

# Receiver Operating Characteristic (ROC) Curves

**Statistics for proteomics**

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William E Fondrie

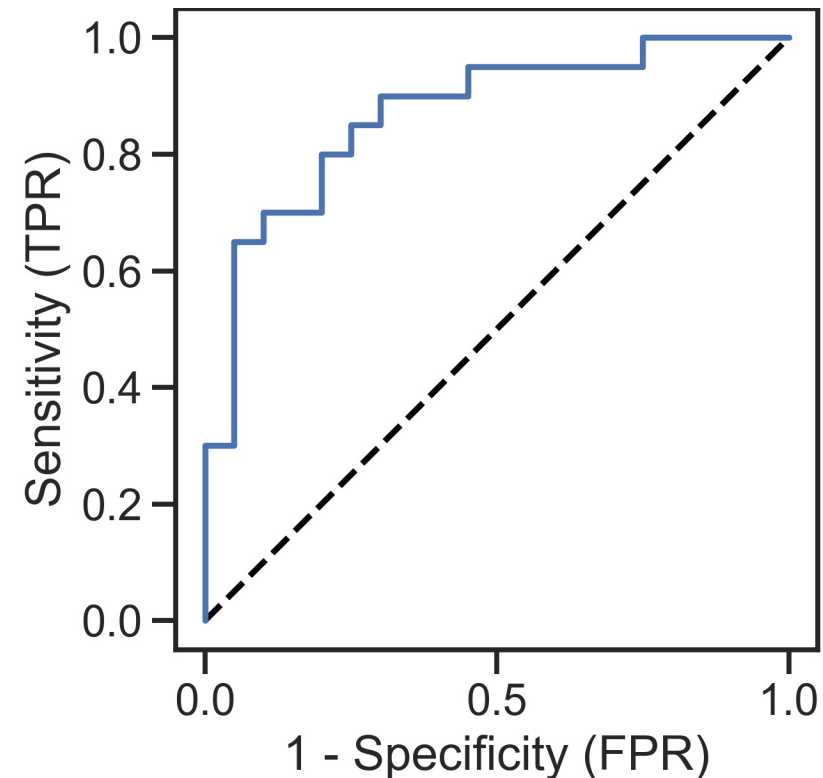
University of Washington

<https://willfondrie.com>



# What are receiver operating characteristic (ROC) curves?

- First used in WWII for the analysis of radio signals
- Assess performance at a binary classification task:
  - Does this patient have this disease?
  - Is this tissue cancerous?
  - Do these two proteins interact?

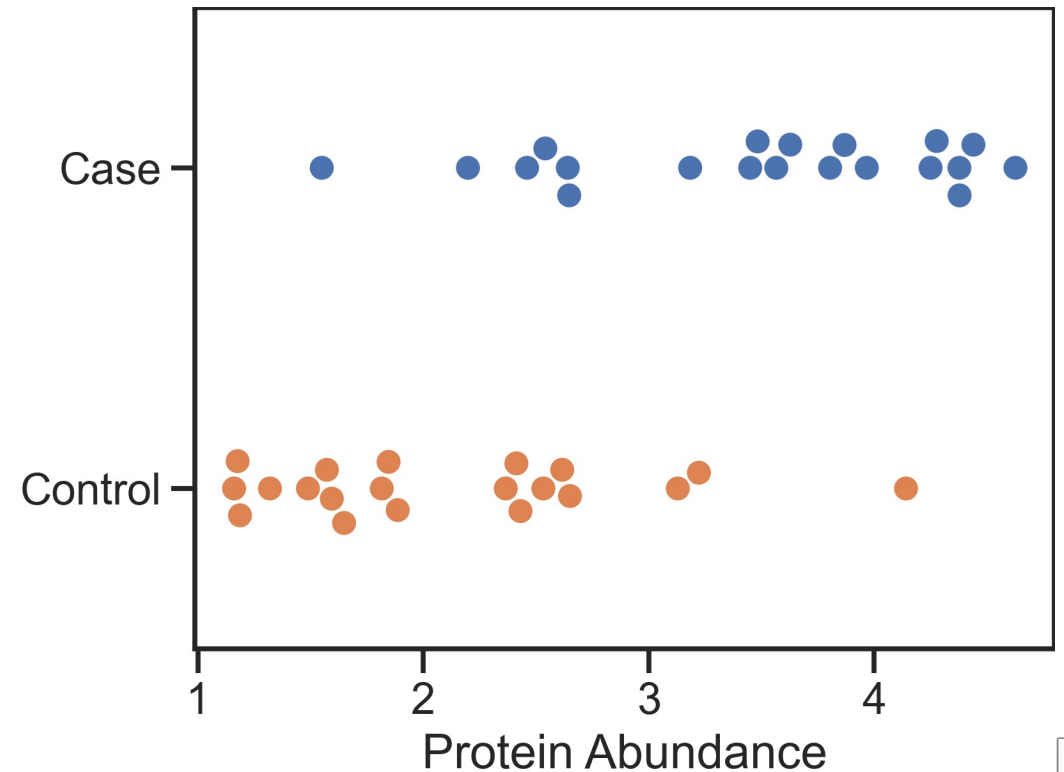


# Our example binary classification task

Predict whether a patient has a disease from the abundance of a single protein biomarker.

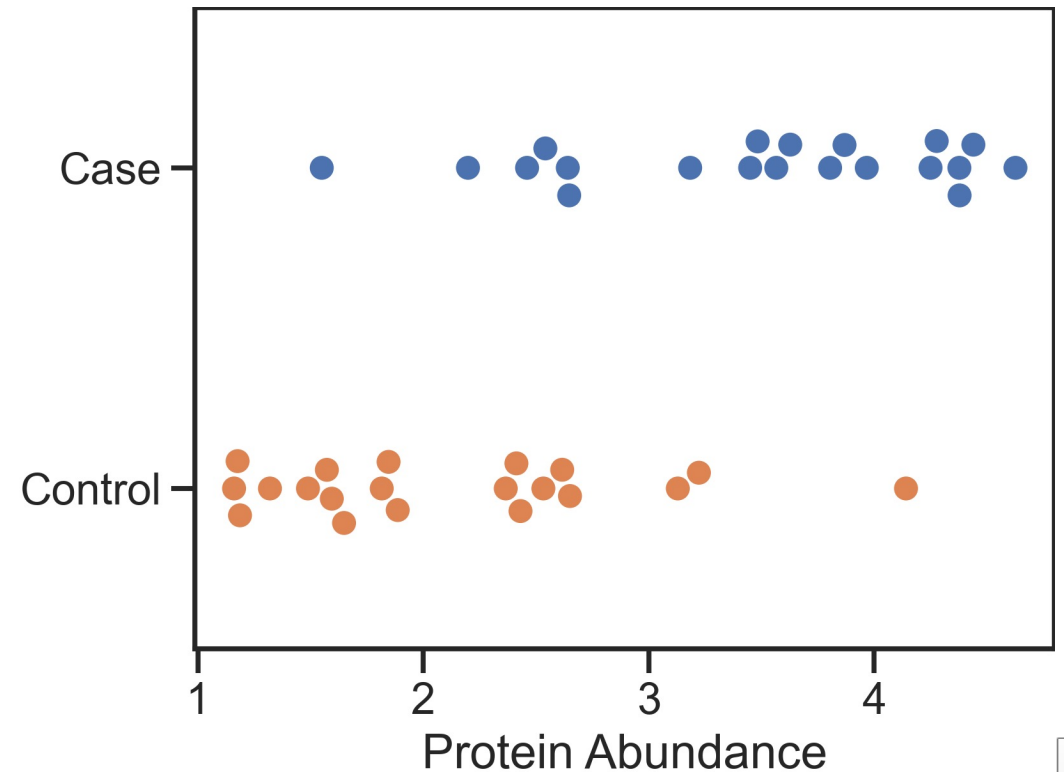
- Measurements from 20 cases and 20 controls

**How well does our biomarker perform?**



# Sensitivity and specificity are useful metrics

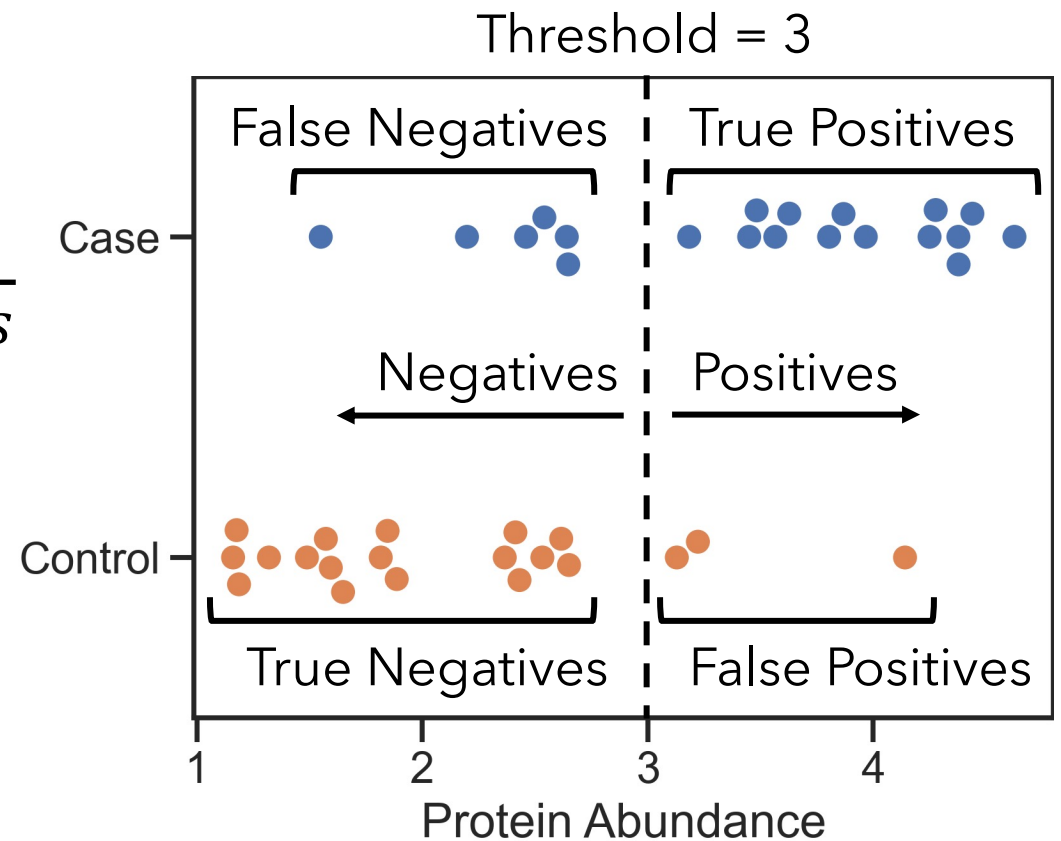
- Sensitivity is the proportion of positives (cases) that are correctly identified.
- Specificity is the proportion of negatives (controls) that are correctly identified.



