



# Tutorial Business Analytics

Tutorial 8: Evaluation  
Decision Sciences & Systems (DSS)  
Department of Informatics  
TU München

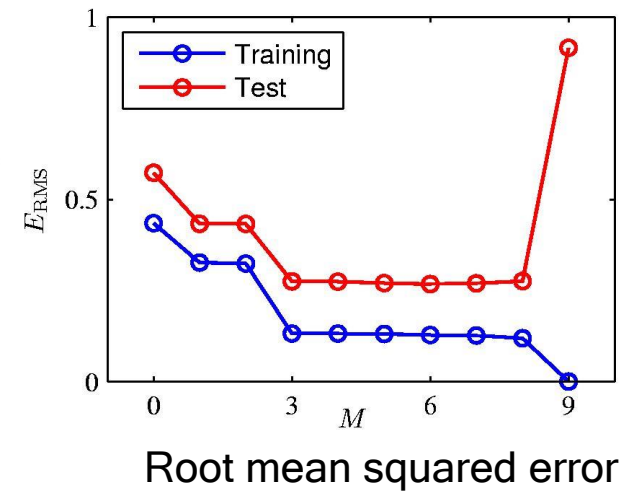
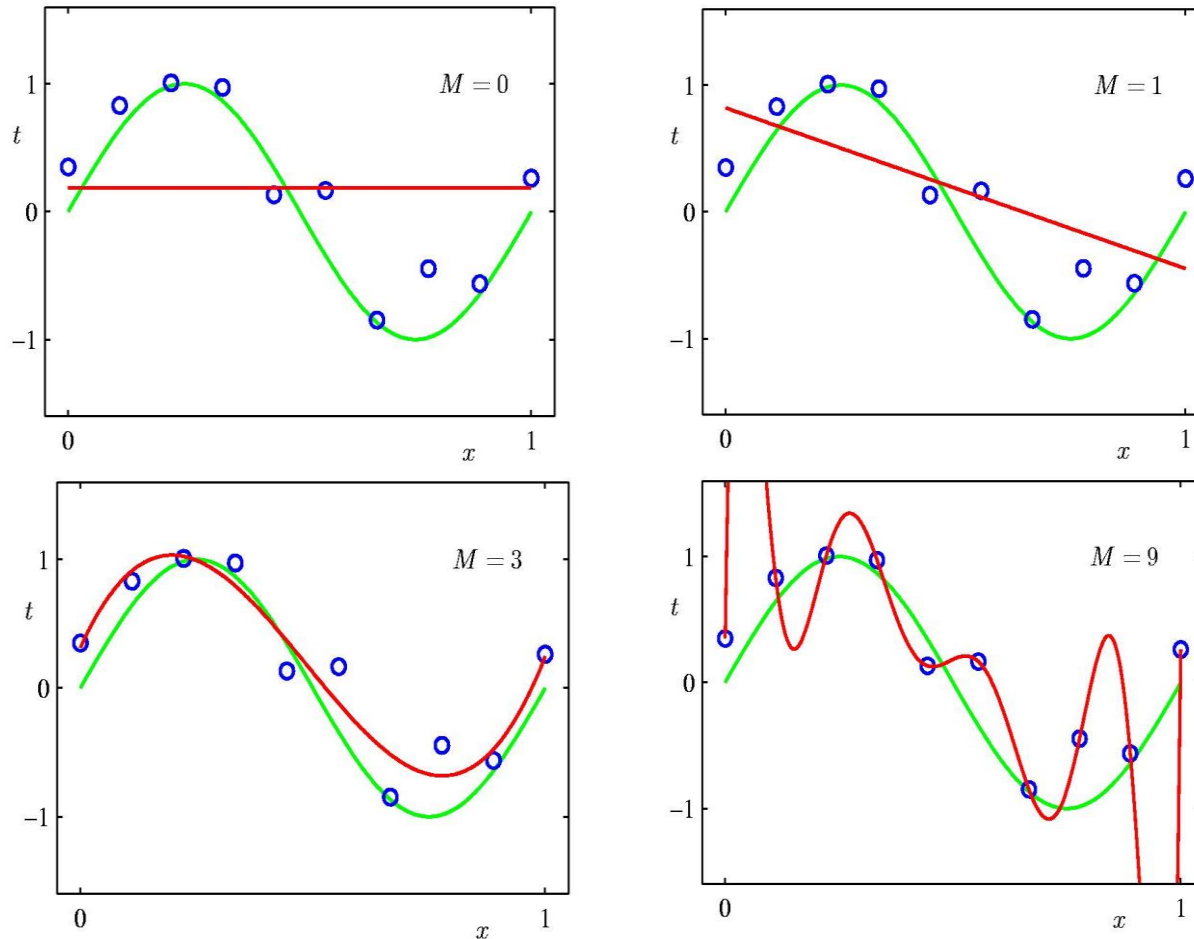
# Tutorial Business Analytics

## Agenda

- Bias-Variance Tradeoff
- Evaluation of classifiers
  - Evaluation on training set
  - Holdout
  - Cross-Validation
- Metrics for classifiers
  - Confusion matrix
  - Gain Curves
  - Lift Curves
  - ROC Curves

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## Bias-Variance Tradeoff



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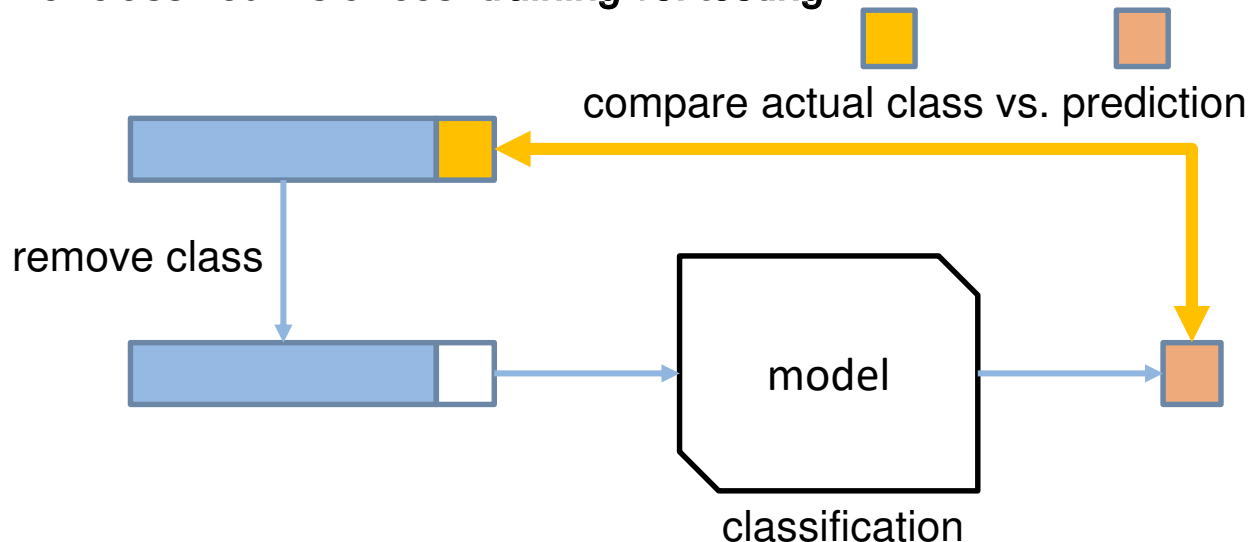
### Evaluation - “How good is your model on new data?”

The model classifies an instance, of which the class is *already known*.  
Then, *compare* model’s prediction *to actual class*.

*Which Instances to use for testing?*

Instances with known class can be used for training and testing

Competition for classified instances: **training vs. testing**

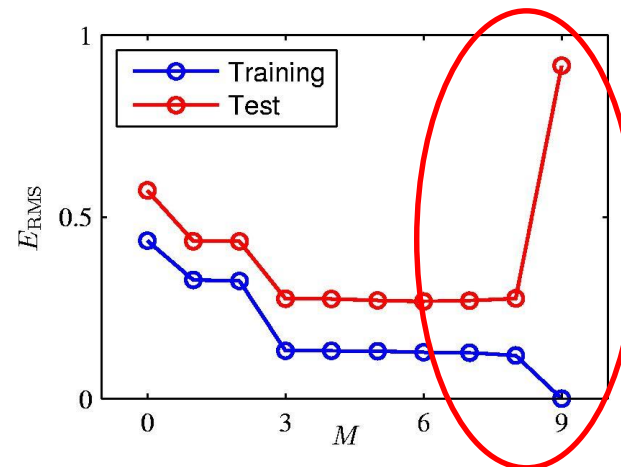


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## Evaluation using training sets

Using training data also for evaluation:

- Overfitting
- Evaluation is too optimistic, the actual error rate is way higher



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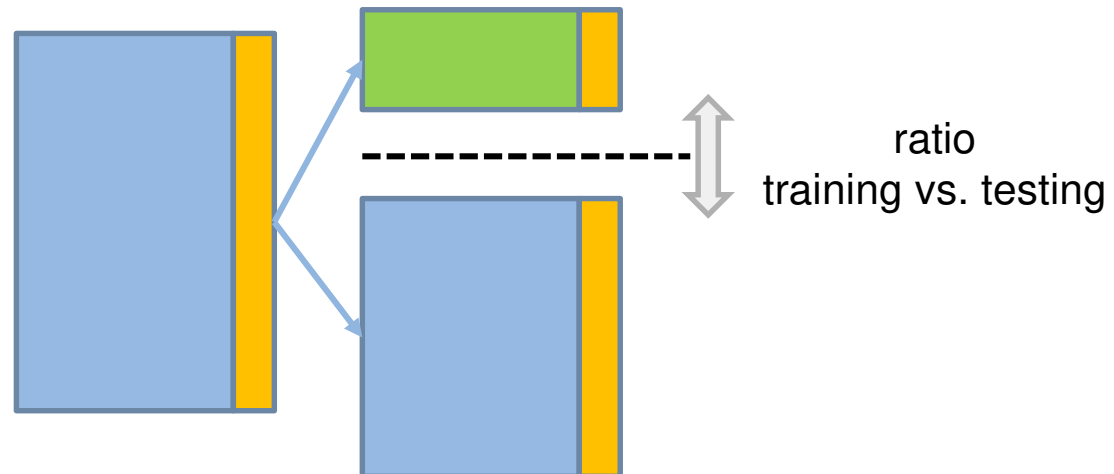
## Holdout

Holding out certain instances for evaluation, not for training:

- Reduces the number of training instances
- Composition (e.g. **distribution of the classes**) influences results

## Solution: Stratified holdout

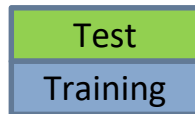
- More reliable results using stratification
- Some instances may never be used for training/testing



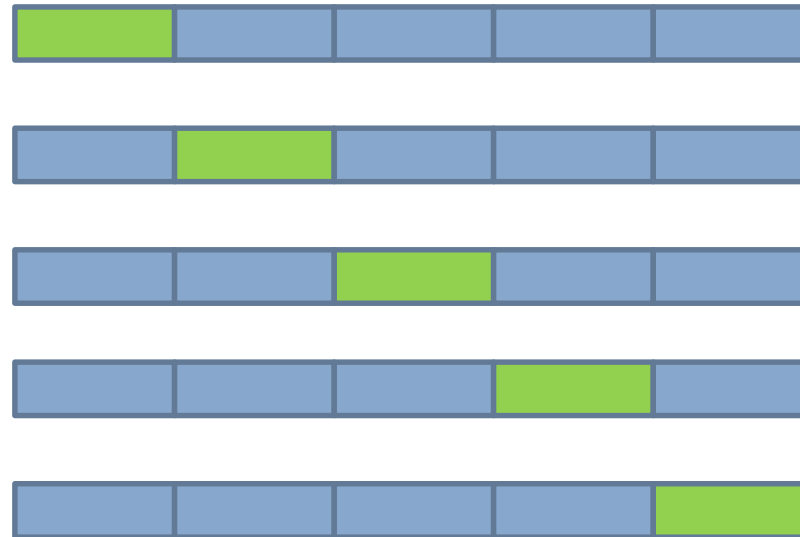
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## k-fold Cross-Validation

- Partition the data set into  $k$  complementary subsets
- Train on  $k - 1$  subsets, test on 1 subset
- Every subset is used  $k - 1$  times for training and 1 time for testing
- Stratification of subsets is possible



Example:  $k = 5$



Average error,

$$\hat{e} = \frac{1}{k} \sum_{i=1}^k e_i$$

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### Leave One Out Validation

- k-fold cross validation with  $k = N$ , where  $N$  is the number of training instances
- Deterministic, a lot of time required for computing
- Extreme class distribution of test data
- Example:  $N = 5000$   
each of which is used 4999 times for training and 1 time for testing  
5000 models need to be trained



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## Statistical tests - t-test

Results of a **validation** may be considered as random chance

- Only *significant* difference between classifiers is interesting

A **paired t-test** checks whether **mean accuracy of two classifiers are significantly different**

- $H_0: d = 0$
- $H_1: d \neq 0$  (depend on question)

$$\bar{d} = \frac{1}{k} \sum_{i=1}^k d_i \quad s_d = \sqrt{\frac{1}{k-1} \sum_i (d_i - \bar{d})^2} \quad t = \frac{\bar{d}}{s_d / \sqrt{k}} \sim t_{k-1}$$

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## Confusion Matrix

Result of an evaluation containing

- the true class
- the class predicted by the classifier

True Class	Predicted Class
1	1
1	0
0	1
0	0

True Positive  
False Negative  
False Positive  
True Negative

Confusion Matrix

		Predicted Class	
		Positive	Negative
True Class	Positive	True Positive	False Negative
	Negative	False Positive	True Negative

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## Predicting Probabilities

Classifiers can predict probabilities instead of classes

- A probability is assigned to every class
- Classify using a cutoff
- Helpful in case errors have different impacts (costs)
  - Minimize cost by selecting a suitable cutoff-value

Example:

- The cost of FN is 5 times the cost of FP
- Adjust the classifier to predict positive rather than negative
- Decrease the cutoff-value

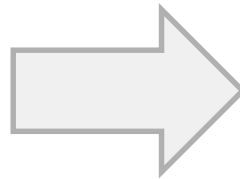
		Predicted Class	
		Positive	Negative
True Class	Positive	0	5
	Negative	1	0

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## Predicting Probabilities

The probability of a positive class is displayed, instead of the predicted class (cutoff-value=0.5)

True Class	Predicted Class
0	0
0	1
1	1
1	0
0	0
1	0
0	0
1	1
0	1
1	0



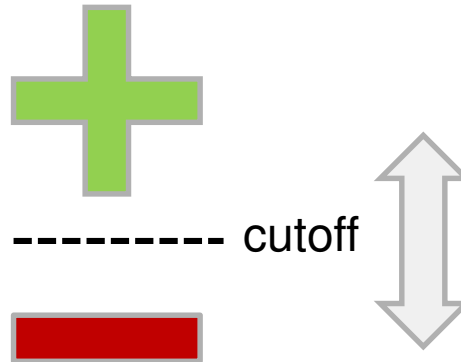
True Class	Predicted Class	Probability
0	0	0.34
0	1	0.56
1	1	0.63
1	0	0.45
0	0	0.47
1	0	0.17
0	0	0.26
1	1	0.89
0	1	0.52
1	0	0.43

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## Predicting Probabilities

- Sort by probability:

True Class	Probability
+	0.89
+	0.63
-	0.56
-	0.52
-	0.47
+	0.45
+	0.43
-	0.34
-	0.26
+	0.17

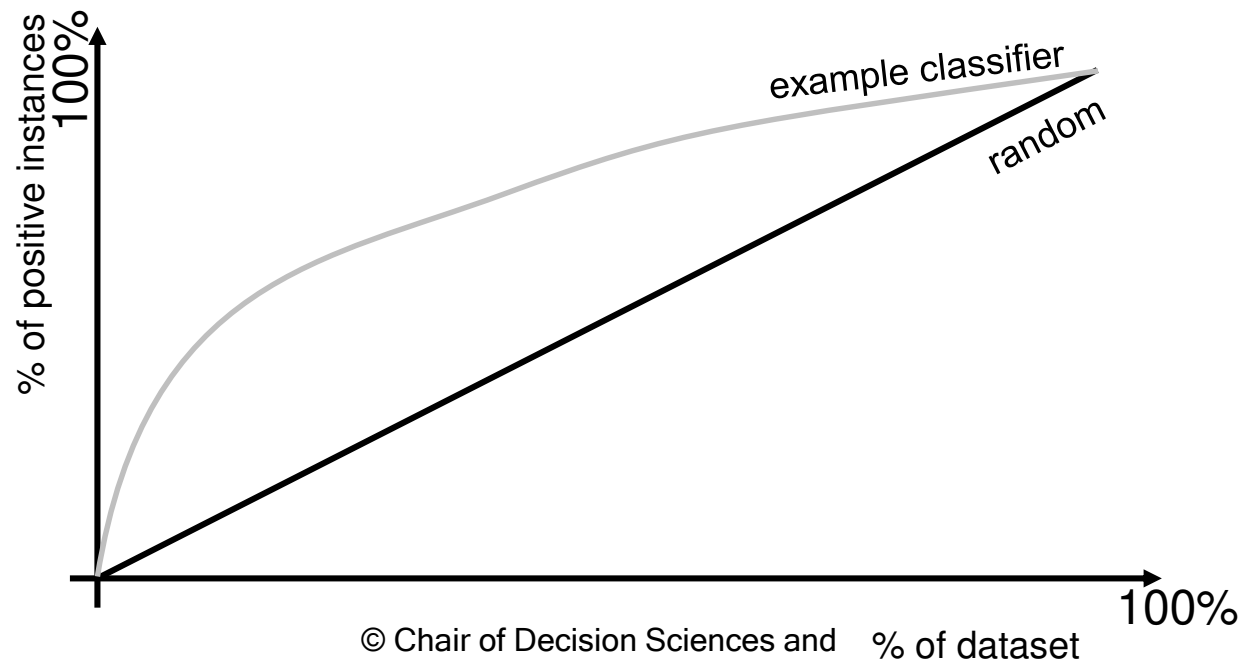


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## Gain Curve

To visualize results of different cutoffs

- Instances are sorted by probability (descending)
- x-axis: percentage of the data set (or number of instances)
- y-axis: percentage (or number) of positive instances in the fraction (subsample)



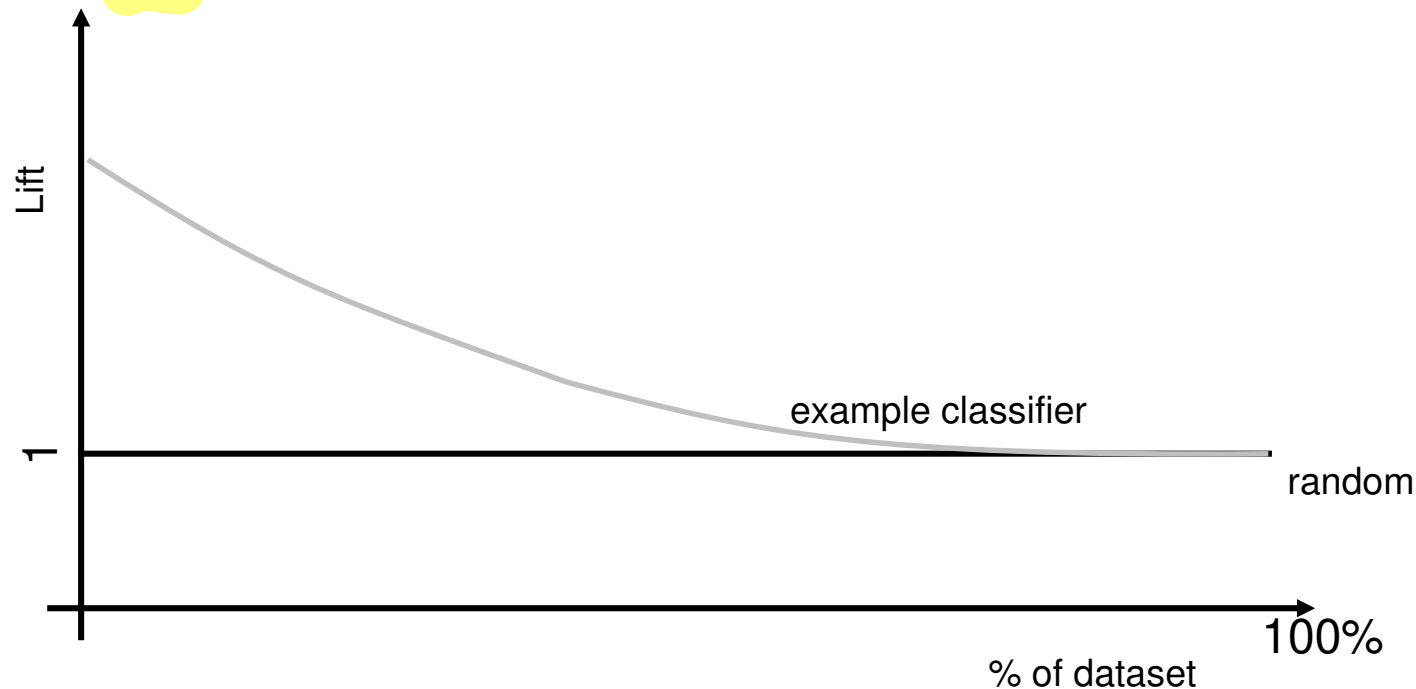
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### Lift Curve

