

Tutorial 2

Decision Tree, Cross-validation, Precision and Recall

Luke Chang

The University of Auckland

Mar. 2021

Objectives

- 1 Evaluation Metrics: Accuracy, Precision, Recall and F1 score
- 2 ROC curve and AUC
- 3 Should you trust the results?
- 4 Parametric Tests VS. Non-parametric Tests
- 5 Regression and Least Square Problem
- 6 Ensemble Methods

Confusion Matrix

Confusion Matrix can be applied to **binary** classification as well as for **multiclass** classification problems.

		Predicted	
		Positive	Negative
Actual	Positive	True Positive	False Negative
	Negative	False Positive	True Negative

- True Positive (TP): Correctly classified.
- True Negative (TN): Correctly rejected.
- False Positive (FP): Incorrectly classified. Type I Error.
- False Negative (FN): Incorrectly rejected. Type II Error.

$$\text{Accuracy} = \frac{\text{TP} + \text{TN}}{\text{TP} + \text{TN} + \text{FP} + \text{FN}}$$

Confusion Matrix

How many selected items are relevant? Selected Elements = TP + FP

$$\text{Precision (P)} = \frac{TP}{TP + FP}$$

How many relevant items are selected? Relevant Elements = TP + FN

$$\text{Recall (R)} = \frac{TP}{TP + FN}$$

F_1 score is the **harmonic mean** between Precision and Recall.

$$F_1 = 2 \times \frac{P \times R}{P + R}$$

Example – Weather Prediction

Build a logistic regression model to predict the weather based on the humidity.

Recorded 10 days in total.

Class	Prediction
P	P
N	P
P	N
P	P
N	P
P	P
N	P
N	N
N	N
P	P

Caveat: A model with high Recall may also has high FPR (Type I Error).

		Predicted		Total
		P	N	
Actual	P	4	1	5
	N	3	2	5
Total		7	3	10

$$\text{Acc.} = \frac{6}{10} = 0.6$$

$$\text{Precision (P)} = \frac{\text{TP}}{\text{TP} + \text{FP}} = \frac{4}{4 + 3} \approx 0.571$$

$$\text{Recall (R)} = \frac{\text{TP}}{\text{TP} + \text{FN}} = \frac{4}{4 + 1} \approx 0.8$$

$$F_1 = 2 \frac{P \times R}{P + R} = 2 \times \frac{0.571 \times 0.8}{0.571 + 0.8} \approx 0.667$$

Precision-Recall (PR) Curve (Optional)

Average precision (AP) summarizes such a plot as the weighted mean of precisions achieved at each threshold.

$$AP = \sum_n (R_n - R_{n-1}) P_n$$

- Where P_n and R_n are the precision and recall at the n -th threshold.
- A pair (P_n, R_k) is referred to as an *operating point*.

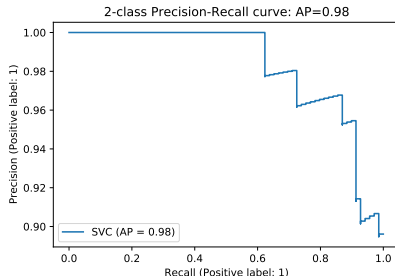


Figure: A SVM classifier trained on the Breast Cancer dataset

Receiver Operating Characteristic (ROC) Curve

- The ROC curve is created by plotting the true positive rate (TPR) against the false positive rate (FPR) at various threshold settings.
- Area Under Curve (AUC): The integration of the ROC function between 0 and 1.

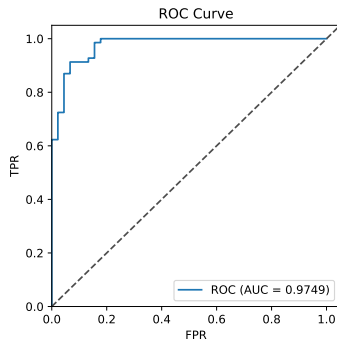


Figure: A SVM classifier trained on the Breast Cancer dataset

Example – Weather Prediction

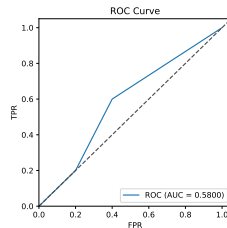
Build a logistic regression model to predict the weather based on the humidity.
Recorded 10 days in total.

Class	Prediction	Thresholds					
		0	0.2	0.4	0.6	0.8	1
P	0.95	1	1	1	1	1	0
N	0.85	1	1	1	1	1	0
P	0.78	1	1	1	1	0	0
P	0.66	1	1	1	1	0	0
N	0.6	1	1	1	1	0	0
P	0.55	1	1	1	0	0	0
N	0.53	1	1	1	0	0	0
N	0.52	1	1	1	0	0	0
N	0.51	1	1	1	0	0	0
P	0.4	1	1	1	0	0	0

Threshold	0	0.2	0.4	0.6	0.8	1
TPR	1	1	1	0.60	0.2	0
FPR	1	1	1	0.4	0.2	0

Sort the results:

Threshold	0	0.2	0.4	0.6	0.8	1
TPR	0	0.2	0.6	1	1	1
FPR	0	0.2	0.4	1	1	1



Should you trust the results?

Parametric Tests VS. Non-parametric Tests

Regression and Least Square Problem

Ensemble Methods