# How do we calculate sensitivity and specificity?

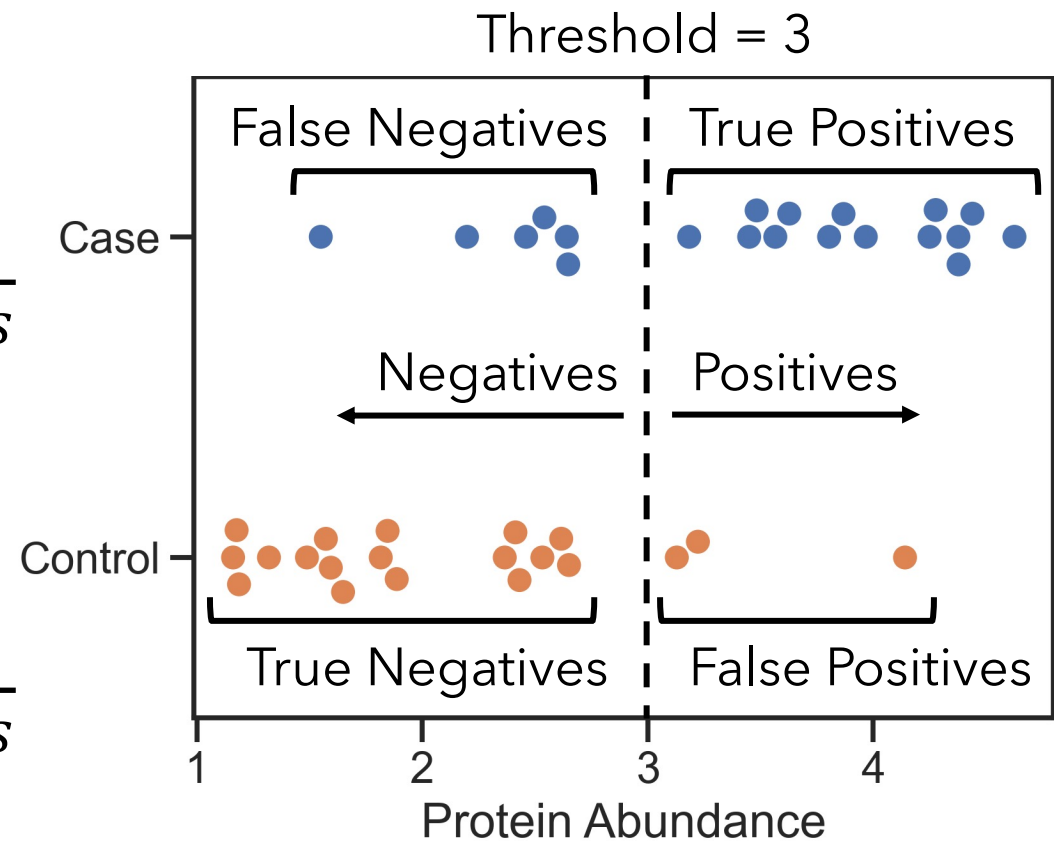
- Sensitivity

$$= \frac{\text{True Positives}}{\text{True Positives} + \text{False Negatives}}$$



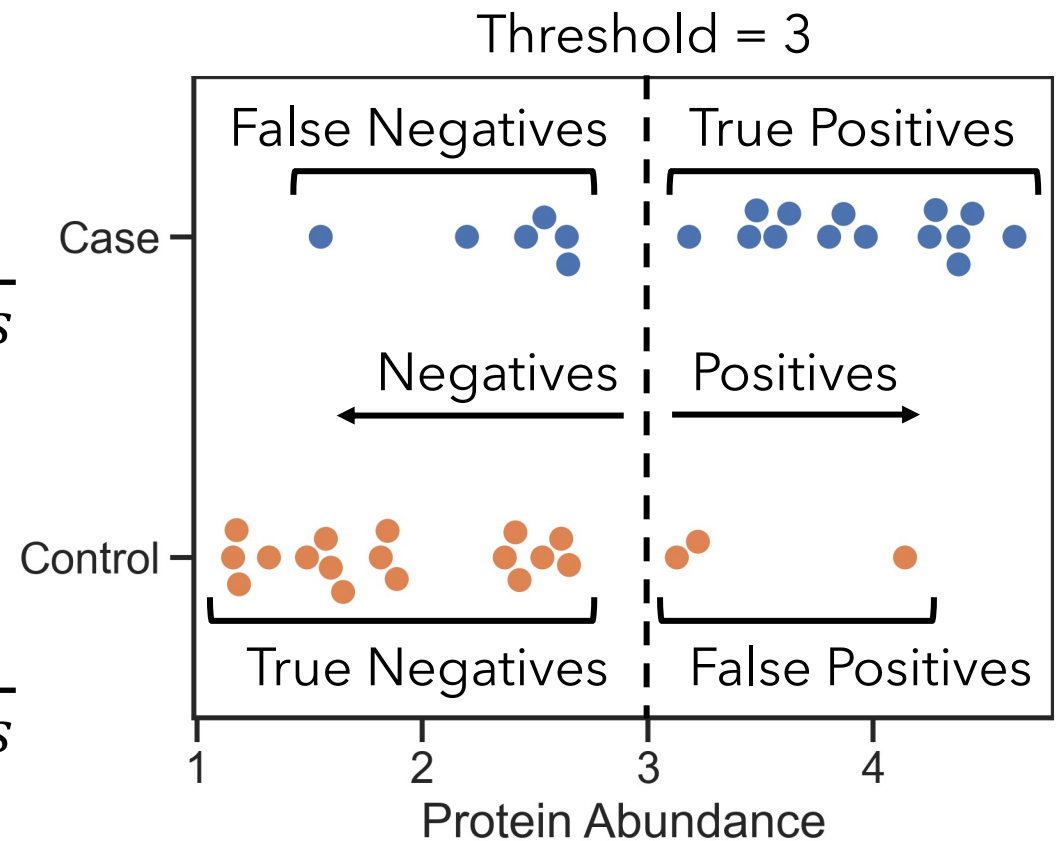
# How do we calculate sensitivity and specificity?

- Sensitivity =  $14 / (14 + 6) = 0.70$   
$$= \frac{\text{True Positives}}{\text{True Positives} + \text{False Negatives}}$$
- Specificity  
$$= \frac{\text{True Negatives}}{\text{True Negatives} + \text{False Positives}}$$

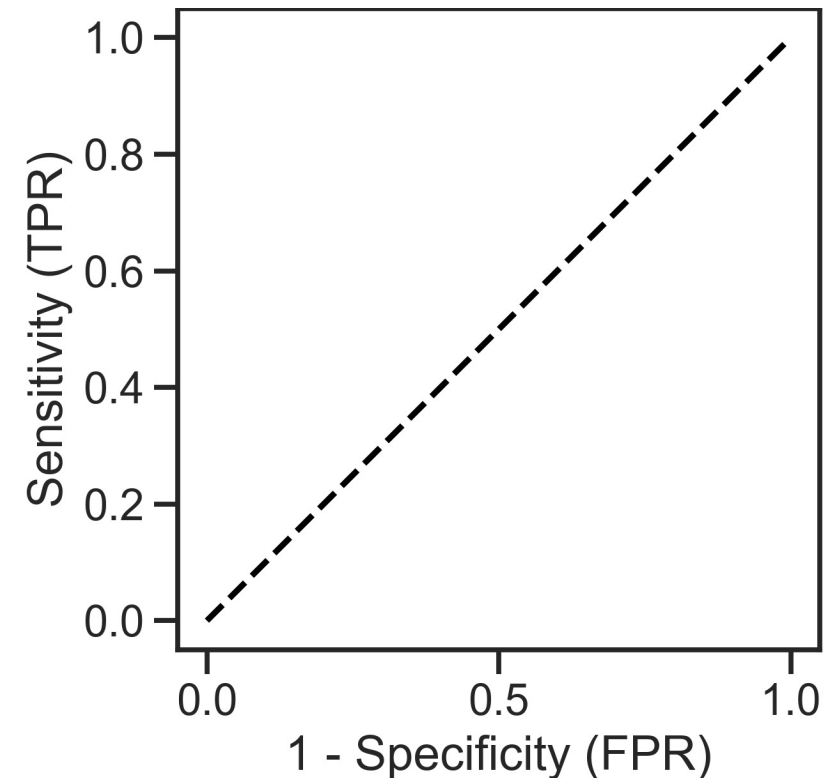
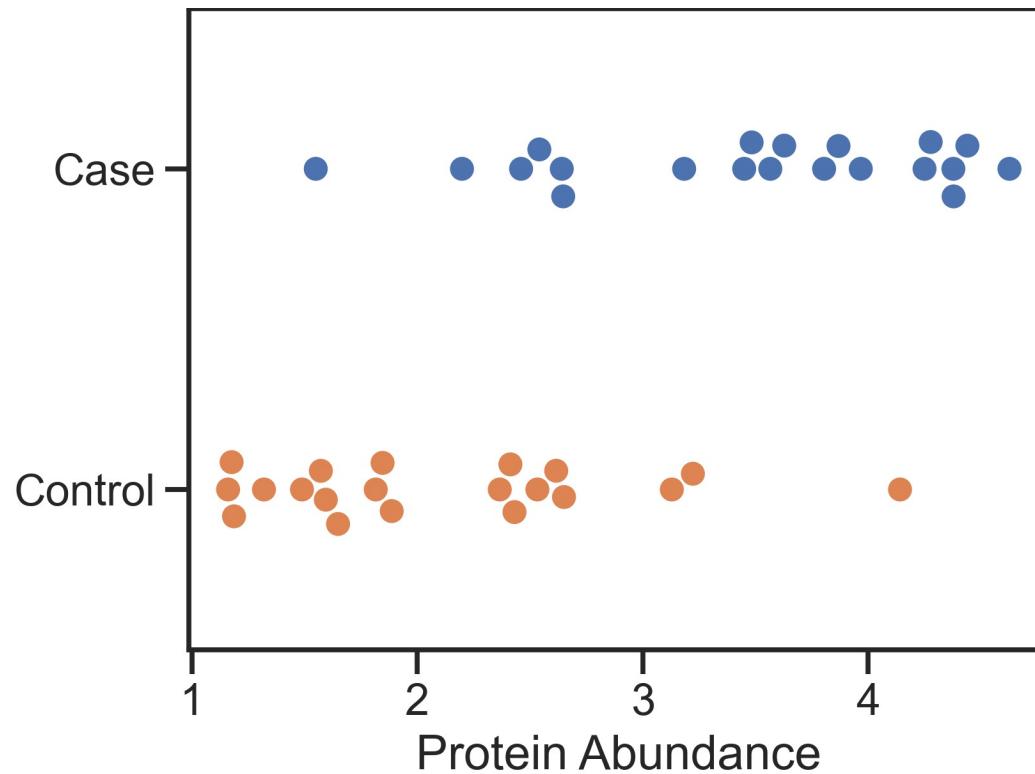


