

Over-fitting and Confusion Matrix

Ernst Henle

ErnstHe@UW.edu

Skype: ernst.predixion

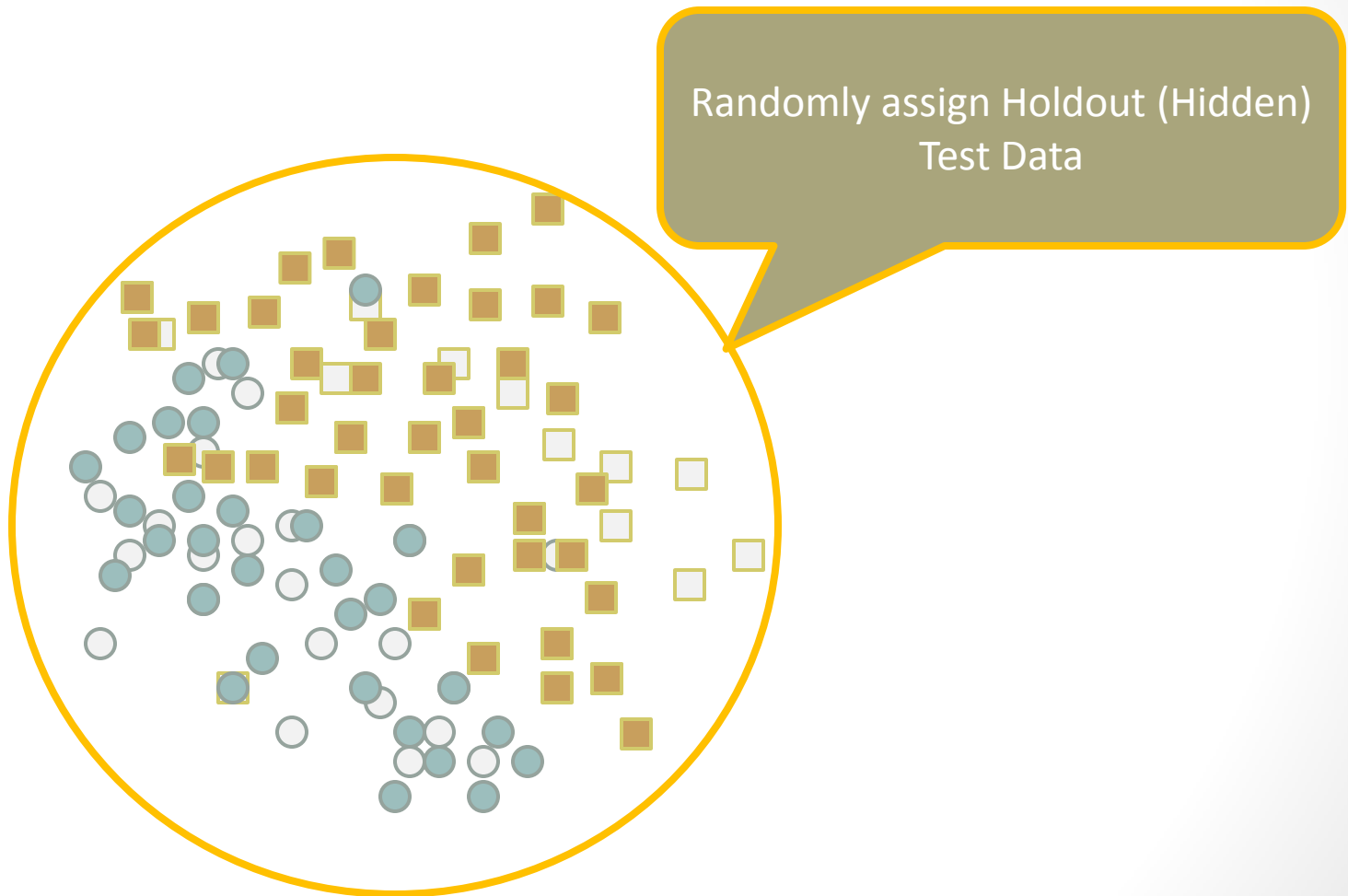
Evaluate Model

- Use an over-fitting example to explain the following concepts:
 - Modeling Data
 - Training Data
 - Test Data
 - Model (Hypothesis)
 - Over-fitting
 - Model Accuracy
 - Confusion Matrix (Classification Matrix)
 - True Positive
 - False Positive
 - True Negative
 - False Negative

