How to Build a Machine Learning Model in **Python**

In the modeling stage of the machine learning process, our goal is to choose and apply the

Learning Objectives

appropriate machine learning approach that works with the data we have and solves the problem that we intend to solve. If our objective is to build a model that predicts a numeric or continuous value, then our problem is known as a regression problem. One of the most common models used in solving regression problems is Linear Regression. By the end of the tutorial, you will have learned: · how to collect, explore and prepare data

· how to build and evaluate a model

1. Collect the Data

import pandas as pd bikes = pd.read_csv("bikes.csv") bikes.head()

In [1]:

```
temperature humidity
                                   windspeed rentals
Out[1]:
              46.716528 0.815969
                                   13.669663
                                                 985
              48.350239 0.800497
                                   15.199782
                                                 801
```

2	34.212394	0.592097	13.247558	1349
3	34.520000	0.623196	11.687963	1562
4	36.800562	0.624643	13.148281	1600
2. Explore the Data				

Data columns (total 4 columns): Column Non-Null Count Dtype

In [2]: bikes.info()

min

25%

6000

2000

0.0

response = 'rentals'

rentals

2

726

727 728 985 801

1349 1562

1600

2114 3095

1341

y = bikes[[response]]

In [7]:

Out[7]:

2.5

3. Prepare the Data

22.602432

46.117264

```
-----
           temperature 731 non-null
         0
                                         float64
         1
           humidity 731 non-null
                                         float64
         2
           windspeed 731 non-null float64
         3
           rentals
                         731 non-null
                                          int64
        dtypes: float64(3), int64(1)
        memory usage: 23.0 KB
       bikes.describe()
In [3]:
Out[3]:
              temperature
                          humidity windspeed
                                                rentals
               731.000000 731.000000 731.000000
        count
                                             731.000000
               59.509553
        mean
                          0.486937
                                    9.238886 4504.348837
          std
               15.486114
                          0.185415
                                   3.379815 1937.211452
```

<class 'pandas.core.frame.DataFrame'> RangeIndex: 731 entries, 0 to 730

```
50%
                 59.758972
                                       9.503508 4548.000000
                             0.502227
                             0.624671 11.814559 5956.000000
                 73.048236
                 90.497028
                             0.972500
                                       21.126627 8714.000000
          max
         %matplotlib inline
In [4]:
         bikes.plot(kind = 'scatter', x = 'temperature', y = 'rentals')
         <AxesSubplot:xlabel='temperature', ylabel='rentals'>
Out[4]:
            8000
```

0.000000

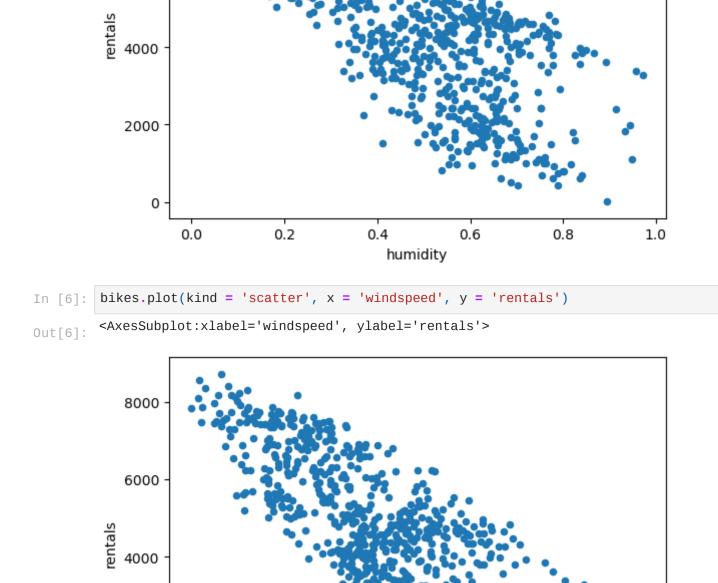
0.353548

0.932208

22.000000

6.863568 3152.000000

```
6000
            4000
            2000
               0
                          30
                                   40
                                            50
                                                     60
                                                              70
                                                                       80
                                                                                90
                 20
                                             temperature
        bikes.plot(kind = 'scatter', x = 'humidity', y = 'rentals')
In [5]:
         <AxesSubplot:xlabel='humidity', ylabel='rentals'>
Out[5]:
            8000
```



729 1796 730 2729

0.815969

0.800497

windspeed

13.669663

15.199782

13.247558

11.687963

13.148281

10.801229

11.829425

 x_{train} , x_{test} , y_{train} , y_{test} = $train_{test_{split}}(x, y, random_{state}$ = 1234)

The model coefficients correspond to the order in which the independent variables are listed in the

y = 3800.68 + 80.35 imes temperature - 4665.74 imes humidity - 196.22 imes windspeed

training data. This means that the equation for the fitted regression line can be written as:

5.0

7.5

10.0

windspeed

12.5

15.0

17.5

20.0

```
731 rows × 1 columns
         predictors = list(bikes.columns)
         predictors.remove(response)
         x = bikes[predictors]
Out[8]:
              temperature humidity
                46.716528
                48.350239
                34.212394 0.592097
                34.520000 0.623196
           4
                36.800562 0.624643
         726
                39.102528 0.482493
                39.031972 0.480433
         728
                39.031972 0.717730
```

```
729
                39.243472 0.523039
                                    12.805314
         730
                35.859472 0.494808
                                    9.346850
        731 rows × 3 columns
In [9]: from sklearn.model_selection import train_test_split
```

model = LinearRegression().fit(x_train, y_train) model.intercept_ In [12]:

array([3800.68469948])

In [10]:

Out[12]:

Out[13]:

In [14]:

Out[14]:

In [15]:

Out[17]:

4. Train the Model

```
model.coef_
In [13]:
                    80.35314543, -4665.73867387,
                                                   -196.21650368]])
         array([[
```

from sklearn.linear_model import LinearRegression

```
With the linear regression equation, we can estimate what our model will predict given any weather
condition. For example, given a temperature of 72^{\circ}F, 22\% humidity and windspeed of 5 miles per
hour, our model would predict:
```

 $7,578 \text{ bikes} \approx 3800.68 + 80.35 \times 72 - 4665.74 \times .22 - 196.22 \times 5$

```
model.score(x_test, y_test)
0.9820623857913312
```

5. Evaluate the Model

```
y_pred = model.predict(x_test)
In [17]: from sklearn.metrics import mean_absolute_error
         mean_absolute_error(y_test, y_pred)
```

194.31620720519678