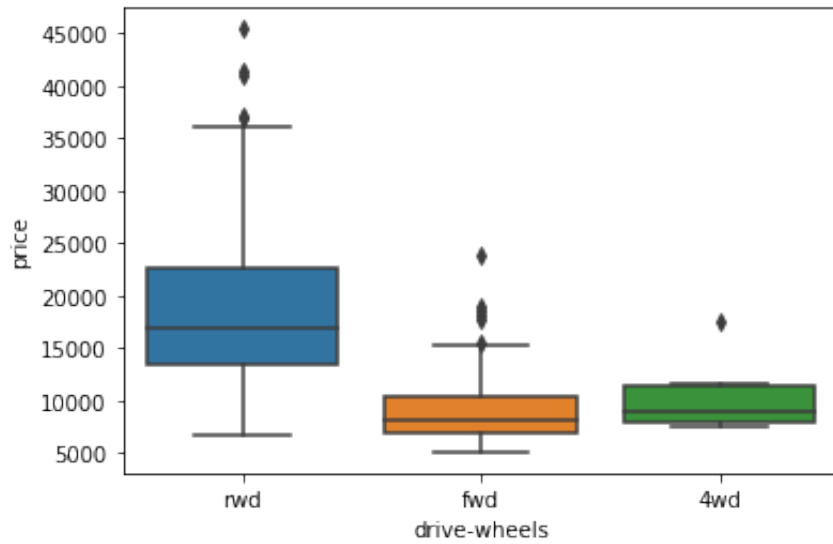
With seaborn boxplot() - comparing “body-style” with “price” and for other categorical variables

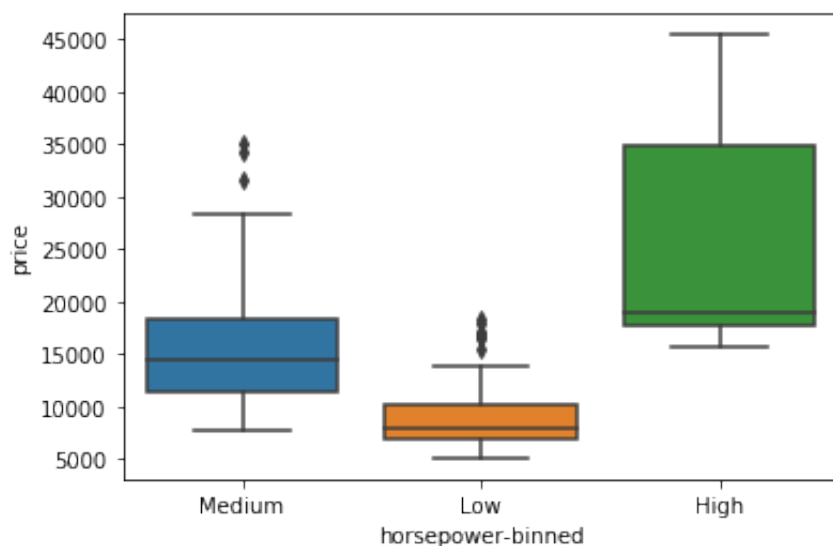
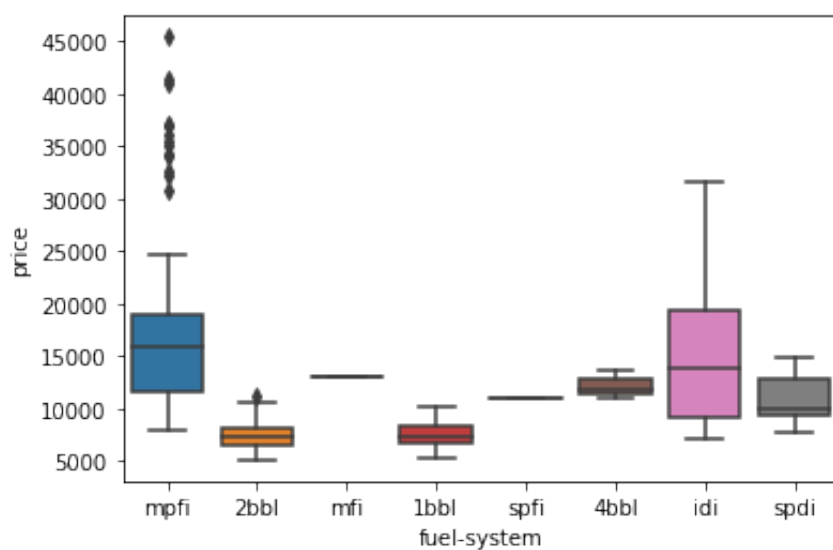
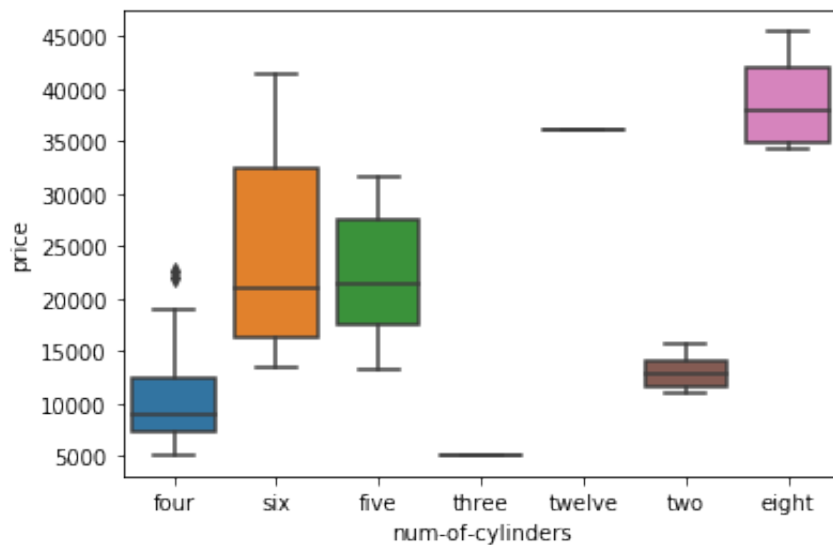
In [22]:

```
for i in cars.describe(include = np.object).columns:
    sns.boxplot(x=cars[i], y=cars.price)
plt.show()
```









Observations :- body-style - We see that the distributions of price between the different body-style categories have a significant overlap, and so body-style would not be a good predictor of price. More or less the price range for every body type car is well defined with rarely any outliers, with the exception of sedan. There are a lot of sport sedans which cost more than the average sedan and cost more and hence are

outliers in the data. drive-wheels - Here we see that the distribution of price between these two engine-location categories, front and rear, are distinct enough to take engine-location as a potential good predictor of price. Majority of the cars have rear wheel drive and have very few outliers. Cars with front wheel or all wheel drive are fairly less and have a very short and well defined range for their price. engine-location - Here we see that the distribution of price between the different drive-wheels categories differs; as such drive-wheels could potentially be a predictor of price. The majority of the cars have enging in the front and the and have a well defined price range. The cars with rear engines and the outliers of cars with front engines are majorly sports cars and that is why they are priced higher overall. horsepower-binned - We can see that as the horsepower of a car increases, the price of the car also increases and the range to define the price of a car with low, medium or high horsepower gets broaded and broader. Using describe() to get descriptive statistics of numeric variables

In [24]:

```
cars.describe()
```

Out[24]:

	symboling	normalized- losses	wheel- base	length	width	height	curb-weight
count	201.000000	201.000000	201.000000	201.000000	201.000000	201.000000	201.000000
mean	0.840796	122.000000	98.797015	0.837102	0.915126	53.766667	2555.6666
std	1.254802	31.99625	6.066366	0.059213	0.029187	2.447822	517.2967
min	-2.000000	65.000000	86.600000	0.678039	0.837500	47.800000	1488.0000
25%	0.000000	101.000000	94.500000	0.801538	0.890278	52.000000	2169.0000
50%	1.000000	122.000000	97.000000	0.832292	0.909722	54.100000	2414.0000
75%	2.000000	137.000000	102.400000	0.881788	0.925000	55.500000	2926.0000
max	3.000000	256.000000	120.900000	1.000000	1.000000	59.800000	4066.0000

Using describe() to get stats of categorical variablesThe default setting of "describe" skips variables of type object. We can apply the method "describe" on the variables of type 'object' as follows:

In [26]:

```
cars.describe(include=np.object)
```

Out[26]:

	make	aspiration	num- of- doors	body- style	drive- wheels	engine- location	engine- type	num-of- cylinders	fuel- system	hor
count	201	201	201	201	201	201	201	201	201	
unique	22	2	2	5	3	2	6	7	8	
top	toyota	std	four	sedan	fwd	front	ohc	four	mpfi	
freq	32	165	115	94	118	198	145	157	92	

Getting unique values in each categorical variable along with their frequency.

