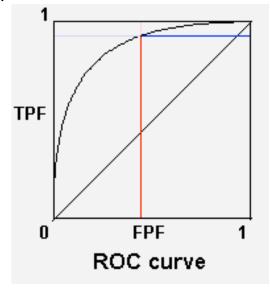


Recall: The percentage of positive labeled instances that were predicted as positive.

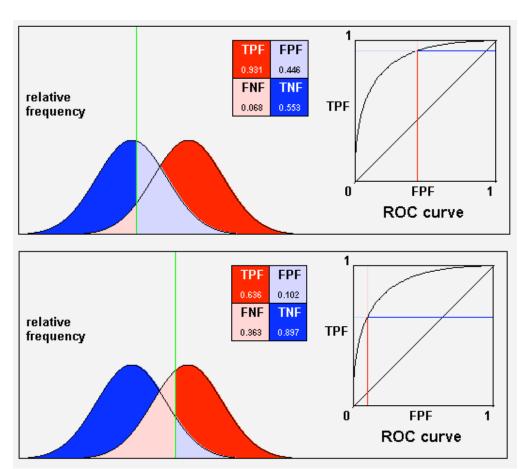
- ROCs (Receiver Operating Characteristic Curves)
  - They are commonly used for decision-making
  - They summarize the performance of the classifier
  - Using them, it is easy to visualize and analyze the output of the classifier
  - They facilitates the comparison of the performance of different methods
  - True Positives Rate (TPR):
    - Fraction of positives (faces), classified as positives.
  - True Negative Rate (TNR):
    - Fraction of *negatives* (non-faces), classified as *negatives*.
  - False Positives Rate (FPR):
    - Fraction of *positives*, classified as *negatives*.
  - False Negative Rate (FNR):
    - Fraction of negatives, classified as positives.

$$TPR + FNR = 1$$
  
 $TNR + FPR = 1$ 



#### ROCs

- A ROC represents the accuracy of the classifier for differentiating the to class.
- They define how much overlapped are the outputs for the two class (the classifier output distribution for the two class).
- When different threshold levels are applied to the classifier output, different operation points are obtained
- The larger the TPR, the larger the FPR



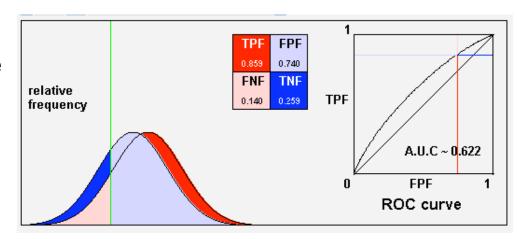
In the example the output of the classifiers each of the class follows a Gaussian distribution for

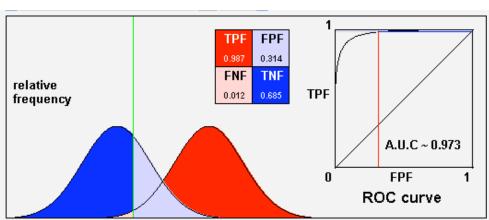
Images from: http://www.anaesthetist.com/mnm/stats/roc/

Javier Ruiz del Solar and Rodrigo Verschae, EVIC 2004

#### ROCs

- The AUC (Area Under the Curve) is used as performance measure in some cases.
- In detection problems is usual the use of the number of false positives, instead of the FPR.

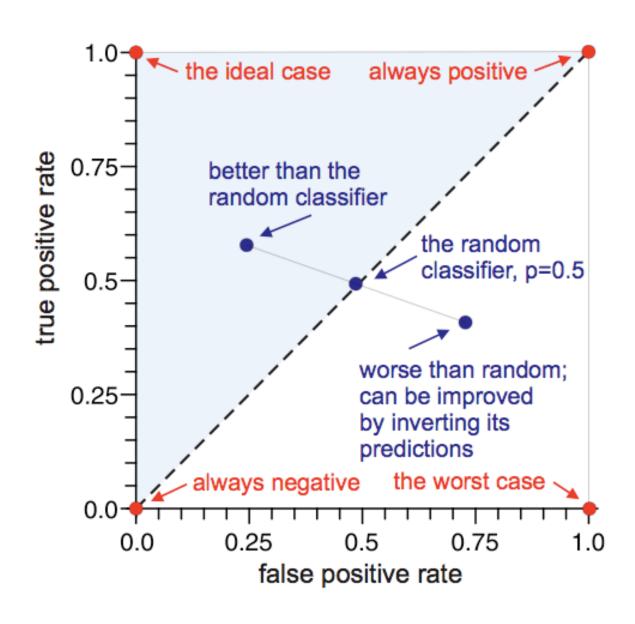




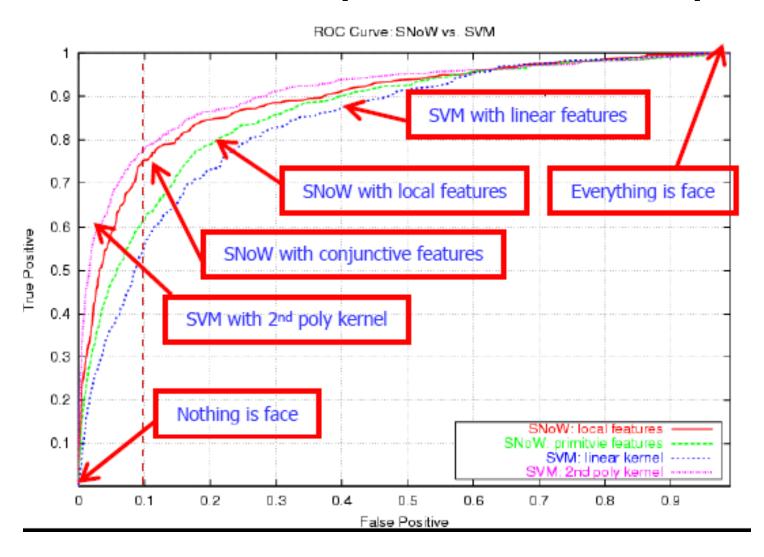
Images from: http://www.anaesthetist.com/mnm/stats/roc/

Javier Ruiz del Solar and Rodrigo Verschae, EVIC 2004

#### **Properties of the ROC space**



## Result comparison example

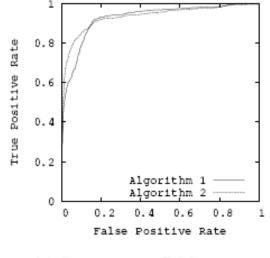


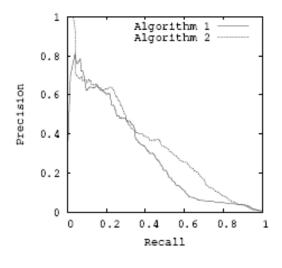
## Precision-Recall Curve

- •ROC curves can present an overly optimistic view of an algorithm's performance if there is a large skew in the class distribution.
- •In this case Precision-Recall curves give a more informative view
- •Looking at PR curves can expose differences between algorithms that are not apparent in ROC space.

•For any dataset, the ROC and PR curves for a given algorithm contain the same

points.





(a) Comparison in ROC space

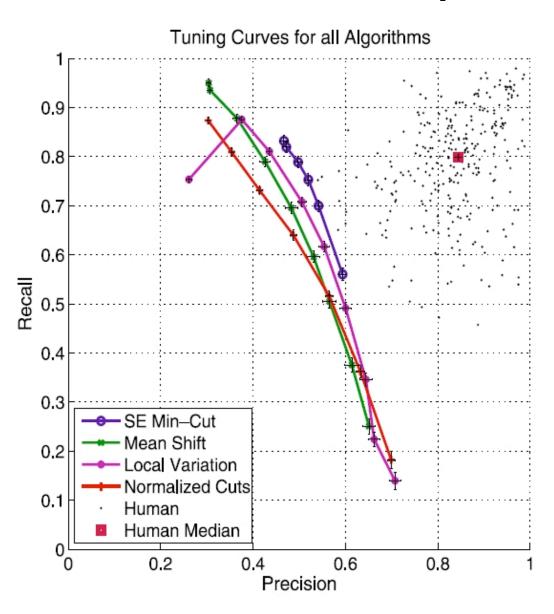
(b) Comparison in PR space

Skewness: A measure of the extent to which a probability distribution of a real-valued random variable "leans" to one side of the mean

*Precision*: The percentage of positive predictions that are correct.

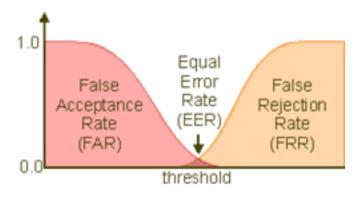
*Recall*: The percentage of positive labeled instances that were predicted as positive.

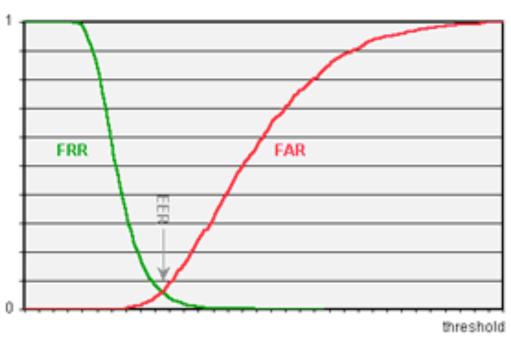
# PR Curve Example



## **FAR-FRR Curve**

Used in biometric contexts.

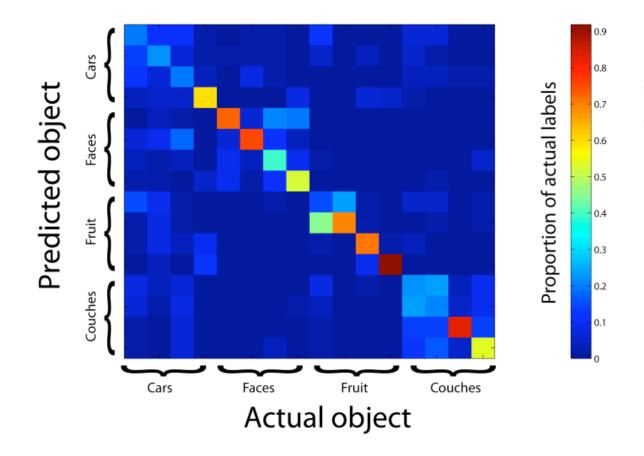




FRR=FNR FAR=FPR

### **Confusion Matrix**

A confusion matrix contains information about actual and predicted classifications done by a classification system. Each column of the matrix represents the instances in a predicted class, while each row represents the instances in an actual class.



The precision and recall values of each class can be computed individually.

## Feature Selection/Extraction

- 1. Measurements and different from features (e.g. pixels versus area).
- 2. Measurement/Feature Filtering
  - Noise filtering
  - Normalization
  - Cleaning (outliers elimination, inconsistent data, duplicate data, missing values)
  - Data discretization
- 3. Feature/Attribute Selection
  - Selection based on attribute characteristics, e.g. variability or correlation
  - Selection using a classifier
  - Elimination versus aggregation Strategies
- 4. Dimensionality Reduction
  - Features/attributes can be correlated and therefore reduced using algorithms such as PCA