ex04-logistic-regression

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```
[1]: import pandas as pd
    from sklearn.model_selection import train_test_split
    from sklearn.linear_model import LogisticRegression
    from sklearn.metrics import confusion_matrix
    from sklearn.datasets import make_classification
    from matplotlib import pyplot as plt
[2]: x, y = make_classification(
        n_samples=100,
        n_features=1,
        n_classes=2,
        n_clusters_per_class=1,
        flip_y=0.03,
        n_informative=1,
        n_redundant=0,
        n_repeated=0,
    print(y)
    [0\ 1\ 0\ 0\ 0\ 1\ 0\ 1\ 0\ 1\ 0\ 1\ 0\ 0\ 0\ 0\ 0\ 0\ 1\ 1\ 1\ 1\ 1\ 1\ 1\ 0\ 1\ 0\ 1\ 1\ 1\ 1\ 0\ 0\ 0\ 1
     1 1 0 0 1 0 0 1 1 1 1 1 1 1 1 0 0 0 1 1 1 1 1 0 0 0 1]
[3]: plt.scatter(x, y, c=y, cmap="rainbow")
    plt.title("scatter plot of Logistic Regression")
    plt.show()
```



```
[9]: df = pd.read_csv("datasets/insurance_data.csv")
[10]: df
[10]:
          age bought_insurance
      0
           22
      1
           25
                                0
      2
           47
                                1
      3
           52
                                0
      4
           46
                                1
      5
           56
                                1
      6
           55
                                0
      7
           60
                                1
      8
           62
                                1
      9
           61
                                1
      10
           18
                                0
      11
           28
                                0
      12
           27
                                0
      13
           29
                                0
      14
           49
                                1
      15
           55
                                1
      16
           25
                                1
      17
           58
                                1
      18
           19
                                0
      19
           18
                                0
      20
           21
                                0
      21
           26
                                0
      22
           40
                                1
      23
           45
                                1
      24
           50
                                1
      25
           54
                                1
      26
           23
                                0
[11]: df.head()
[11]:
         age bought_insurance
      0
          22
      1
          25
                               0
      2
          47
                               1
      3
          52
                               0
      4
          46
                               1
[12]: plt.scatter(df.age, df.bought_insurance, marker="+", color="red")
```

[12]: <matplotlib.collections.PathCollection at 0x7f4ce327e510>

```
[13]: df.shape
[13]: (27, 2)
[14]: x_train, x_test, y_train, y_test = train_test_split(
          df[["age"]], df.bought_insurance, test_size=0.1
[15]: x_test
[15]:
          age
      21
           26
      5
           56
      18
           19
[16]: model = LogisticRegression()
[17]: model.fit(x_train, y_train)
[17]: LogisticRegression()
[18]: model.predict(x_test)
```

```
[18]: array([0, 1, 0])
[19]: model.score(x_test, y_test)
[19]: 1.0
[20]: model.predict_proba(x_test)
[20]: array([[0.82649559, 0.17350441],
             [0.09917355, 0.90082645],
             [0.91983146, 0.08016854]])
[21]:
      df.describe()
[21]:
                   age
                        bought_insurance
             27.000000
                                27.000000
      count
      mean
             39.666667
                                 0.518519
      std
             15.745573
                                 0.509175
                                 0.000000
      min
             18.000000
      25%
             25.000000
                                 0.000000
      50%
                                 1.000000
             45.000000
      75%
             54.500000
                                 1.000000
             62.000000
                                 1.000000
      max
```