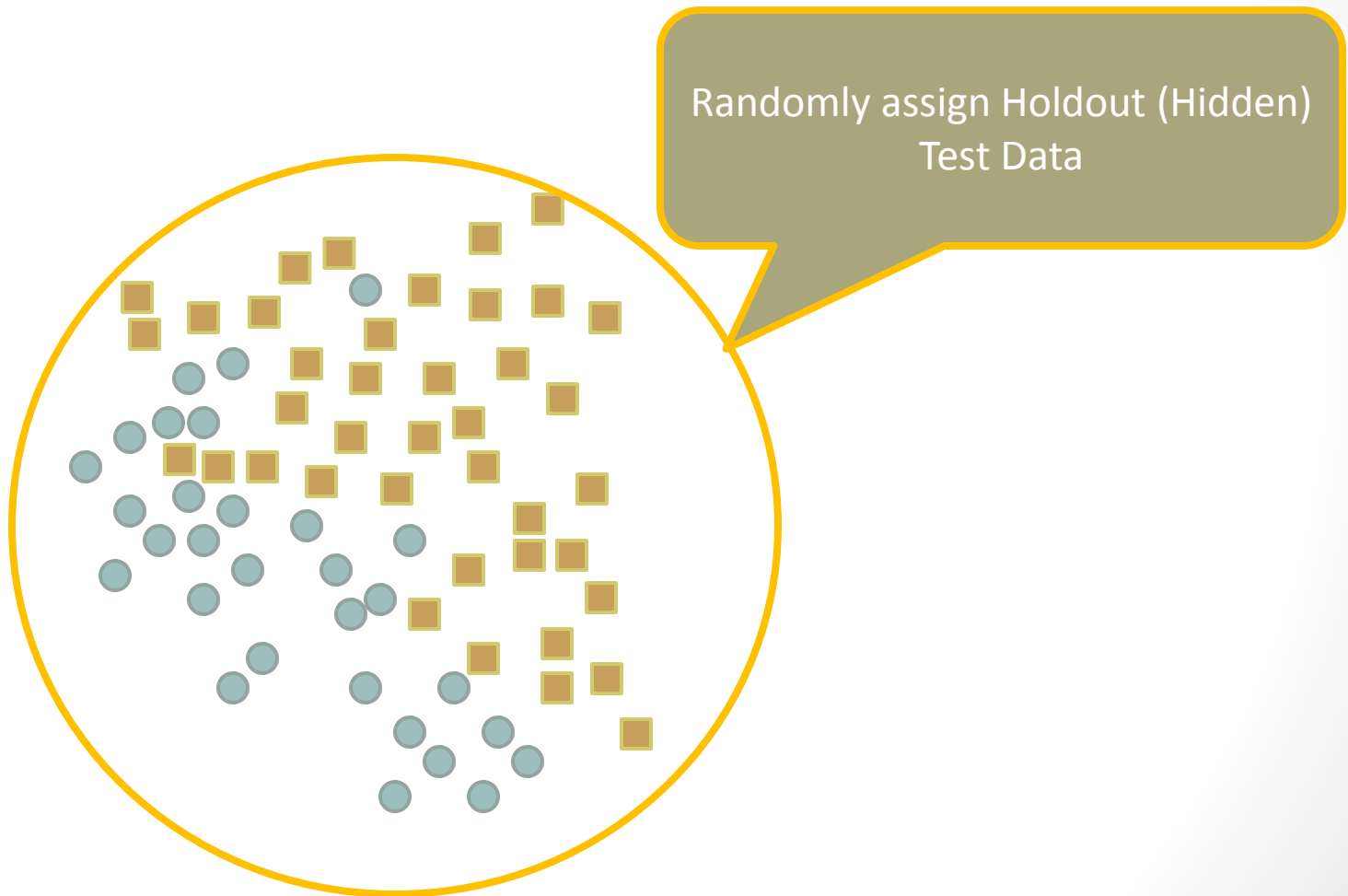
Evaluate Model: All Data



Evaluate Model: Test Data



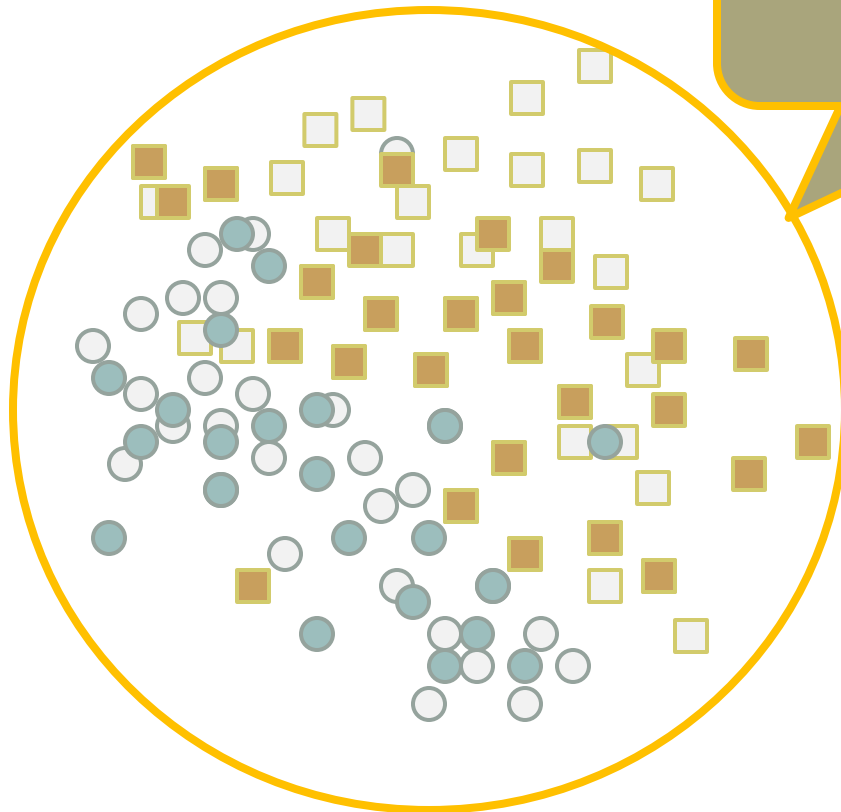
Evaluate Model: Test Data



Evaluate Model: All Data

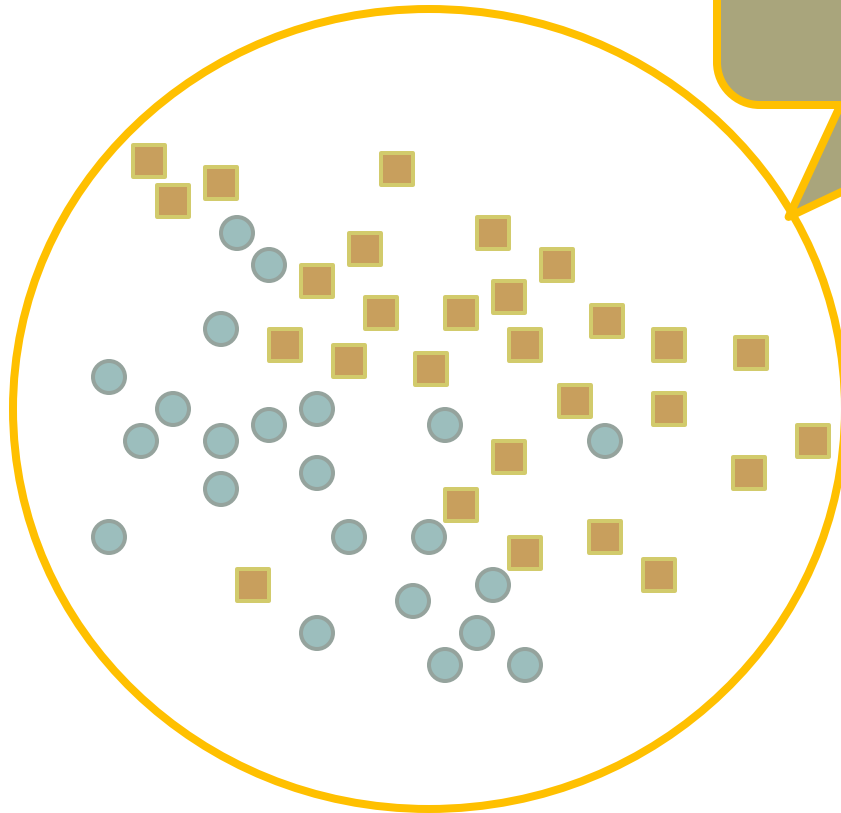


Evaluate Model: Training Data



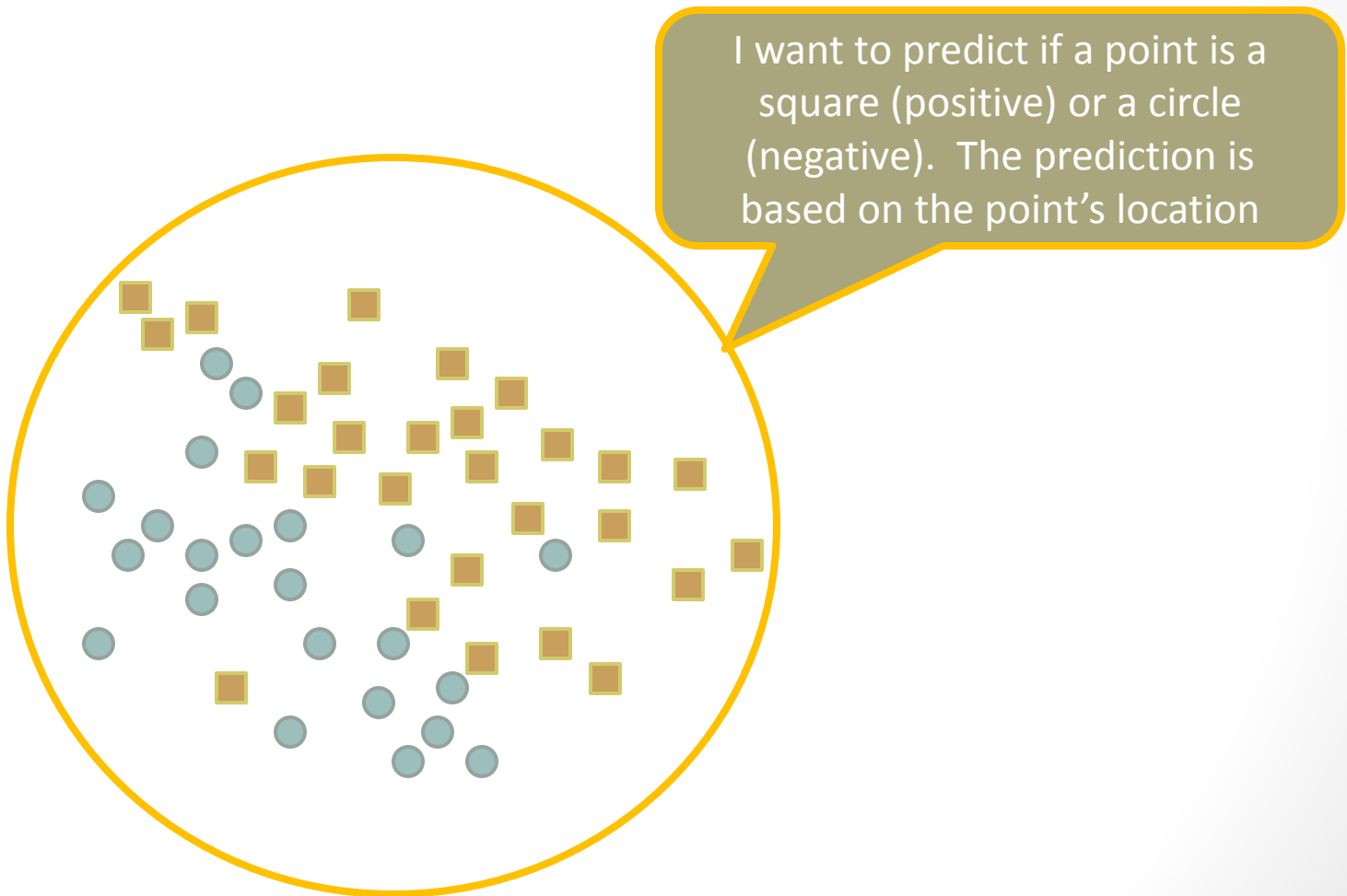
The Data that is not Test Data is
used for Training

