

Practicing Logistic Regression Model

Predicting whether a person will buy insurance.

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

```
In [2]: df = pd.read_csv(r'D:\NEEL_FOLDER\Data Science\Logistics_Regression_codebasic
        \insurance_data.csv')
        df.head()
```

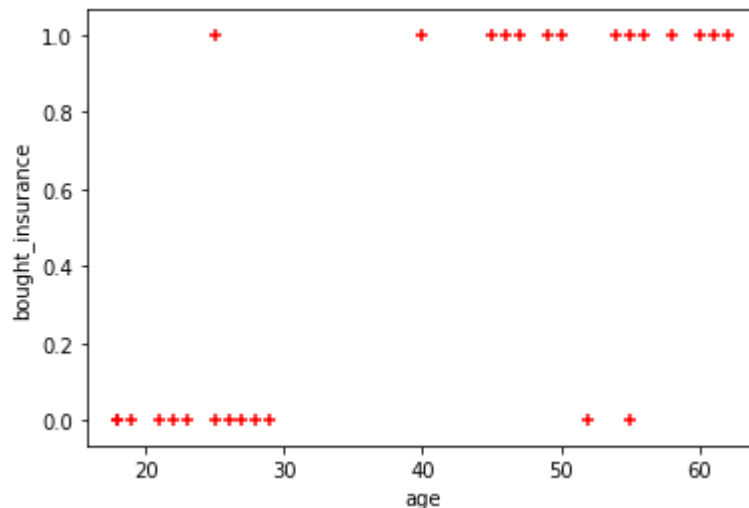
Out[2]:

	age	bought_insurance
0	22	0
1	25	0
2	47	1
3	52	0
4	46	1

Plot scatter plot to see data distribution.

```
In [3]: plt.xlabel('age')
        plt.ylabel('bought_insurance')
        plt.scatter(df.age, df.bought_insurance, marker='+', color='red')
```

Out[3]: <matplotlib.collections.PathCollection at 0x8bd1f30>



Use train test split method to split our dataset

```
In [4]: from sklearn.model_selection import train_test_split
```

```
In [5]: X_train, X_test, y_train, y_test = train_test_split(df[['age']],df.bought_insurance,train_size=0.8)
```

```
In [6]: X_test
```

Out[6]:

	age
8	62
24	50
22	40
13	29
6	55
2	47

```
In [7]: from sklearn.linear_model import LogisticRegression
```

```
In [8]: model = LogisticRegression()
```

```
In [9]: model.fit(X_train,y_train)
```

C:\Users\admin\anaconda3\lib\site-packages\sklearn\linear_model\logistic.py:432: FutureWarning: Default solver will be changed to 'lbfgs' in 0.22. Specify a solver to silence this warning.
FutureWarning)

Out[9]: LogisticRegression(C=1.0, class_weight=None, dual=False, fit_intercept=True, intercept_scaling=1, l1_ratio=None, max_iter=100, multi_class='warn', n_jobs=None, penalty='l2', random_state=None, solver='warn', tol=0.0001, verbose=0, warm_start=False)

```
In [10]: X_test
```

Out[10]:

	age
8	62
24	50
22	40
13	29
6	55
2	47

```
In [11]: y_predicted = model.predict(X_test)
```

```
In [12]: model.predict_proba(X_test)
```

```
Out[12]: array([[0.15905366, 0.84094634],
                [0.26936303, 0.73063697],
                [0.39134422, 0.60865578],
                [0.54243394, 0.45756606],
                [0.21824002, 0.78175998],
                [0.30343382, 0.69656618]])
```

```
In [13]: model.score(X_test,y_test)
```

```
Out[13]: 0.8333333333333334
```

```
In [14]: y_predicted
#the prediction says person with age of 62, 50, 40, 55 & 47 have bought insurance.
#Age 29 have not bought insurance.
```

```
Out[14]: array([1, 1, 1, 0, 1, 1], dtype=int64)
```

```
In [15]: model.coef_
```

```
Out[15]: array([[0.05561913]])
```

```
In [16]: model.intercept_
```

```
Out[16]: array([-1.78309979])
```

Lets defined sigmoid function now and do the math with hand

```
In [17]: import math
def sigmoid(x):
    return 1 / (1 + math.exp(-x))
```

```
In [18]: def prediction_function(age):
          z = 0.042 * age - 1.53 # 0.04150133 ~ 0.042 and -1.52726963 ~ -1.53
          y = sigmoid(z)
          return y
```

```
In [19]: age = 35
          prediction_function(age)
          #0.485 is less than 0.5 which means person with 35 age will not buy insurance
```

```
Out[19]: 0.4850044983805899
```

```
In [20]: age = 43  
         prediction_function(age)  
         #0.485 is more than 0.5 which means person with 43 will buy the insurance
```

Out[20]: 0.568565299077705

Source:Codebasic youtube channel