# How do we calculate sensitivity and specificity?

- Sensitivity =  $14 / (14 + 6) = 0.70$   
$$= \frac{\text{True Positives}}{\text{True Positives} + \text{False Negatives}}$$
- Specificity =  $17 / (17 + 3) = 0.85$   
$$= \frac{\text{True Negatives}}{\text{True Negatives} + \text{False Positives}}$$



ROC curves show the sensitivity and specificity across all thresholds

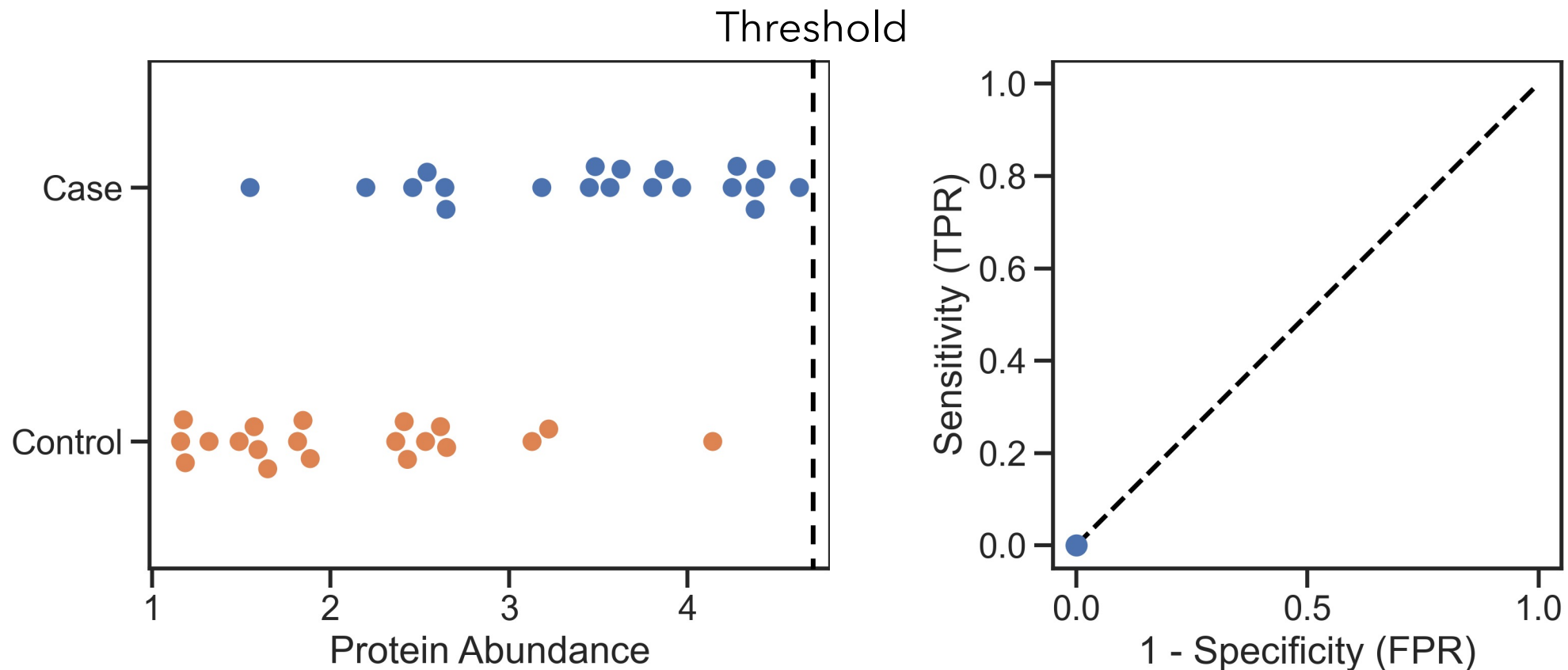


TPR = True Positive Rate, FPR = False Positive Rate