Evaluate Model: Training Data

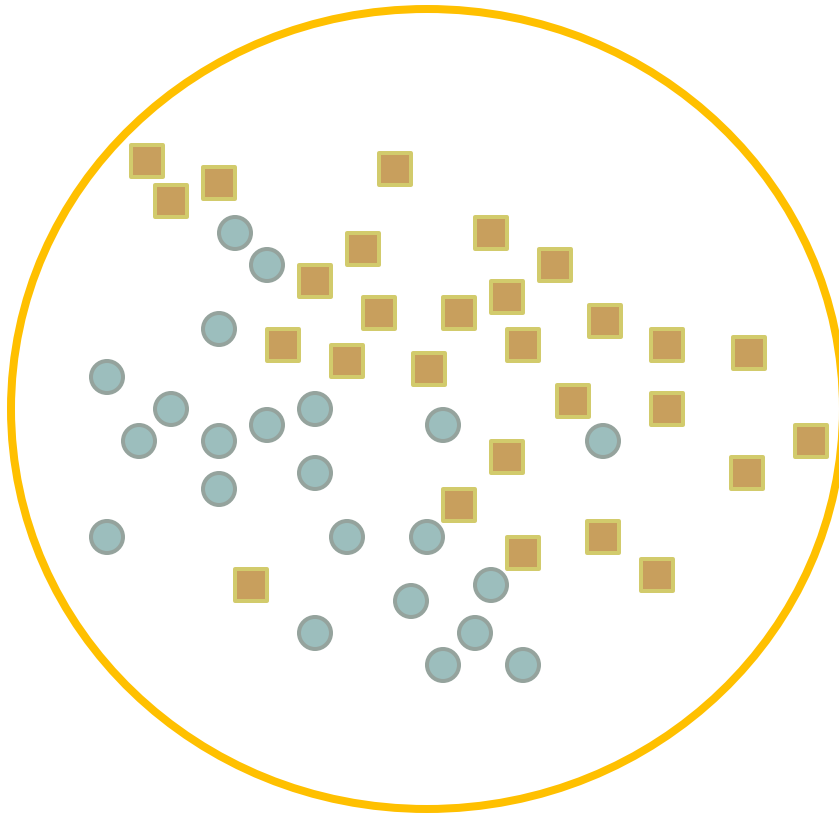


The Data that is not Test Data is
used for Training

Evaluate Model: Training



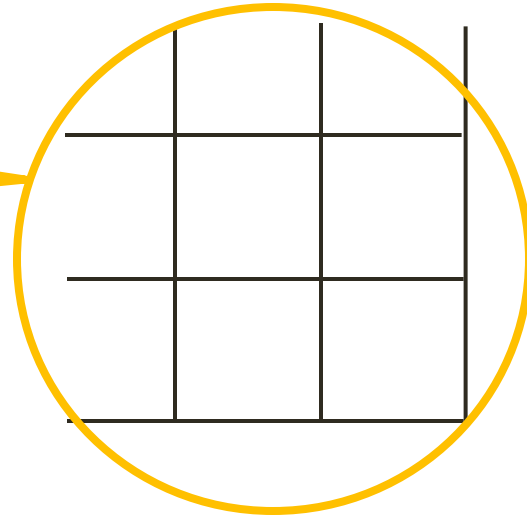
Evaluate Model: Training

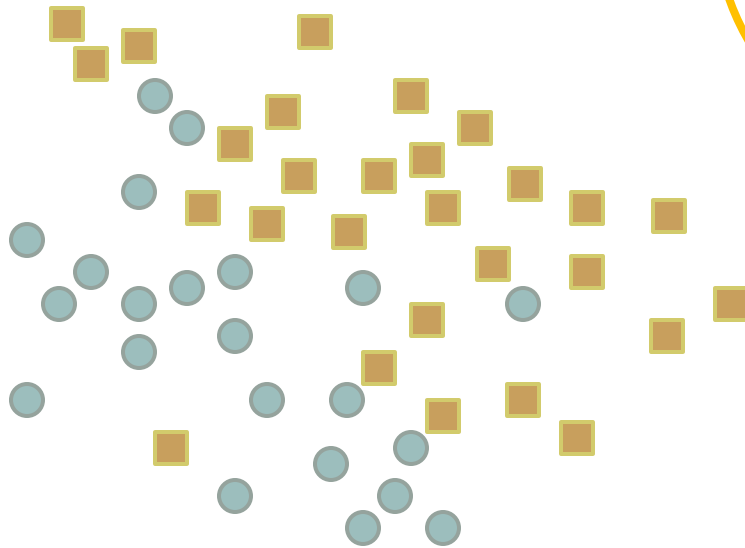


$\text{isSquare} \sim \text{xLocation} + \text{yLocation}$

Evaluate Model: Confusion Matrix

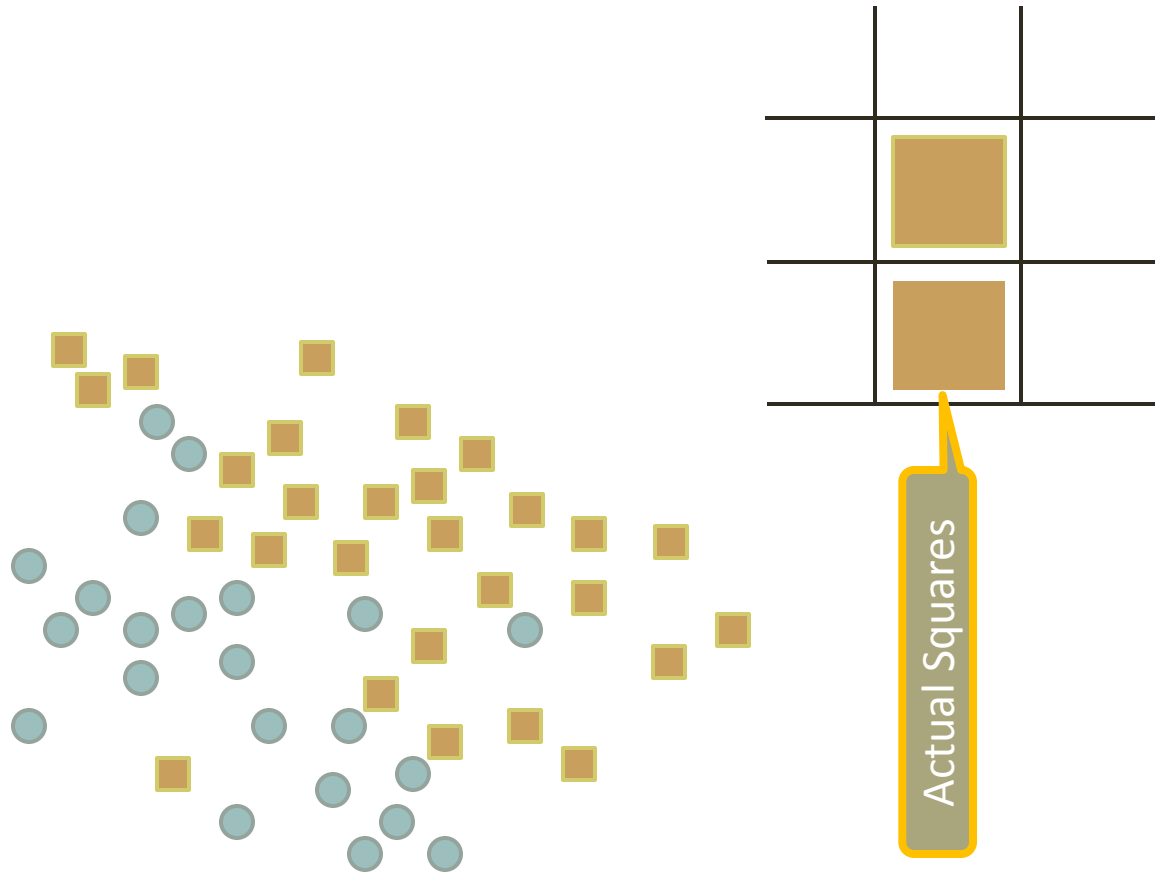
Confusion Matrix (Classification Matrix):
Compare Squares and Circles with
Predicted Squares and Circles





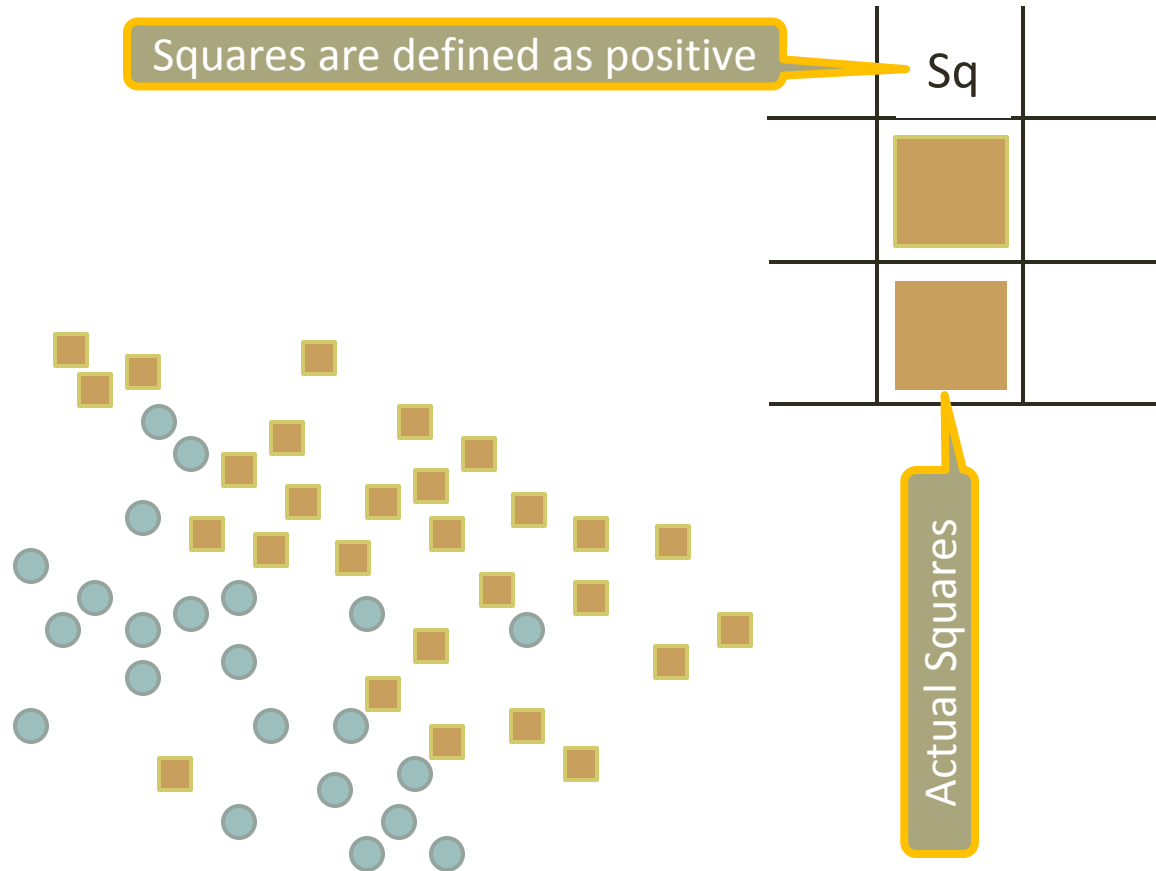
$\text{isSquare} \sim \text{xLocation} + \text{yLocation}$

