

Intro to ML

December 15th, 2021

教學助理 教學意見調查

WE NEED YOUR FEEDBACK

即日起
~1/2止

就是現在!!

- ① 登入「校務資訊系統」。



- ② 點選「教學助理評量問卷」即可填寫。

問卷填寫時間：110/12/10-111/1/2止。



完成全部問卷填寫的同學，就有機會抽到NTHU限量小禮唷！
同學現在就拿起手機掃描QR Code前往填寫問卷！

主辦單位：教務處教學發展中心 | 聯絡資訊：03-5716200 | <http://ctld.nthu.edu.tw/>

110上期末教學意見調查

懶人包

2021.12.10~2022.01.10上午9:00

教務處課務組



抽獎獎品有哪些？

品名	數量
iPad Air 64GB Wi-Fi	1
Apple Watch S7 41mm搭 配運動錶帶	4
BIRDYEDGE G3 (輕量)電動 滑板車	4

填卷截止後一週內，以電腦亂數抽出獲獎者並公布於課務組網頁。

求關注幫高調

懇請幫忙
分享宣傳



《班級鼓勵獎》獎金有多少？

填卷率高的班級發給鼓勵金(獲獎率約10%)。

首獎：共1班 獎金 5,000元
貳獎：共1班 獎金 4,000元
參獎：共1班 獎金 3,000元
特別獎：共15班 獎金各2,000元

★首獎、貳獎、參獎全校限各1班。

去哪查排名？

課務組網頁
每個工作天更新
最新排名



比賽得獎金！

加碼抽大獎！

CHAPTER 20:

Design and Analysis of ML Experiments

Algorithm Preference

- Criteria (Application-dependent):
 - Misclassification error, or risk (loss functions)
 - Training time/space complexity
 - Testing time/space complexity
 - Interpretability
 - Easy programmability
- Cost-sensitive learning

Guidelines for ML experiments

- A. Aim of the study
- B. Selection of the response variable
- C. Choice of factors and levels
- D. Choice of experimental design
- E. Performing the experiment
- F. Statistical Analysis of the Data
- G. Conclusions and Recommendations

Resampling and K -Fold Cross-Validation

- The need for multiple training/validation sets
 $\{X_i, V_i\}_i$: Training/validation sets of fold i
- K -fold cross-validation: Divide X into k , $X_i, i=1, \dots, K$

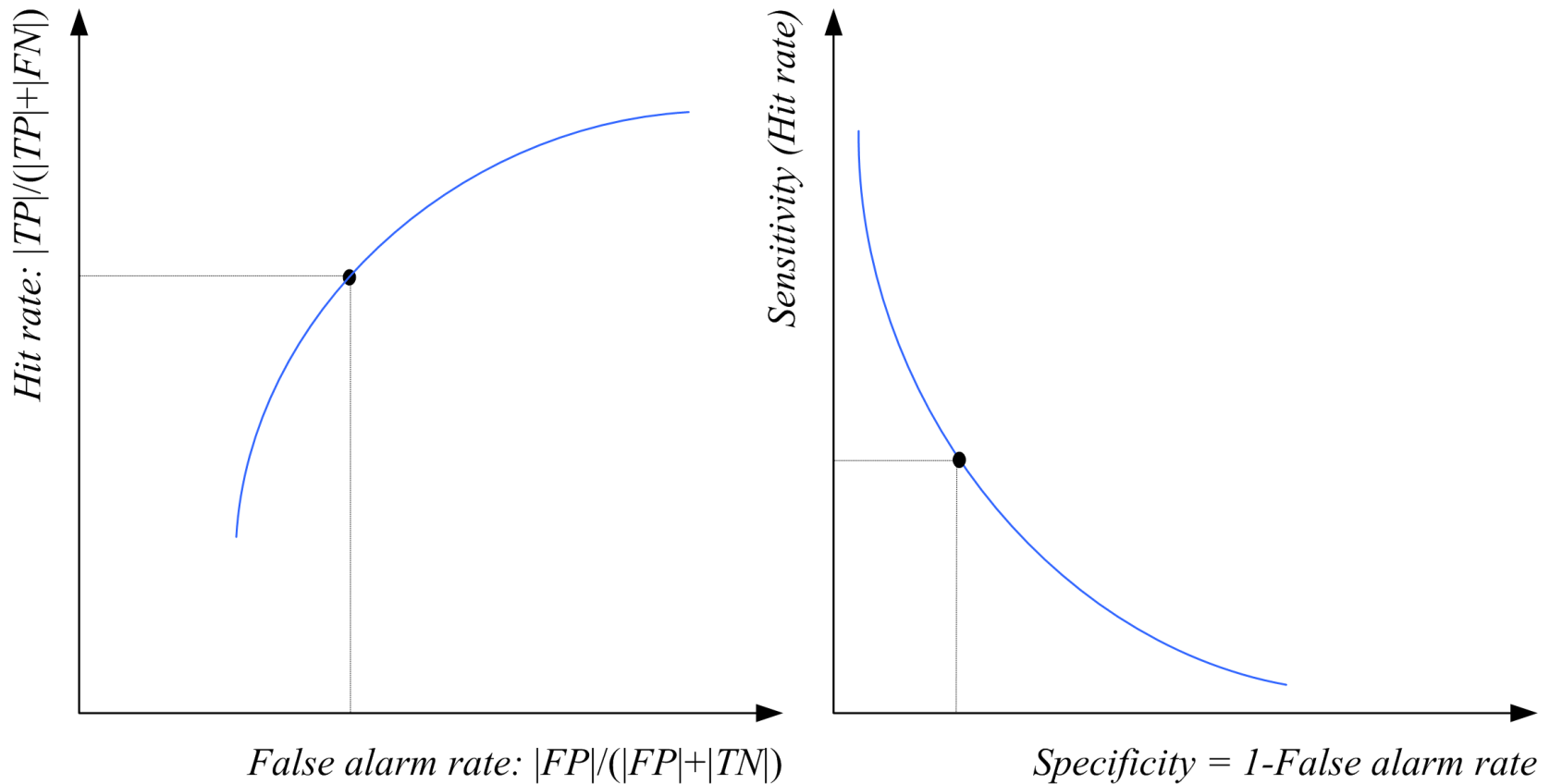
$$\begin{aligned} \mathcal{V}_1 &= \mathcal{X}_1 & \mathcal{T}_1 &= \mathcal{X}_2 \cup \mathcal{X}_3 \cup \dots \cup \mathcal{X}_K \\ \mathcal{V}_2 &= \mathcal{X}_2 & \mathcal{T}_2 &= \mathcal{X}_1 \cup \mathcal{X}_3 \cup \dots \cup \mathcal{X}_K \\ &\vdots & & \\ \mathcal{V}_K &= \mathcal{X}_K & \mathcal{T}_K &= \mathcal{X}_1 \cup \mathcal{X}_2 \cup \dots \cup \mathcal{X}_{K-1} \end{aligned}$$

