Logistic Regression(Binary Classifier)

July 15, 2022

0.1 Predicting whether a person will buy a life insurance or not based on their age

0.1.1 Import libraries

```
[1]: import pandas as pd import matplotlib.pyplot as plt %matplotlib inline
```

0.1.2 data exploration

```
[4]: df = pd.read_csv('insurance_data.csv')
```

```
[5]: # show the top five from the data df.head()
```

```
[5]:
         age
              bought_insurance
          22
         25
                               0
     1
     2
         47
                               1
     3
         52
                               0
     4
          46
                               1
```

```
[6]: # show the bottom five from the data df.head()
```

```
[6]:
              bought_insurance
         age
     0
          22
                                0
     1
          25
                                0
     2
          47
                                1
     3
          52
                                0
     4
          46
                                1
```

```
[8]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 27 entries, 0 to 26
Data columns (total 2 columns):
# Column Non-Null Count Dtype
```

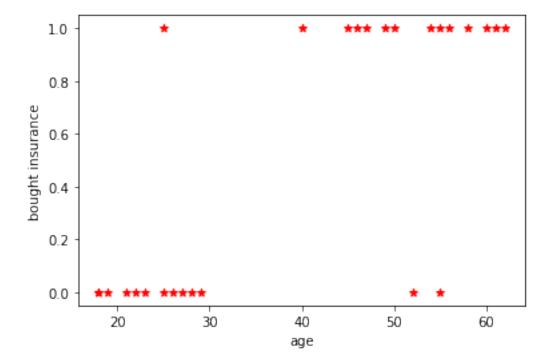
0 age 27 non-null int64 1 bought_insurance 27 non-null int64

dtypes: int64(2)

memory usage: 560.0 bytes

```
[12]: plt.scatter(df.age, df.bought_insurance, marker = '*', color = 'red')
    plt.xlabel('age')
    plt.ylabel('bought insurance')
```

[12]: Text(0, 0.5, 'bought insurance')



0.1.3 spliting the dataset

```
[13]: from sklearn.model_selection import train_test_split

[26]: x_train, x_test, y_train, y_test = train_test_split(df[['age']], df.

⇒bought_insurance, test_size = 0.2, random_state = 2)
```

[27]: y_train

[27]: 6 0 3 0 17 1 24 1

```
14
            1
      4
            1
      10
            0
      5
            1
      19
            0
      1
            0
      2
            1
      7
            1
      23
            1
      21
            0
      18
            0
      11
            0
      22
            1
      26
            0
      13
            0
      15
            1
      8
            1
      Name: bought_insurance, dtype: int64
     0.1.4 Predicting the model
[28]: # here we use Logistic regression technique to slove binary classifier
      from sklearn.linear_model import LogisticRegression
      model = LogisticRegression()
[31]: model.fit(X_train, y_train)
[31]: LogisticRegression()
[32]: x_test
[32]:
          age
      25
           54
      0
           22
      9
           61
      20
           21
      16
           25
      12
           27
[34]: y_predict = model.predict(x_test)
      y_predict
[34]: array([1, 0, 1, 0, 0, 0], dtype=int64)
[37]: y_test
```

```
[37]: 25
            1
      0
      9
            1
      20
            0
      16
            1
      12
      Name: bought_insurance, dtype: int64
[36]: model.predict_proba(x_test)
[36]: array([[0.11926354, 0.88073646],
             [0.95178274, 0.04821726],
             [0.04355309, 0.95644691],
             [0.95844557, 0.04155443],
             [0.92522499, 0.07477501],
             [0.90062385, 0.09937615]])
[38]: # check the accurary of our model
      model.score(x_test, y_test)
[38]: 0.8333333333333333
     0.1.5 Optional
[39]: \# model.coef_ indicates value of m in y = m*x + b equation
      model.coef
[39]: array([[0.15568883]])
[40]: | \# model.intercept_indicates value of b in y = m*x + b equation
      model.intercept_
[40]: array([-6.40777392])
[41]: # sigmoid function
      import math
      def sigmoid(z):
          return 1 / (1 + math.exp(-z))
[42]: # predicting the age by using sigmoid function
      def prediction_function(age):
          z = 0.156*age + -6.41
          y = sigmoid(z)
          return y
[43]: age = 45
      prediction_function(age)
```

```
[43]: 0.6479408020806501

0.647 is more than 0.5 so the person will buy the insurance

[45]: age = 31
    prediction_function(age)

[45]: 0.1716469075389188

0.171 is less than 0.5 so the person will not buy the insurance

[]:
```