

ACIT4830 – Special Robotics and Control Subject

Evaluating Model Performance

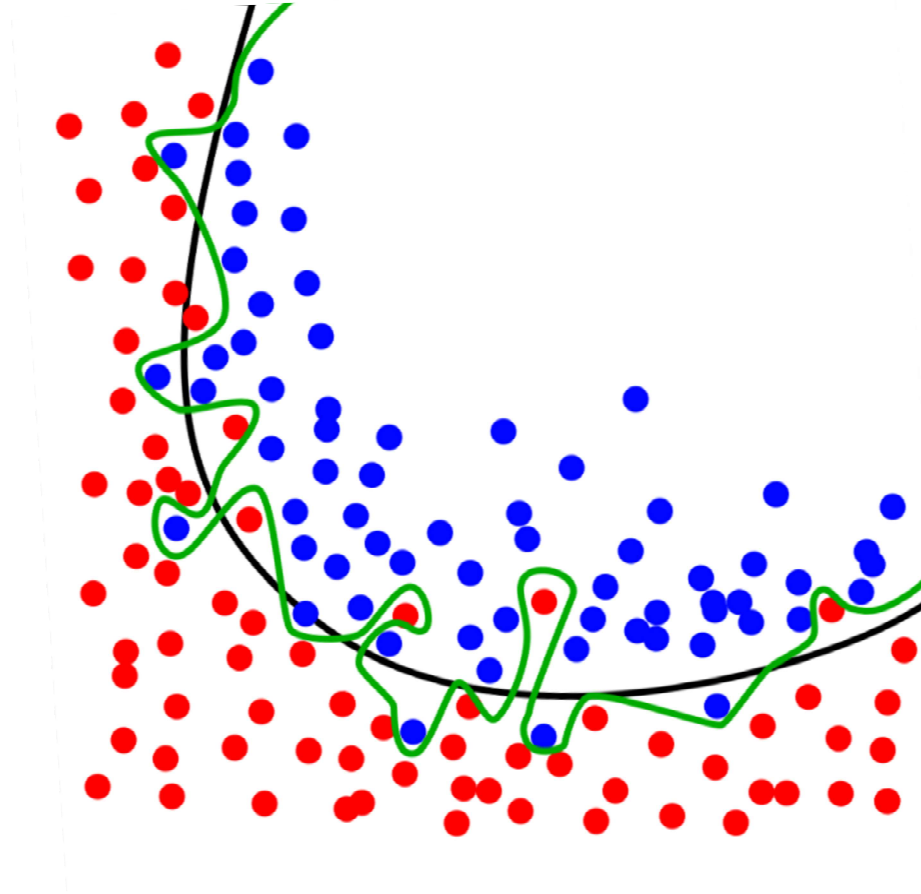
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Content

(Chapter 10 – evaluating model performance)

- Kappa statistic
 - Sensitivity, Specificity
 - Precision, Recall, F-measure
 - PR-curve
 - Hold out method
 - K-fold Cross Validation
-
- Hands-on exercise

overfitting vs. well-performing model



Confusion Matrix (spam example)

		Predicted to be Spam	
		no	yes
Actually Spam	no	<div>TN</div> <div>True Negative</div>	<div>FP</div> <div>False Positive</div>
	yes	<div>FN</div> <div>False Negative</div>	<div>TP</div> <div>True Positive</div>

		Predicted to be Spam	
		no	yes
Actually Spam	no	TN True Negative	FP False Positive
	yes	FN False Negative	TP True Positive

$$\text{accuracy} = \frac{TP + TN}{TP + TN + FP + FN}$$

$$\text{error rate} = \frac{FP + FN}{TP + TN + FP + FN} = 1 - \text{accuracy}$$

Kappa statistic

$$k = \frac{\Pr(a) - \Pr(e)}{1 - \Pr(e)}$$

$\Pr(a)$: actual agreement between classifier & true values

$\Pr(e)$: expected agreement by chance alone

- Poor agreement = Less than 0.20
- Fair agreement = 0.20 to 0.40
- Moderate agreement = 0.40 to 0.60
- Good agreement = 0.60 to 0.80
- Very good agreement = 0.80 to 1.00