Evaluate Model: Confusion Matrix



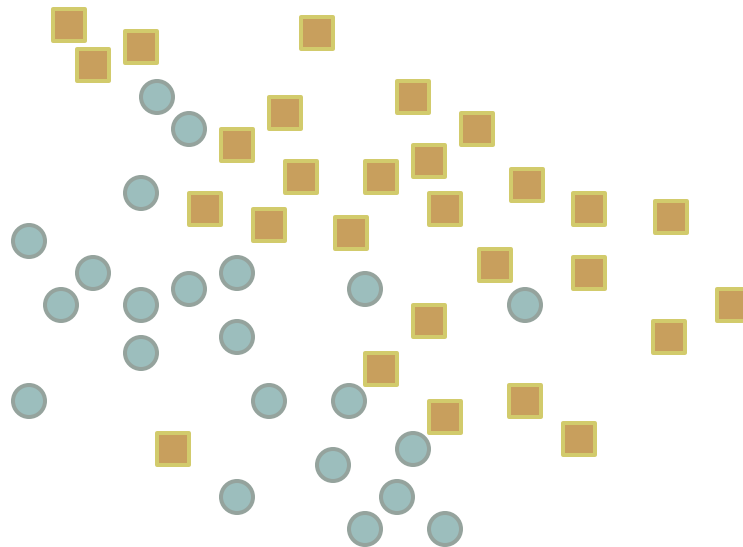
$\text{isSquare} \sim \text{xLocation} + \text{yLocation}$





Evaluate Model: Confusion Matrix



$$\text{isSquare} \sim \text{xLocation} + \text{yLocation}$$

Evaluate Model: Confusion Matrix

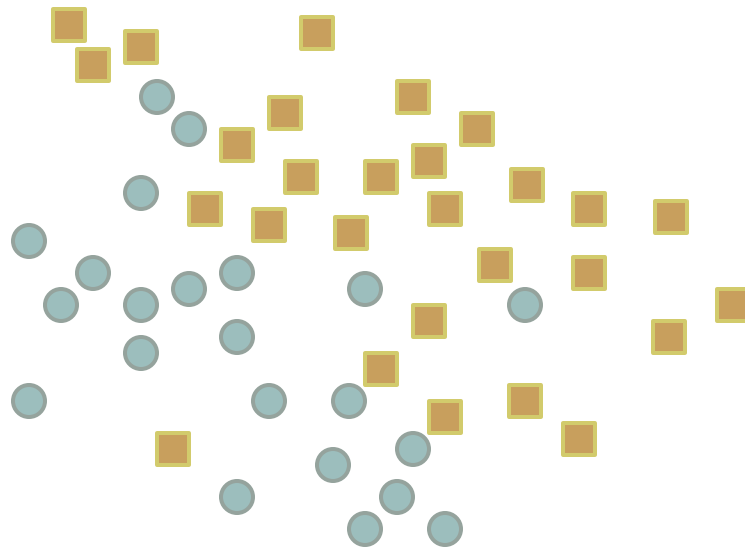


	Sq	
		
		





Actual Circles

$\text{isSquare} \sim \text{xLocation} + \text{yLocation}$

Evaluate Model: Confusion Matrix



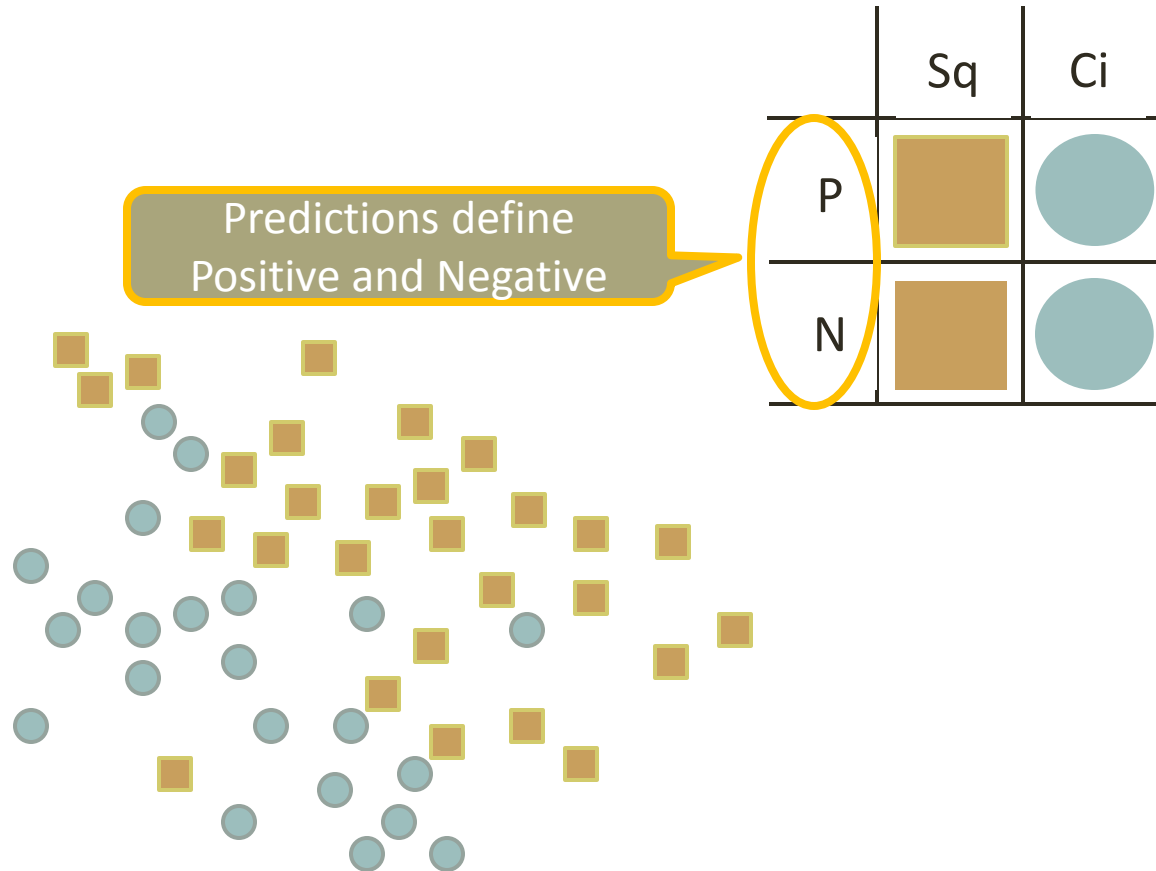
Circles are defined as negative

	Sq	Ci
		
		

