

Applied Text Analytics & Natural Language Processing

with Dr. Mahdi Roozbahani
& Wafa Louhichi

Classification Evaluation

These slides are inspired based on slides from Professors Guy Lebanon, Jeffrey Heer, John Stasko, Christos Faloutsos, Polo Chau, Mahdi Roozbahani



Learning Objectives

In this lesson, you will learn about the goodness of classification

- Confusion matrix
- Accuracy
- RoC-AUC curve

Classification Performance Confusion Matrix

		Predicted Class		
		Sport	News	Politics
Actual Class	Sport	5	3	0
	News	2	3	1
	Politics	0	2	11

https://en.wikipedia.org/wiki/Confusion_matrix



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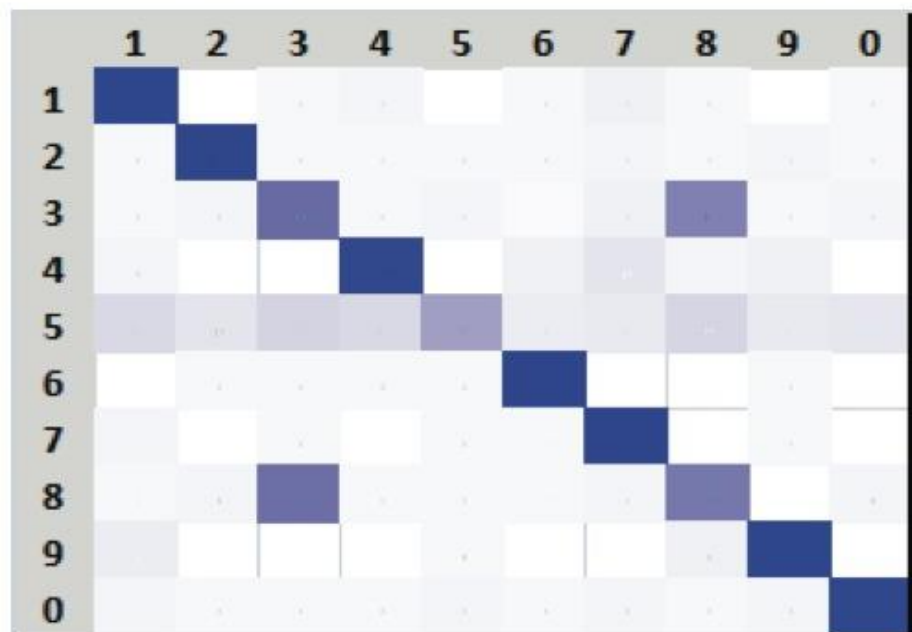


A better way to visualize Confusion Matrix

	1	2	3	4	5	6	7	8	9	0
1	91	0	1	2	0	1	3	1	0	1
2	1	89	1	1	1	1	2	1	2	1
3	1	2	48	1	2	0	3	40	1	2
4	2	0	0	83	0	3	7	2	3	0
5	10	7	12	10	30	4	5	11	5	6
6	0	1	1	1	1	95	0	0	1	0
7	2	0	1	0	1	1	94	0	1	0
8	1	2	47	1	1	1	2	43	0	2
9	4	0	0	0	1	0	0	3	92	0
0	2	1	1	1	2	1	2	1	2	87




Hard to spot trends and patterns





Much easier!

Confusion Matrix

- Very important: Find out what “**positive** and **negative**” means

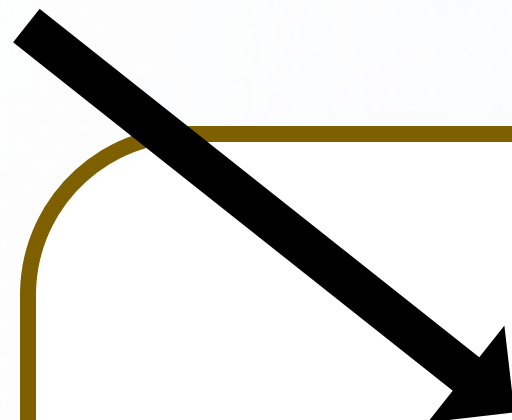


		Predicated	
		Sport	News
Actual	Sport	5	3
	News	2	4




$$\text{Accuracy} = \frac{\text{True positive} + \text{True negative}}{\text{total observations}}$$

