

Section 4M. Receiver Operating Characteristics

Statistics for Data Science

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Threshold Classifier

How do we choose a classifier? We should take into account the cost of a FP versus the cost of a FN (e.g., medical errors). In the medical case, the cost of FNs is much higher than the cost of FPs; hence, we would like to reduce the FNR at expense of increasing the FPR...

- Remember: A (binary) Bayesian classifier satisfies

$$C(\mathbf{x}) = \begin{cases} 1 & \text{when } \Pr(Y = 1|X = \mathbf{x}) \geq 1/2 \\ 0 & \text{when } \Pr(Y = 1|X = \mathbf{x}) < 1/2 \end{cases}$$

This classifier minimizes the *classification error rate*, which is cost optimal only when the cost of a FP is the same as the cost of a FN

- To consider problems in which these costs are different, we define a classifier parameterized by a threshold $\tau \in [0, 1]$

$$C_{\tau}(\mathbf{x}) = \begin{cases} 1 & \text{when } \Pr(Y = 1|X = \mathbf{x}) \geq \tau \\ 0 & \text{when } \Pr(Y = 1|X = \mathbf{x}) < \tau \end{cases}$$

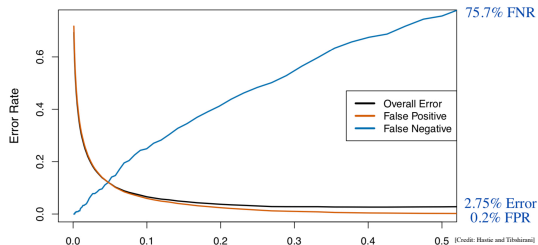
Threshold Classifier: Example

How do FNR and FPR change as a function of $\tau \in (0, 1/2)$ for the LDA classifier applied to the Default dataset?

- In the LDA binary classifier, we have that $C_\tau(\mathbf{x}) = 1$ when (Bayes' theorem)

$$\Pr(Y = 1|X = \mathbf{x}) = \frac{f_1(\mathbf{x}) \pi_1}{f_0(\mathbf{x}) \pi_0 + f_1(\mathbf{x}) \pi_1} \geq \tau$$

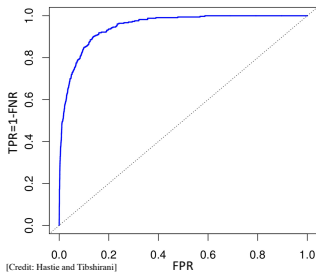
- The evolution of different error rates as we vary τ are shown below



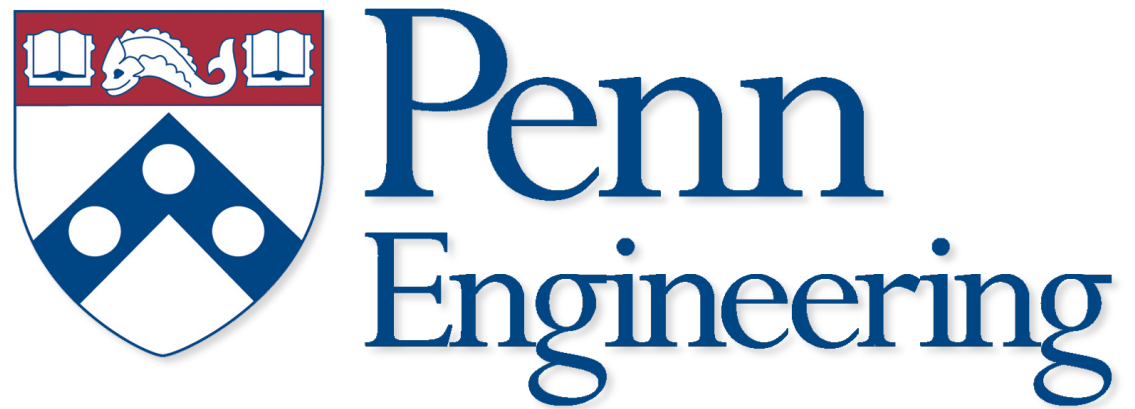
Receiver Operating Characteristics

As we have seen before, we can choose τ such that our classifier obtains a lower number of FNs by sacrificing FPs. It is convenient to plot the relationship between FNR and FPR in a plot...

- **Receiver Operating Characteristics (ROC):** For historical reasons, this relationship is typically plotted in terms of FPR and $\text{TPR}=1-\text{FNR}$.



The Area Under the ROC Curve (AUC) summarizes the overall performance of the parameterized classifier. A better classifier results in a higher AUC.



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