assignment_25.1

January 20, 2019

In [1]:
```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn import datasets

# this allows plots to appear directly in the notebook
%matplotlib inline
```

0.1 Load Boston Dataset

In [2]:
```
boston = datasets.load_boston()
bos = pd.DataFrame(boston.data)
```

In [3]:
```
boston.keys()
```

Out[3]:
```
dict_keys(['data', 'target', 'feature_names', 'DESCR'])
```

In [4]:
```
boston.data.shape
```

Out[4]:
```
(506, 13)
```

In [5]:
```
print( boston.feature_names )
```

```
['CRIM' 'ZN' 'INDUS' 'CHAS' 'NOX' 'RM' 'AGE' 'DIS' 'RAD' 'TAX' 'PTRATIO'
 'B' 'LSTAT']
```

In [6]:
```
bos.head()
```

Out[6]:

	0	1	2	3	4	5	6	7	8	9	10	\
0	0.00632	18.0	2.31	0.0	0.538	6.575	65.2	4.0900	1.0	296.0	15.3	
1	0.02731	0.0	7.07	0.0	0.469	6.421	78.9	4.9671	2.0	242.0	17.8	
2	0.02729	0.0	7.07	0.0	0.469	7.185	61.1	4.9671	2.0	242.0	17.8	
3	0.03237	0.0	2.18	0.0	0.458	6.998	45.8	6.0622	3.0	222.0	18.7	
4	0.06905	0.0	2.18	0.0	0.458	7.147	54.2	6.0622	3.0	222.0	18.7	

```
       11    12
0  396.90  4.98
1  396.90  9.14
2  392.83  4.03
3  394.63  2.94
4  396.90  5.33
```

0.2 Add columns as feature names and PRICE from target

```
In [7]: bos.columns = boston.feature_names

In [8]: bos['PRICE']= boston.target
        bos.head()
```

Out[8]:

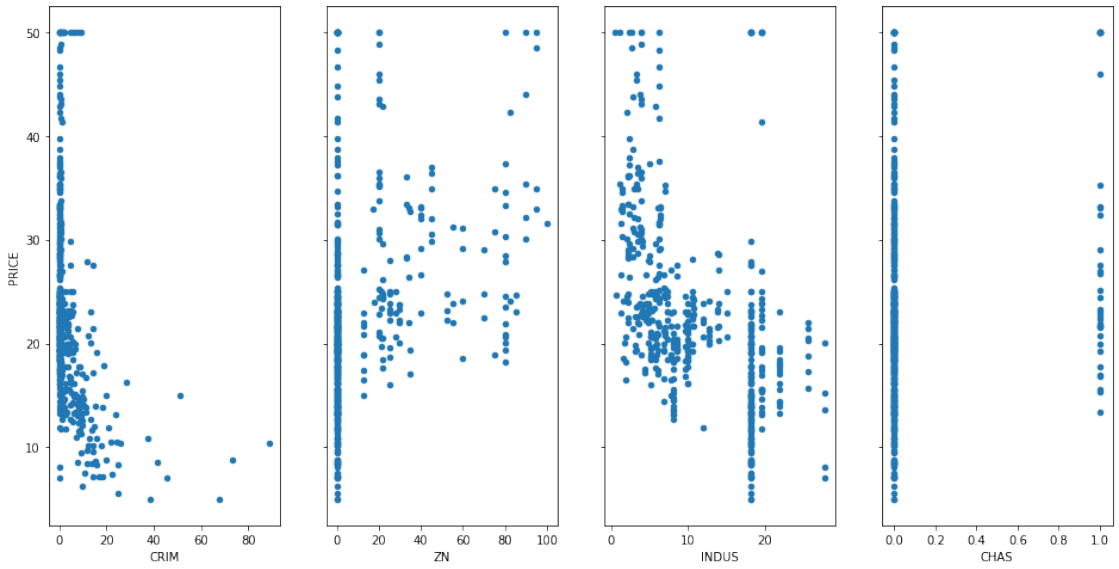
	CRIM	ZN	INDUS	CHAS	NOX	RM	AGE	DIS	RAD	TAX	\
0	0.00632	18.0	2.31	0.0	0.538	6.575	65.2	4.0900	1.0	296.0	
1	0.02731	0.0	7.07	0.0	0.469	6.421	78.9	4.9671	2.0	242.0	
2	0.02729	0.0	7.07	0.0	0.469	7.185	61.1	4.9671	2.0	242.0	
3	0.03237	0.0	2.18	0.0	0.458	6.998	45.8	6.0622	3.0	222.0	
4	0.06905	0.0	2.18	0.0	0.458	7.147	54.2	6.0622	3.0	222.0	