Confusion Matrix

- Very important: Find out what “**positive** and **negative**” means



		Predicated	
		Sport	News
Actual	Sport	5	3
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Performance Metrics

- Very important: Find out what “**positive** and **negative**” means

“False Alarm” easy to remember in security applications



Terminology and derivations from a confusion matrix

true positive (TP)

eqv. with hit

true negative (TN)

eqv. with correct rejection

false positive (FP)

eqv. with false alarm, Type I error

false negative (FN)

eqv. with miss, Type II error

sensitivity or true positive rate (TPR)

eqv. with hit rate, recall

$$TPR = \frac{TP}{P} = \frac{TP}{TP + FN}$$

specificity (SPC) or true negative rate (TNR)

$$SPC = \frac{TN}{N} = \frac{TN}{FP + TN}$$

precision or positive predictive value (PPV)

$$PPV = \frac{TP}{TP + FP}$$

recall (recall)

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

negative predictive value (NPV)

$$NPV = \frac{TN}{TN + FN}$$

fall-out or false positive rate (FPR)

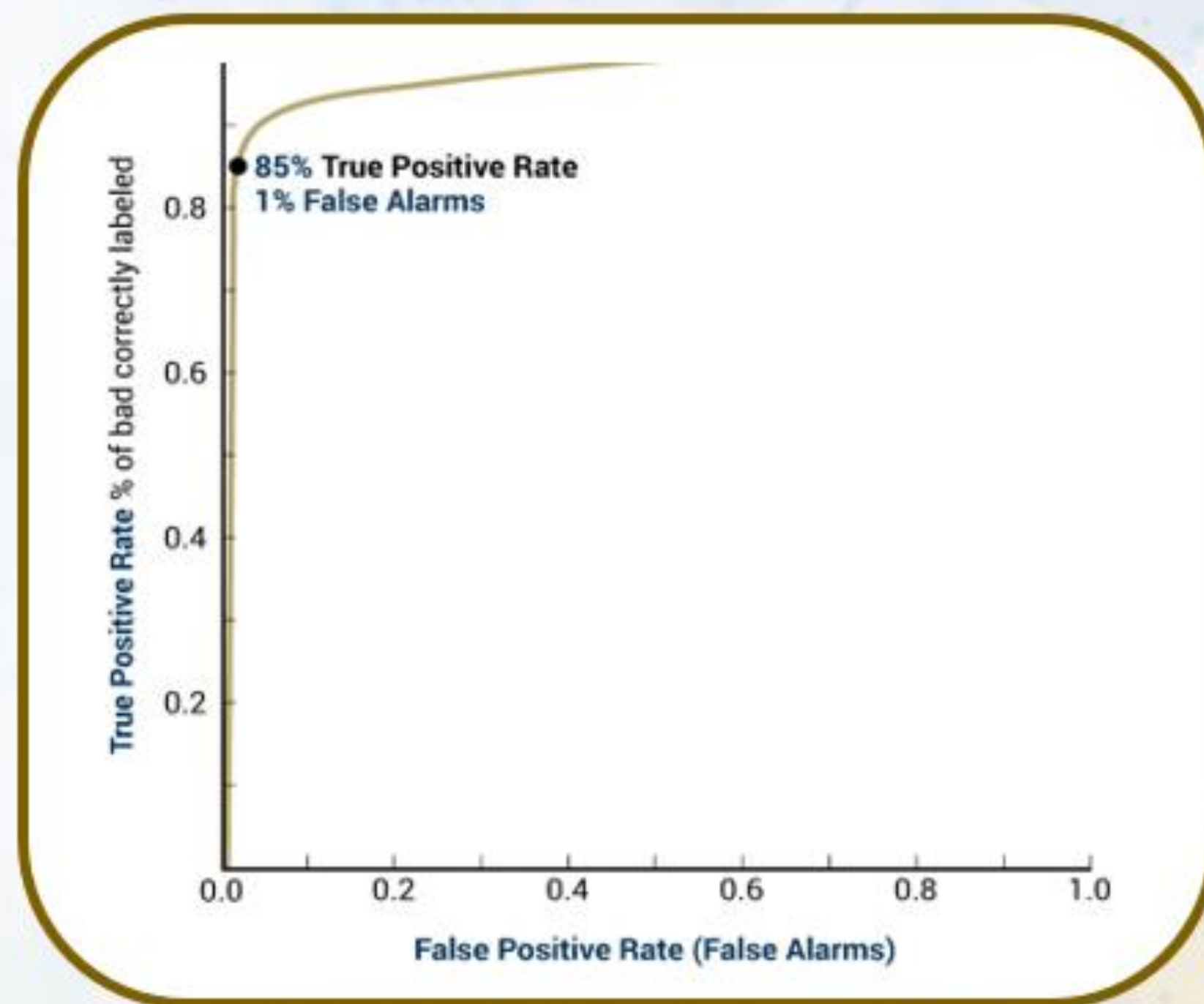
$$FPR = \frac{FP}{N} = \frac{FP}{FP + TN} = 1 - SPC$$