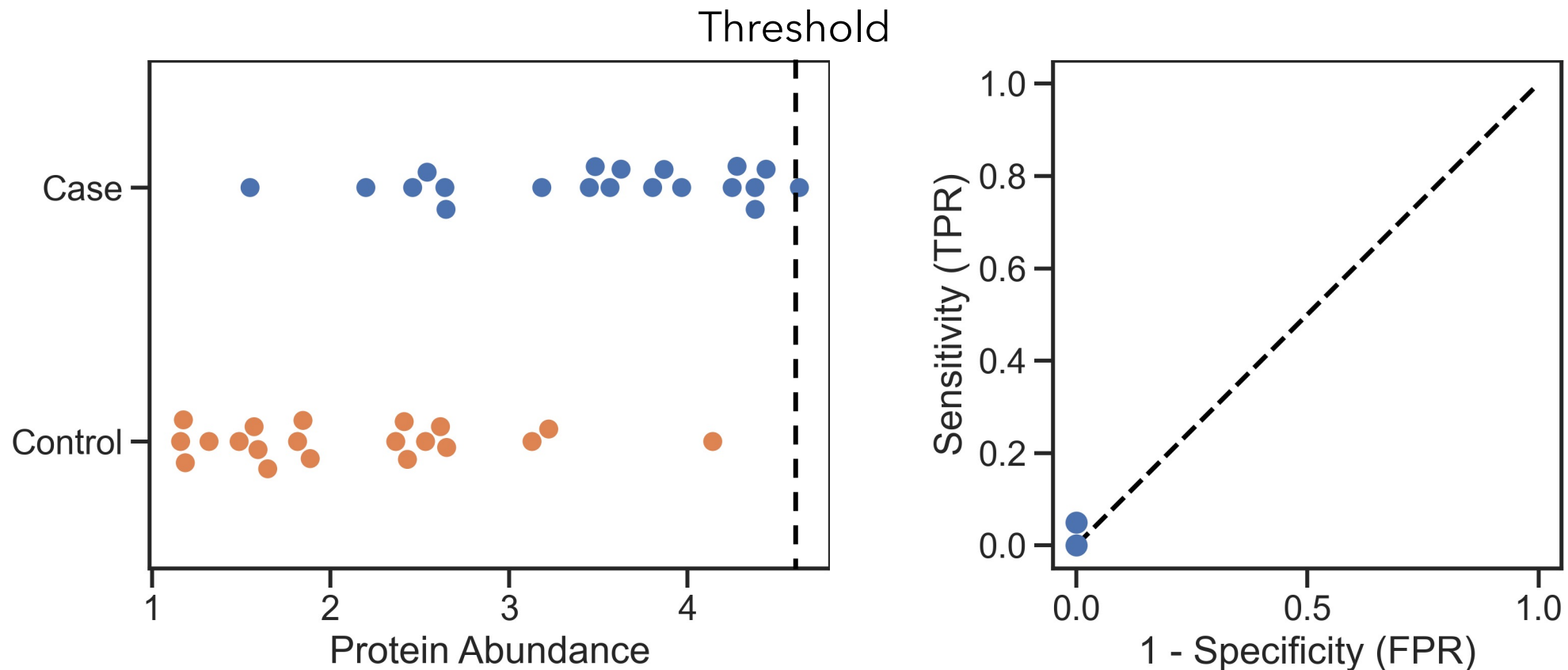
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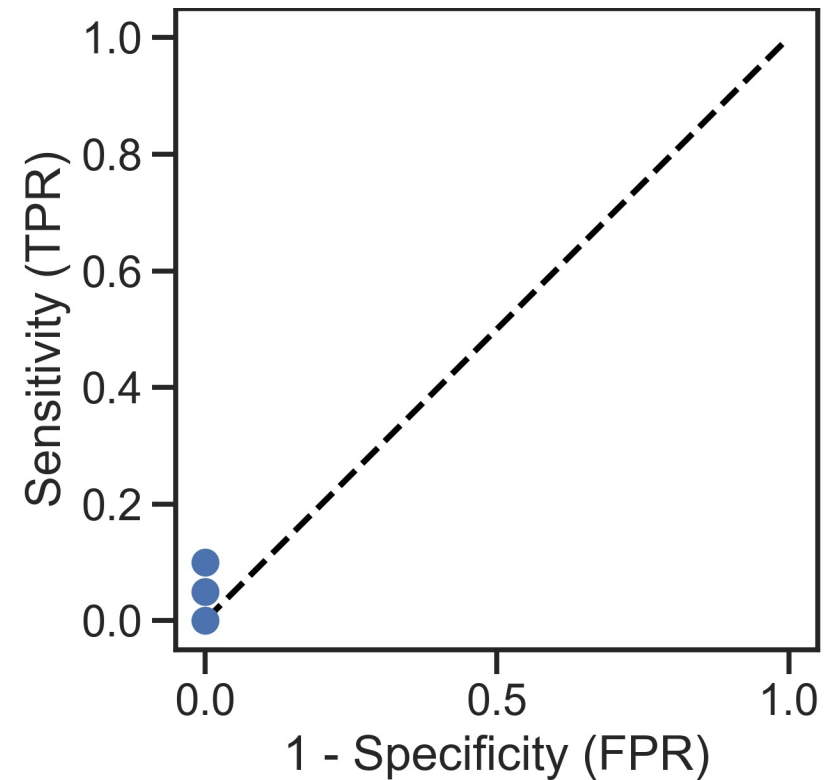
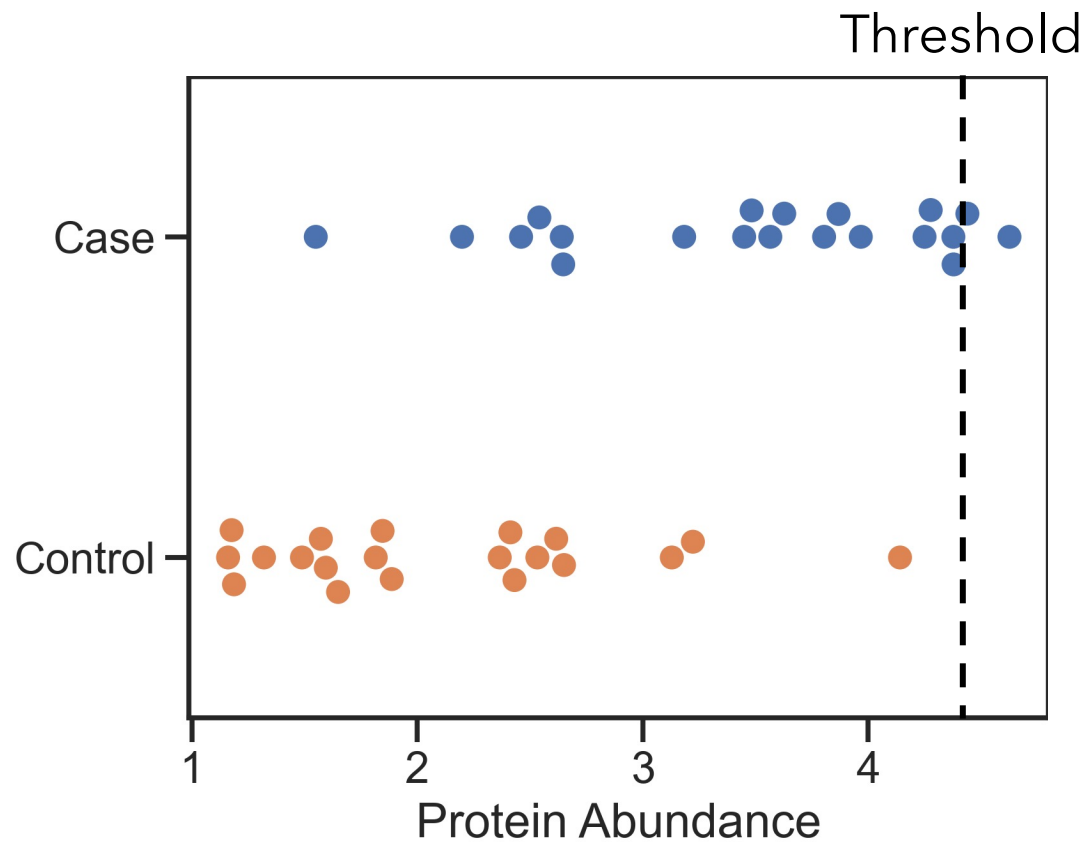
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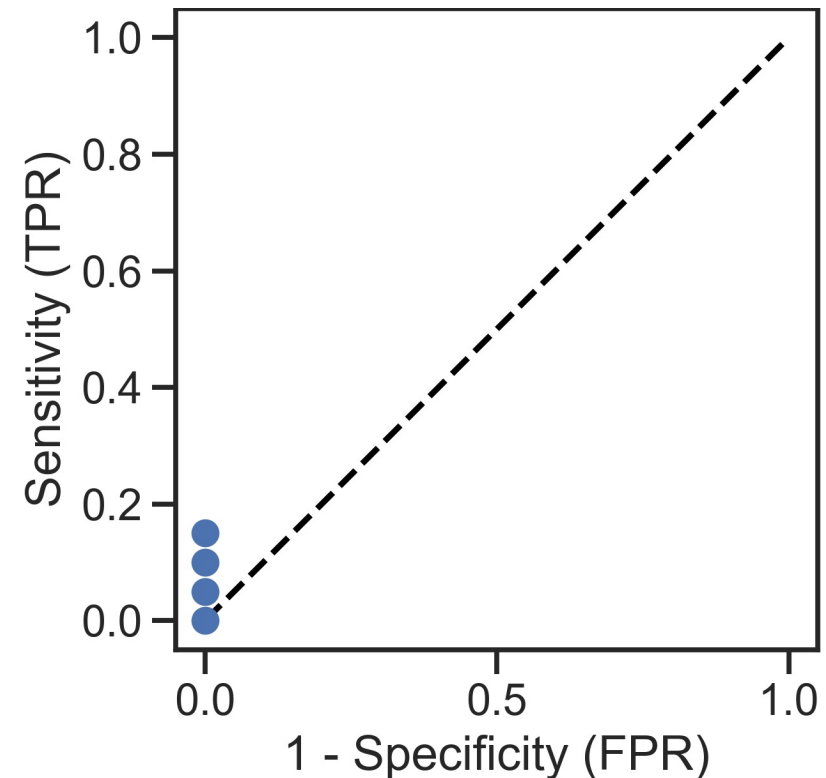
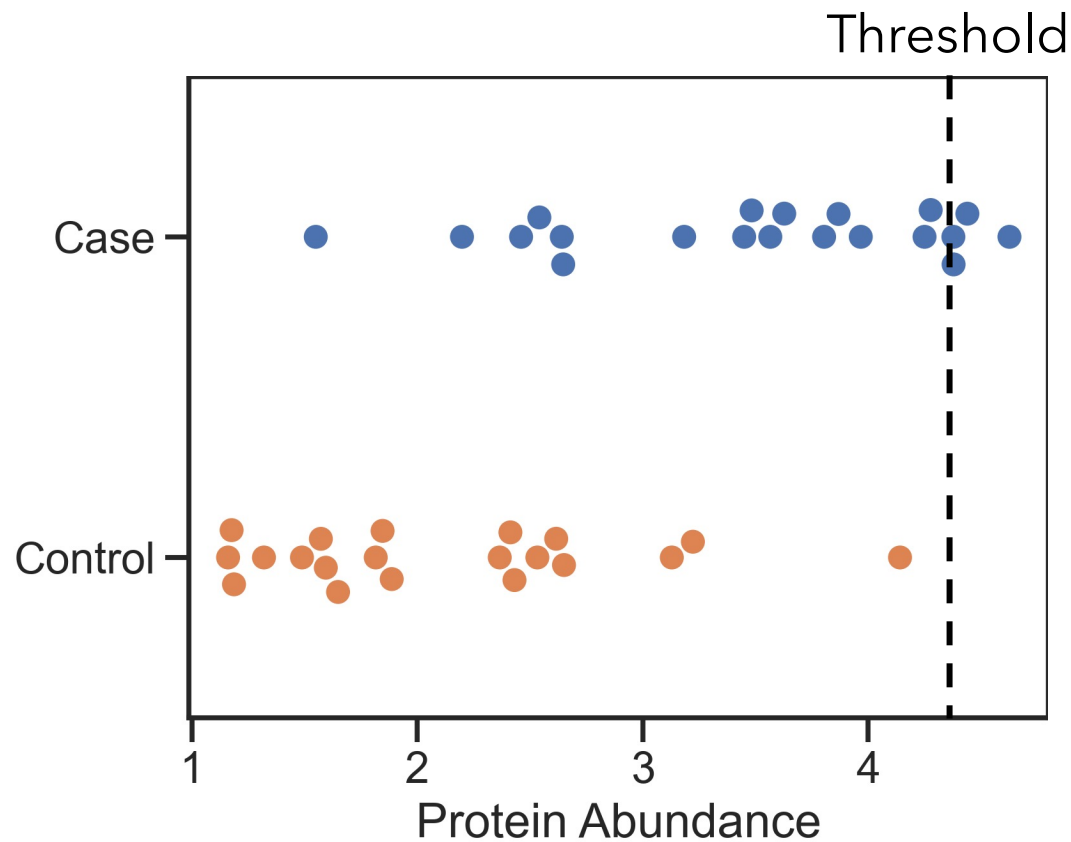
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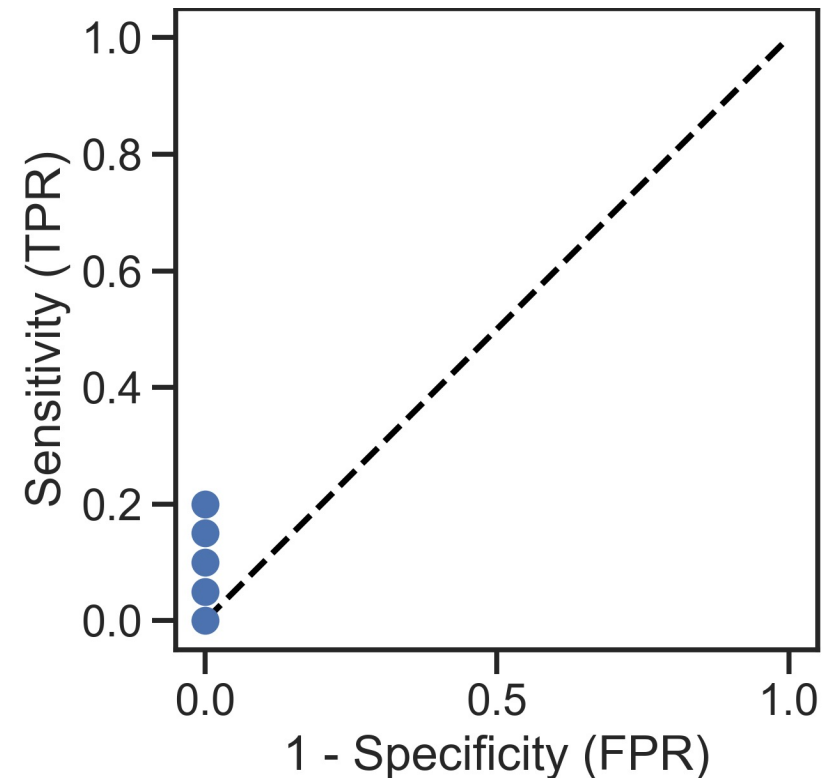
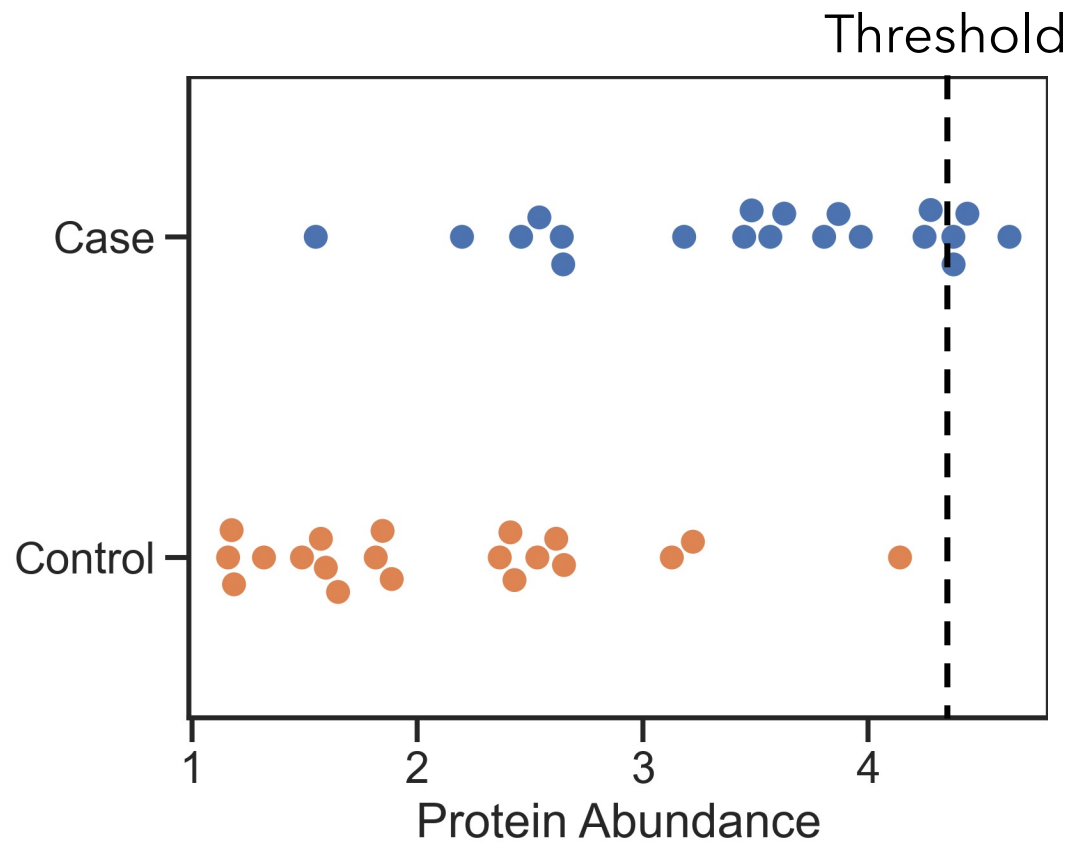
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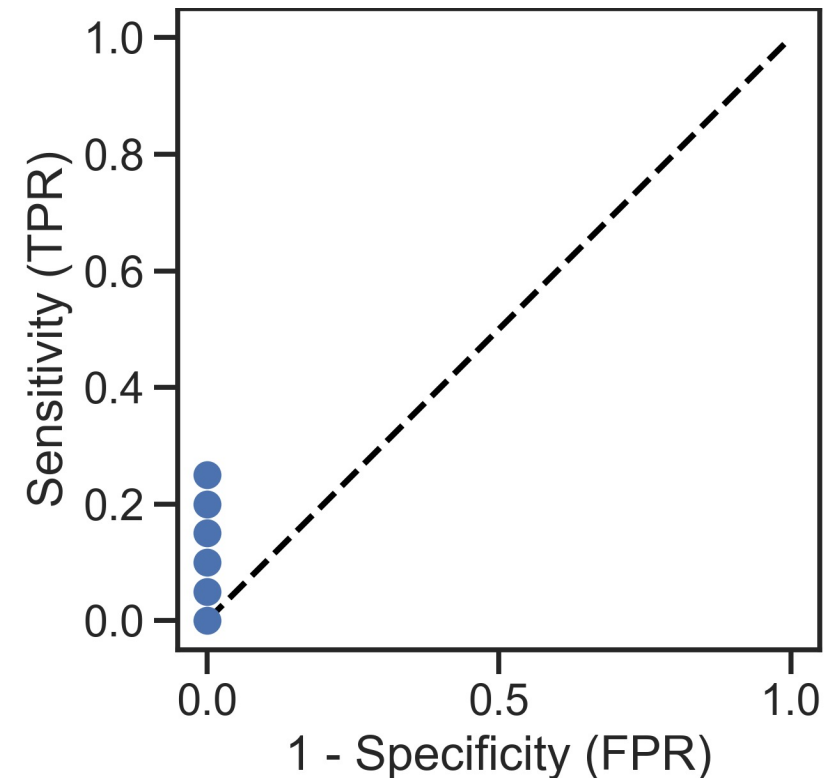
