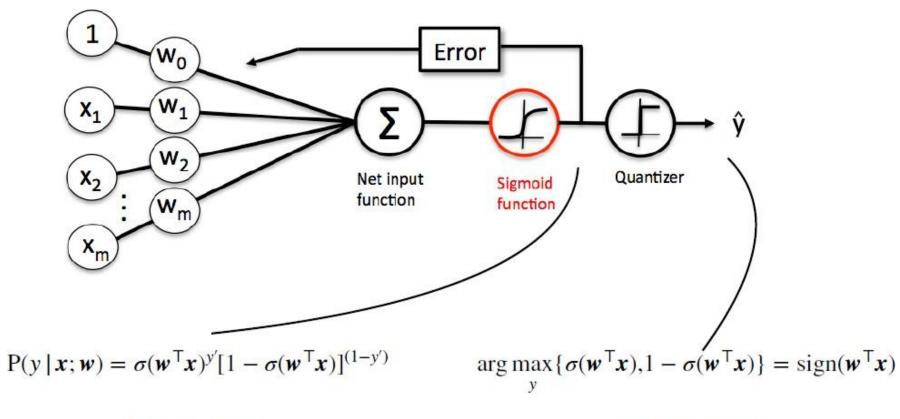
# Lab6: Logistic Regression and Metrics

Chia-Han Lu & DataLab 2022.10.05

- Common Evaluation Metrics for Binary Classification
  - Confusion Matrix
  - Soft Classifiers ROC Curve

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# Logistic Regression



Soft prediction

Label prediction

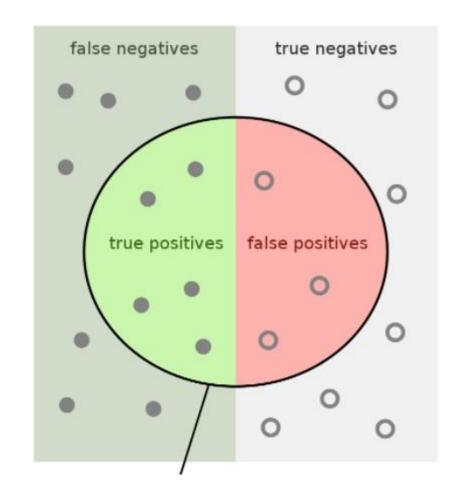
# Logistic Regression + Regularization

• 把Regularization Term加到Loss內, 讓模型在學weights的時候, 會傾向選擇比較簡單的模型。

- Common Evaluation Metrics for Binary Classification
  - Confusion Matrix
  - Soft Classifiers ROC Curve

• It is important to know how the model make wrong prediction.

• In binary classification, confusion matrix is a common tool to analyze the predictions.

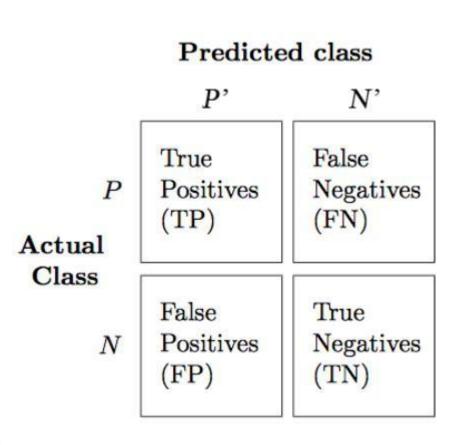


• It is important to know how the model make wrong prediction.

- •e.g. 檢測絕症
  - 寧可讓多一點人到TP & FP, 也不要讓FN很高。

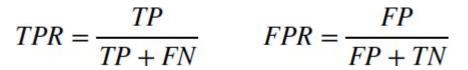
Other metrics we can use:

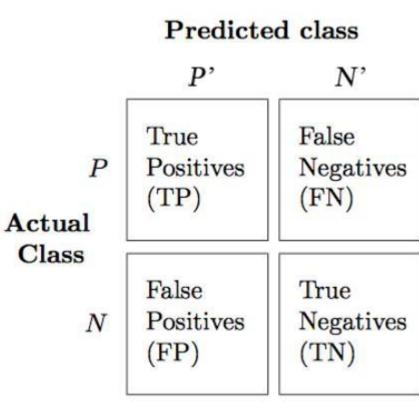
$$TPR = \frac{TP}{TP + FN}$$
  $FPR = \frac{FP}{FP + TN}$ 



• It is important to know how the model make wrong prediction.