Visualizing Classification Performance using **ROC curve** (Receiver Operating Characteristic)

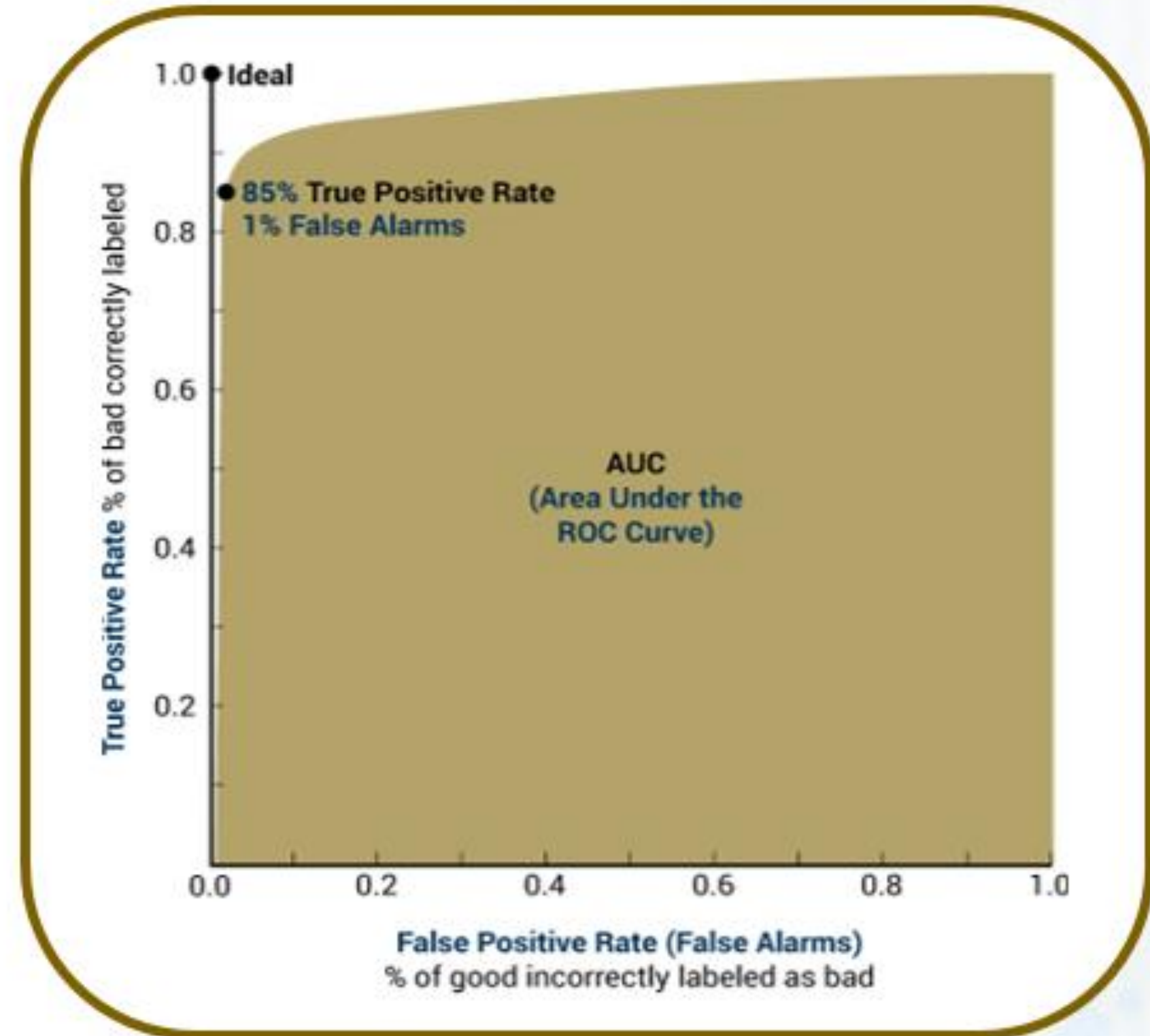
Polonium's ROC Curve

Positive class: malware

Negative class: benign



