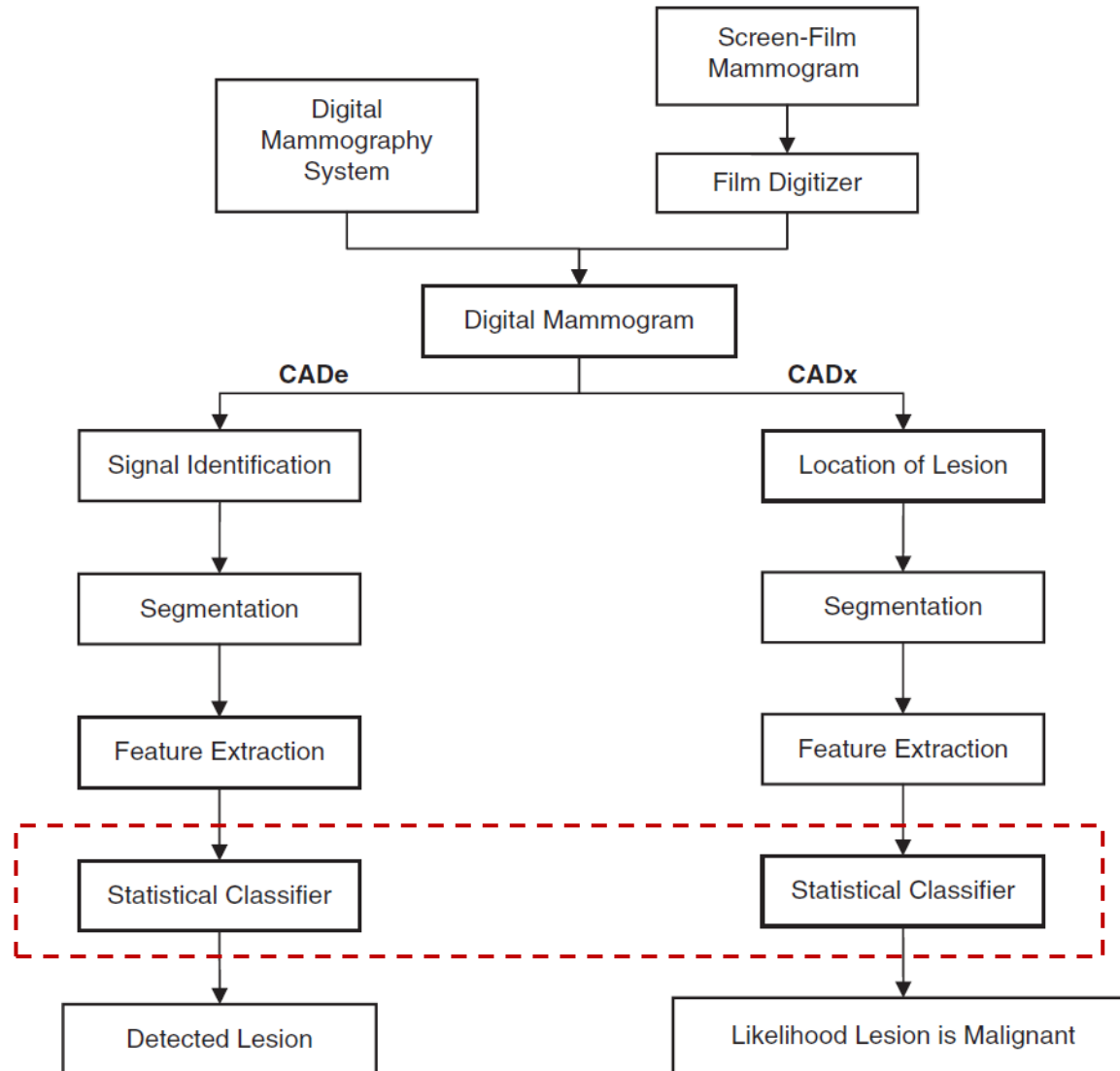


ADVANCED TOPICS IN BIOMEDICAL ENGINEERING

Topic 6: Classification

CAD



Classification Assumptions

- There is a visible difference between normal and abnormal images
 - ▣ If you cannot see consistent differences, you cannot program a CAD system to see them
 - ▣ Training set with normal and abnormal cases of interest
- Features selected describe such difference effectively
 - ▣ Irrelevant features will only confuse and misguide the CAD system
- Normal and abnormal cases form distinct clusters that are somewhat apart according to a distance measure
 - ▣ Cases from the same pathology are represented by points in feature space that with smaller distance separation than with cases from other pathologies
 - ▣ “Intra-cluster” distance is significantly smaller than “Inter-cluster” distance