Performance Measures

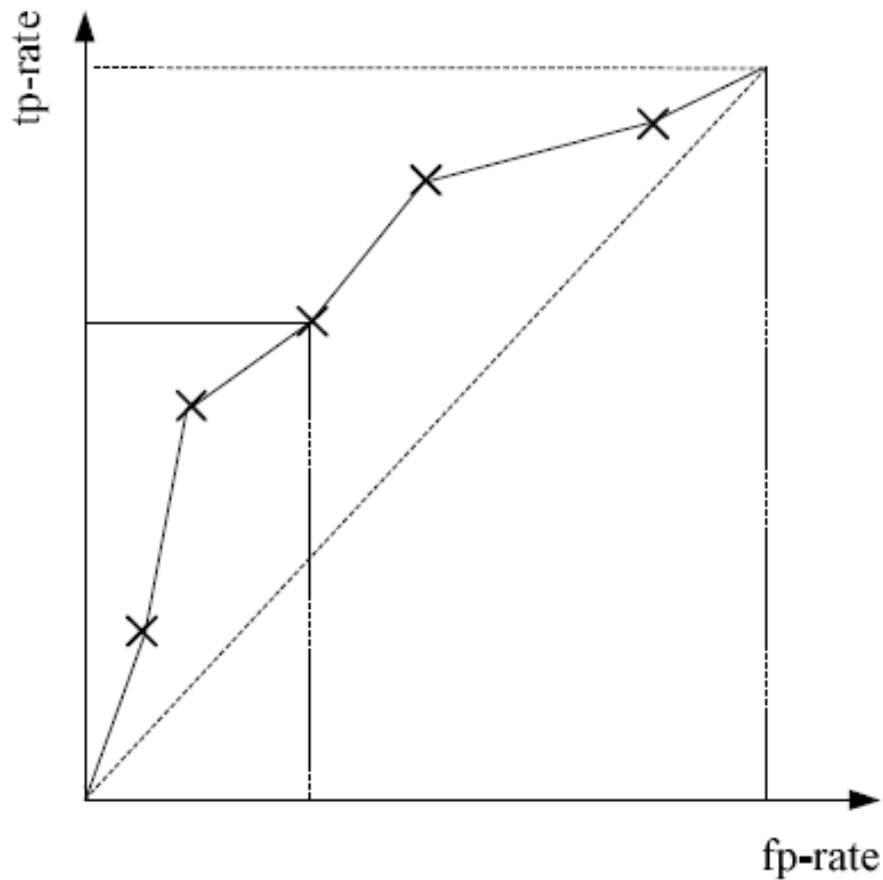
	Predicted class	
True Class	Yes	No
Yes	TP: True Positive	FN: False Negative
No	FP: False Positive	TN: True Negative

- Error rate = # of errors / # of instances = $(FN+FP) / N$
- Recall = # of found positives / # of positives
= $TP / (TP+FN)$ = sensitivity = hit rate
- Precision = # of found positives / # of found
= $TP / (TP+FP)$
- Specificity = $TN / (TN+FP)$
- False alarm rate = $FP / (FP+TN)$ = 1 - Specificity