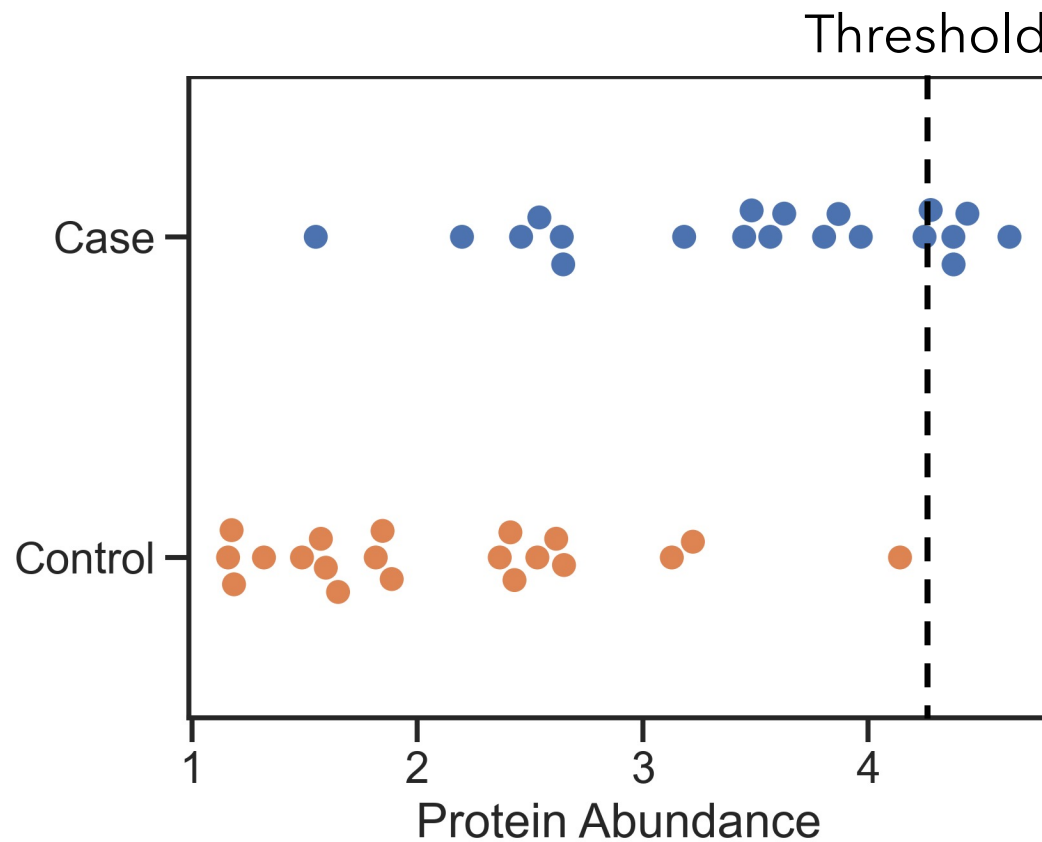
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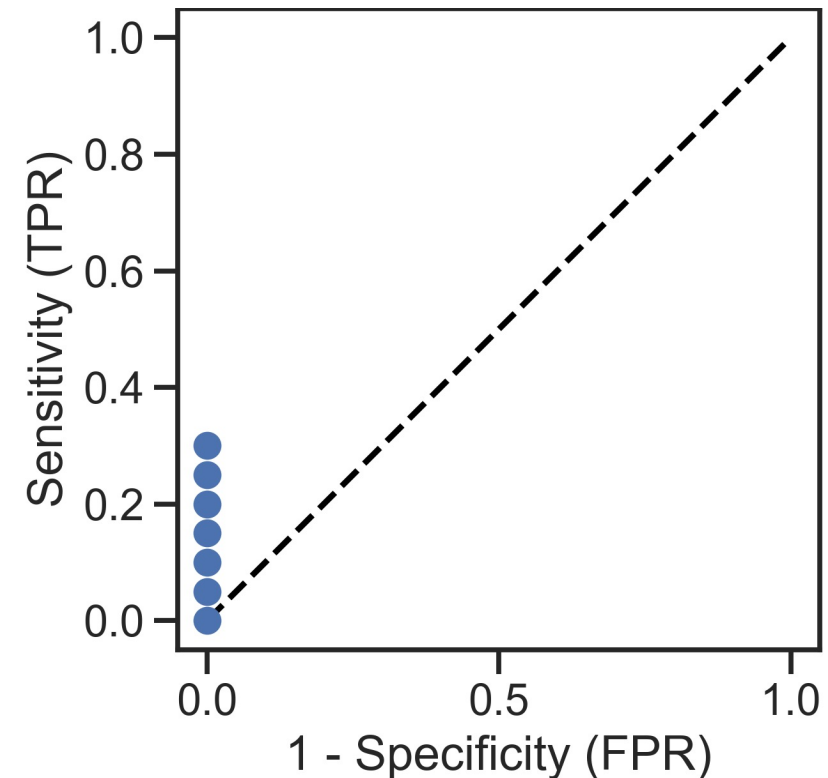
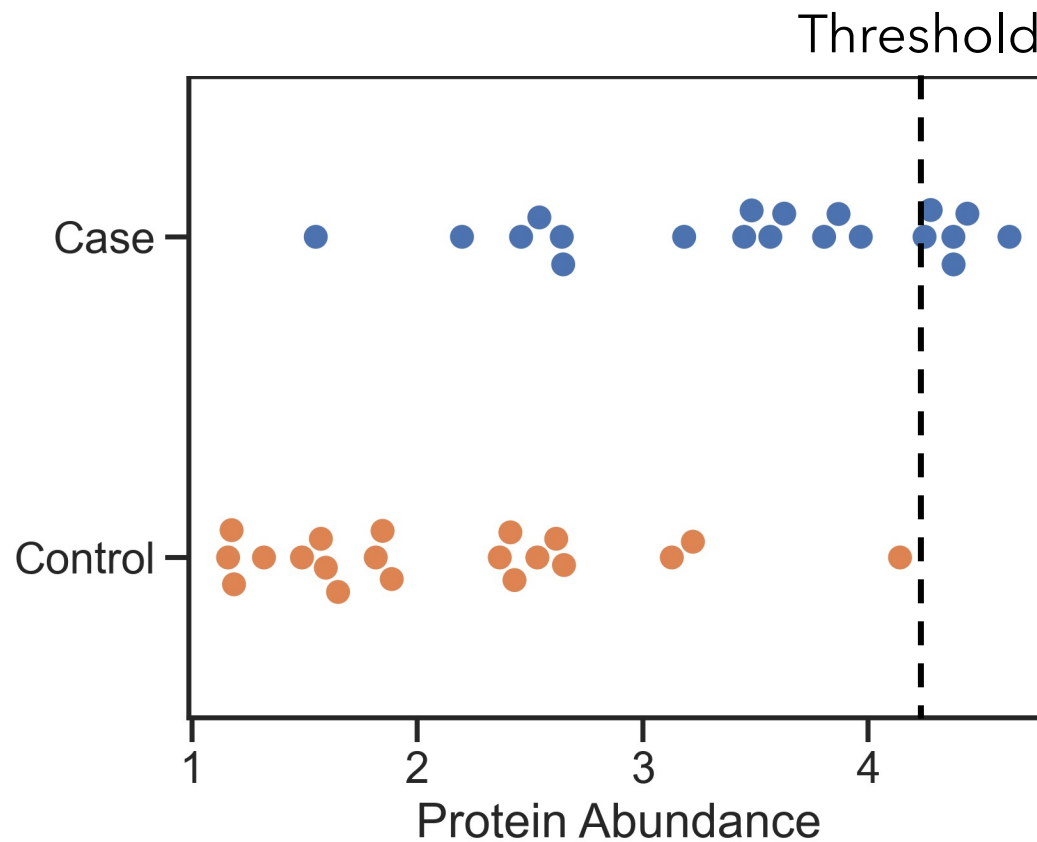
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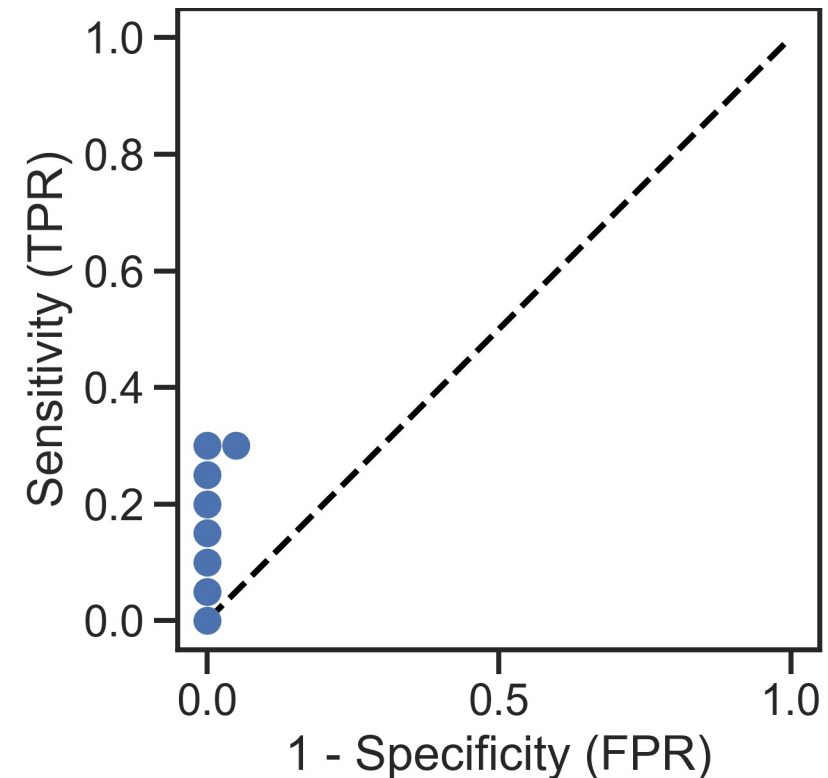
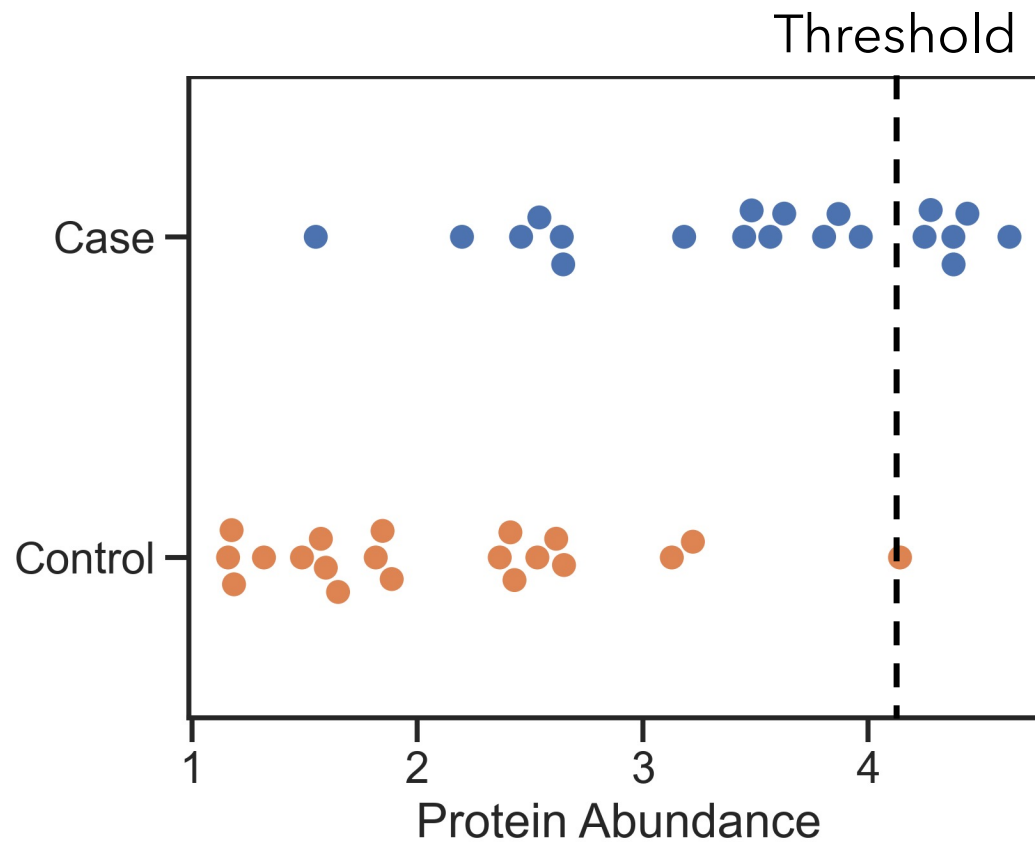
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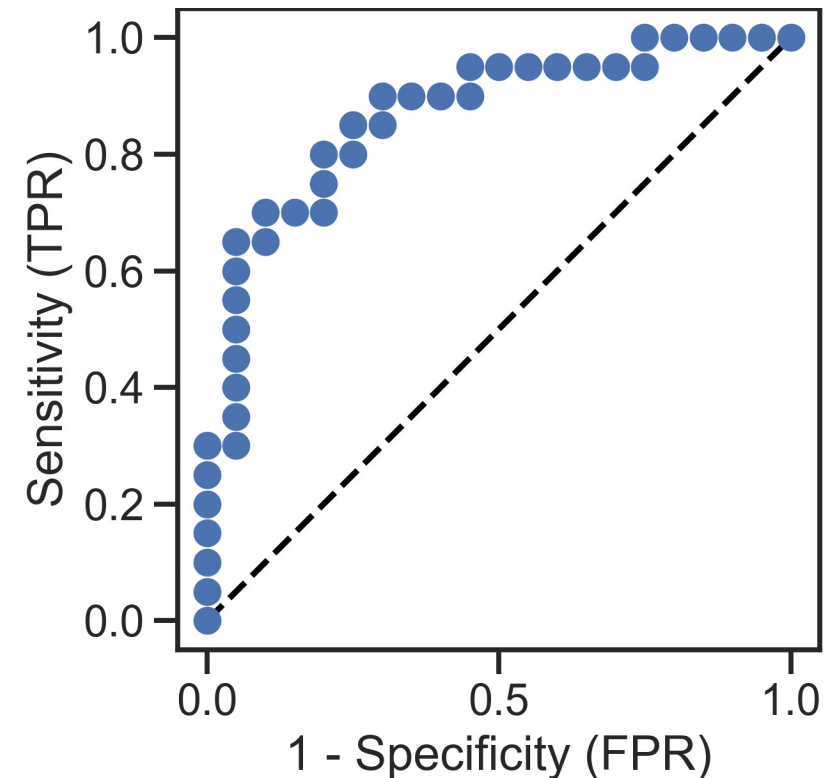
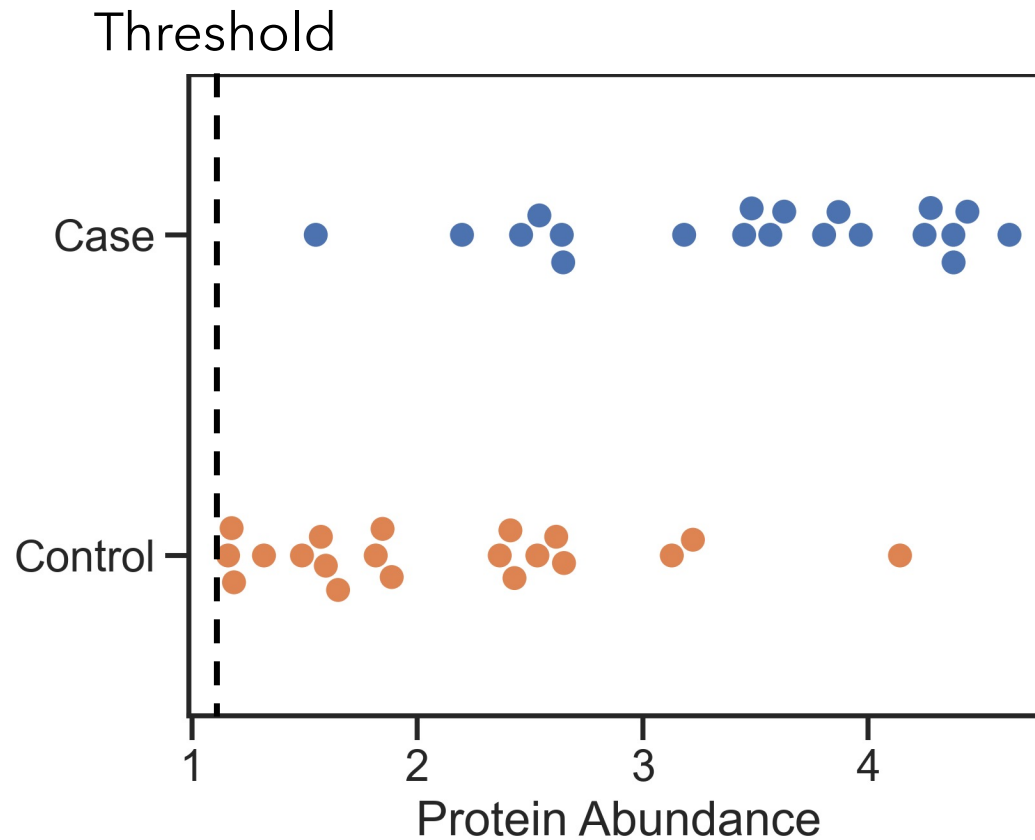


