

Section 4L. Confusion Matrix

Statistics for Data Science

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Confusion Matrix

The ***confusion matrix*** is defined as

	$y_i = 0$	$y_i = 1$	
$C(\mathbf{x}_i) = 0$	$ TN $	$ FN $	$ TN + FN $
$C(\mathbf{x}_i) = 1$	$ FP $	$ TP $	$ FP + TP $
	$ TN + FP $	$ FN + TP $	N

Confusion Matrix: LDA Practice

We train an LDA classifier $C_{\text{LDA}}(x)$ using the Default dataset and show results below:

- ▶ The confusion matrix takes the form

	$y_i = 0$	$y_i = 1$	
$C(\mathbf{x}_i) = 0$	9644	252	9896
$C(\mathbf{x}_i) = 1$	23	81	104
	9667	333	10000

- ▶ The classification error rate (in the training dataset) is

$$\text{Err}_{\text{LDA}} = \frac{23 + 252}{10000} = 2.75\%$$

- ▶ Is this a good classification rate? For example, the *null classifier* $C_{\text{Null}}(x_i) = 0$ has $|FP| = 0$ and $|FN| = 333$; hence,

$$\text{Err}_{\text{Null}} = \frac{333}{10000} = 3.33\%$$

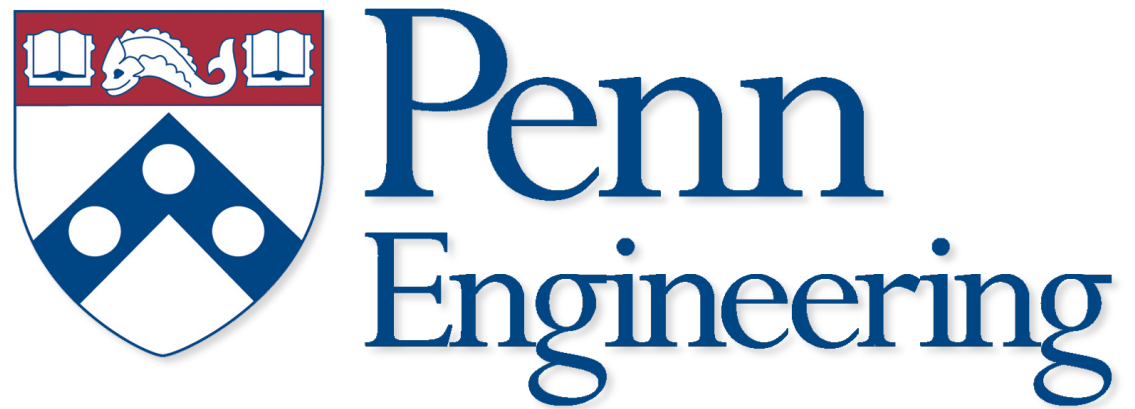
Classification Error Rates

Observations on the LDA classifier

- ▶ For those samples for which $y_i = 0$, we make an error rate of $23/9667 = 0.2\%$
- ▶ However, for those samples for which $y_i = 1$, our error rate is $252/333 = 75.7\%$

Hence, the classification error rate alone is not a good measure of performance...

- ▶ We can define more refined versions of the classification rate:
 - ▶ *True Positive Rate*: $TPR = |TP| / (|TP| + |FN|)$
 - ▶ *True Negative Rate*: $TNR = |TN| / (|TN| + |FP|)$
 - ▶ *False Positive Rate*: $FPR = |FP| / (|FP| + |TN|) = 1 - TNR$
 - ▶ *False Negative Rate*: $FNR = |FN| / (|FN| + |TP|) = 1 - TPR$



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