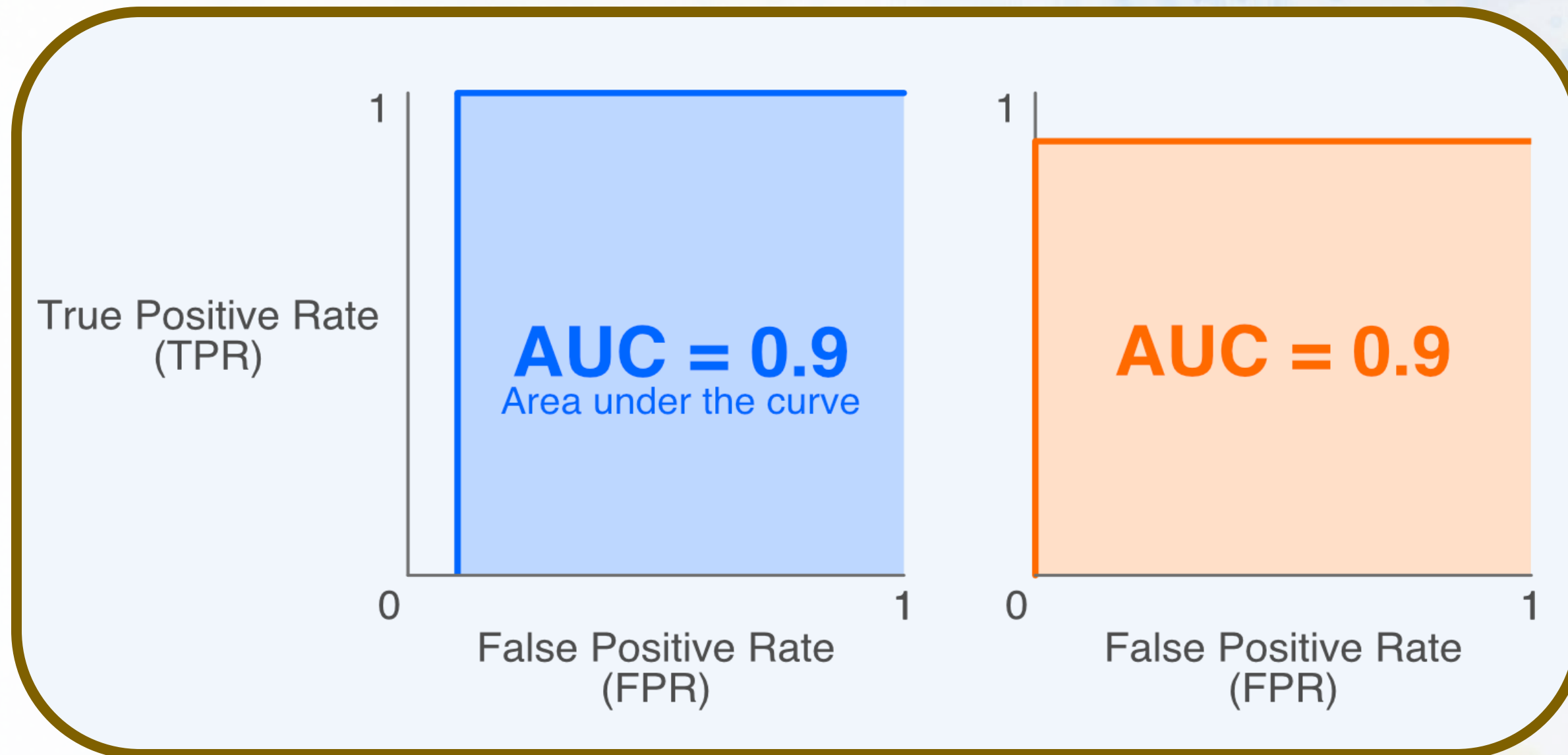
Measuring Classification Performance



If a machine learning algorithm achieves
0.9 AUC (out of 1.0),

that's a great algorithm, right?

Be Careful with AUC!



Summary

- Confusion matrix
- ROC and AUC curve