		Predicted to be Spam	
		no	yes
Actually Spam	no	<div>TN</div> <div>True Negative</div>	<div>FP</div> <div>False Positive</div>
	yes	<div>FN</div> <div>False Negative</div>	<div>TP</div> <div>True Positive</div>

$$\text{specificity} = \frac{\text{TN}}{\text{TN} + \text{FP}}$$

(True Negative rate)

$$\text{sensitivity} = \frac{\text{TP}}{\text{TP} + \text{FN}}$$

(True Positive rate)

		Predicted to be Spam	
		no	yes
Actually Spam	no	<div>TN</div> <div>True Negative</div>	<div>FP</div> <div>False Positive</div>
	yes	<div>FN</div> <div>False Negative</div>	<div>TP</div> <div>True Positive</div>

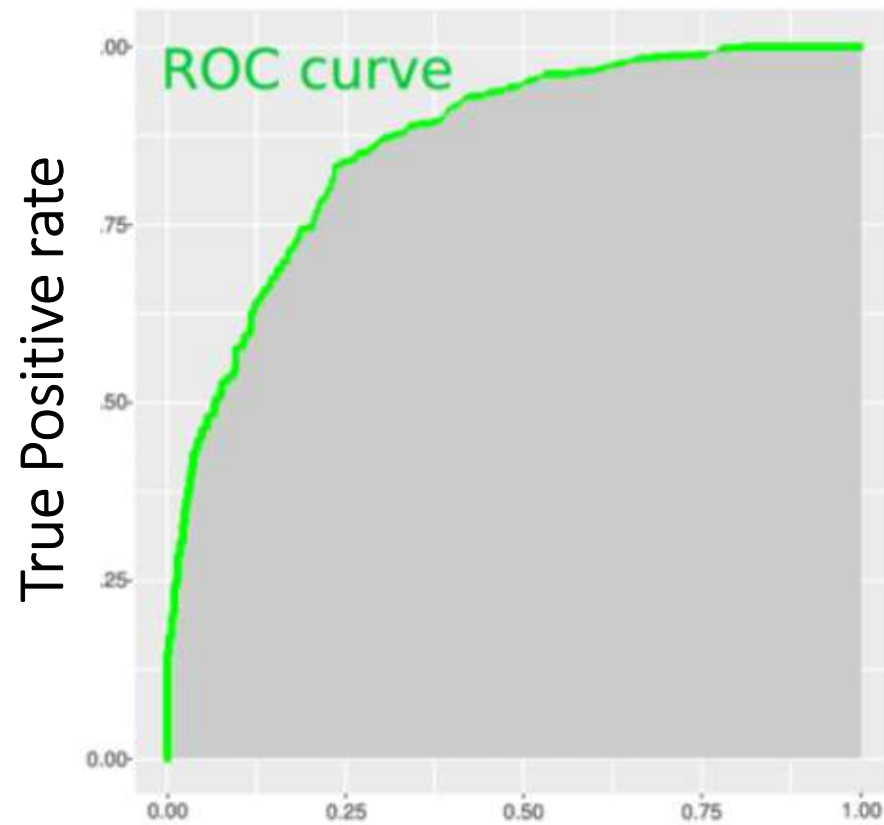
$$\text{recall} = \frac{TP}{TP + FN}$$

(what proportion of **actual** positives are correctly classified?)