	PTRATIO	B	LSTAT	PRICE
0	15.3	396.90	4.98	24.0
1	17.8	396.90	9.14	21.6
2	17.8	392.83	4.03	34.7
3	18.7	394.63	2.94	33.4
4	18.7	396.90	5.33	36.2

0.3 Display scatter plots betweebn features

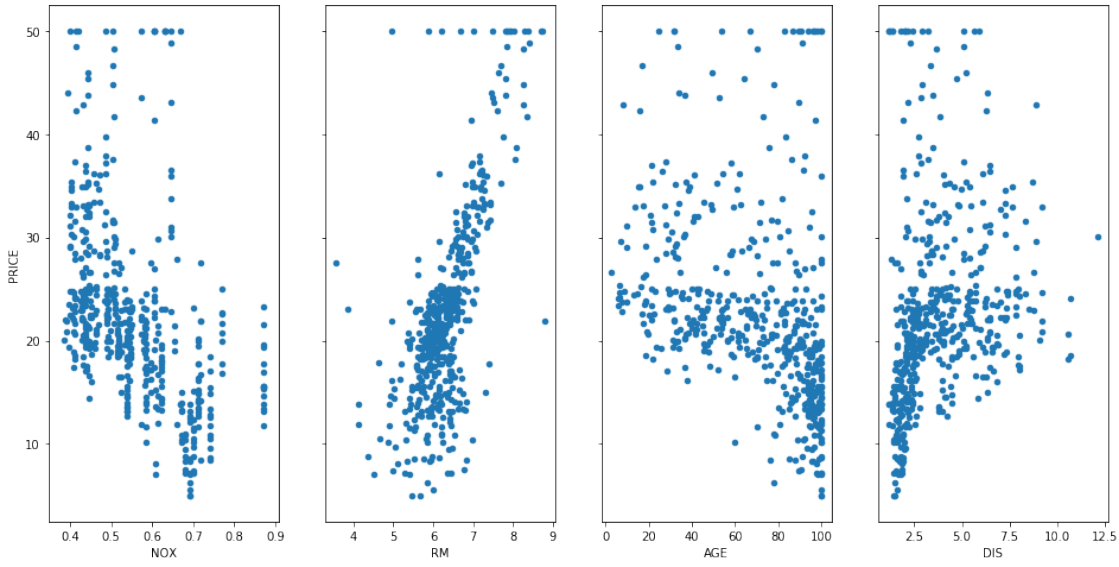
```
In [9]: fig, axs = plt.subplots(1, 4, sharey=True)
        bos.plot(kind='scatter', x='CRIM', y='PRICE', ax=axs[0], figsize=(16, 8))
        bos.plot(kind='scatter', x='ZN', y='PRICE', ax=axs[1])
        bos.plot(kind='scatter', x='INDUS', y='PRICE', ax=axs[2])
        bos.plot(kind='scatter', x='CHAS', y='PRICE', ax=axs[3])

Out[9]: <matplotlib.axes._subplots.AxesSubplot at 0x297bc54d9b0>
```