In [28]:

```
for i in cars.describe(include=np.object).columns:
    print(i)
    print(cars[i].value_counts())
    print("")
    print("")
```

```
make
toyota      32
nissan      18
mazda       17
mitsubishi  13
honda       13
subaru      12
volkswagen  12
peugot      11
volvo       11
dodge       9
bmw         8
mercedes-benz 8
plymouth    7
audi        6
saab        6
porsche     4
chevrolet   3
alfa-romero 3
jaguar      3
renault     2
isuzu       2
mercury     1
Name: make, dtype: int64
```

```
aspiration
```

```
std      165
turbo    36
Name: aspiration, dtype: int64
```

```
num-of-doors
four     115
two      86
Name: num-of-doors, dtype: int64
```

```
body-style
sedan    94
hatchback 68
wagon    25
hardtop   8
convertible 6
Name: body-style, dtype: int64
```

```
drive-wheels
fwd      118
rwd      75
4wd      8
Name: drive-wheels, dtype: int64
```

```
engine-location
front    198
rear      3
Name: engine-location, dtype: int64
```

```
engine-type
ohc      145
ohcf     15
ohcv     13
dohc     12
l        12
rotor     4
Name: engine-type, dtype: int64
```

```
num-of-cylinders
four     157
six      24
five     10
two       4
eight     4
twelve    1
three     1
Name: num-of-cylinders, dtype: int64
```

```

fuel-system
mpfi      92
2bbl      64
idi       20
1bbl      11
spdi      9
4bbl      3
mfi       1
spfi      1
Name: fuel-system, dtype: int64

```

```

horsepower-binned
Low        115
Medium     62
High       23
Name: horsepower-binned, dtype: int64

```

Value-counts is a good way of understanding how many units of each characteristic/variable we have. We can apply the "value_counts" method on the column 'drive-wheels'. Don't forget the method "value_counts" only works on Pandas series, not Pandas Dataframes. As a result, we only include one bracket "df['drive-wheels']" not two brackets "df[['drive-wheels']]". Using groupby() to get the average price of "drive-wheels" wrt "price"

In [29]:

```

for i in cars.describe(include=np.object).columns:
    print(cars.groupby(i)['price'].mean())
    print("")
    print("")

```

```

make
alfa-romero      15498.333333
audi             17859.166667
bmw             26118.750000
chevrolet        6007.000000
dodge            7875.444444
honda            8184.692308
isuzu            8916.500000
jaguar          34600.000000
mazda           10652.882353
mercedes-benz    33647.000000
mercury          16503.000000
mitsubishi       9239.769231
nissan           10415.666667
peugot           15489.090909
plymouth         7963.428571
porsche          31400.500000
renault          9595.000000
saab             15223.333333
subaru           8541.250000

```

```
toyota          9885.812500
volkswagen      10077.500000
volvo           18063.181818
Name: price, dtype: float64
```

```
aspiration
std          12542.181818
turbo        16254.805556
Name: price, dtype: float64
```

```
num-of-doors
four         13498.034783
two          12818.127907
Name: price, dtype: float64
```

```
body-style
convertible   21890.500000
hardtop       22208.500000
hatchback     9957.441176
sedan         14459.755319
wagon         12371.960000
Name: price, dtype: float64
```

```
drive-wheels
4wd           10241.000000
fwd            9244.779661
rwd           19757.613333
Name: price, dtype: float64
```

```
engine-location
front         12884.085859
rear          34528.000000
Name: price, dtype: float64
```

```
engine-type
dohc          18116.416667
l             14627.583333
ohc           11567.358621
ohcf          13738.600000
ohcv          25098.384615
rotor         13020.000000
Name: price, dtype: float64
```

```
num-of-cylinders
eight         38900.000000
five          22007.600000
```

```
four      10303.197452
six       23671.833333
three     5151.000000
twelve    36000.000000
two       13020.000000
Name: price, dtype: float64
```

```
fuel-system
1bbl      7555.545455
2bbl      7433.203125
4bbl     12145.000000
idi       15838.150000
mfi       12964.000000
mpfi      17605.141304
spdi      10990.444444
spfi      11048.000000
Name: price, dtype: float64
```

```
horsepower-binned
High      25899.130435
Low       9011.704348
Medium    15913.016129
Name: price, dtype: float64
```

The "groupby" method groups data by different categories. The data is grouped based on one or several variables and analysis is performed on the individual groups. From our data, it seems rear-wheel drive vehicles are, on average, the most expensive, while 4-wheel and front-wheel are approximately the same in price.

In [30]:

```
cars.groupby(['drive-wheels', 'body-style'])['price'].mean()
```

Out[30]:

```
drive-wheels  body-style
4wd           hatchback    7603.000000
              sedan       12647.333333
              wagon       9095.750000
fwd           convertible  11595.000000
              hardtop     8249.000000
              hatchback   8396.387755
              sedan       9811.800000
              wagon       9997.333333
rwd           convertible  23949.600000
              hardtop     24202.714286
              hatchback   14337.777778
              sedan       21711.833333
              wagon       16994.222222
```

Name: price, dtype: float64

- We can see that there are no convertibles or hardtops in the 4 wheel drive and the average price of sedan cars is the most in a 4 wheel drive system. - We can see that convertibles have a higher overall price average in forward wheel drive, while hatchback and hardtop and sedan and wagon are relatively close in their average pricing. - We can see that rear wheel drive convertibles are the most expensive closely followed by the hardtop cars

Creating Pivot tables: Using pivot() on the result of step 19 to get “drive-wheels” as index and “body-style” as columns.

In [41]:

```
a = cars[['drive-wheels', 'body-style', 'price']]
pivot = a.groupby(['drive-wheels', 'body-style'], as_index=False).mean()
pivot_table = pivot.pivot(index = 'drive-wheels', columns = 'body-style')
pivot_table
```

Out[41]:

	price				
body-style	convertible	hardtop	hatchback	sedan	wagon
drive-wheels					
4wd	NaN	NaN	7603.000000	12647.333333	9095.750000
fwd	11595.0	8249.000000	8396.387755	9811.800000	9997.333333
rwd	23949.6	24202.714286	14337.777778	21711.833333	16994.222222

In []:

In [42]:

```
col = cars.describe(include=np.object).columns
redundant = []
for i in col:
    redundant.append(i)
    for j in col:
        if i==j or j in redundant:
            continue
        else:
            a = cars.groupby([i,j])['price'].mean()
            print(a, end='\n\n')
            a = cars[[i,j,'price']]
            pivot = a.groupby([i,j], as_index=False).mean()
            pivot = pivot.pivot(index = i, columns = j)
            pivot.replace(np.nan, '-', inplace= True)
            print('Pivot Table')
            print(pivot)
            print("")
            print("", end='\n\n')
```

make	aspiration	
alfa-romero	std	15498.333333
audi	std	16656.000000
	turbo	23875.000000
bmw	std	26118.750000
chevrolet	std	6007.000000
dodge	std	6900.000000
	turbo	9826.333333
honda	std	8184.692308
isuzu	std	8916.500000
jaguar	std	34600.000000
mazda	std	10652.882353
mercedes-benz	std	38900.000000
	turbo	28394.000000
mercury	turbo	16503.000000
mitsubishi	std	7314.714286
	turbo	11485.666667
nissan	std	9869.588235
	turbo	19699.000000
peugot	std	14649.000000
	turbo	16189.166667
plymouth	std	7004.600000
	turbo	10360.500000
porsche	std	31400.500000
renault	std	9595.000000
saab	std	13642.500000
	turbo	18385.000000
subaru	std	7954.200000
	turbo	11476.500000
toyota	std	9859.612903
	turbo	10698.000000
volkswagen	std	9759.000000

```

          turbo      11670.000000
volvo      std      16197.500000
          turbo      20302.000000

