

— Confusion Matrices



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**What's your “muddiest point”
surrounding confusion matrices?**
Come off mic!





Solo Exercise:

Confusion Matrices

1. Assume Beatles is “positive” and Queen is “negative”. Label this confusion matrix with TP, TN, FP, FN.

	Predicted Queen	Predicted Beatles
Actual Queen	5657	794
Actual Beatles	1116	5452

2. Calculate the *specificity*. Interpret it.
3. Calculate the *sensitivity (recall)*. Interpret it.
4. Calculate the *accuracy*. Interpret it.



Solo Exercise:

Confusion Matrices

	Predicted Queen	Predicted Beatles
Actual Queen	5657	794
Actual Beatles	1116	5452

1. Which number represents our Type I errors?
2. Which number represents our Type II errors?
3. Which type of error is worse *in this case*?

****BONUS**** Calculate the false positive and false negative rates!



Solo Exercise:

SOLUTIONS

1. TN FP - FN TP
 - $tp = 5452$
 - $fn = 1116$
 - $tn = 5657$
 - $fp = 794$
2. Calculate specificity (TN / All N) -- $5657 / (5657 + 794) = 0.8769183072391877$
 - Our model correctly predicted 88% of the posts belonging to the Queen subreddit.
3. Calculate sensitivity (TP / All P) -- $5452 / (5452 + 1116) = .0.8300852618757613$
 - Our model correctly predicted 83% of the posts belonging to the Beatles subreddit.



1. Calculate accuracy $(TP+TN) / All \rightarrow (5657 + 5452) / (5657 + 5452 + 1116 + 794) = 0.8532913434211536$
 - Our model got 85% of our predictions correct.
2. Type I error: False positive $\rightarrow 794$
 - False positive rate = $1 - \text{specificity} = 12\%$
 - Our model incorrectly predicted 12% of the posts belonging to the Queen subreddit
3. Type II error: False negative $\rightarrow 1116$
 - False negative rate = $1 - \text{sensitivity}$
 - Our model incorrectly predicted 17% of the posts belonging to the Beatles subreddit.

Confusion Matrix and ROC Curve

		Predicted Class	
		No	Yes
Observed Class	No	TN	FP
	Yes	FN	TP

TN True Negative
FP False Positive
FN False Negative
TP True Positive

Model Performance

Accuracy $= (TN+TP)/(TN+FP+FN+TP)$

Precision $= TP/(FP+TP)$

Sensitivity $= TP/(TP+FN)$

Specificity $= TN/(TN+FP)$

SeNsitivity

When it is actually "yes" how often it predicts "Yes"

$TP / All\ P$

$TP / (TP + FN)$

SPecificity

When it is actually "No", how often it is "No"

$TN / All\ N$

$TN / (TN + FP)$

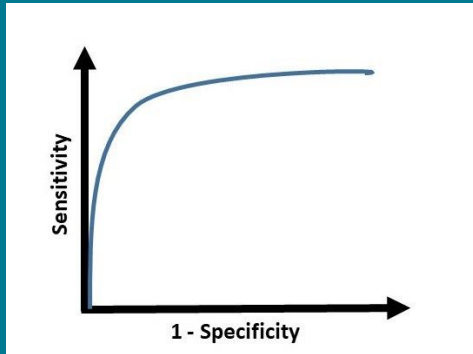


Type I Error

I False Positive Rate

1 - specificity

***the x-axis of ROC-AUC curve!
(the y-axis is the sensitivity AKA True Positive Rate!)



Type II Error

II False Negative Rate

1 - sensitivity

Hypothesis Testing

Type I Error = incorrectly reject the null hypothesis and accept H_A
***worse, so threshold is .05 generally

Type II Error = incorrectly accept the null hypothesis and reject H_A

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**"Everything should be as simple as it
can be, but not simpler."**

-Einstein