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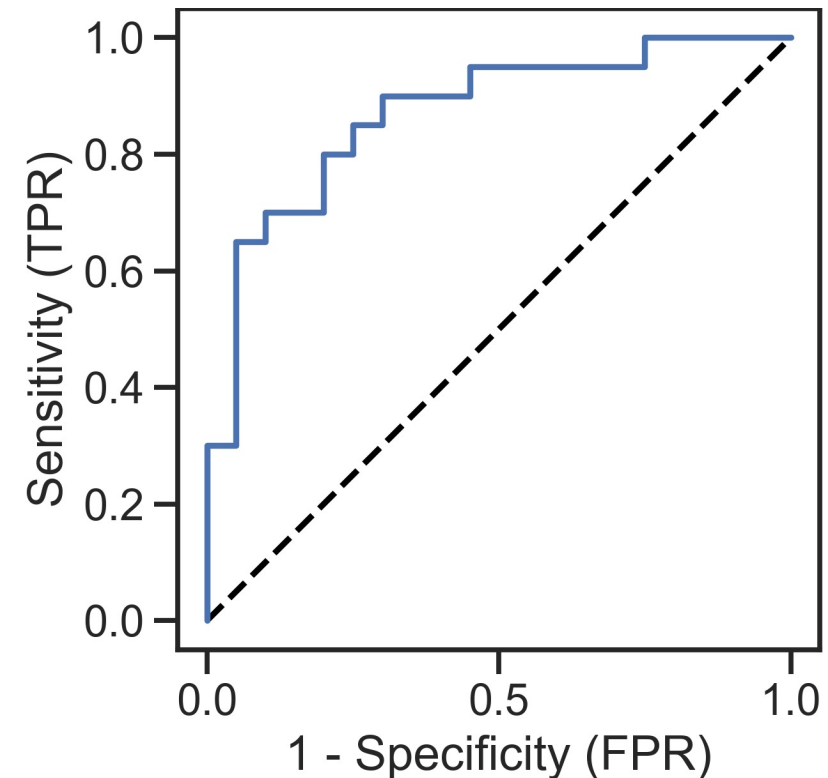
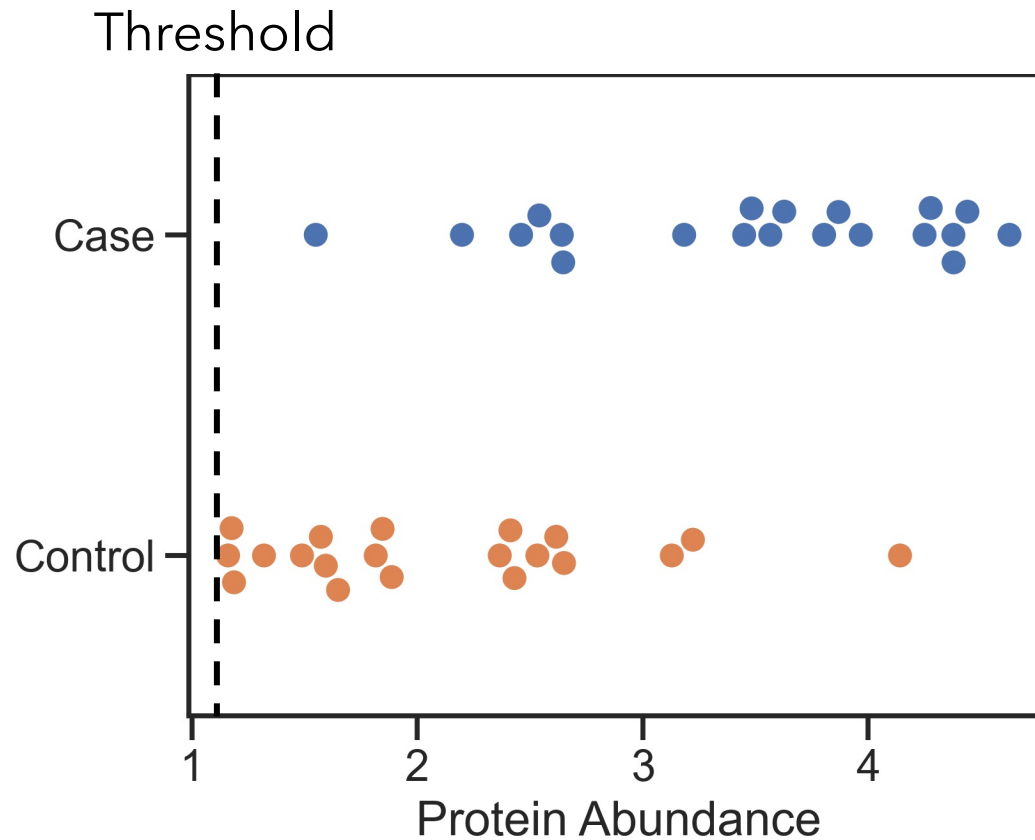
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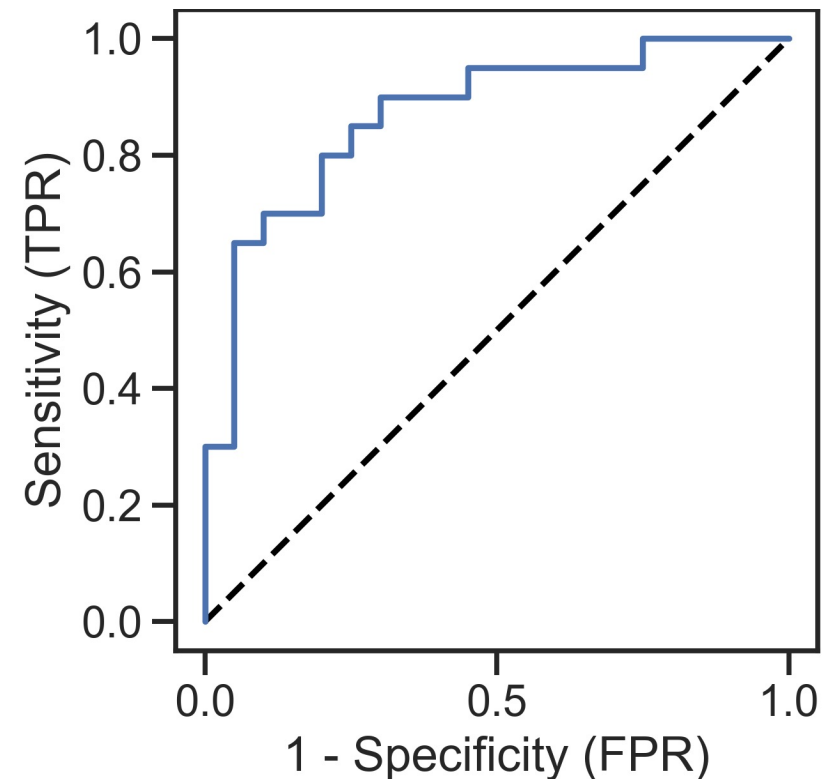