Actual Circles

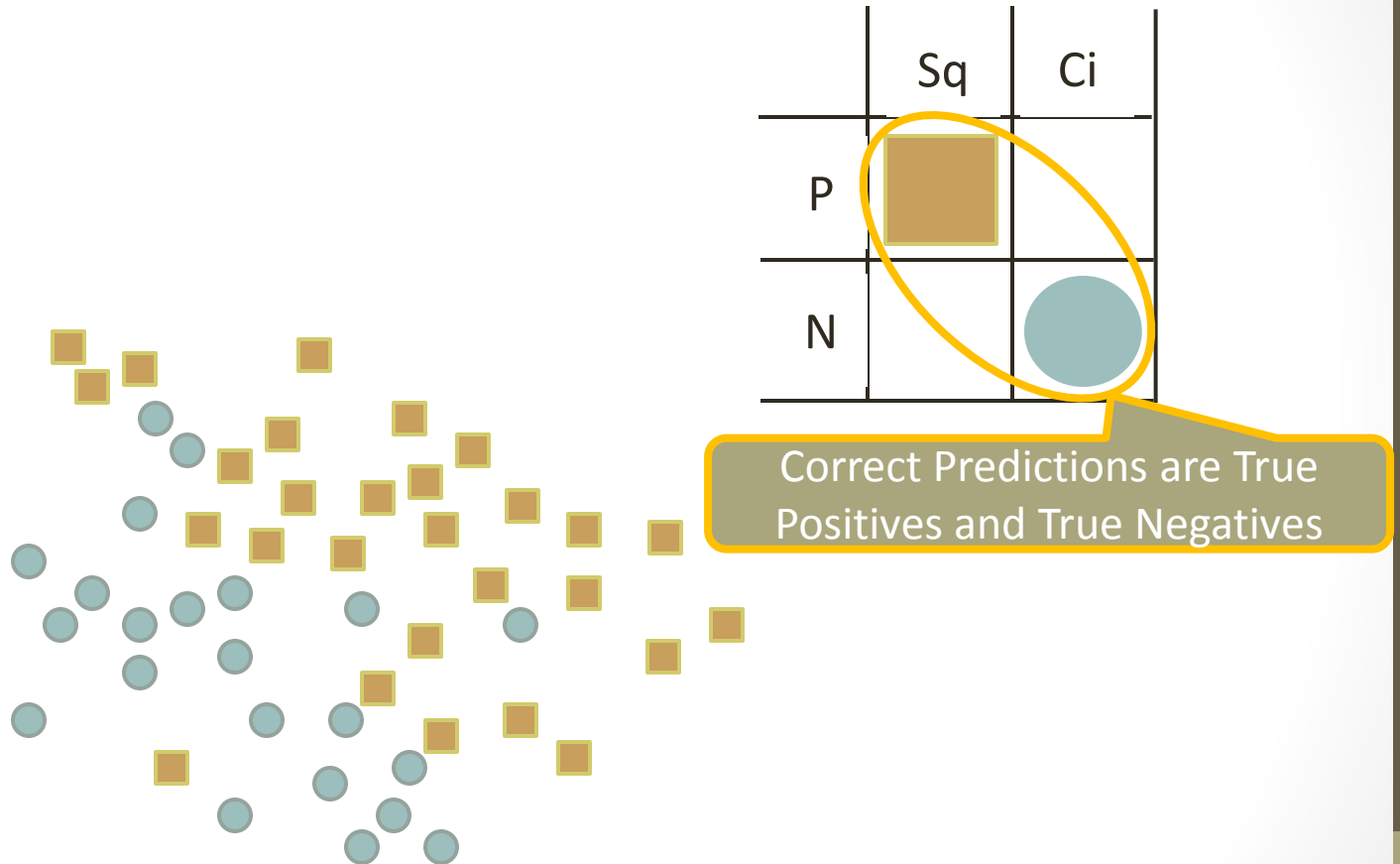
$\text{isSquare} \sim \text{xLocation} + \text{yLocation}$

Evaluate Model: Confusion Matrix



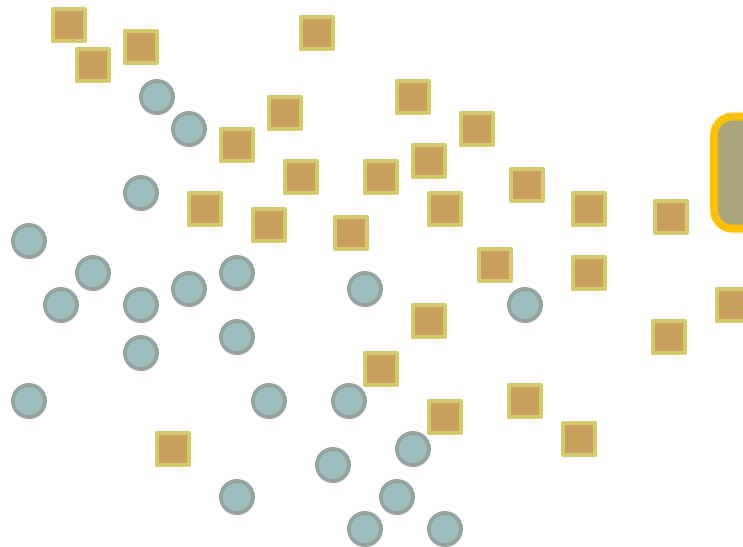
$$\text{isSquare} \sim \text{xLocation} + \text{yLocation}$$



Evaluate Model: Confusion Matrix



$\text{isSquare} \sim \text{xLocation} + \text{yLocation}$

Evaluate Model: Confusion Matrix






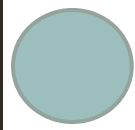
	Sq	Ci
P		
N		

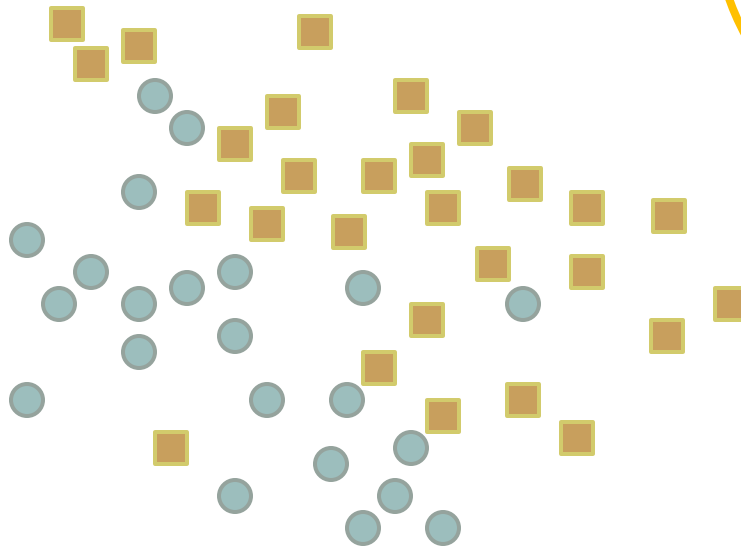
Incorrect Predictions are False Positives and False Negatives

$$\text{isSquare} \sim \text{xLocation} + \text{yLocation}$$

Evaluate Model: Confusion Matrix

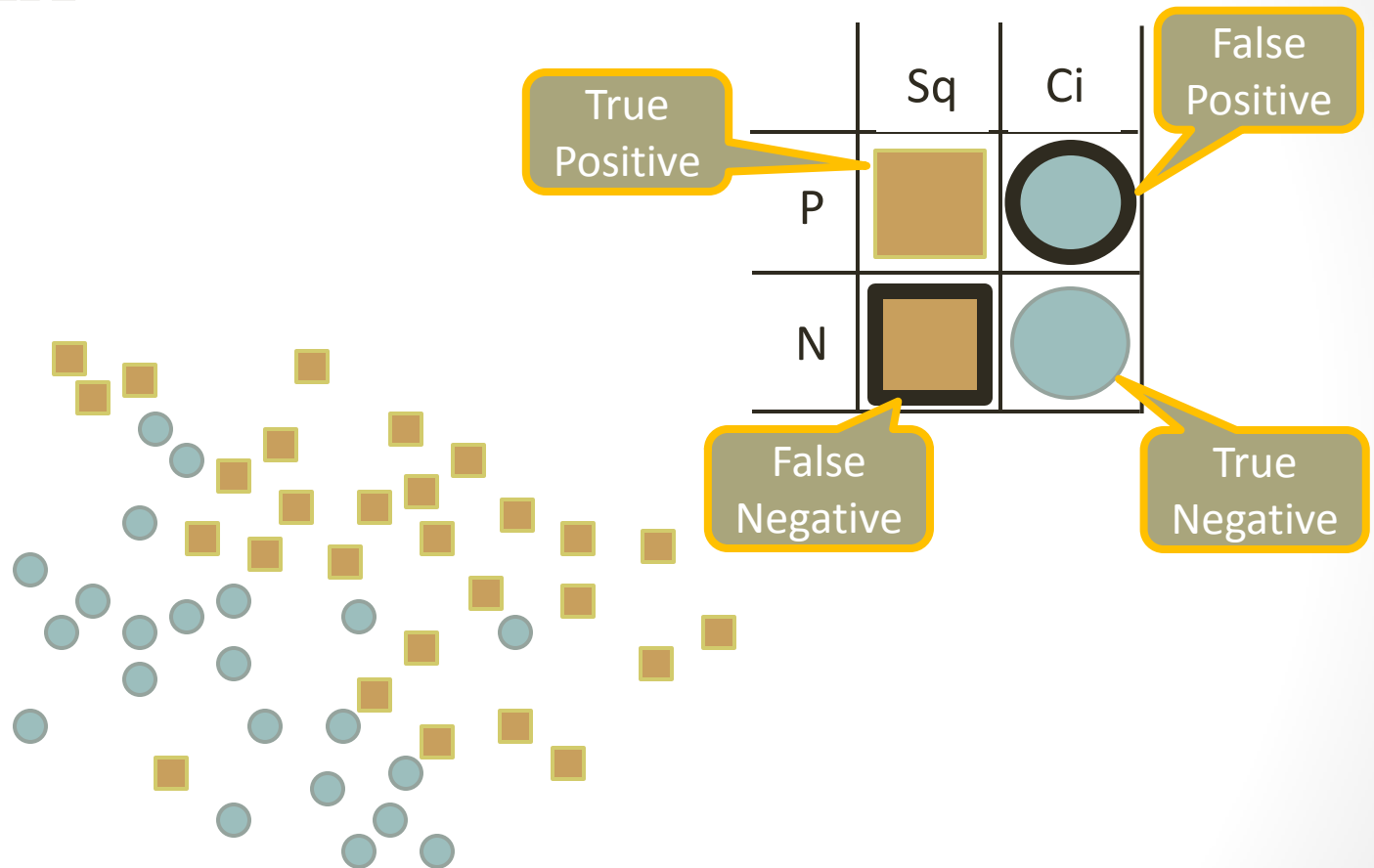
Confusion Matrix (Classification Matrix):
Vertical are actual classes
Horizontal are predicted classes

