```
import numpy as np
 In [2]: import matplotlib.pyplot as plt
          import seaborn as sns
 In [3]: %matplotlib inline
In [ ]:
In [4]: data1 = pd.read_csv("air_index_ml.csv")
 In [5]: data1
 Out[5]:
                 humidity wind_speed wind_direction visibility_in_miles dew_point temperature clouds_all air_pollution_index
                                                                                                         121
              0
                      89
                                            329
                                                                             288.28
                                                                                          40
              1
                      67
                                 3
                                            330
                                                             1
                                                                      1
                                                                             289.36
                                                                                          75
                                                                                                         178
              2
                                 3
                                                                             289.58
                                                                                          90
                                                                                                         113
                      66
                                            329
              3
                      66
                                 3
                                            329
                                                                      5
                                                                             290.13
                                                                                          90
              4
                      65
                                 3
                                            329
                                                                             291.14
                                                                                          75
                                                                                                         281
                                             ...
                                                                                ...
           33745
                      85
                                 1
                                            328
                                                                             287.88
                                                                                          90
           33746
                      70
                                 1
                                             24
                                                             2
                                                                             286.95
                                                                                          90
           33747
                      70
                                 1
                                             24
                                                                             286.95
                                                                                          90
                                                                                                         138
           33748
                      66
                                 1
                                             42
                                                             6
                                                                      6
                                                                             285.75
                                                                                          90
                                                                                                         229
           33749
                      63
                                 1
                                             27
                                                                             285.75
                                                                                          90
                                                                                                         140
          33750 rows × 8 columns
In [6]: data1.shape
 Out[6]: (33750, 8)
 In [7]: data1.info()
          <class 'pandas.core.frame.DataFrame'>
          RangeIndex: 33750 entries, 0 to 33749
          Data columns (total 8 columns):
               Column
           #
                                      Non-Null Count Dtype
          - - -
               -----
                                       -----
               humidity
                                       33750 non-null int64
           1
               wind_speed
                                       33750 non-null int64
               wind_direction
                                       33750 non-null int64
           2
               visibility_in_miles 33750 non-null int64
           3
           4
               dew_point
                                       33750 non-null int64
           5
               temperature
                                       33750 non-null float64
               clouds_all
                                       33750 non-null int64
               air_pollution_index 33750 non-null int64
          dtypes: float64(1), int64(7)
          memory usage: 2.1 MB
In [8]: data1.describe()
 Out[8]:
                             wind_speed wind_direction visibility_in_miles
                                                                      dew_point temperature
                                                                                              clouds_all air_pollution_i
           count 33750.000000 33750.000000 33750.000000
                                                        33750.000000 33750.000000 33750.000000 33750.000000
                                                                                              38.871734
                   16.852248
                                2.055792
                                            99.841088
                                                            2.570021
                                                                        2.570021
                                                                                  13.415256
             std
                   13.000000
                                0.000000
                                             0.000000
                                                            1.000000
                                                                        1.000000
                                                                                   0.000000
                                                                                               0.000000
            min
            25%
                   60.000000
                                2.000000
                                           130.000000
                                                            3.000000
                                                                        3.000000
                                                                                 271.720000
                                                                                               1.000000
                                                            5.000000
            50%
                   72.000000
                                3.000000
                                           200.000000
                                                                        5.000000
                                                                                 280.150000
                                                                                              64.000000
                                                                                 290.620000
            75%
                   85.000000
                                5.000000
                                           290.000000
                                                            7.000000
                                                                        7.000000
                                                                                              90.000000
                               16.000000
                  100.000000
                                           360.000000
                                                            9.000000
                                                                        9.000000
                                                                                 308.240000
                                                                                             100.000000
 In [9]: data1.columns
 Out[9]: Index(['humidity', 'wind_speed', 'wind_direction', 'visibility_in_miles',
                  'dew_point', 'temperature', 'clouds_all', 'air_pollution_index'],
                dtype='object')
In [10]: sns.distplot(data1["air_pollution_index"])
Out[10]: <matplotlib.axes._subplots.AxesSubplot at 0x1892785b888>
           0.0035
           0.0030