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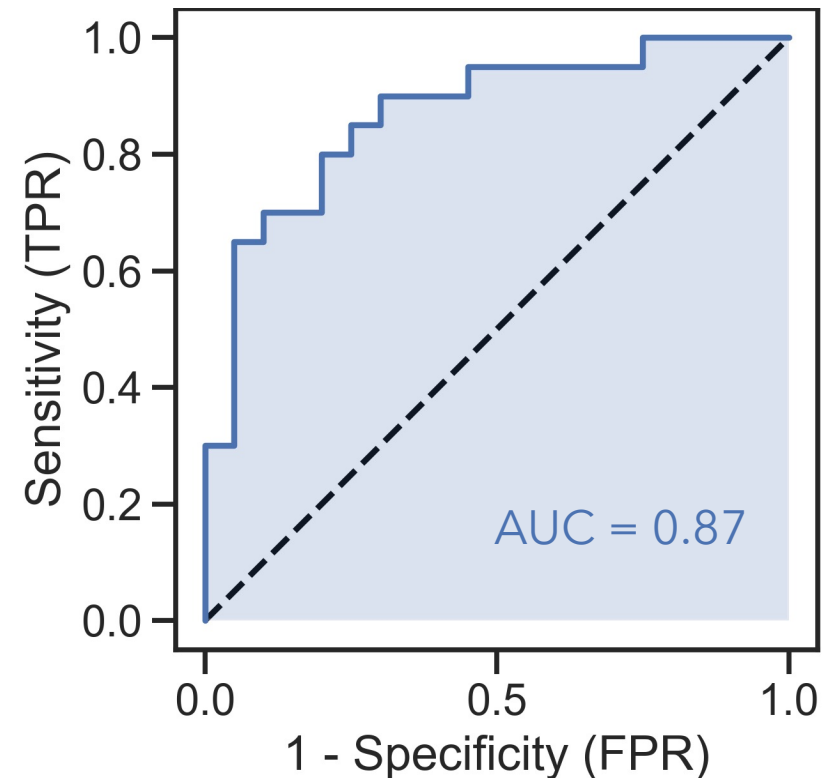
# How do we evaluate ROC curves?