```

Name: price, dtype: float64

Pivot Table

	price	
	std	turbo
make		
alfa-romero	15498.3	-
audi	16656	23875
bmw	26118.8	-
chevrolet	6007	-
dodge	6900	9826.33
honda	8184.69	-
isuzu	8916.5	-
jaguar	34600	-
mazda	10652.9	-
mercedes-benz	38900	28394
mercury	-	16503
mitsubishi	7314.71	11485.7
nissan	9869.59	19699
peugot	14649	16189.2
plymouth	7004.6	10360.5
porsche	31400.5	-
renault	9595	-
saab	13642.5	18385
subaru	7954.2	11476.5
toyota	9859.61	10698
volkswagen	9759	11670
volvo	16197.5	20302

make	num-of-doors	
alfa-romero	two	15498.333333
audi	four	18381.000000
	two	15250.000000
bmw	four	26047.000000
	two	26238.333333
chevrolet	four	6575.000000
	two	5723.000000
dodge	four	7601.800000
	two	8217.500000
honda	four	9335.000000
	two	7465.750000
isuzu	four	6785.000000
	two	11048.000000
jaguar	four	33900.000000
	two	36000.000000
mazda	four	11436.750000
	two	9956.111111
mercedes-benz	four	32108.800000

	two	36210.666667
mercury	two	16503.000000
mitsubishi	four	8434.000000
	two	9597.888889
nissan	four	9954.555556
	two	10876.777778
peugot	four	15489.090909
plymouth	four	7362.750000
	two	8764.333333
porsche	two	31400.500000
renault	four	9295.000000
	two	9895.000000
saab	four	15433.333333
	two	15013.333333
subaru	four	9191.222222
	two	6591.333333
toyota	four	9359.888889
	two	10562.000000
volkswagen	four	10450.625000
	two	9331.250000
volvo	four	18063.181818

Name: price, dtype: float64

Pivot Table

	price	
num-of-doors	four	two
make		
alfa-romero	-	15498.3
audi	18381	15250
bmw	26047	26238.3
chevrolet	6575	5723
dodge	7601.8	8217.5
honda	9335	7465.75
isuzu	6785	11048
jaguar	33900	36000
mazda	11436.8	9956.11
mercedes-benz	32108.8	36210.7
mercury	-	16503
mitsubishi	8434	9597.89
nissan	9954.56	10876.8
peugot	15489.1	-
plymouth	7362.75	8764.33
porsche	-	31400.5
renault	9295	9895
saab	15433.3	15013.3
subaru	9191.22	6591.33
toyota	9359.89	10562
volkswagen	10450.6	9331.25
volvo	18063.2	-

make body-style

alfa-romero	convertible	14997.500000
	hatchback	16500.000000
audi	sedan	17647.000000
	wagon	18920.000000
bmw	sedan	26118.750000
chevrolet	hatchback	5723.000000
	sedan	6575.000000
dodge	hatchback	7819.800000
	sedan	7619.666667
	wagon	8921.000000
honda	hatchback	7054.428571
	sedan	9945.000000
	wagon	7295.000000
isuzu	hatchback	11048.000000
	sedan	6785.000000
jaguar	sedan	34600.000000
mazda	hatchback	10085.000000
	sedan	11464.142857
mercedes-benz	convertible	35056.000000
	hardtop	36788.000000
	sedan	33074.000000
	wagon	28248.000000
mercury	hatchback	16503.000000
mitsubishi	hatchback	9597.888889
	sedan	8434.000000
nissan	hardtop	8249.000000
	hatchback	14409.000000
	sedan	8604.555556
	wagon	9915.666667
peugot	sedan	15758.571429
	wagon	15017.500000
plymouth	hatchback	8130.500000
	sedan	7150.500000
	wagon	8921.000000
porsche	convertible	37028.000000
	hardtop	33278.000000
	hatchback	22018.000000
renault	hatchback	9895.000000
	wagon	9295.000000
saab	hatchback	15013.333333
	sedan	15433.333333
subaru	hatchback	6591.333333
	sedan	9070.600000
	wagon	9342.000000
toyota	convertible	17669.000000
	hardtop	9762.333333
	hatchback	9616.000000
	sedan	9542.200000
	wagon	9836.000000
volkswagen	convertible	11595.000000
	hatchback	9980.000000
	sedan	9673.888889
	wagon	12290.000000

```

volvo      sedan      18726.875000
           wagon      16293.333333

```

```
Name: price, dtype: float64
```

```
Pivot Table
```

	price				
body-style	convertible	hardtop	hatchback	sedan	wagon
make					
alfa-romero	14997.5	-	16500	-	-
audi	-	-	-	17647	18920
bmw	-	-	-	26118.8	-
chevrolet	-	-	5723	6575	-
dodge	-	-	7819.8	7619.67	8921
honda	-	-	7054.43	9945	7295
isuzu	-	-	11048	6785	-
jaguar	-	-	-	34600	-
mazda	-	-	10085	11464.1	-
mercedes-benz	35056	36788	-	33074	28248
mercury	-	-	16503	-	-
mitsubishi	-	-	9597.89	8434	-
nissan	-	8249	14409	8604.56	9915.67
peugot	-	-	-	15758.6	15017.5
plymouth	-	-	8130.5	7150.5	8921
porsche	37028	33278	22018	-	-
renault	-	-	9895	-	9295
saab	-	-	15013.3	15433.3	-
subaru	-	-	6591.33	9070.6	9342
toyota	17669	9762.33	9616	9542.2	9836
volkswagen	11595	-	9980	9673.89	12290
volvo	-	-	-	18726.9	16293.3

make	drive-wheels	
alfa-romero	rwd	15498.333333
audi	4wd	17450.000000
	fwd	17941.000000
bmw	rwd	26118.750000
chevrolet	fwd	6007.000000
dodge	fwd	7875.444444
honda	fwd	8184.692308
isuzu	rwd	8916.500000
jaguar	rwd	34600.000000
mazda	fwd	8399.545455
	rwd	14784.000000
mercedes-benz	rwd	33647.000000
mercury	rwd	16503.000000
mitsubishi	fwd	9239.769231
nissan	fwd	8812.333333
	rwd	18432.333333
peugot	rwd	15489.090909
plymouth	fwd	7163.333333
	rwd	12764.000000

porsche	rwd	31400.500000
renault	fwd	9595.000000
saab	fwd	15223.333333
subaru	4wd	9560.400000
	fwd	7813.285714
toyota	4wd	8338.000000
	fwd	8253.000000
	rwd	11973.000000
volkswagen	fwd	10077.500000
volvo	rwd	18063.181818

Name: price, dtype: float64

Pivot Table

	price		
drive-wheels	4wd	fwd	rwd
make			
alfa-romero	-	-	15498.3
audi	17450	17941	-
bmw	-	-	26118.8
chevrolet	-	6007	-
dodge	-	7875.44	-
honda	-	8184.69	-
isuzu	-	-	8916.5
jaguar	-	-	34600
mazda	-	8399.55	14784
mercedes-benz	-	-	33647
mercury	-	-	16503
mitsubishi	-	9239.77	-
nissan	-	8812.33	18432.3
peugot	-	-	15489.1
plymouth	-	7163.33	12764
porsche	-	-	31400.5
renault	-	9595	-
saab	-	15223.3	-
subaru	9560.4	7813.29	-
toyota	8338	8253	11973
volkswagen	-	10077.5	-
volvo	-	-	18063.2

make	engine-location	
alfa-romero	front	15498.333333
audi	front	17859.166667
bmw	front	26118.750000
chevrolet	front	6007.000000
dodge	front	7875.444444
honda	front	8184.692308
isuzu	front	8916.500000
jaguar	front	34600.000000
mazda	front	10652.882353
mercedes-benz	front	33647.000000
mercury	front	16503.000000

mitsubishi	front	9239.769231
nissan	front	10415.666667
peugot	front	15489.090909
plymouth	front	7963.428571
porsche	front	22018.000000
	rear	34528.000000
renault	front	9595.000000
saab	front	15223.333333
subaru	front	8541.250000
toyota	front	9885.812500
volkswagen	front	10077.500000
volvo	front	18063.181818