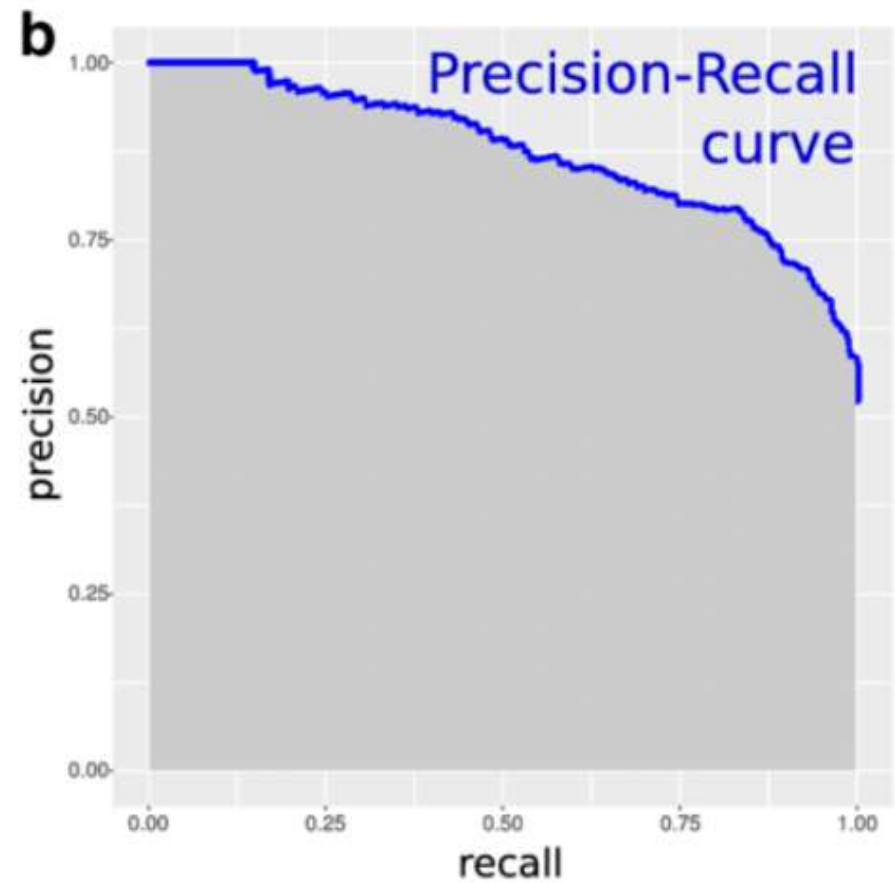
$$\text{precision} = \frac{TP}{TP + FP}$$

(what proportion of **predicted** positives are truly positive?)

$$\text{F - measure} = \frac{2 \times \text{precision} \times \text{recall}}{\text{recall} + \text{precision}} = \frac{2 \times TP}{2 \times TP + FP + FN}$$



False Positive rate
1 – specificity



sensitivity

caret package – performance evaluation

Confusion Matrix and Statistics

	Reference	
Prediction	ham	spam
ham	1202	29
spam	5	154

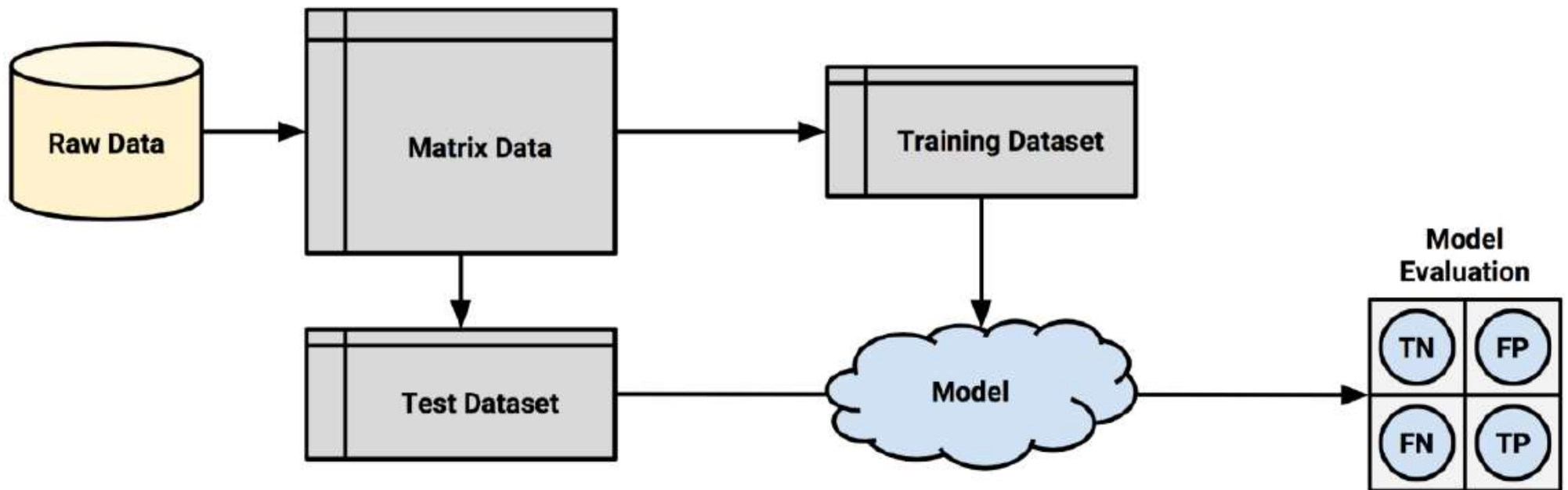
Accuracy : 0.9755
95% CI : (0.966, 0.983)
No Information Rate : 0.8683
P-Value [Acc > NIR] : < 2.2e-16

