Multiple Linear Regression

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In [5]: # conventional way to import pandas
    import pandas as pd
    # conventional way to import seaborn
    import seaborn as sns
    # conventional way to import numpy
    import numpy as np

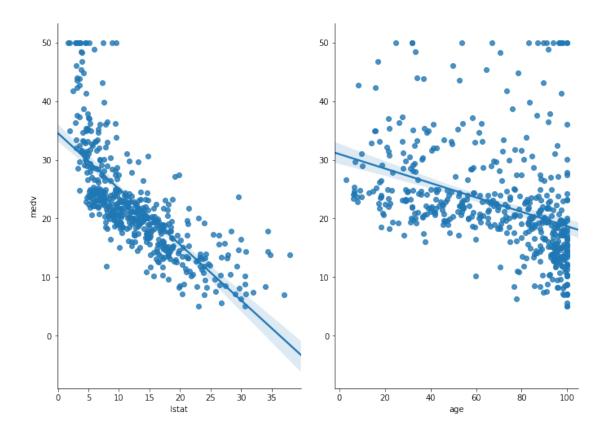
from sklearn import metrics
    import matplotlib.pyplot as plt

data = pd.read_csv("https://raw.github.com/vincentarelbundock/Rdatasets/master/csv/MAS:
    # create a Python list of feature names
    feature_cols = ['lstat', 'age']

# use the list to select a subset of the original DataFrame
    X = data[feature_cols].values

# select a Series from the DataFrame
    y = data['medv'].values
```

Viewing the data in seaborn to get a sense of it - MEDV Median value of owner-occupied homes in \$1000's - LSTAT lower status of the population - AGE proportion of owner-occupied units built prior to 1940



In [7]: # import model
 from sklearn.linear_model import LinearRegression

instantiate
 linreg = LinearRegression()

fit the model to the training data (learn the coefficients)
 linreg.fit(X, y)

print the intercept and coefficients
 print('intercept: \n', linreg.intercept_)

The coefficients
 print('Coefficients: \n', # pair the feature names with the coefficients
 list(zip(feature_cols, linreg.coef_)))

make predictions on the testing set
 y_pred = linreg.predict(X)

calculate RMSE using scikit-learn
 np.sqrt(metrics.mean_squared_error(y, y_pred))

```
# Explained variance score: 1 is perfect prediction
    print('Variance score: %.2f' % metrics.r2_score(y, y_pred))
intercept:
    33.2227605318
Coefficients:
    [('lstat', -1.0320685641826013), ('age', 0.034544338571646085)]
Variance score: 0.55
```

The obtained the predictions produced by linear regression. Look at the residuals, the difference between the real target set and the predicted target set:

