

# Visualization for Classification

## ROC, AUC, Confusion Matrix

Duen Horng (Polo) Chau

Associate Professor

Associate Director, MS Analytics

Machine Learning Area Leader, College of Computing

Georgia Tech

Partly based on materials by

Professors Guy Lebanon, Jeffrey Heer, John Stasko, Christos Faloutsos, Parishit Ram, Alex Gray

# Visualizing Classification Performance

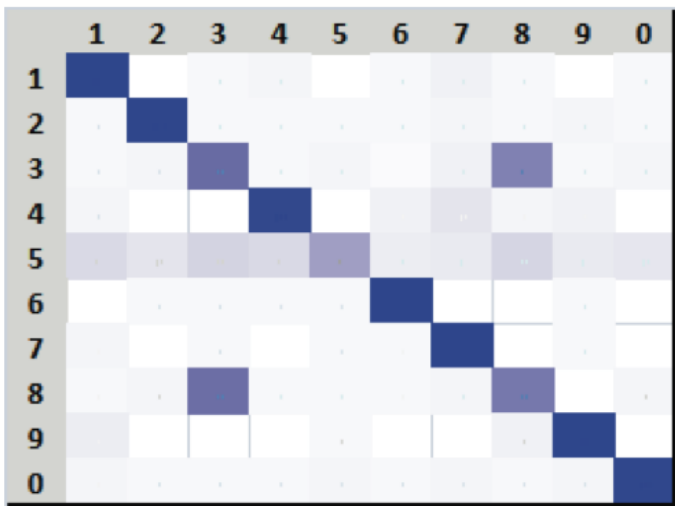
Confusion matrix

		Predicted class		
		Cat	Dog	Rabbit
Actual class	Cat	5	3	0
	Dog	2	3	1
	Rabbit	0	2	11

	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

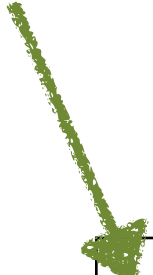


Much easier!

**Figure 2. Representations of confusion matrix for a handwritten digit classification task. (top) standard confusion matrix; (bottom) heat-map confusion matrix. It is much easier to identify underlying patterns in the visual representation; 3 and 8 are often misclassified as each other and 5 is misclassified as many different numbers.**

Very important:  
Find out what “**positive**” means

Predicated



	Cat	Dog
Cat	5	3
Dog	2	4

Actual

Very important:  
Find out what  
“positive” means

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

“False Alarm”

easy to remember  
in security  
applications



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)

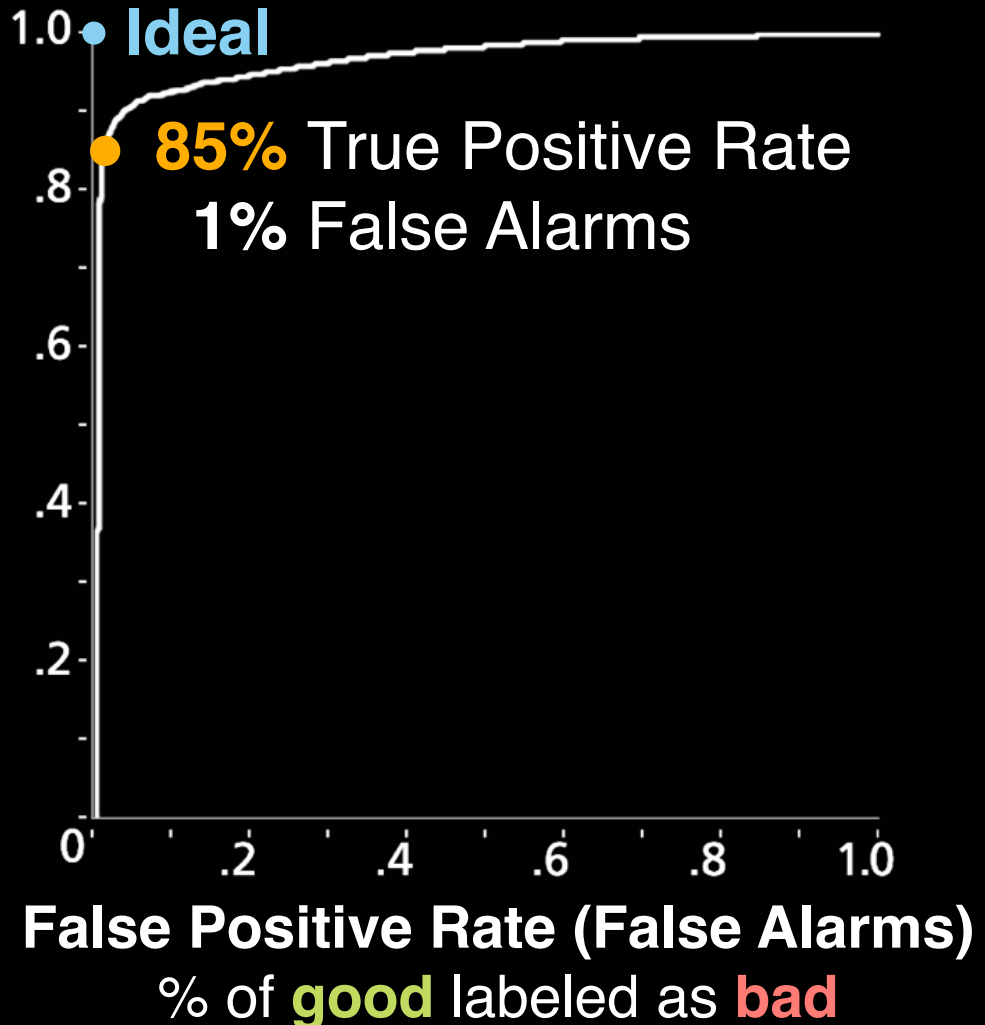
**Visualizing Classification Performance**  
using  
**ROC curve**  
(Receiver Operating Characteristic)

# Polonium's ROC Curve

Positive class: malware

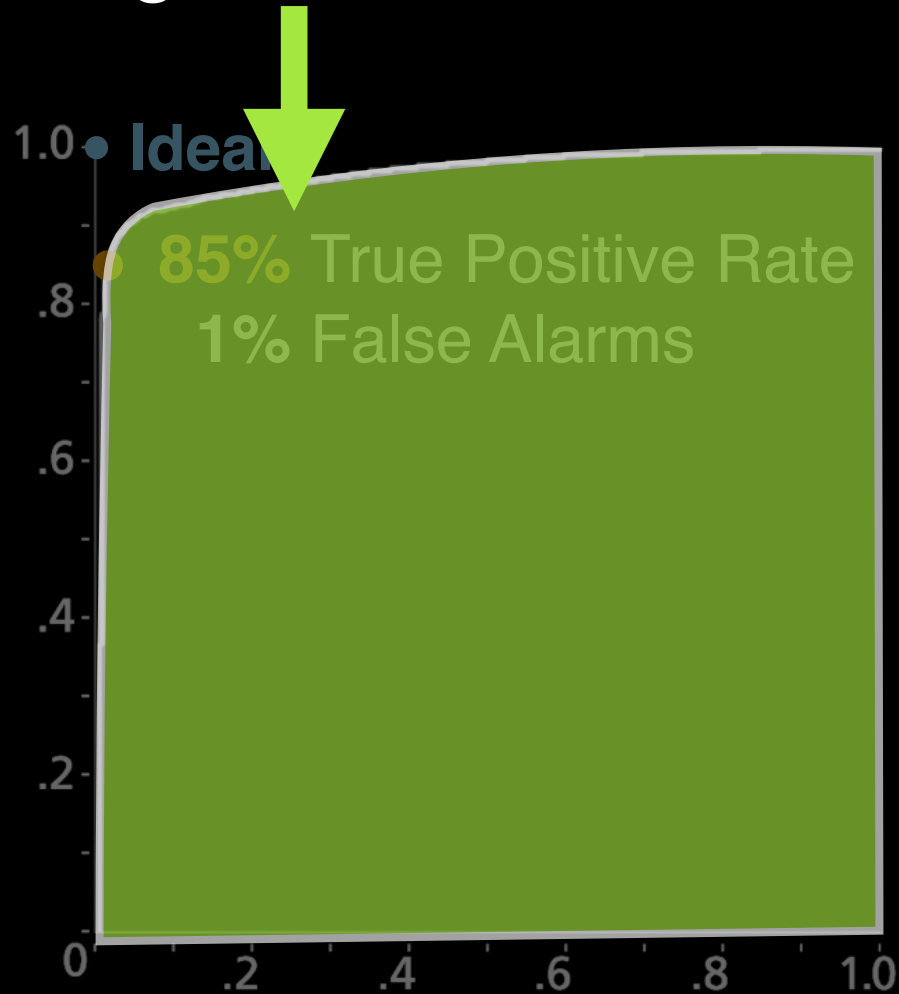
Negative class: benign

**True Positive Rate**  
% of **bad** correctly labeled



# Measuring Classification Performance

using **AUC** (Area under the curve)