- The dashed line indicates the performance of randomly guessing.



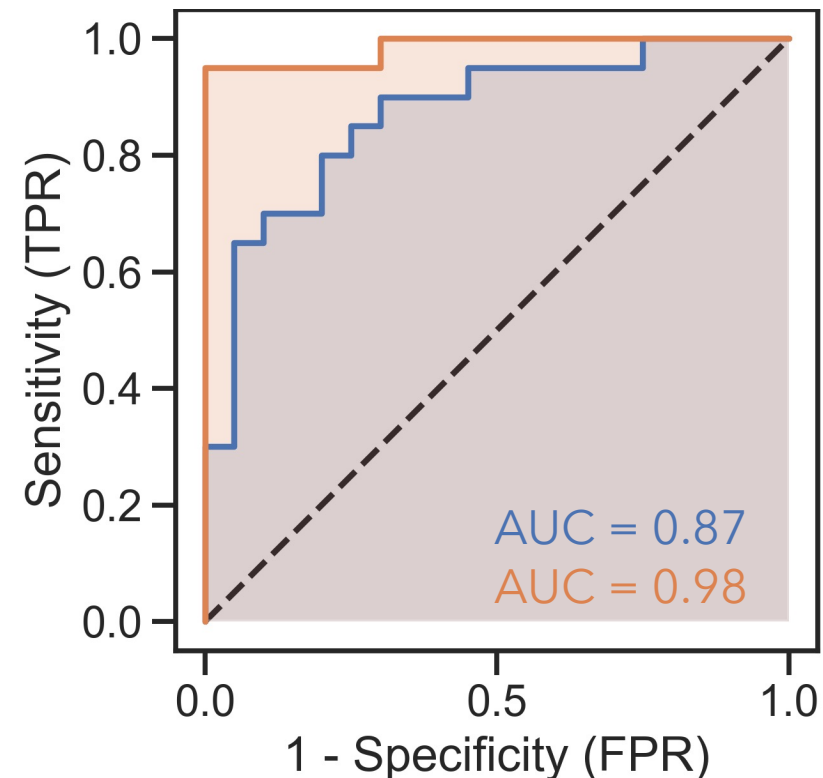
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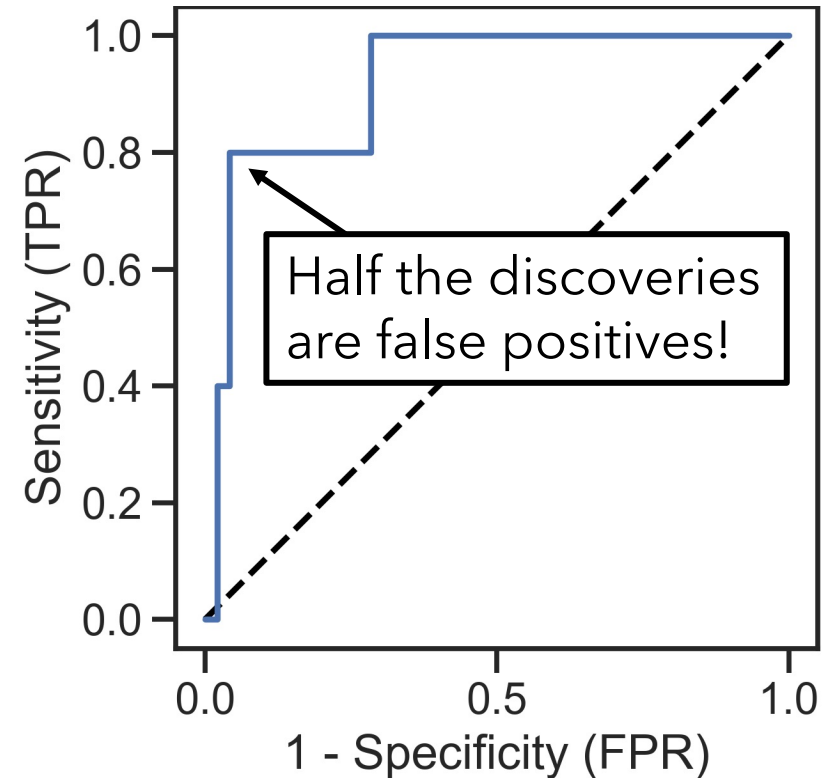
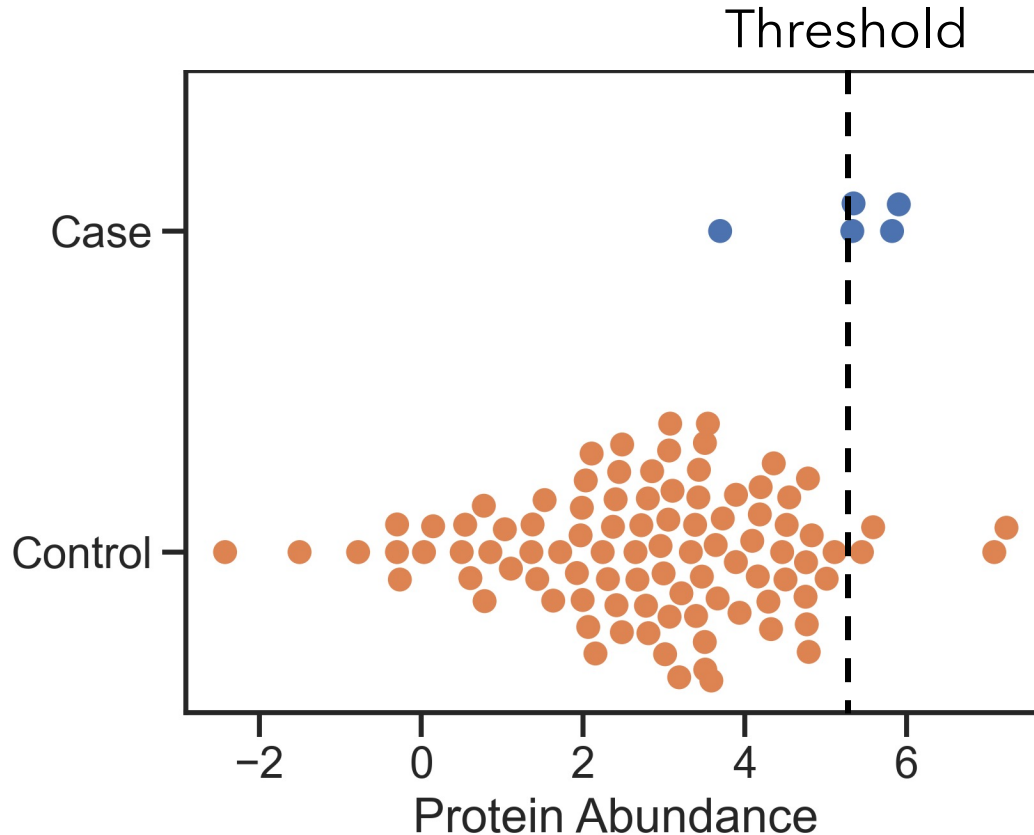
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# ROC curves can be misleading when the classes are imbalanced

What if we had only 5 cases and 95 controls in our data?



# There are many other metrics we can use

		Predicted			
		+	-		
Actual	+	TP Type I error	FN Type II error	Sensitivity (recall) TP/●	False negative rate FN/●
	-	FP Type I error	TN	False positive rate FP/●	Specificity TN/●
		Precision TP/■	False omission rate FN/■	Accuracy ( TP + TN )/( ● + ● )	
		FDR FP/■	Negative predictive value TN/■	$F_1$ score $2TP/( 2TP + FP + FN )$	