- •e.g. 檢測絕症
  - 寧可讓多一點人到TP & FP, 也不要讓FN很高。
  - i.e. TPR higher





• Precision(PRE), Recall

$$PRE = \frac{TP}{TP + FP},$$
 (the higher, the better)
$$REC = \frac{TP}{TP + FN} = TPR.$$
 (the higher, the better)

Actual Class

• F 1 COPRE \* REC)  $F_1 = 2 \frac{(PRE * REC)}{PRE + REC}, \text{ (the higher, the better)}$ 

#### Predicted class

P' N'True
Positives
(TP)

False
Negatives
(FN)

False
True

False Positives (FP)

True Negatives (TN)

Actual

Class

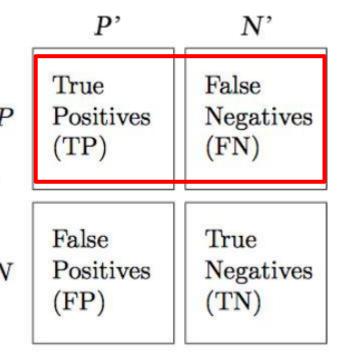
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### Predicted class



- Common Evaluation Metrics for Binary Classification
  - Confusion Matrix
  - Soft Classifiers ROC Curve

### **ROC Curve**

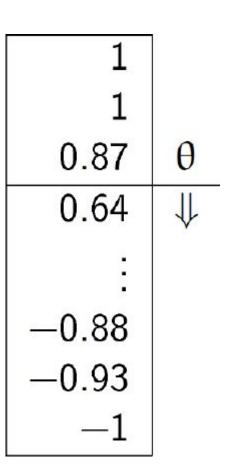
 ROC curve analyze the performance for every threshold in soft classifiers.

• In X-axis: FPR

$$FPR = \frac{FP}{FP + TN}$$

• In Y-axis: TPR

$$TPR = \frac{TP}{TP + FN}$$



### **ROC Curve**

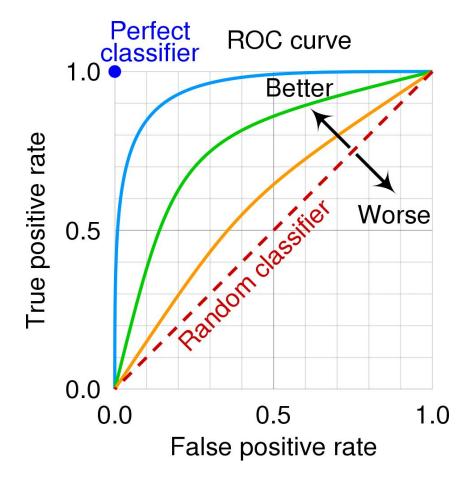
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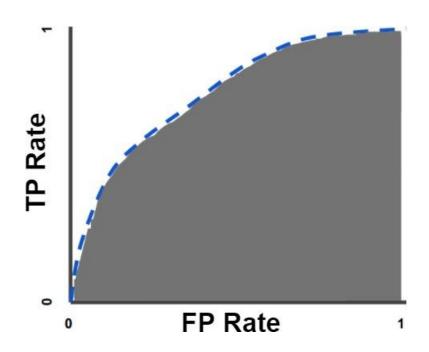
• In Y-axis: TPR

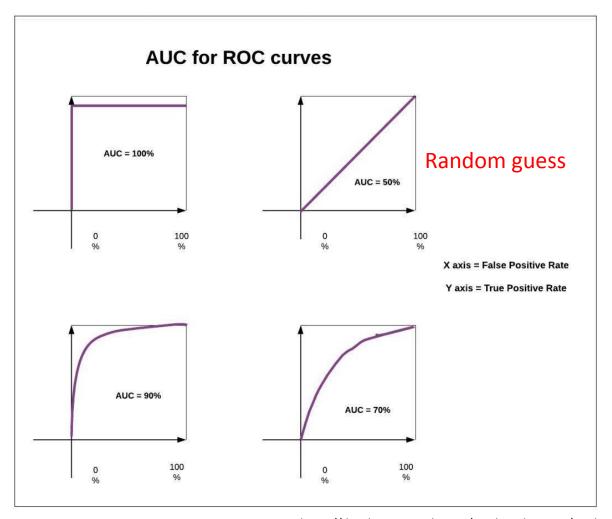
$$TPR = \frac{TP}{TP + FN}$$



## AUC

- AUC Area Under the ROC Curve.
  - ROC can be quantified using AUC.





https://developers.google.com/machine-learning/crash-course/classification/roc-and-auc https://medium.com/acing-ai/what-is-auc-446a71810df9

### Homework

Homework: Lab06

- Lab06: Logistic Regression, Metrics

• Bonus: Lab07 && Lab08

- Lab07: Support Vector Machine, k-Nearest Neighbors

- Lab08: Cross Validation, Ensemble

### Reference

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