Name: price, dtype: float64

Pivot Table

		price	
engine-location		front	rear
make			
alfa-romero	15498.333333		-
audi	17859.166667		-
bmw	26118.750000		-
chevrolet	6007.000000		-
dodge	7875.444444		-
honda	8184.692308		-
isuzu	8916.500000		-
jaguar	34600.000000		-
mazda	10652.882353		-
mercedes-benz	33647.000000		-
mercury	16503.000000		-
mitsubishi	9239.769231		-
nissan	10415.666667		-
peugot	15489.090909		-
plymouth	7963.428571		-
porsche	22018.000000	34528	
renault	9595.000000		-
saab	15223.333333		-
subaru	8541.250000		-
toyota	9885.812500		-
volkswagen	10077.500000		-
volvo	18063.181818		-

make	engine-type	
alfa-romero	dohc	14997.500000
	ohcv	16500.000000
audi	ohc	17859.166667
bmw	ohc	26118.750000
chevrolet	l	5151.000000
	ohc	6435.000000
dodge	ohc	7875.444444
honda	ohc	8184.692308
isuzu	ohc	8916.500000

jaguar	dohc	33900.000000
	ohcv	36000.000000
mazda	ohc	9924.538462
	rotor	13020.000000
mercedes-benz	ohc	28394.000000
	ohcv	38900.000000
mercury	ohc	16503.000000
mitsubishi	ohc	9239.769231
nissan	ohc	7565.666667
	ohcv	16115.666667
peugot	l	15489.090909
plymouth	ohc	7963.428571
porsche	ohc	22018.000000
	ohcf	34528.000000
renault	ohc	9595.000000
saab	dohc	18385.000000
	ohc	13642.500000
subaru	ohcf	8541.250000
toyota	dohc	13805.333333
	ohc	8981.307692
volkswagen	ohc	10077.500000
volvo	ohc	17721.000000
	ohcv	21485.000000

Name: price, dtype: float64

Pivot Table

	price					
engine-type	dohc	l	ohc	ohcf	ohcv	rotor
make						
alfa-romero	14997.5	-	-	-	16500	-
audi	-	-	17859.2	-	-	-
bmw	-	-	26118.8	-	-	-
chevrolet	-	5151	6435	-	-	-
dodge	-	-	7875.44	-	-	-
honda	-	-	8184.69	-	-	-
isuzu	-	-	8916.5	-	-	-
jaguar	33900	-	-	-	36000	-
mazda	-	-	9924.54	-	-	13020
mercedes-benz	-	-	28394	-	38900	-
mercury	-	-	16503	-	-	-
mitsubishi	-	-	9239.77	-	-	-
nissan	-	-	7565.67	-	16115.7	-
peugot	-	15489.1	-	-	-	-
plymouth	-	-	7963.43	-	-	-
porsche	-	-	22018	34528	-	-
renault	-	-	9595	-	-	-
saab	18385	-	13642.5	-	-	-
subaru	-	-	-	8541.25	-	-
toyota	13805.3	-	8981.31	-	-	-
volkswagen	-	-	10077.5	-	-	-
volvo	-	-	17721	-	21485	-

make	num-of-cylinders	
alfa-romero	four	14997.500000
	six	16500.000000
audi	five	18641.000000
	four	13950.000000
bmw	four	16677.500000
	six	29265.833333
chevrolet	four	6435.000000
	three	5151.000000
dodge	four	7875.444444
honda	four	8184.692308
isuzu	four	8916.500000
jaguar	six	33900.000000
	twelve	36000.000000
mazda	four	9924.538462
	two	13020.000000
mercedes-benz	eight	38900.000000
	five	28394.000000
mercury	four	16503.000000
mitsubishi	four	9239.769231
nissan	four	7565.666667
	six	16115.666667
peugot	four	15489.090909
plymouth	four	7963.428571
porsche	four	22018.000000
	six	34528.000000
renault	four	9595.000000
saab	four	15223.333333
subaru	four	8541.250000
toyota	four	9012.500000
	six	15999.000000
volkswagen	five	13295.000000
	four	9785.000000
volvo	four	17193.333333
	six	21977.500000

Name: price, dtype: float64

Pivot Table

	price						
num-of-cylinders	eight	five	four	six	three	twelve	t
wo							
make							
alfa-romero	-	-	14997.5	16500	-	-	
-							
audi	-	18641	13950	-	-	-	
-							
bmw	-	-	16677.5	29265.8	-	-	
-							
chevrolet	-	-	6435	-	5151	-	
-							
dodge	-	-	7875.44	-	-	-	
-							

honda	-	-	8184.69	-	-	-
-						
isuzu	-	-	8916.5	-	-	-
-						
jaguar	-	-	-	33900	-	36000
-						
mazda	-	-	9924.54	-	-	-
20						130
mercedes-benz	38900	28394	-	-	-	-
-						
mercury	-	-	16503	-	-	-
-						
mitsubishi	-	-	9239.77	-	-	-
-						
nissan	-	-	7565.67	16115.7	-	-
-						
peugot	-	-	15489.1	-	-	-
-						
plymouth	-	-	7963.43	-	-	-
-						
porsche	-	-	22018	34528	-	-
-						
renault	-	-	9595	-	-	-
-						
saab	-	-	15223.3	-	-	-
-						
subaru	-	-	8541.25	-	-	-
-						
toyota	-	-	9012.5	15999	-	-
-						
volkswagen	-	13295	9785	-	-	-
-						
volvo	-	-	17193.3	21977.5	-	-
-						

make	fuel-system	
alfa-romero	mpfi	15498.333333
audi	mpfi	17859.166667
bmw	mpfi	26118.750000
chevrolet	2bbl	6007.000000
dodge	2bbl	6900.000000
	mfi	12964.000000
	mpfi	8257.500000
honda	1bbl	7555.545455
	2bbl	10345.000000
	mpfi	12945.000000
isuzu	2bbl	6785.000000
	spfi	11048.000000
jaguar	mpfi	34600.000000
mazda	2bbl	8160.000000
	4bbl	12145.000000

	idi	14569.500000
	mpfi	16962.500000
mercedes-benz	idi	28394.000000
	mpfi	38900.000000
mercury	mpfi	16503.000000
mitsubishi	2bbl	6987.333333
	spdi	11170.428571
nissan	2bbl	7608.090909
	idi	7099.000000
	mpfi	16115.666667
peugot	idi	15797.000000
	mpfi	15232.500000
plymouth	2bbl	7004.600000
	spdi	10360.500000
porsche	mpfi	31400.500000
renault	mpfi	9595.000000
saab	mpfi	15223.333333
subaru	2bbl	7423.000000
	mpfi	10777.750000
toyota	2bbl	7504.153846
	idi	8794.666667
	mpfi	12025.500000
volkswagen	idi	9777.500000
	mpfi	10227.500000
volvo	idi	22470.000000
	mpfi	17622.500000

Name: price, dtype: float64

Pivot Table

	price					
\	1bbl	2bbl	4bbl	idi	mfi	mpfi
fuel-system						
spdi						
make						
alfa-romero	-	-	-	-	-	15498.3
-						
audi	-	-	-	-	-	17859.2
-						
bmw	-	-	-	-	-	26118.8
-						
chevrolet	-	6007	-	-	-	-
-						
dodge	-	6900	-	-	12964	8257.5
-						
honda	7555.55	10345	-	-	-	12945
-						
isuzu	-	6785	-	-	-	-
-						
jaguar	-	-	-	-	-	34600
-						
mazda	-	8160	12145	14569.5	-	16962.5
-						
mercedes-benz	-	-	-	28394	-	38900