	Sq	Ci
P		
N		



$\text{isSquare} \sim \text{xLocation} + \text{yLocation}$

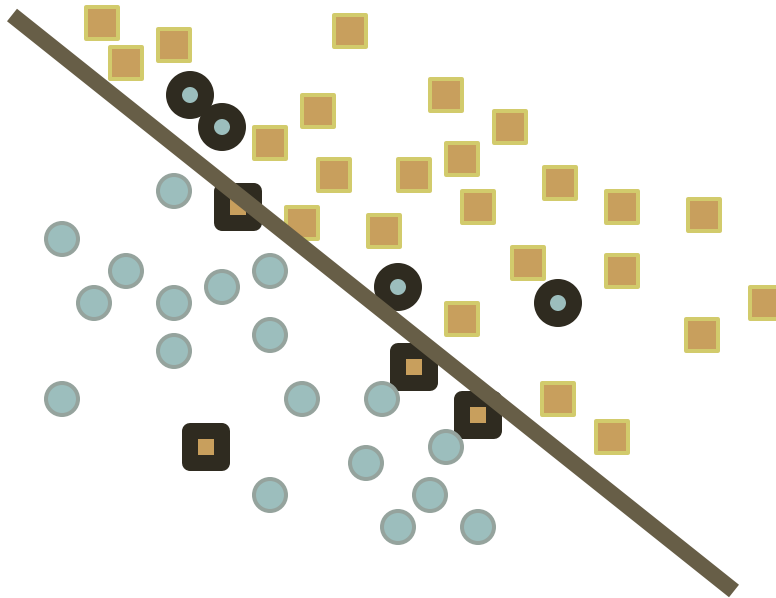
Evaluate Model: Confusion Matrix



$\text{isSquare} \sim \text{xLocation} + \text{yLocation}$

Evaluate Model : Train Model 1

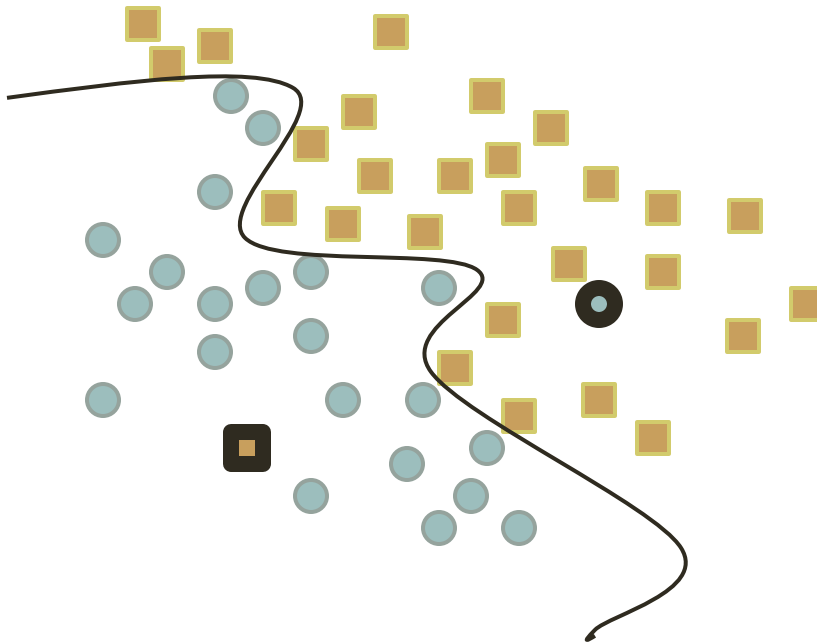
	Sq	Ci
P	36	4
N	4	26



isSquare \sim xLocation + yLocation

Evaluate Model : Train Model 2

	Sq	Ci
P	39	1
N	1	29



$\text{isSquare} \sim \text{xLocation} + \text{yLocation}$

Evaluate Model : Train Model 3

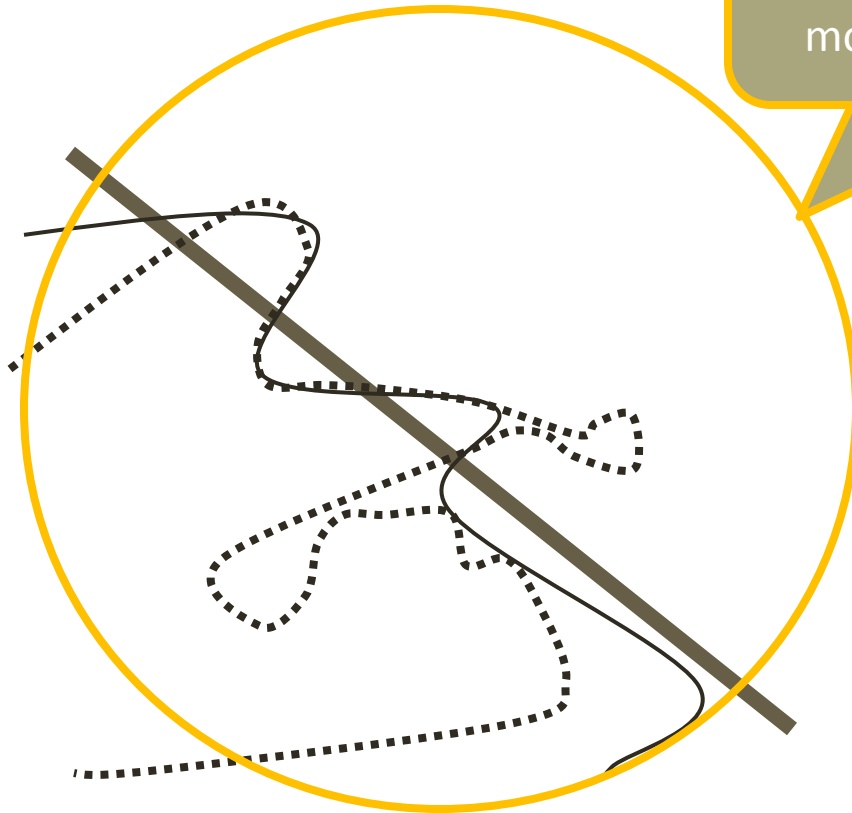
	Sq	Ci
P	40	0
N	0	30



$\text{isSquare} \sim x\text{Location} + y\text{Location}$

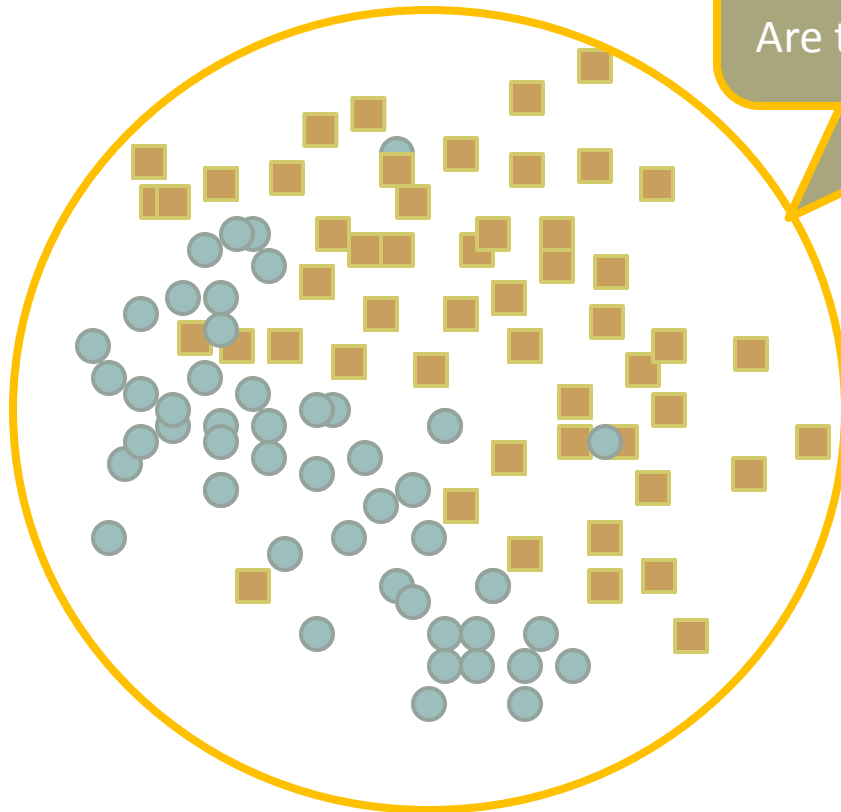
Evaluate Model : 3 Models

These models are based on training data. In these cases, models are called hypotheses.



$$\text{isSquare} \sim \text{xLocation} + \text{yLocation}$$

Evaluate Model : All Data



Training data overlaid on test data.
Visual comparison of data sets.
Are the distributions comparable?