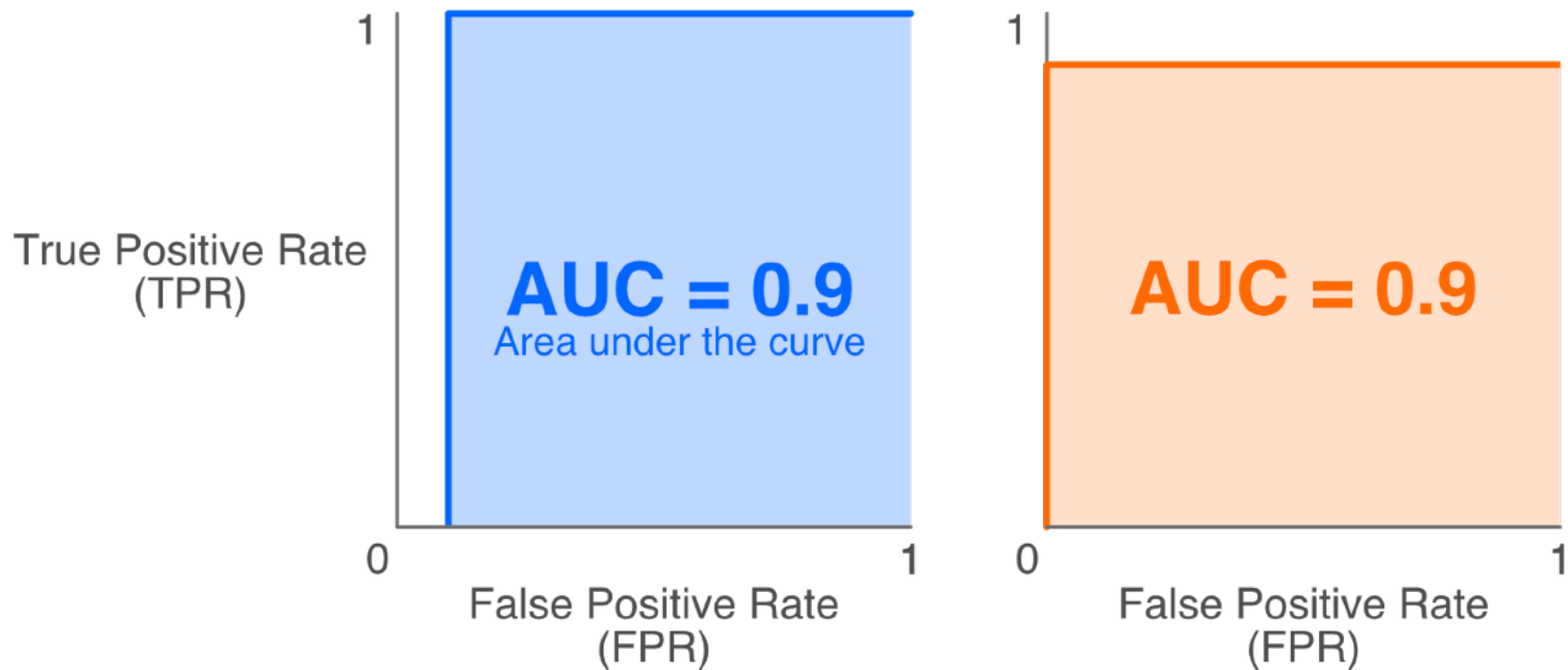




If a machine learning algorithm  
achieves **0.9 AUC** (out of 1.0),

that's a great algorithm, right?

# Be Careful with AUC!



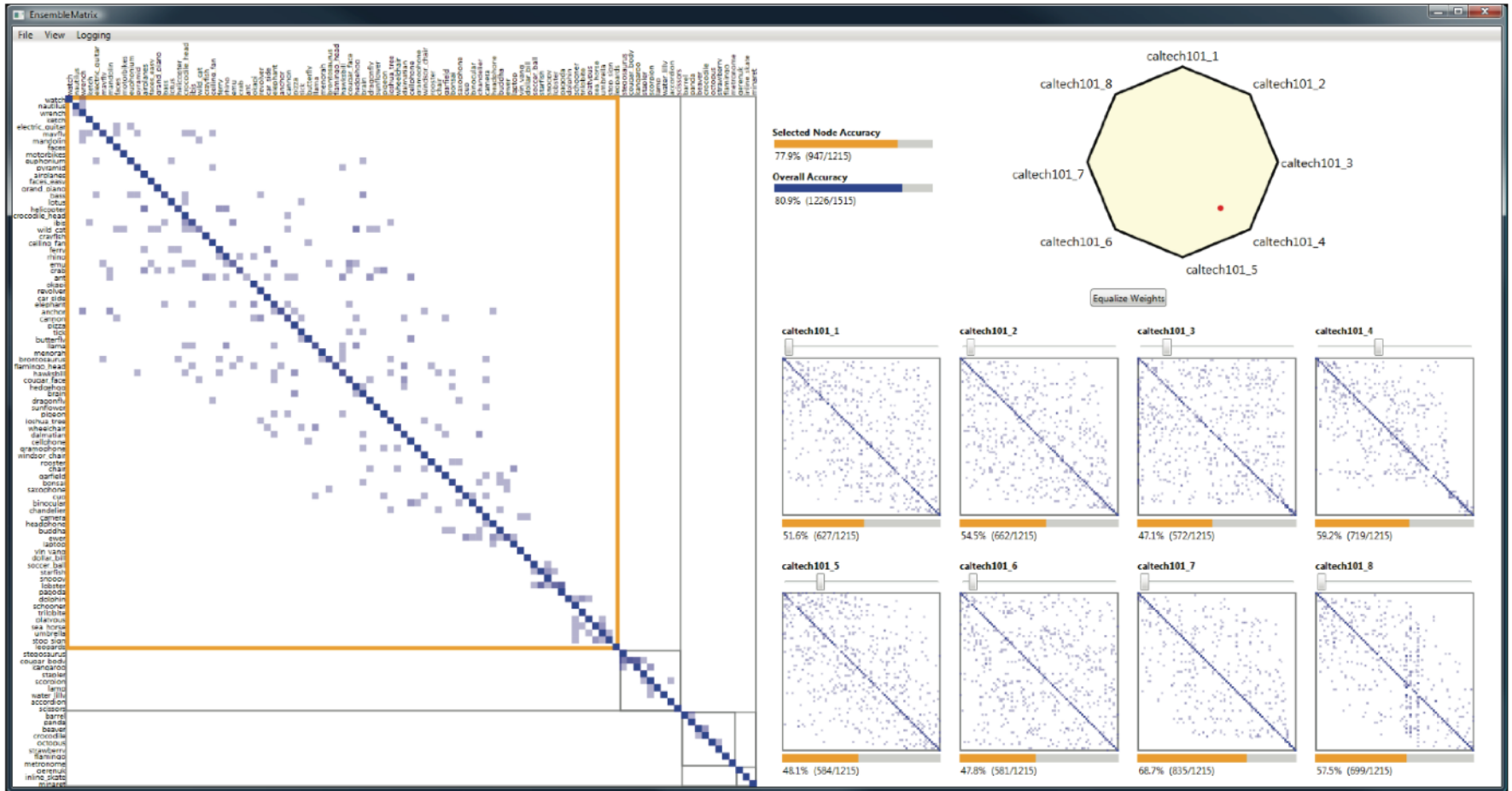
# Weights in combined models

Bagging / Random forests

- Majority voting

**Let people play with the weights?**

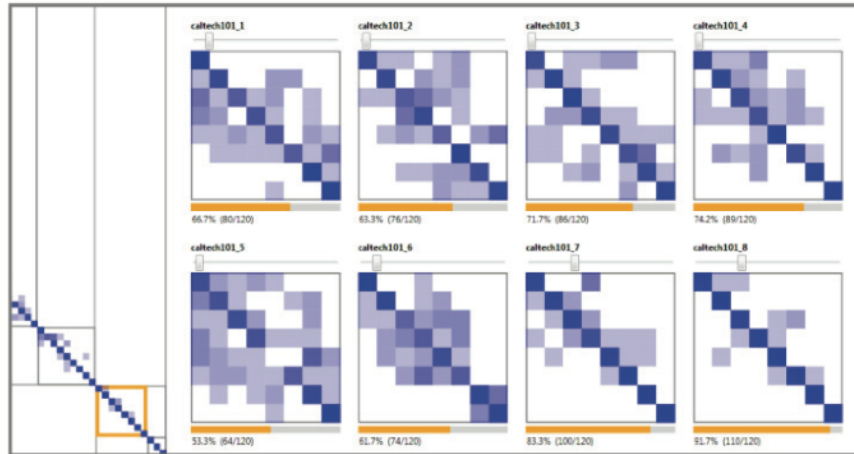
# EnsembleMatrix



**Figure 1. Primary view in EnsembleMatrix. Confusion matrices of component classifiers are shown in thumbnails on the right. The matrix on the left shows the confusion matrix of the current ensemble classifier built by the user.**

<http://research.microsoft.com/en-us/um/redmond/groups/cue/publications/CHI2009-EnsembleMatrix.pdf>

# Improving performance



**Figure 3.** After partitioning the matrix, selecting a partition, outlined in orange, causes the thumbnails to display only the data instances in that partition. The component classifiers demonstrate very different behavior in this partition, including clustering and large differences in accuracy.

- Adjust the weights of the individual classifiers
- Data partition to separate problem areas
  - Adjust weights just for these individual parts
- Caveat: evaluation used one dataset

<http://research.microsoft.com/en-us/um/redmond/groups/cue/publications/CHI2009-EnsembleMatrix.pdf>