-					
mercury	-	-	-	-	16503
-					
mitsubishi	-	6987.33	-	-	- 1
1170.4					
nissan	-	7608.09	-	7099	- 16115.7
-					
peugot	-	-	-	15797	- 15232.5
-					
plymouth	-	7004.6	-	-	- 1
0360.5					
porsche	-	-	-	-	- 31400.5
-					
renault	-	-	-	-	- 9595
-					
saab	-	-	-	-	- 15223.3
-					
subaru	-	7423	-	-	- 10777.8
-					
toyota	-	7504.15	-	8794.67	- 12025.5
-					
volkswagen	-	-	-	9777.5	- 10227.5
-					
volvo	-	-	-	22470	- 17622.5
-					

fuel-system	spfi
make	
alfa-romero	-
audi	-
bmw	-
chevrolet	-
dodge	-
honda	-
isuzu	11048
jaguar	-
mazda	-
mercedes-benz	-
mercury	-
mitsubishi	-
nissan	-
peugot	-
plymouth	-
porsche	-
renault	-
saab	-
subaru	-
toyota	-
volkswagen	-
volvo	-

make	horsepower-binned	
alfa-romero	Medium	15498.333333
audi	Medium	17859.166667
bmw	High	36318.333333
	Low	16677.500000
	Medium	22213.333333
chevrolet	Low	6007.000000
dodge	Low	6900.000000
	Medium	9826.333333
honda	Low	8184.692308
isuzu	Low	8916.500000
jaguar	High	33900.000000
mazda	Low	9811.600000
	Medium	16962.500000
	High	43180.000000
mercedes-benz	Medium	30469.333333
	High	16503.000000
mercury	High	16503.000000
mitsubishi	Low	6987.333333
	Medium	11170.428571
	High	18432.333333
nissan	Low	7565.666667
	Medium	13799.000000
	High	15223.000000
peugot	Low	15223.000000
	Medium	18150.000000
plymouth	Low	7004.600000
	Medium	10360.500000
porsche	High	34528.000000
	Medium	22018.000000
renault	Medium	9595.000000
saab	High	18385.000000
	Medium	13642.500000
subaru	Low	7954.200000
	Medium	11476.500000
toyota	High	15999.000000
	Low	8251.000000
	Medium	10916.250000
volkswagen	Low	9785.000000
	Medium	13295.000000
volvo	High	18805.000000
	Medium	17785.000000

Name: price, dtype: float64

Pivot Table

horsepower-binned	price		
	High	Low	Medium
make			
alfa-romero	-	-	15498.3
audi	-	-	17859.2
bmw	36318.3	16677.5	22213.3
chevrolet	-	6007	-
dodge	-	6900	9826.33
honda	-	8184.69	-

isuzu	-	8916.5	-
jaguar	33900	-	-
mazda	-	9811.6	16962.5
mercedes-benz	43180	-	30469.3
mercury	16503	-	-
mitsubishi	-	6987.33	11170.4
nissan	18432.3	7565.67	13799
peugot	-	15223	18150
plymouth	-	7004.6	10360.5
porsche	34528	-	22018
renault	-	-	9595
saab	18385	-	13642.5
subaru	-	7954.2	11476.5
toyota	15999	8251	10916.2
volkswagen	-	9785	13295
volvo	18805	-	17785

aspiration	num-of-doors	
std	four	12509.847826
	two	12582.931507
turbo	four	17450.782609
	two	14138.846154

Name: price, dtype: float64

Pivot Table

		price	
num-of-doors		four	two
aspiration			
std		12509.847826	12582.931507
turbo		17450.782609	14138.846154

aspiration	body-style	
std	convertible	21890.500000
	hardtop	21356.000000
	hatchback	9312.089286
	sedan	13785.210526
	wagon	10973.600000
turbo	hardtop	28176.000000
	hatchback	12969.083333
	sedan	17307.833333
	wagon	17965.400000

Name: price, dtype: float64

Pivot Table

		price			
body-style	convertible	hardtop	hatchback	sedan	wag
on					
aspiration					
std	21890.5	21356.0	9312.089286	13785.210526	10973

```
.6
turbo          - 28176.0 12969.083333 17307.833333 17965
.4
```

```
aspiration drive-wheels
std         4wd         9829.166667
           fwd         8724.029412
           rwd         19660.245614
turbo       4wd         11476.500000
           fwd         12564.562500
           rwd         20065.944444
```

Name: price, dtype: float64

Pivot Table

	price		
drive-wheels	4wd	fwd	rwd
aspiration			
std	9829.166667	8724.029412	19660.245614
turbo	11476.500000	12564.562500	20065.944444

```
aspiration engine-location
std         front         12135.037037
           rear         34528.000000
turbo       front         16254.805556
```

Name: price, dtype: float64

Pivot Table

	price	
engine-location	front	rear
aspiration		
std	12135.037037	34528
turbo	16254.805556	-

```
aspiration engine-type
std         dohc         18062.700000
           l             13066.000000
           ohc          10572.091667
           ohcf         14086.615385
           ohcv         25548.333333
           rotor        13020.000000
turbo       dohc         18385.000000
           l             16189.166667
           ohc          16344.640000
           ohcf         11476.500000
           ohcv         19699.000000
```

Name: price, dtype: float64

Pivot Table

		price				
\	engine-type	dohc	1	ohc	ohcf	
aspiration	std	18062.7	13066.000000	10572.091667	14086.615385	25
	548.333333					
	turbo	18385.0	16189.166667	16344.640000	11476.500000	19
	699.000000					

engine-type	rotor
aspiration	
std	13020
turbo	-

aspiration	num-of-cylinders	
std	eight	38900.000000
	five	16525.000000
	four	9469.132812
	six	23907.045455
	three	5151.000000
	twelve	36000.000000
	two	13020.000000
turbo	five	27490.200000
	four	13984.586207
	six	21084.500000

Name: price, dtype: float64

Pivot Table

		price				
\	num-of-cylinders	eight	five	four	six	three
aspiration	std	38900	16525.0	9469.132812	23907.045455	5151
	36000					
	turbo	-	27490.2	13984.586207	21084.500000	-
	-					

num-of-cylinders	two
aspiration	
std	13020
turbo	-

aspiration	fuel-system
std	1bbl 7555.545455

	2bbl	7433.203125
	4bbl	12145.000000
	idi	9670.571429
	mpfi	17771.384615
	spdi	9279.000000
	spfi	11048.000000
turbo	idi	19159.153846
	mfi	12964.000000
	mpfi	16678.928571
	spdi	11204.375000

Name: price, dtype: float64

Pivot Table

	price					
\	1bbl	2bbl	4bbl	idi	mfi	
fuel-system						
mpfi						
aspiration						
std	7555.55	7433.2	12145	9670.571429	-	17771.38
4615						
turbo	-	-	-	19159.153846	12964	16678.92
8571						

fuel-system	spdi	spfi
aspiration		
std	9279.000	11048
turbo	11204.375	-

aspiration	horsepower-binned	
std	High	29143.312500
	Low	8629.186916
	Medium	15703.512195
turbo	High	18483.857143
	Low	14127.875000
	Medium	16322.047619

Name: price, dtype: float64

Pivot Table

	price		
horsepower-binned	High	Low	Medium
aspiration			
std	29143.312500	8629.186916	15703.512195
turbo	18483.857143	14127.875000	16322.047619

num-of-doors	body-style	
four	hatchback	8372.000000
	sedan	14490.687500
	wagon	12371.960000

two	convertible	21890.500000
	hardtop	22208.500000
	hatchback	10230.793103
	sedan	14283.000000

Name: price, dtype: float64

Pivot Table

	price				
body-style	convertible	hardtop	hatchback	sedan	wagon
num-of-doors					
four	-	-	8372.000000	14490.6875	12372
two	21890.5	22208.5	10230.793103	14283.0000	-

num-of-doors	drive-wheels	
four	4wd	10617.857143
	fwd	9789.072464
	rwd	20577.000000
two	4wd	7603.000000
	fwd	8478.326531
	rwd	18869.944444

Name: price, dtype: float64

Pivot Table

	price		
drive-wheels	4wd	fwd	rwd
num-of-doors			
four	10617.857143	9789.072464	20577.000000
two	7603.000000	8478.326531	18869.944444

num-of-doors	engine-location	
four	front	13498.034783
two	front	12033.433735
	rear	34528.000000

Name: price, dtype: float64

Pivot Table

	price	
engine-location	front	rear
num-of-doors		
four	13498.034783	-
two	12033.433735	34528

num-of-doors	engine-type	
four	dohc	23572.000000
	l	15489.090909
	ohc	12420.083333
	ohcf	9191.222222

	ohcv	23004.333333
two	dohc	14219.571429
	l	5151.000000
	ohc	10393.114754
	ohcf	20559.666667
	ohcv	26893.285714
	rotor	13020.000000

Name: price, dtype: float64

Pivot Table

	price			
\				
engine-type	dohc	l	ohc	oh
cf				
num-of-doors				
four	23572.000000	15489.090909	12420.083333	9191.2222
22				
two	14219.571429	5151.000000	10393.114754	20559.6666
67				

engine-type	ohcv	rotor
num-of-doors		
four	23004.333333	-
two	26893.285714	13020

num-of-doors	num-of-cylinders	
four	eight	37572.000000
	five	22081.250000
	four	10897.586957
	six	22915.538462
two	eight	40228.000000
	five	21713.000000
	four	9461.907692
	six	24565.636364
	three	5151.000000
	twelve	36000.000000
	two	13020.000000

Name: price, dtype: float64

Pivot Table

	price			
\				
num-of-cylinders	eight	five	four	six th
ree twelve				
num-of-doors				
four	37572.0	22081.25	10897.586957	22915.538462
-				
two	40228.0	21713.00	9461.907692	24565.636364
151	36000			5

```

num-of-cylinders    two
num-of-doors
four                -
two                13020

```

```

num-of-doors  fuel-system
four          1bbl          8432.500000
              2bbl          7807.864865
              idi           16100.764706
              mpfi          17043.309091
              spdi          9279.000000
two           1bbl          7054.428571
              2bbl          6919.777778
              4bbl          12145.000000
              idi           14350.000000
              mfi           12964.000000
              mpfi          18440.297297
              spdi          11479.428571
              spfi          11048.000000

```

Name: price, dtype: float64

Pivot Table

```

\
fuel-system      price
num-of-doors
four            1bbl      2bbl      4bbl      idi      mfi
two            7054.428571 6919.777778 12145    14350.000000 12964

```

```

fuel-system      mpfi      spdi      spfi
num-of-doors
four            17043.309091 9279.000000 -
two            18440.297297 11479.428571 11048

```

```

num-of-doors  horsepower-binned
four          High          25715.909091
              Low           9743.577465
              Medium        17503.181818
two           High          26067.083333
              Low           7830.727273
              Medium        14103.517241

```

Name: price, dtype: float64

Pivot Table

```

horsepower-binned  price
num-of-doors      High      Low      Medium

```

four	25715.909091	9743.577465	17503.181818
two	26067.083333	7830.727273	14103.517241

body-style	drive-wheels	
convertible	fwd	11595.000000
	rwd	23949.600000
hardtop	fwd	8249.000000
	rwd	24202.714286
hatchback	4wd	7603.000000
	fwd	8396.387755
	rwd	14337.777778
sedan	4wd	12647.333333
	fwd	9811.800000
	rwd	21711.833333
wagon	4wd	9095.750000
	fwd	9997.333333
	rwd	16994.222222

Name: price, dtype: float64

Pivot Table

	price		
drive-wheels	4wd	fwd	rwd
body-style			
convertible	- 11595.000000	23949.600000	
hardtop	- 8249.000000	24202.714286	
hatchback	7603 8396.387755	14337.777778	
sedan	12647.3 9811.800000	21711.833333	
wagon	9095.75 9997.333333	16994.222222	

body-style	engine-location	
convertible	front	18863.000000
	rear	37028.000000
hardtop	front	18518.666667
	rear	33278.000000
hatchback	front	9957.441176
sedan	front	14459.755319
wagon	front	12371.960000

Name: price, dtype: float64

Pivot Table

	price	
engine-location	front	rear
body-style		
convertible	18863.000000	37028
hardtop	18518.666667	33278
hatchback	9957.441176	-
sedan	14459.755319	-
wagon	12371.960000	-

body-style	engine-type	
convertible	dohc	14997.500000
	ohc	14632.000000
	ohcf	37028.000000
	ohcv	35056.000000
hardtop	ohc	13142.400000
	ohcf	33278.000000
	ohcv	45400.000000
hatchback	dohc	15061.000000
	l	5151.000000
	ohc	9001.153846
	ohcf	6591.333333
	ohcv	17949.250000
	rotor	13020.000000
sedan	dohc	22281.600000
	l	15758.571429
	ohc	13134.070423
	ohcf	9070.600000
	ohcv	26604.500000
wagon	dohc	15750.000000
	l	15017.500000
	ohc	12114.133333
	ohcf	9342.000000
	ohcv	14399.000000

Name: price, dtype: float64

Pivot Table

	price				
engine-type	dohc	l	ohc	ohcf	ohc
v rotor					
body-style					
convertible	14997.5	-	14632.000000	37028.000000	35056.0
0 -					
hardtop	-	-	13142.400000	33278.000000	45400.0
0 -					
hatchback	15061	5151	9001.153846	6591.333333	17949.2
5 13020					
sedan	22281.6	15758.6	13134.070423	9070.600000	26604.5
0 -					
wagon	15750	15017.5	12114.133333	9342.000000	14399.0
0 -					

body-style	num-of-cylinders	
convertible	eight	35056.000000
	four	14814.750000
	six	37028.000000
hardtop	eight	45400.000000
	five	28176.000000
	four	9384.000000

	six	33278.000000
hatchback	four	9044.245614
	six	17392.166667
	three	5151.000000
sedan	two	13020.000000
	eight	37572.000000
	five	20676.000000
	four	10891.591549
	six	25387.538462
wagon	twelve	36000.000000
	five	23584.000000
	four	11046.761905
	six	15074.500000

Name: price, dtype: float64

Pivot Table

	price				
num-of-cylinders	eight	five	four	six	three
twelve					
two					
body-style					
convertible	35056	-	14814.750000	37028.000000	-
-					
hardtop	45400	28176	9384.000000	33278.000000	-
-					
hatchback	-	-	9044.245614	17392.166667	5151
-					
13020					
sedan	37572	20676	10891.591549	25387.538462	-
36000	-				
wagon	-	23584	11046.761905	15074.500000	-
-					
-					

body-style	fuel-system	
convertible	mpfi	21890.500000
hardtop	2bbl	8249.000000
	idi	28176.000000
	mpfi	23540.500000
hatchback	1bbl	7054.428571
	2bbl	6979.678571
	4bbl	12145.000000
	idi	7788.000000
	mfi	12964.000000
sedan	mpfi	14185.150000
	spdi	11479.428571
	spfi	11048.000000
	1bbl	8811.666667
	2bbl	7684.038462
	idi	14774.400000
	mpfi	18600.479167
wagon	spdi	9279.000000
	1bbl	7295.000000
	2bbl	8028.888889

idi 19727.666667

mpfi 14213.416667

Name: price, dtype: float64

Pivot Table

	price					
\	1bbl	2bbl	4bbl	idi	mfi	mpfi
fuel-system						
spdi						
body-style						
convertible	-	-	-	-	-	21890.500000
-						
hardtop	-	8249	-	28176	-	23540.500000
-						
hatchback	7054.43	6979.68	12145	7788	12964	14185.150000
11479.4						
sedan	8811.67	7684.04	-	14774.4	-	18600.479167
9279						
wagon	7295	8028.89	-	19727.7	-	14213.416667
-						

fuel-system	spfi
body-style	
convertible	-
hardtop	-
hatchback	11048
sedan	-
wagon	-

body-style	horsepower-binned	
convertible	High	37028.000000
	Low	11595.000000
	Medium	20680.000000
hardtop	High	37318.666667
	Low	8249.000000
	Medium	14365.750000
hatchback	High	17500.857143
	Low	7704.627907
	Medium	12405.611111
sedan	High	28949.000000
	Low	9686.851852
	Medium	17608.172414
wagon	High	17350.000000
	Low	10132.062500
	Medium	16069.428571