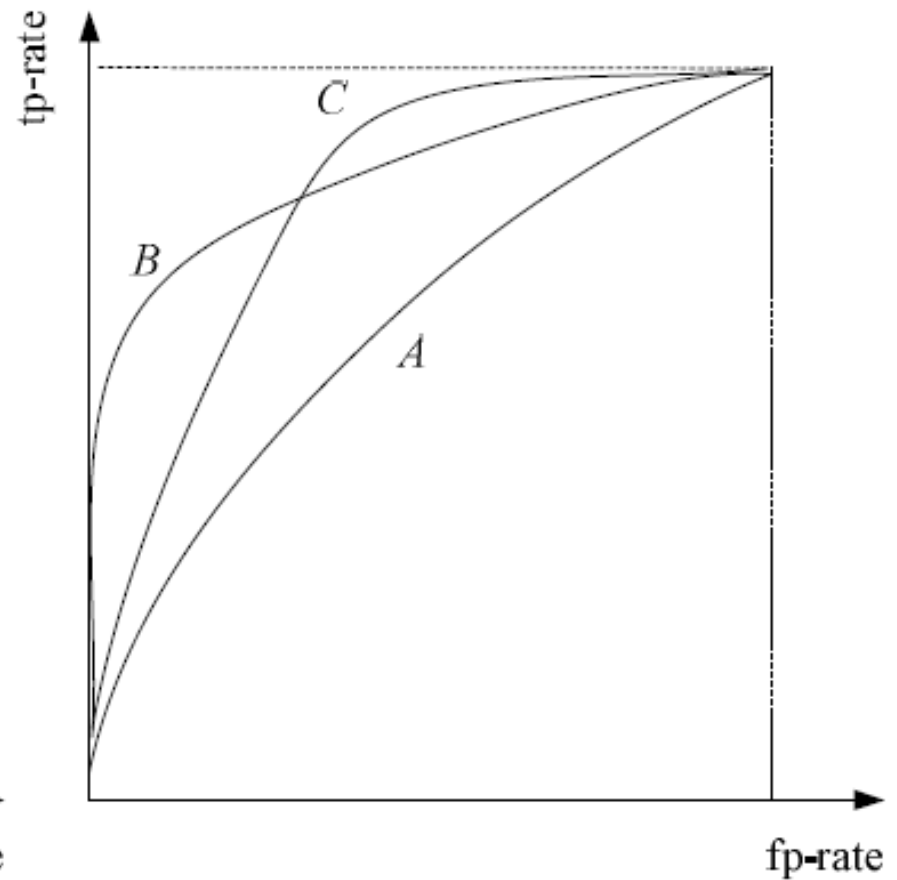
Plotting over Different Decision Thresholds



ROC curve

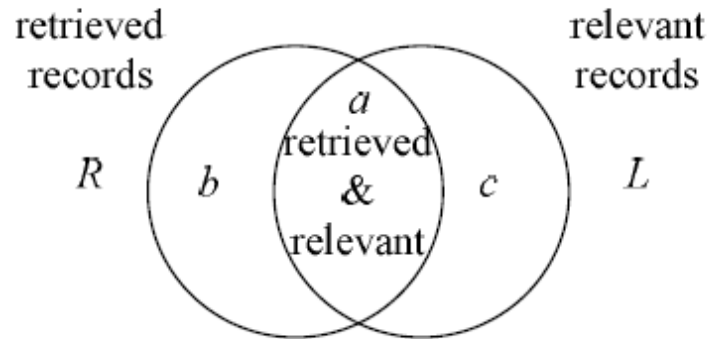


(a) Example ROC curve



(b) Different ROC curves for different classifiers

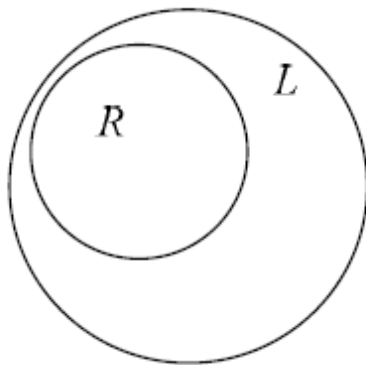
Precision and Recall



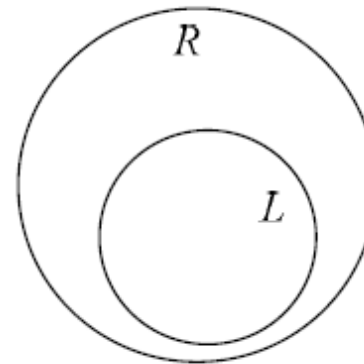
$$\text{Precision: } \frac{a}{a + b}$$

$$\text{Recall: } \frac{a}{a + c}$$

(a) Precision and recall



(b) Precision = 1



(c) Recall = 1