Classifiers: Model

□ Parametric classification

- ▣ Assumed a certain distribution for data clusters (e.g., Gaussian)
- ▣ Estimates model parameters from the data
- ▣ Uses this a priori information to design the classification method and estimate its parameters
- ▣ Good if model is correct but bad if not (difficult to know ahead)

□ Non-Parametric classification

- ▣ Does not assume or impose any model on the data
- ▣ “Model-Free” or “Data-Dependent”

Classifiers: Learning

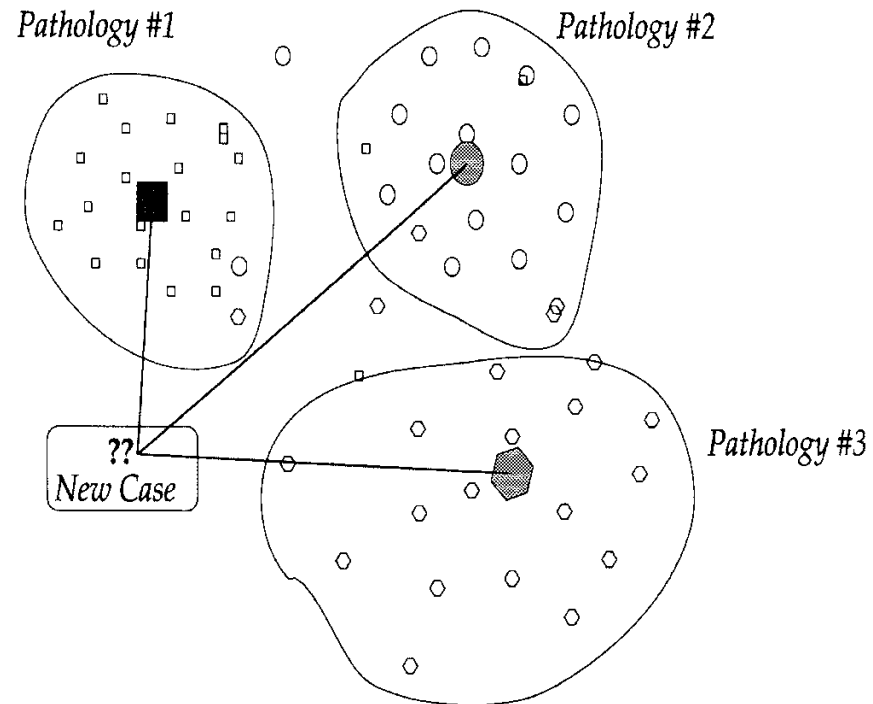
- **Supervised classifiers**
 - ▣ Classifiers are trained with data samples of known labels
 - ▣ Number of clusters known a priori and cannot be changed
 - ▣ Relies on “trainer” to provide the correct information

- **Unsupervised classifiers**
 - ▣ Classifiers are trained with data samples with unknown labels
 - ▣ Discover underlying clusters according to particular criteria
 - ▣ Interesting to identify new clusters representing sub-classes within large normal or pathological classes

Minimum Distance Classifier

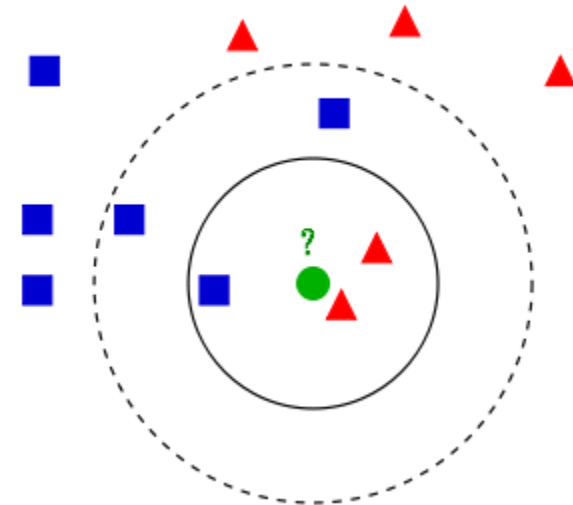
- Idea: Compute the distance between unknown case and the center of known clusters and assign the case to the closest cluster

- ▣ How to compute cluster centers
- ▣ Distance computation
- ▣ Simple but not good in many cases



K-Voting Nearest Neighbor

- Assumption: an unknown case is likely to belong to the class of its neighbors in space
- Idea: Find the K neighbors of the unknown case and make a majority vote among them and assign the case to the pathology of the majority
 - ▣ Select K in such a way to avoid “ties”: e.g., odd number for binary classification
 - ▣ Special case: $K=1$ – Nearest Neighbor Classifier
 - ▣ Matlab function “knnclassify”

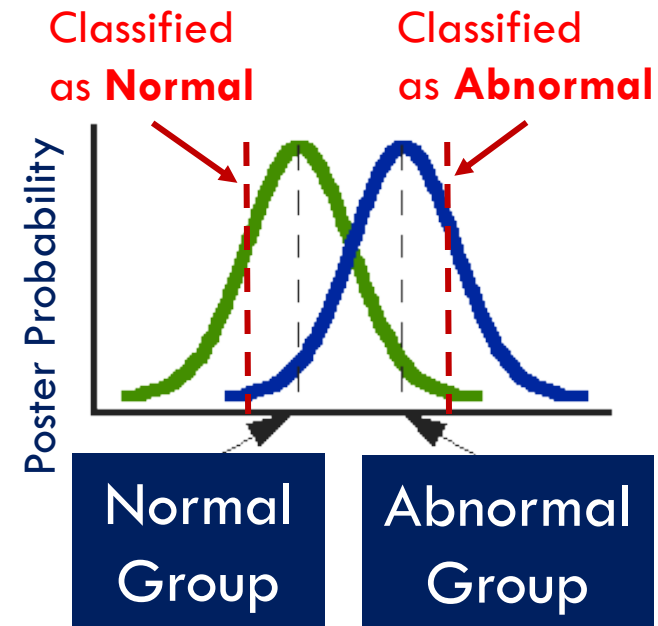


Naïve Bayesian Classifier

- Idea: Assign the unknown case to the class of largest posterior probability

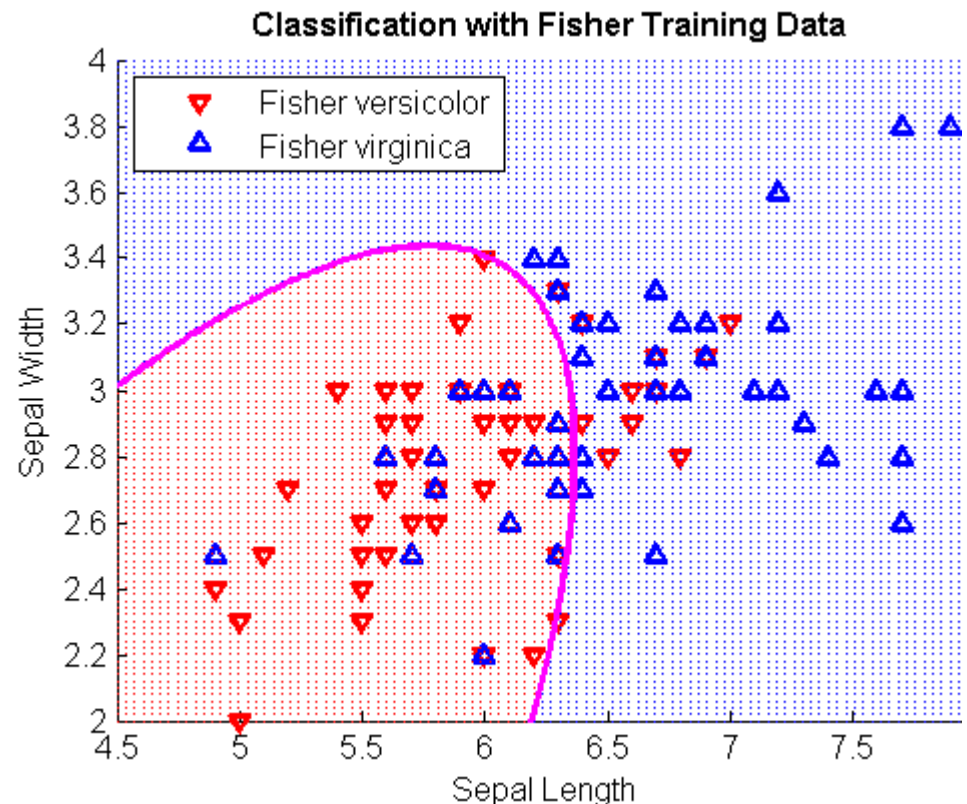
$$P(y | x) = \frac{P(x | y)P(y)}{P(x)}$$

- $P(y)$ is the *a priori* probability of y
- $P(x | y)$ is the likelihood function;
- $P(x)$ is the marginal probability
- $P(y | x)$ is the *a posteriori* probability



Discriminant Analysis

- Idea: Find the best separation surface between clusters based
- Generalization of Naïve Bayesian classifier which is a linear version of it
 - Linear: LDA
 - Quadratic: QDA
- Matlab function “classify”



Things to Watch for

- Feature normalization is important and can improve training and classification results
- Independent training and testing data is a must
 - ▣ Half and half for example
 - ▣ Leave-one-out cross validation can be used for small data sets

Assignments

- Apply 2 classification methods to the selected features obtained from previous tasks and compare their results