Name: price, dtype: float64

Pivot Table

	price		
horsepower-binned	High	Low	Medium

body-style			
convertible	37028.000000	11595.000000	20680.000000
hardtop	37318.666667	8249.000000	14365.750000
hatchback	17500.857143	7704.627907	12405.611111
sedan	28949.000000	9686.851852	17608.172414
wagon	17350.000000	10132.062500	16069.428571

drive-wheels	engine-location	
4wd	front	10241.000000
fwd	front	9244.779661
rwd	front	19142.180556
	rear	34528.000000

Name: price, dtype: float64

Pivot Table

		price	
engine-location		front	rear
drive-wheels			
4wd		10241.000000	-
fwd		9244.779661	-
rwd		19142.180556	34528

drive-wheels	engine-type	
4wd	ohc	11375.333333
	ohcf	9560.400000
fwd	dohc	18385.000000
	l	5151.000000
	ohc	9074.980952
	ohcf	7813.285714
	ohcv	13799.000000
rwd	dohc	18062.700000
	l	15489.090909
	ohc	18655.891892
	ohcf	34528.000000
	ohcv	28488.200000
	rotor	13020.000000

Name: price, dtype: float64

Pivot Table

		price				
engine-type		dohc	l	ohc	ohcf	ohc
v rotor						
drive-wheels						
4wd		-	-	11375.333333	9560.400000	
-	-					
fwd		18385	5151	9074.980952	7813.285714	1379
9	-					
rwd		18062.7	15489.1	18655.891892	34528.000000	28488.
2	13020					

drive-wheels	num-of-cylinders	
4wd	five	17450.000000
	four	9211.142857
fwd	five	17810.000000
	four	8764.091743
	six	13799.000000
rwd	three	5151.000000
	eight	38900.000000
	five	28394.000000
	four	14581.414634
	six	25082.238095
	twelve	36000.000000
	two	13020.000000

Name: price, dtype: float64

Pivot Table

		price					
num-of-cylinders		eight	five	four	six	three	twel
ve	two						
drive-wheels							
4wd		-	17450.0	9211.142857	-	-	
-	-						
fwd		-	17810.0	8764.091743	13799	5151	
-	-						
rwd		38900	28394.0	14581.414634	25082.2	-	360
00	13020						

drive-wheels	fuel-system	
4wd	2bbl	8305.000000
	mpfi	13467.666667
fwd	1bbl	7555.545455
	2bbl	7341.410714
	idi	9265.333333
	mfi	12964.000000
	mpfi	12550.060606
rwd	spdi	10768.750000
	2bbl	7693.666667
	4bbl	12145.000000
	idi	21215.909091
	mpfi	20805.678571
	spdi	12764.000000
	spfi	11048.000000

Name: price, dtype: float64

Pivot Table

		price				
\	fuel-system	1bbl	2bbl	4bbl	idi	mfi

```

mpfi
drive-wheels
4wd          -  8305.000000      -      -      -  13467.6
66667
fwd          7555.55  7341.410714      -  9265.33  12964  12550.0
60606
rwd          -  7693.666667  12145  21215.9      -  20805.6
78571

```

```

fuel-system   spdi   spfi
drive-wheels
4wd           -      -
fwd           10768.8  -
rwd           12764   11048

```

```

drive-wheels  horsepower-binned
4wd           Low           8305.000000
              Medium        13467.666667
fwd           High          18385.000000
              Low           8003.644444
              Medium        12837.923077
rwd           High          26614.761905
              Low           13724.650000
              Medium        18558.121212

```

Name: price, dtype: float64

Pivot Table

```

             price
horsepower-binned  High      Low      Medium
drive-wheels
4wd               -  8305.000000  13467.666667
fwd              18385  8003.644444  12837.923077
rwd              26614.8 13724.650000  18558.121212

```

```

engine-location  engine-type
front            dohc        18116.416667
                l           14627.583333
                ohc         11567.358621
                ohcf        8541.250000
                ohcv        25098.384615
                rotor       13020.000000
rear             ohcf       34528.000000

```

Name: price, dtype: float64

Pivot Table

```

             price
engine-type     dohc      l      ohc      ohcf      ohcv      rot
or

```

```

engine-location
front      18116.4  14627.6  11567.4   8541.25  25098.4  130
20
rear      -      -      -   34528.00      -
-

```

```

engine-location  num-of-cylinders
front           eight      38900.000000
                five      22007.600000
                four      10303.197452
                six      22120.952381
                three      5151.000000
                twelve    36000.000000
                two      13020.000000
rear           six      34528.000000

```

Name: price, dtype: float64

Pivot Table

```

                price
num-of-cylinders  eight      five      four      six three twel
ve      two
engine-location
front      38900  22007.6  10303.2  22120.952381  5151  360
00  13020
rear      -      -      -   34528.000000      -
-      -

```

```

engine-location  fuel-system
front           1bbl      7555.545455
                2bbl      7433.203125
                4bbl      12145.000000
                idi      15838.150000
                mfi      12964.000000
                mpfi     17034.707865
                spdi     10990.444444
                spfi     11048.000000
rear           mpfi     34528.000000

```

Name: price, dtype: float64

Pivot Table

```

                price
\
fuel-system      1bbl      2bbl      4bbl      idi      mfi      m
pfi
engine-location
front      7555.55  7433.2  12145  15838.1  12964  17034.707
865
rear      -      -      -      -      -      34528.000
000

```

fuel-system	spdi	spfi
engine-location		
front	10990.4	11048
rear	-	-

engine-location	horsepower-binned	
front	High	24604.800000
	Low	9011.704348
	Medium	15913.016129
rear	High	34528.000000

Name: price, dtype: float64

Pivot Table

	price		
horsepower-binned	High	Low	Medium
engine-location			
front	24604.8	9011.7	15913
rear	34528.0	-	-

engine-type	num-of-cylinders	
dohc	four	14266.833333
	six	21966.000000
l	four	15489.090909
	three	5151.000000
ohc	five	22007.600000
	four	9836.921875
	six	28295.000000
ohcf	four	8541.250000
	six	34528.000000
ohcv	eight	38900.000000
	six	16834.875000
	twelve	36000.000000
rotor	two	13020.000000

Name: price, dtype: float64

Pivot Table

	price						
num-of-cylinders	eight	five	four	six	three	twelve	two
engine-type							
dohc	-	-	14266.8	21966	-	-	-
-							
l	-	-	15489.1	-	5151	-	-
-							
ohc	-	22007.6	9836.92	28295	-	-	-
-							
ohcf	-	-	8541.25	34528	-	-	-

```

-
ohcv          38900          -          -    16834.9          -    36000
-
rotor          -          -          -          -          -          -    1
3020

```

```

engine-type  fuel-system
dohc         mpfi          18116.416667
1            2bbl          5151.000000
            idi          15797.000000
            mpfi          15232.500000
ohc          1bbl          7555.545455
            2bbl          7476.181818
            idi          15851.866667
            mfi          12964.000000
            mpfi          15514.377358
            spdi          10990.444444
            spfi          11048.000000
ohcf         2bbl          7423.000000
            mpfi          20956.428571
ohcv         mpfi          25098.384615
rotor        4bbl          12145.000000
            mpfi          15645.000000

```

Name: price, dtype: float64

Pivot Table

```

           price
\
fuel-system  1bbl    2bbl    4bbl      idi    mfi      mpfi
spdi
engine-type
dohc         -        -        -        -        -    18116.416667
-
1            -      5151        -    15797        -    15232.500000
-
ohc          7555.55  7476.18        -    15851.9  12964  15514.377358
10990.4
ohcf         -      7423        -        -        -    20956.428571
-
ohcv         -        -        -        -        -    25098.384615
-
rotor        -        -    12145        -        -    15645.000000
-

```

```

fuel-system  spfi
engine-type
dohc         -
1            -
ohc          11048
ohcf         -