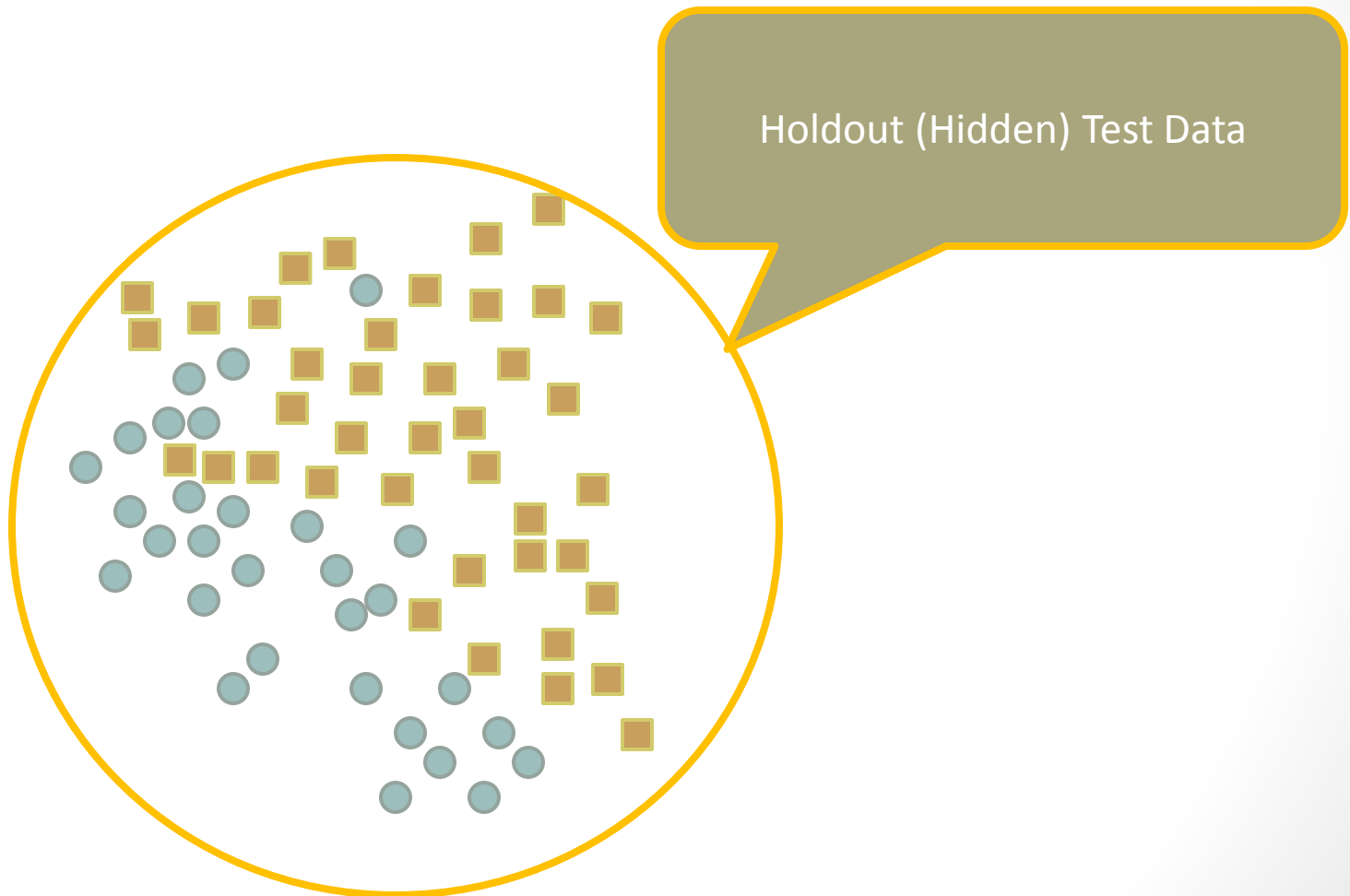
$$\text{isSquare} \sim \text{xLocation} + \text{yLocation}$$

Evaluate Model : Training Data



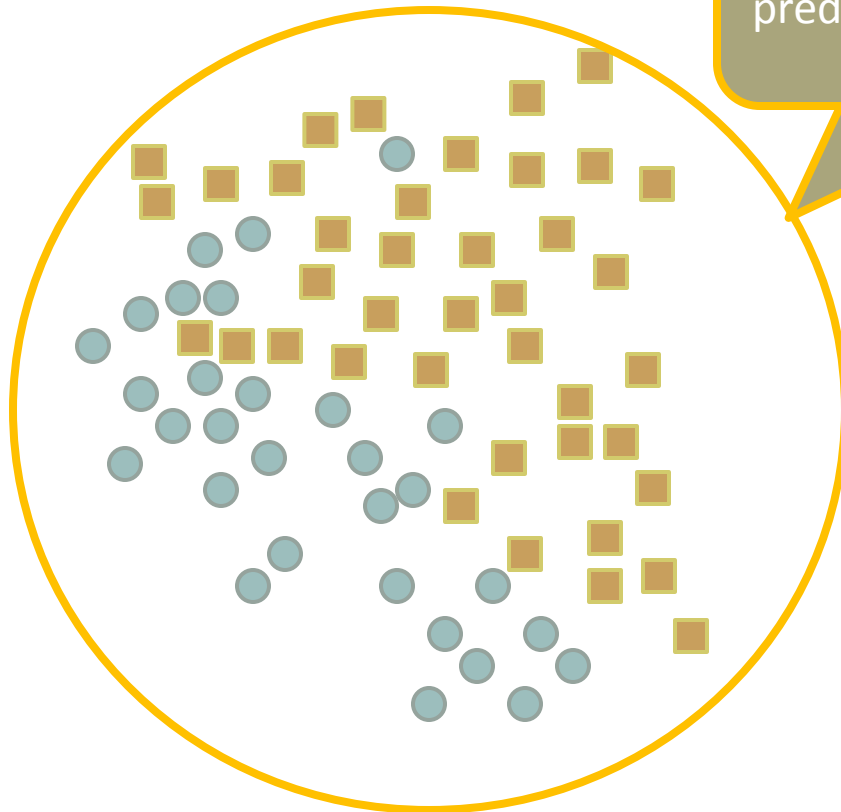
$$\text{isSquare} \sim \text{xLocation} + \text{yLocation}$$