           0.0025
           0.0020
           0.0015
           0.0010
           0.0005
           0.0000
                          50
                               100
                                     150
                                          200
                                                250
                                                      300
                                air_pollution_index
In [47]: | sns.heatmap(data1.corr())
Out[47]: <matplotlib.axes._subplots.AxesSubplot at 0x189350777c8>
                  humidity
                                                                - 0.8
                wind_speed
              wind_direction
                                                                - 0.6
            visibility_in_miles
                                                                - 0.4
                 dew_point
               temperature
                                                                0.2
                 clouds_all
                                                                - 0.0
           air_pollution_index
In [48]: from sklearn.preprocessing import StandardScaler
         scaler = StandardScaler()
In [52]: scaler.fit(data1.drop('air_pollution_index',axis=1))
Out[52]: StandardScaler(copy=True, with_mean=True, with_std=True)
In [56]: | scaled_features = scaler.transform(data1.drop('air_pollution_index', axis=1))
In [58]: | df_feat = pd.DataFrame(scaled_features, columns=data1.columns[:-1])
In [59]: df_feat
Out[59]:
                 humidity wind_speed wind_direction visibility_in_miles dew_point temperature clouds_all
                                                                            0.612030 -0.269063
              0 1.055720
                            -0.670405
                                         1.297362
                                                       -1.552441 -1.552441
              1 -0.249763
                            -0.183967
                                         1.307378
                                                       -1.552441 -1.552441
                                                                            0.692536
                                                                                     0.631348
                            -0.183967
                                                       -1.163334 -1.163334
              2 -0.309103
                                         1.297362
                                                                            0.708936
                                                                                     1.017238
              3 -0.309103
                           -0.183967
                                         1.297362
                                                        0.003989
                                                                 0.003989
                                                                            0.749934
                                                                                     1.017238
              4 -0.368443
                           -0.183967
                                         1.297362
                                                        0.782204
                                                                 0.782204
                                                                            0.825223
                                                                                     0.631348
                                                                 0.782204
           33745 0.818359
                            -1.156843
                                         1.287346
                                                        0.782204
                                                                            0.582212
                                                                                     1.017238
           33746 -0.071743
                           -1.156843
                                        -1.757537
                                                       -1.163334 -1.163334
                                                                            0.512887
                                                                                     1.017238
           33747 -0.071743
                           -1.156843
                                        -1.757537
                                                       -1.552441 -1.552441
                                                                            0.512887
                                                                                     1.017238
           33748 -0.309103
                           -1.156843
                                        -1.577248
                                                        0.393097
                                                                 0.393097
                                                                            0.423436
                                                                                     1.017238
                           -1.156843
           33749 -0.487124
                                        -1.727489
                                                        1.171312 1.171312
                                                                            0.423436 1.017238
          33750 rows × 7 columns
In [60]: | X = df_feat[['humidity', 'wind_speed', 'wind_direction', 'visibility_in_miles',
                  'dew_point', 'temperature', 'clouds_all']]
In [61]: y = data1["air_pollution_index"]
In [ ]:
In [62]: from sklearn.model_selection import train_test_split
In [63]: | X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.4, random_state=101)
In [64]: from sklearn.linear_model import LinearRegression
In [65]: | lm = LinearRegression()
In [66]: lm.fit(X_train,y_train)
Out[66]: LinearRegression(copy_X=True, fit_intercept=True, n_jobs=None, normalize=False)
In [67]: | lm.coef_
Out[67]: array([ 0.2249568 , 0.03240539, 0.85865452, -0.01772103, -0.01772103,
                   0.32577829, -0.17129599])
In [68]: | cdf = pd.DataFrame(lm.coef_, X.columns, columns=["Coeff"])
In [69]: cdf
Out[69]:
                             Coeff
                  humidity
                          0.224957
                         0.032405
               wind_speed
             wind_direction 0.858655
                dew_point -0.017721
               temperature 0.325778
                clouds_all -0.171296
In [ ]:
In [70]: predictions = lm.predict(X_test)
In [71]: predictions
Out[71]: array([154.6314371 , 155.20855255, 154.79561539, ..., 155.53065861,
                  154.59754095, 154.60629112])
In [72]: | y_test
Out[72]: 2606
                    187
          7281
                     41
          17448
                    152
          29785
                    130
          16654
                    120
          1378
                    195
          22034
                    32
          14634
                    177
          12211
                    247
          2509
                    114
          Name: air_pollution_index, Length: 13500, dtype: int64
In [73]: plt.scatter(y_test, predictions)
Out[73]: <matplotlib.collections.PathCollection at 0x189366b8c08>
           156
           154
           152
           150
           148
           146
                     50
                            100
                                  150
                                                 250
                                                       300
                                         200
In [74]: sns.distplot((y_test - predictions))
Out[74]: <matplotlib.axes._subplots.AxesSubplot at 0x18936b55448>
           0.0035
           0.0030
           0.0025
           0.0020
           0.0015
           0.0010
           0.0005
                -200 -150 -100
                                                100
                                                     150
                               -50
                                           50
                                air_pollution_index
In [75]: from sklearn import metrics
In [76]: metrics.mean_absolute_error(y_test, predictions)
Out[76]: 72.30634223695697
In [77]: | metrics.mean_squared_error(y_test, predictions)
Out[77]: 6959.427398165745
In [78]: | np.sqrt(metrics.mean_squared_error(y_test,predictions))
Out[78]: 83.42318261829709
In [ ]:
In [79]: | from sklearn.metrics import mean_squared_error
In [80]: def plot_learning_curves(model,x,y):
              X_train, X_val, y_train, y_val = train_test_split(X, y, test_size=0.4)
              val_errors = []
              for m in range(1,len(X_train)):
                   model.fit(X_train[:m],y_train[:m])
                   y_train_predict = model.predict(X_train[:m])
                   y_val_predict = model.predict(X_val)
                   val_errors.append(mean_squared_error(y_val_predict,y_val))
              plt.plot(np.sqrt(val_errors), "r-+", linewidth=2, label='validation')
              plt.legend()
In [81]: lin_reg = LinearRegression()
          plot_learning_curves(lin_reg, X, y)
                                                validation
           350
           300
           250
           200
           150
           100
                   2500 5000 7500 10000 12500 15000 17500 20000
In [ ]:
In [ ]:
In [ ]:
In [ ]:
 In [ ]:
 In [ ]:
```

In [1]: import pandas as pd

20

74

81

33750.00

83.73

10.00

83.00

155.00

228.00

299.00