```

```
ohcv      -
rotor     -
```

```
engine-type  horsepower-binned
dohc         High          21070.750000
             Medium        12207.750000
1            Low          14307.363636
             Medium        18150.000000
ohc          High          25981.857143
             Low           8384.483516
             Medium        15583.106383
ohcf         High          34528.000000
             Low           7954.200000
             Medium        11476.500000
ohcv         High          28331.400000
             Medium        21231.714286
rotor        Low          12145.000000
             Medium        15645.000000
```

```
Name: price, dtype: float64
```

```
Pivot Table
```

horsepower-binned	price		
	High	Low	Medium
engine-type			
dohc	21070.8	-	12207.750000
1	-	14307.4	18150.000000
ohc	25981.9	8384.48	15583.106383
ohcf	34528	7954.2	11476.500000
ohcv	28331.4	-	21231.714286
rotor	-	12145	15645.000000

```
num-of-cylinders  fuel-system
eight             mpfi          38900.000000
five              idi           28394.000000
                  mpfi          17750.000000
four              1bbl          7555.545455
                  2bbl          7469.428571
                  idi           12047.800000
                  mfi           12964.000000
                  mpfi          13338.140351
                  spdi          10990.444444
                  spfi          11048.000000
six               idi           22470.000000
                  mpfi          23724.086957
three             2bbl          5151.000000
twelve            mpfi          36000.000000
two               4bbl          12145.000000
                  mpfi          15645.000000
```

```
Name: price, dtype: float64
```

Pivot Table

		price					
\		1bbl	2bbl	4bbl	idi	mfi	mpfi
fuel-system	spdi						
num-of-cylinders							
eight		-	-	-	-	-	38900
-							
five		-	-	-	28394	-	17750
-							
four		7555.55	7469.43	-	12047.8	12964	13338.1
10990.4							
six		-	-	-	22470	-	23724.1
-							
three		-	5151	-	-	-	-
-							
twelve		-	-	-	-	-	36000
-							
two		-	-	12145	-	-	15645
-							

fuel-system	spfi
num-of-cylinders	
eight	-
five	-
four	11048
six	-
three	-
twelve	-
two	-

num-of-cylinders	horsepower-binned	
eight	High	43180.000000
	Medium	34620.000000
five	Medium	22007.600000
four	High	18281.333333
	Low	8961.801802
	Medium	12828.850000
six	High	26642.133333
	Medium	18721.333333
three	Low	5151.000000
two	Low	12145.000000
	Medium	15645.000000

Name: price, dtype: float64

Pivot Table

		price		
horsepower-binned		High	Low	Medium
num-of-cylinders				

eight	43180	-	34620
five	-	-	22007.6
four	18281.3	8961.8	12828.9
six	26642.1	-	18721.3
three	-	5151	-
two	-	12145	15645

fuel-system	horsepower-binned	
1bbl	Low	7555.545455
2bbl	Low	7433.203125
4bbl	Low	12145.000000
idi	Low	12047.800000
	Medium	27209.200000
mfi	Medium	12964.000000
mpfi	High	25899.130435
	Low	11871.904762
	Medium	15716.659574
spdi	Medium	10990.444444
spfi	Low	11048.000000

Name: price, dtype: float64

Pivot Table

	price		
horsepower-binned	High	Low	Medium
fuel-system			
1bbl	-	7555.55	-
2bbl	-	7433.2	-
4bbl	-	12145	-
idi	-	12047.8	27209.2
mfi	-	-	12964
mpfi	25899.1	11871.9	15716.7
spdi	-	-	10990.4
spfi	-	11048	-

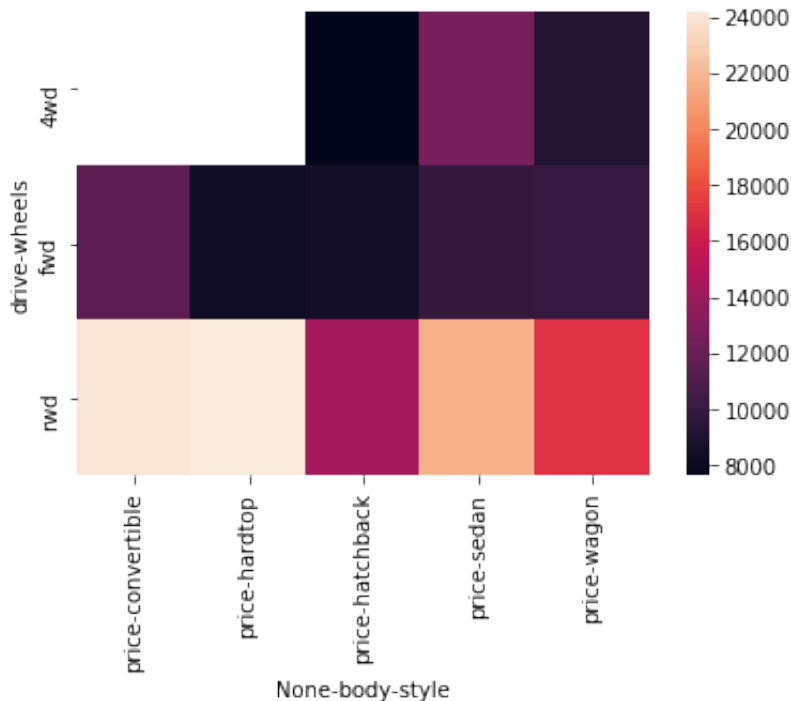
Observations :- We can see that there are no cars with medium or low horsepower which have their engine in rear. These are majorly the sports cars of segment. We also see that of all the segments, the cars having engine in the front and with low horsepower have the lowest average price of any car.

In [43]:

```
sns.heatmap(pivot_table)
```

Out[43]:

```
<matplotlib.axes._subplots.AxesSubplot at 0x7fed5d10bb10>
```



The heatmap plots the target variable (price) proportional to colour with respect to the variables 'drive-wheel' and 'body-style' in the vertical and horizontal axis respectively. This allows us to visualize how the price is related to 'drive-wheel' and 'body-style'. Using Pearson coefficient The Pearson Correlation measures the linear dependence between two variables X and Y. The resulting coefficient is a value between -1 and 1 inclusive, where: 1: Total positive linear correlation. 0: No linear correlation, the two variables most likely do not affect each other. -1: Total negative linear correlation. Wheel-base vs Price

In [45]:

```
pearson_coeff, p_val = ss.pearsonr(cars['wheel-base'], cars['price'])
print(pearson_coeff, p_val)
```

```
0.584641822265508 8.076488270733218e-20
```

Since the p-value is < 0.001 , the correlation between wheel-base and price is statistically significant, although the linear relationship isn't extremely strong (~ 0.585). Since ANOVA analyzes the difference between different groups of the same variable, the groupby function will come in handy. Because the ANOVA algorithm averages the data automatically, we do not need to take the average before hand. Let's see if different types 'drive-wheels' impact 'price', we group the data.

In [47]:

```
a = cars[['drive-wheels', 'body-style', 'price']]
a = a[['drive-wheels', 'price']].groupby(['drive-wheels'])
a
f_val, p_val = stats.f_oneway(a.get_group('fwd')['price'],
                              a.get_group('4wd')['price'],
                              a.get_group('rwd')['price'])
print(f_val, p_val)
```

67.95406500780398 3.3945443577151245e-23

This is a great result, with a large F test score showing a strong correlation and a P value of almost 0 implying almost certain statistical significance. Separately: ANOVA RESULTS

In [51]:

```
print("fwd and rwd")
f_val, p_val = stats.f_oneway(a.get_group('fwd')['price'], a.get_group('rwd')['price'])

print("ANOVA results: F=", f_val, ", P =", p_val)

print("4wd and rwd")
f_val, p_val = stats.f_oneway(a.get_group('4wd')['price'], a.get_group('rwd')['price'])

print("ANOVA results: F=", f_val, ", P =", p_val)

print("4wd and fwd")
f_val, p_val = stats.f_oneway(a.get_group('4wd')['price'], a.get_group('fwd')['price'])

print("ANOVA results: F=", f_val, ", P =", p_val)
```

fwd and rwd

ANOVA results: F= 130.5533160959111 , P = 2.2355306355677845e-23

4wd and rwd

ANOVA results: F= 8.580681368924756 , P = 0.004411492211225333

4wd and fwd

ANOVA results: F= 0.665465750252303 , P = 0.41620116697845666

We now have a better idea of what our data looks like and which variables are important to take into account when predicting the car price. We have narrowed it down to the following variables: Continuous numerical variables: Length Width Curb-weight Engine-size Horsepower City-mpg Highway-mpg Wheel-base Bore Categorical variables: Drive-wheels