Displays the factor between the classifier and random value for every part of the gain curve

- x-axis: percentage of the data set (or number of instances)

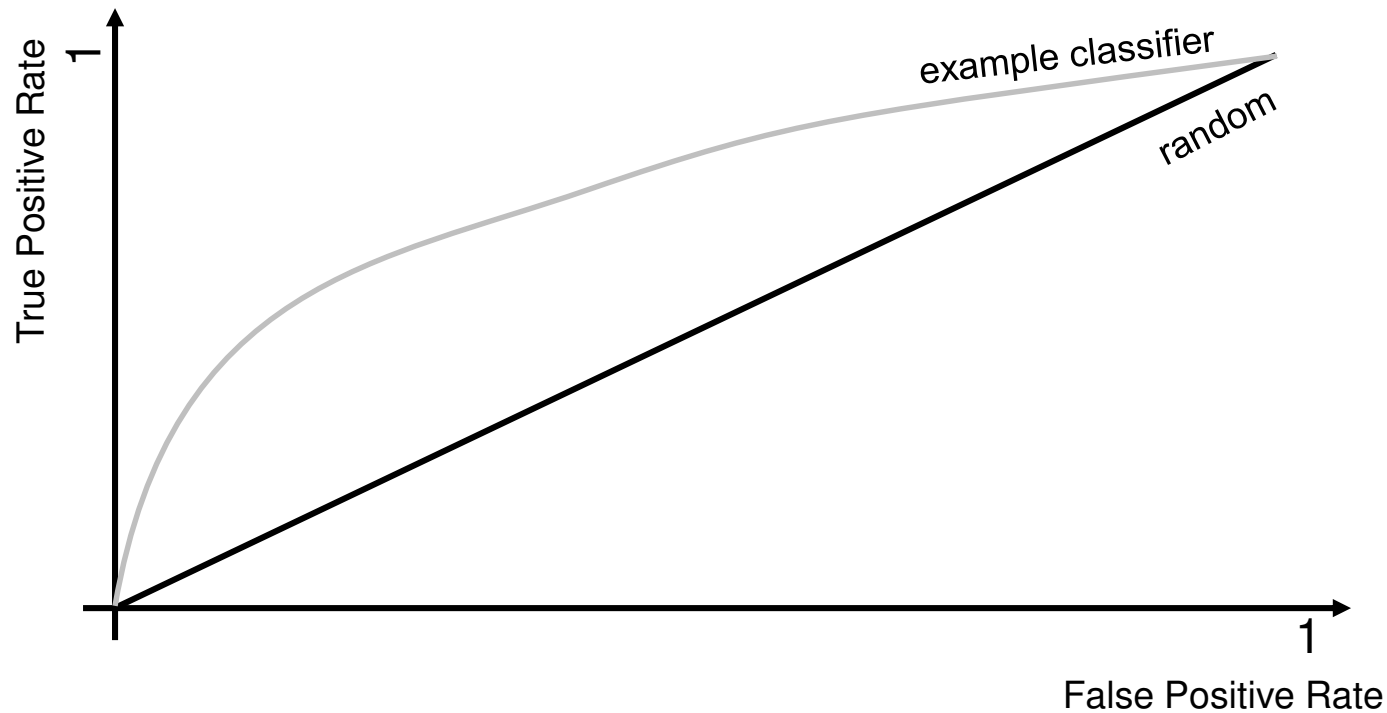
- y-axis:  $\frac{\text{Gain at } x}{x}$



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## ROC Curve

Displays the ratio of False Positive Rate and True Positive Rate





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