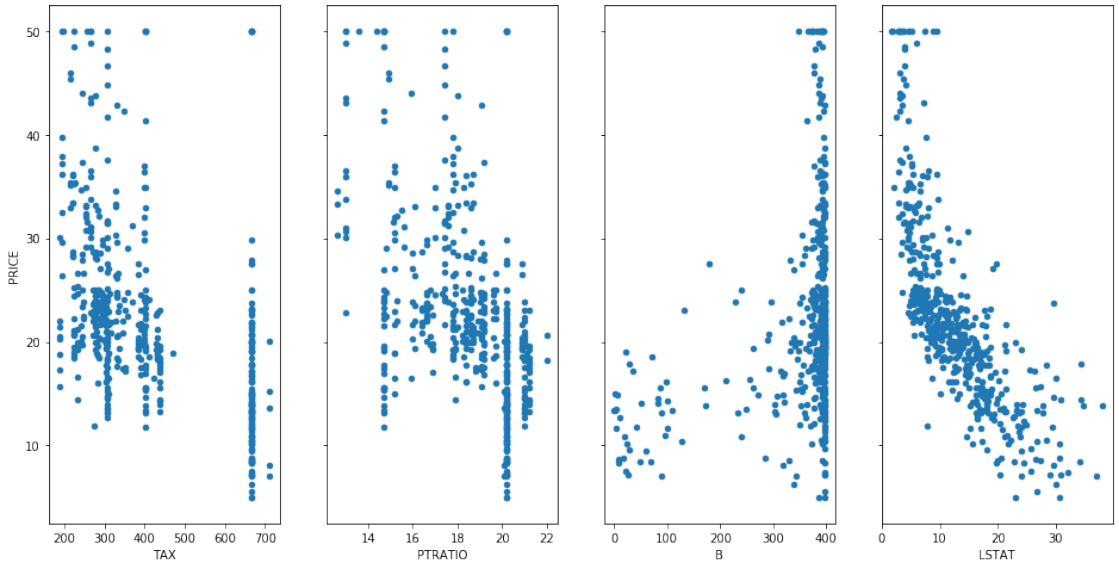

```
In [10]: fig, axs = plt.subplots(1, 4, sharey=True)
         bos.plot(kind='scatter', x='NOX', y='PRICE', ax=axs[0],  figsize=(16, 8))
         bos.plot(kind='scatter', x='RM', y='PRICE', ax=axs[1])
         bos.plot(kind='scatter', x='AGE', y='PRICE', ax=axs[2])
         bos.plot(kind='scatter', x='DIS', y='PRICE', ax=axs[3])
```

```
Out[10]: <matplotlib.axes._subplots.AxesSubplot at 0x297bc6e2860>
```



```
In [11]: fig, axs = plt.subplots(1, 4, sharey=True)
         bos.plot(kind='scatter', x='TAX', y='PRICE', ax=axs[0],  figsize=(16, 8))
         bos.plot(kind='scatter', x='PTRATIO', y='PRICE', ax=axs[1])
         bos.plot(kind='scatter', x='B', y='PRICE', ax=axs[2])
         bos.plot(kind='scatter', x='LSTAT', y='PRICE', ax=axs[3])
```

```
Out[11]: <matplotlib.axes._subplots.AxesSubplot at 0x297bc86ab00>
```



0.4 Split dataset into train and test

```
In [12]: from sklearn.model_selection import train_test_split
         import numpy as np
```

```
In [13]: # create X and y
         feature_cols = [ 'CRIM', 'ZN', 'CHAS', 'NOX', 'RM', 'DIS', 'RAD', 'TAX', 'PTRATIO', '
         X_Train,X_Test,y_Train,y_Test = train_test_split(bos[feature_cols ],bos['PRICE'],test_
```

```
In [14]: ## Use RandomForestRegressor to fit the model and calculate accuracy
```

```
In [15]: from sklearn.ensemble import RandomForestRegressor
         model = RandomForestRegressor()

         # Fitting to the model
         model.fit(X_Train, y_Train)
         y_predict = model.predict(X_Test)
         print("Prediction values: ", y_predict)
```

```
Prediction values:  [24.1  26.03 22.74 13.23 20.49 20.72 21.11 20.23 18.69 19.06 10.03 15.32
 15.58  8.71 47.62 33.95 20.35 34.12 25.54 20.31 23.83 22.77 20.15 23.66
```

```
 20.69 18.49 18.66 16.9  45.74 19.17 14.64 19.32 21.18 21.36 23.06 18.32
  8.25 24.23 14.44 14.54 22.56 21.34 21.85 14.75 23.17 23.08 18.94 17.69
 14.8  24.15 16.51 18.18 19.28 40.12 14.34 19.6  20.93 18.67 20.21 20.6
 21.88 19.55 33.76 28.88 19.05 30.09 17.25 19.01 17.58 20.08 19.91 22.96
 27.27 31.61 26.75  8.86 45.13 21.3  23.13 19.6  26.04 19.28 19.45 42.68
 41.4  24.56 21.95 14.54 24.23 15.88 18.86 13.71 23.28 31.67 21.25 22.24
 14.29 23.26 14.7  19.82 24.07 20.79 28.33 20.42 28.94 20.04  8.73 18.42
 21.01 22.54 35.32 12.09 19.68 18.62 17.   20.73  9.8  18.67  9.94 48.33
 31.31  9.62 18.3  20.05 20.94 18.48 34.04 17.93 21.35 35.15 12.94  9.89
 14.   21.16 13.06 34.21 19.91 15.14 25.28  9.01 10.71 19.74 32.29 23.99
 24.32 17.32 34.61 33.35 11.79  8.33 29.82 24.58]
```

In [16]:
```
print("Accuracy of RandomForestRegressor is: ",model.score(X_Test,y_Test))
```

```
Accuracy of RandomForestRegressor is:  0.7847288235111113
```