# Logistic Regression

李孟潔

111522030

## Result

Loss function 採用 cross entropy loss:

$$-\sum \{y^{(i)}\log(\hat{y}^{(i)}) + (1-y^{(i)})\log(1-\hat{y}^{(i)})\}$$

• 透過隨機梯度下降(SGD)的方式來找最佳theta:

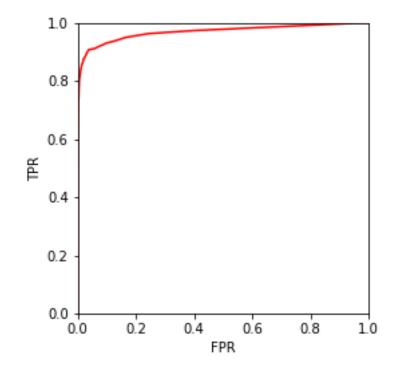
$$\boldsymbol{\theta}^{(k+1)} \coloneqq \boldsymbol{\theta}^{(k)} - \alpha \boldsymbol{g}$$

$$\theta_i^{(k+1)} = \theta_i^{(k)} - \alpha \frac{\partial J(\boldsymbol{\theta})}{\partial \theta_i} \Big|_{\boldsymbol{\theta} = \boldsymbol{\theta}^{(k)}}$$

$$\theta_i^{(k+1)} = \theta_i^{(k)} - \alpha \left. \frac{\partial J(\boldsymbol{\theta})}{\partial \theta_i} \right|_{\boldsymbol{\theta} = \boldsymbol{\theta}^{(k)}} \qquad \frac{\frac{\partial}{\partial \theta_i} J(\boldsymbol{\theta}) = \frac{1}{n} \sum_j \left( \hat{y}^{(j)} - y^{(j)} \right) x_i^{(j)},}{\frac{1}{n} \text{ is usually ignored}}$$

### Result

藉由不同的Threshold來找到TPR跟FPR的關係,就能畫出ROC curve。藉由觀察ROC curve來看分類的情形,當ROC curve底下的面積越大,代表預測的準確度越高!



### FPR:

False Positive Rate, 當Specificity越高, FPR越低, model越能正確判斷負樣本

#### TPR:

True Positive Rate, 又稱Sensitivity or Recall, TPR 越高,model越能正確判斷正樣本

### Result

· 藉由上述方法找到的theta:

```
[[ 0. ]

[ -9.60770821]

[ 0.44053858]

[ 16.9074481 ]

[ 0.80578103]

[ -5.81422069]

[ 2.75235616]

[-11.04834491]

[ 3.6384396 ]]
```

 Accuracy, precision and recall for training data and the testing data:

```
Logreg train accuracy: 0.975081
Logreg train precision: 0.937838
Logreg train recall: 0.796785
Logreg test accuracy: 0.977763
Logreg test precision: 0.943838
Logreg test recall: 0.787760
```