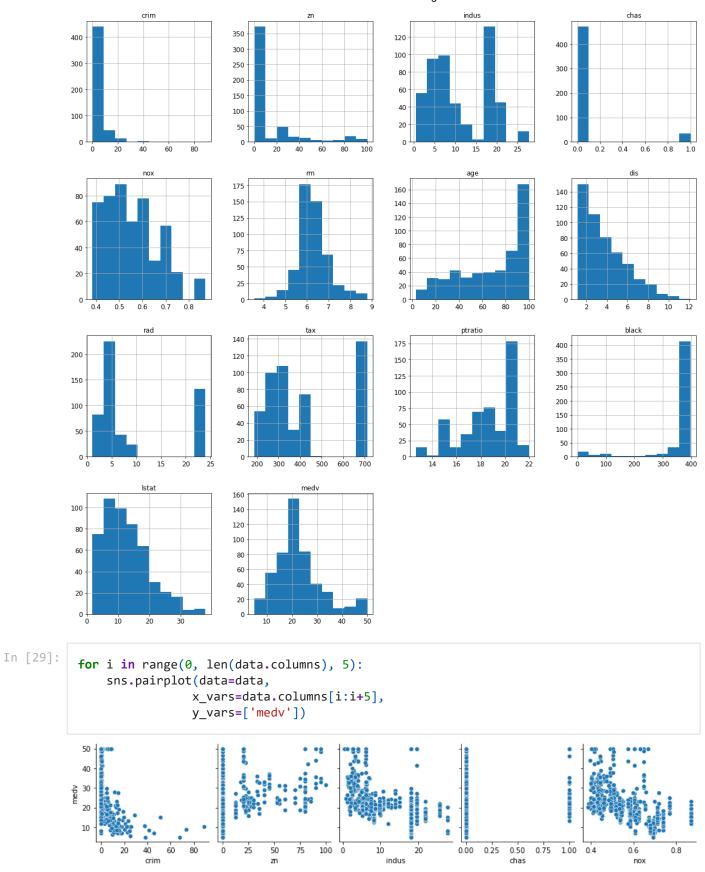
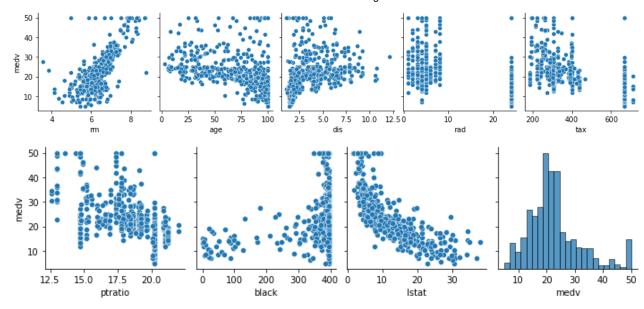
```
In [26]:
           import pandas as pd
           import numpy as np
           import matplotlib.pyplot as plt
           import seaborn as sns
           %matplotlib inline
           from sklearn.model selection import train test split
           from sklearn.linear model import LinearRegression
           from sklearn import metrics
           from sklearn.metrics import r2 score
In [27]:
           data = pd.read csv('Boston.csv')
           data = data.drop(['Unnamed: 0'],axis=1)
           data.head(5)
Out[27]:
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                       zn indus chas
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                                                                                  black Istat medv
                                        nox
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          0 0.00632
                                             6.575
                                                                     296
                      18.0
                            2.31
                                       0.538
                                                    65.2 4.0900
                                                                  1
                                                                             15.3
                                                                                  396.90
                                                                                         4.98
                                                                                                24.0
             0.02731
                       0.0
                            7.07
                                       0.469
                                             6.421
                                                    78.9 4.9671
                                                                  2
                                                                     242
                                                                             17.8
                                                                                  396.90
                                                                                         9.14
                                                                                                21.6
             0.02729
                       0.0
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                                                                             17.8 392.83
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             0.06905
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                            2.18
                                    0 0.458 7.147 54.2 6.0622
                                                                  3 222
                                                                             18.7 396.90
                                                                                         5.33
                                                                                                36.2
In [28]:
           data.hist(figsize = (20,20),xlabelsize=12, ylabelsize=12)
Out[28]: array([[<AxesSubplot:title={'center':'crim'}>,
                   <AxesSubplot:title={'center':'zn'}>,
                   <AxesSubplot:title={'center':'indus'}>,
                   <AxesSubplot:title={'center':'chas'}>],
                  [<AxesSubplot:title={'center':'nox'}>,
                   <AxesSubplot:title={'center':'rm'}>,
                   <AxesSubplot:title={'center':'age'}>,
<AxesSubplot:title={'center':'dis'}>],
                  [<AxesSubplot:title={'center':'rad'}>,
                   <AxesSubplot:title={'center':'tax'}>,
                   <AxesSubplot:title={'center':'ptratio'}>,
                   <AxesSubplot:title={'center':'black'}>],
                  [<AxesSubplot:title={'center':'lstat'}>,
                   <AxesSubplot:title={'center':'medv'}>, <AxesSubplot:>,
                   <AxesSubplot:>]], dtype=object)
```

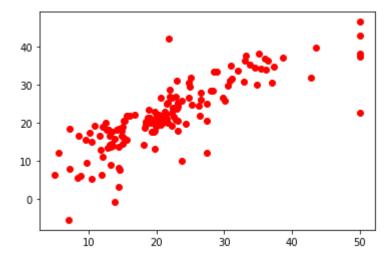




```
In [30]: y=data['medv']
  data = data.drop('medv',axis = 1)
```

```
In [31]: x_train,x_test,y_train,y_test=train_test_split(data,y,test_size=0.3)
    reg=LinearRegression()
    reg.fit(x_train,y_train)
    y_pred = reg.predict(x_test)
    plt.scatter(y_test, y_pred, c = 'red')
```

Out[31]: <matplotlib.collections.PathCollection at 0x21be858e710>



```
print('MAE', metrics.mean_absolute_error(y_test, y_pred))
print('MSE', metrics.mean_squared_error(y_test, y_pred))
print('RMSE', np.sqrt(metrics.mean_squared_error(y_test, y_pred)))
print('R squared error', r2_score(y_test, y_pred))
```

MAE 3.7129656398788566 MSE 29.390148626967523 RMSE 5.4212681751567615 R squared error 0.66708551054697