Evaluate Model : Test Data



$$\text{isSquare} \sim \text{xLocation} + \text{yLocation}$$

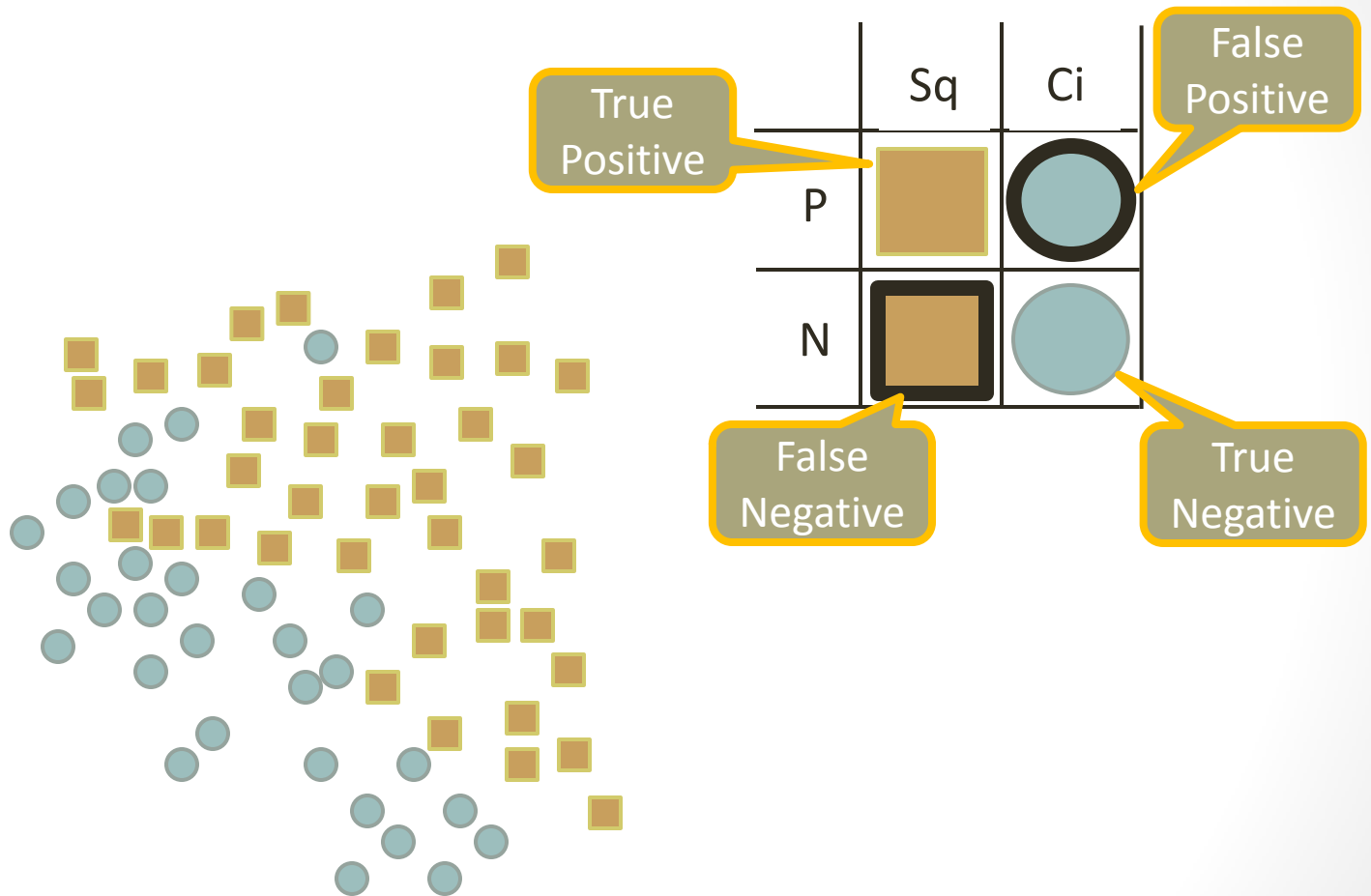
Evaluate Model : Test Data



In the test data set:
I want to test if a square is
predicted as positive and if a circle
is predicted as negative

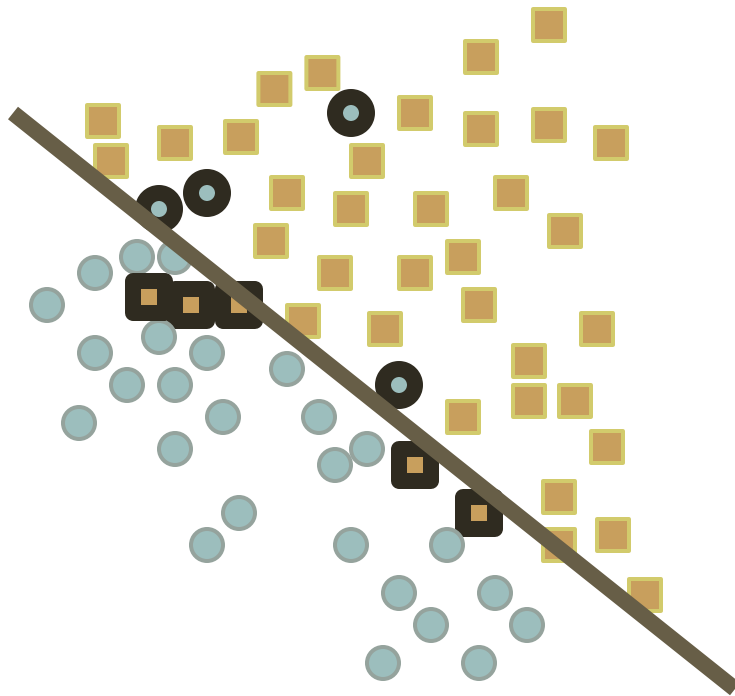
$$\text{isSquare} \sim \text{xLocation} + \text{yLocation}$$

Evaluate Model : Test Data



$$\text{isSquare} \sim x\text{Location} + y\text{Location}$$

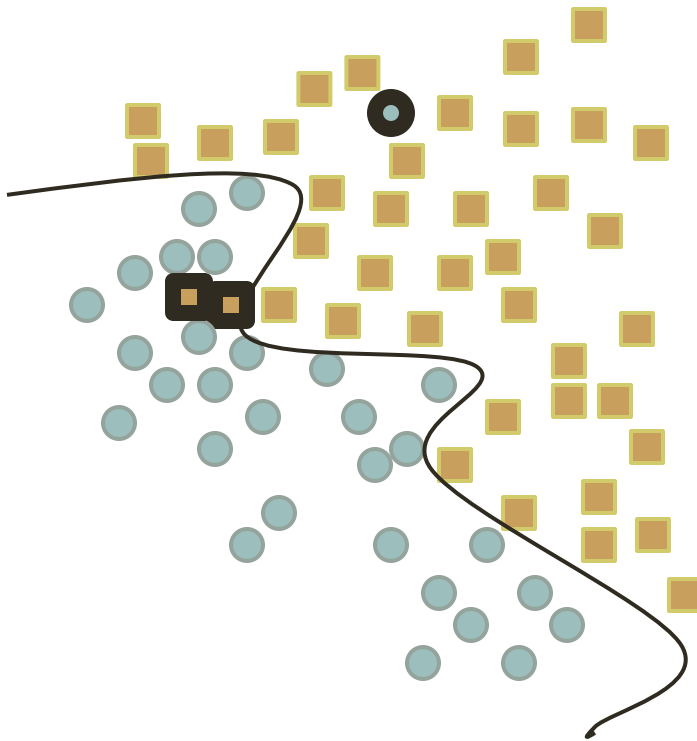
Evaluate Model : Test Model 1



	Sq	Ci
P	35	4
N	5	26

$\text{isSquare} \sim x\text{Location} + y\text{Location}$

Evaluate Model : Test Model 2

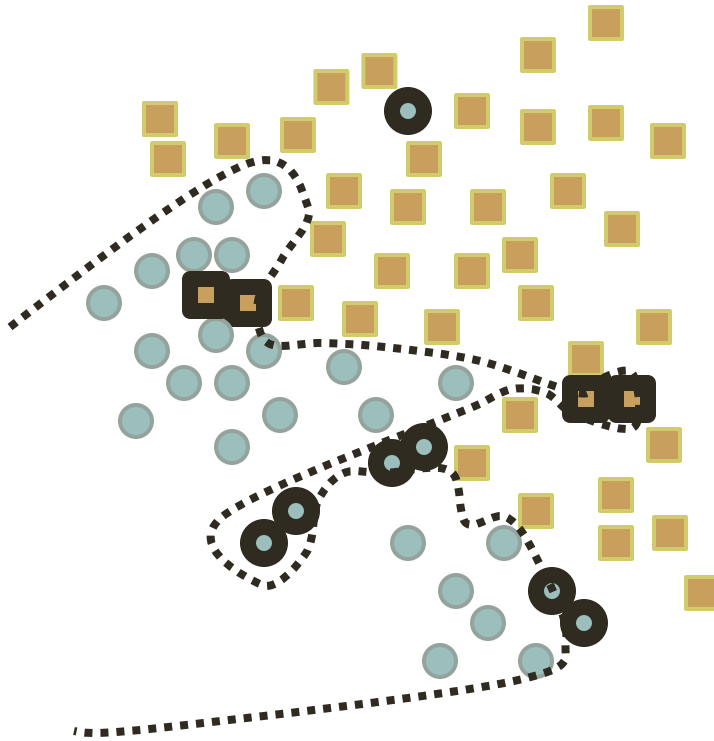


	Sq	Ci
P	38	1
N	2	29

$\text{isSquare} \sim \text{xLocation} + \text{yLocation}$

Evaluate Model : Test Model 3

	Sq	Ci
P	36	7
N	4	23



$\text{isSquare} \sim \text{xLocation} + \text{yLocation}$

Relate a Confusion Matrix with an ROC chart

- Optional exercise in Predixion Insight: Open up a synced Classification (Confusion) Matrix and ROC chart.
 - Set the threshold of the Classification Matrix to 0, 0.5, and 1. How do these thresholds compare to the FPR and TPR on the ROC chart?
 - Set the FPR on the ROC chart to 0, 0.5, and 1. What are the TPR on the ROC chart? How does the threshold of the classification matrix change?
 - Open up a cost chart. Set the readmission penalty to 3X the cost of the intervention cost. What is the optimal threshold? What is the FPR?

Assignment (1)

1. Why are performance metrics better on training data than on test data?
2. How do you determine which data are training data and which data are test data?
3. Beware, this problem contains irrelevant data while some important numbers are not explicitly presented. A model was trained on **300** individuals where **149** had the cold and **151** were healthy. The model was tested on **100** individuals where **10** were ill. The model correctly predicted that **85** of the healthy individuals were indeed healthy and correctly predicted that **7** of the ill individuals were indeed ill. The other predictions were incorrect. Consult Wikipedia: http://en.wikipedia.org/wiki/Precision_and_recall and construct a confusion matrix and then calculate the following:
 - a) Sensitivity
 - b) Specificity
 - c) Accuracy
 - d) Precision
 - e) Recall

Assignment (2)

4. The probability threshold for a classification varies in an ROC chart from 0 to 1.
 - a) What point of the graph corresponds to a threshold of zero?
 - b) What point of the graph corresponds to a threshold of one?
 - c) What point of the graph corresponds to a threshold of 0.5?
(trick question)
5. A Classification is tested on 1000 cases. In the middle of its ROC chart, where the false positive rate is 0.4, the true positive rate is 0.8. The accuracy is 0.7.
 - a) What does the confusion matrix look like?
 - b) What can you say about the probability threshold at that point?
(trick question)

Over-fitting and Confusion Matrix