Kappa : 0.8867
McNemar's Test P-Value : 7.998e-05

Sensitivity : 0.8415
Specificity : 0.9959
Pos Pred Value : 0.9686
Neg Pred Value : 0.9764
Prevalence : 0.1317
Detection Rate : 0.1108
Detection Prevalence : 0.1144

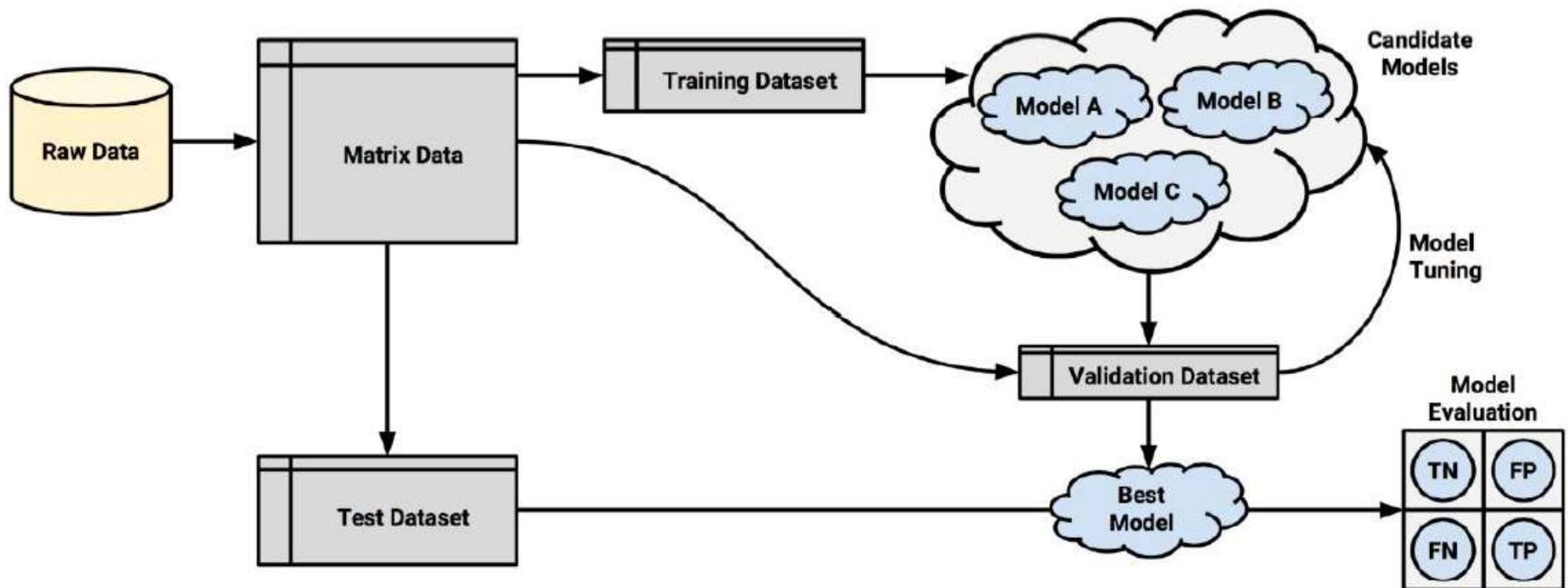
'Positive' class : spam

Hold out method



Typically: 2/3 Training, 1/3 Test (70:30 - 80:20 training: test split)

Hold out with Validation



Typically: 50% Training, 25% Validation, 25% Test

Cross-validation

The repeated holdout is the basis of a technique known as **k-fold cross-validation** (or k-fold CV), which has become the industry standard for estimating model performance. But rather than taking repeated random samples that could potentially use the same record more than once, k-fold CV randomly divides the data into k completely separate random partitions called folds.

Standard $k = 10$