

# Appendix

This appendix lists the functions (m-files) developed by the authors and used in the examples in this book. Functions used that are part of MATLAB's commercial distribution have been omitted; the reader is referred to the respective MATLAB manuals.

In the following list, functions are ordered alphabetically by chapter. For further function details, including descriptions of input and output arguments, refer to MATLAB's help utility. Also see the complete source code of the listed m-files, provided as part of the software on the companion website.

## Chapter 1

**bayes\_classifier** Bayesian classification rule for  $c$  classes, modeled by Gaussian distributions (also used in [Chapter 2](#)).

**comp\_gauss\_dens\_val** Computes the value of a Gaussian distribution at a specific point (also used in [Chapter 2](#)).

**compute\_error** Computes the error of a classifier based on a data set (also used in [Chapter 4](#)).

**em\_alg\_function** EM algorithm for estimating the parameters of a mixture of normal distributions, with diagonal covariance matrices.

**EM\_pdf\_est** EM estimation of the pdfs of  $c$  classes. It is assumed that the pdf of each class is a mixture of Gaussians and that the respective covariance matrices are diagonal.

**euclidean\_classifier** Euclidean classifier for the case of  $c$  classes.

**Gaussian\_ML\_estimate** Maximum Likelihood parameters estimation of a multivariate Gaussian distribution.

**generate\_gauss\_classes** Generates a set of points that stem from  $c$  classes, given the corresponding a priori class probabilities and assuming that each class is modeled by a Gaussian distribution (also used in [Chapter 2](#)).

**k\_nn\_classifier**  $k$ -nearest neighbor classifier for  $c$  classes (also used in [Chapter 4](#)).

**knn\_density\_estimate**  $k$ -nn-based approximation of a pdf at a given point.

**mahalanobis\_classifier** Mahalanobis classifier for  $c$  classes.

**mixt\_model** Generates a set of data vectors that stem from a mixture of normal distributions (also used in [Chapter 2](#)).

**mixt\_value** Computes the value of a pdf that is given as a mixture of normal distributions, at a given point.

**mixture\_Bayes** Bayesian classification rule for  $c$  classes, whose pdf's are mixtures of normal distributions.

**Parzen\_gauss\_kernel** Parzen approximation of a pdf using a Gaussian kernel.

**plot\_data** Plotting utility, capable of visualizing 2-dimensional data sets that consist of, at most, 7 classes.

**Auxiliary functions** gauss.

## Chapter 2

**base\_clas\_coord** Implements a specific weak classifier.

**base\_clas\_coord\_out** Computes the output of the weak classifier implemented by the `base_clas_coord` function.

**boost\_clas\_coord** Generation of a “strong” classifier, using the Adaboost algorithm, that utilizes weak classifiers generated by the `base_clas_coord` function.

**boost\_clas\_coord\_out** Computes the output of a “strong” classifier  $B$  as a weighted sum of the outputs of the weak classifiers.

**CalcKernel** Computes the value of a kernel function between two points.

**kernel\_perce** Implements the kernel perceptron algorithm.

**NN\_evaluation** Returns the classification error of a neural network based on a data set.

**NN\_training** Returns a trained multilayer perceptron.

**perce** Realizes the perceptron learning rule, in a batch mode.

**perce\_online** Realizes the online perceptron learning rule.

**plot\_kernel\_perce\_reg** Plots the decision boundary that is generated by the kernel perceptron algorithm.

**plot\_NN\_reg** Plots the decision boundary that is formed by a neural network.

**SMO2** Generates a SVM classifier using either Platt’s algorithm or one of its two modifications proposed by Keerthi.

**SSErr** Generates the linear classifier that optimizes the sum of error squares criterion.

**svcplot\_book** Support Vector Machine plotting utility. It plots the decision regions, the decision surfaces and the margin obtained by a SVM classifier.

## Chapter 3

**cut\_cylinder\_3D** Generates a cut cylinder in the 3-dimensional space.

**im\_point** Performs the projection of a vector on the subspace spanned by the first  $m$  principal components, that result after performing kernel PCA on a data set.

**K\_fun** Computes the value of a kernel function (polynomial or exponential) for two vectors.

**kernel\_PCA** Performs kernel PCA based on a given set of data vectors.

**lapl\_eig** Performs Laplacian eigenmap based on a given data set.

**pca\_fun** Performs Principal Component Analysis (PCA) based on a data set.

**plot\_orig\_trans\_kPCA** Plots, in different figures, (a) the data points and the classifier in the original (2-dimensional) data space and (b) the projections of the data points and the classifier in the space spanned by the two most significant principal components, as they are computed using the kernel PCA method.

**scatter\_mat** Computes the within scatter matrix, the between scatter matrix and the mixture scatter matrix for a  $c$ -class classification problem, based on a given data set.

**spiral\_3D** Creates a 3-dimensional Archimedes spiral.

**svd\_fun** Performs Singular Value Decomposition (SVD) of a matrix.

## Chapter 4

**compositeFeaturesRanking** Scalar feature ranking that takes into account the cross-correlation coefficient.

**divergence** Computes the divergence between two classes.

**divergenceBhata** Computes the Bhattacharyya distance between two classes.

**exhaustiveSearch** Exhaustive search for the best feature combination, depending on the adopted class separability measure.

**Fisher** Computes Fisher's discriminant ratio of a scalar feature in a 2-class problem.

**normalizeMnmx** Performs MinMax normalization in a given interval  $[l\ r]$ .

**normalizeSoftmax** Performs Softmax normalization in the interval  $[0\ 1]$ .

**normalizeStd** Performs data normalization to zero mean and standard deviation equal to 1.

**plotData** Plotting utility for class data.

**plotHist** Plots the histograms of two classes for the same feature.

**ROC** Plots the ROC curve and computes the area under the curve.

**ScalarFeatureSelection Ranking** Features are treated individually and are ranked according to the adopted class separability criterion.

**ScatterMatrices** Class separability measure, which is computed using the within-class and mixture scatter matrices.

**SequentialBackward Selection** Feature vector selection by means of the Sequential Backward Selection technique.

**SequentialForward FloatingSelection** Feature vector selection by means of the Sequential Forward Floating Selection technique.

**SequentialForward Selection** Feature vector selection by means of the Sequential Forward Selection technique.

**simpleOutlierRemoval** Removes outliers from a normally distributed data set by means of the thresholding method.

## Chapter 5

**BackTracking** Performs backtracking on a matrix of node predecessors and returns the best path. This function is also used in [Chapter 6](#).

**DTWItakura** Computes the Dynamic Time Warping cost between two feature sequences, based on the standard Itakura local constraints.

**DTWItakuraEndp** Similar to *DTWItakura*, with the addition that endpoints constraints are allowed in the test sequence.

**DTWSakoe** Computes the Dynamic Time Warping cost between two feature sequences, based on the Sakoe-Chiba local constraints.

**DTWSakoeEndp** Similar to *DTWSakoe*, with the addition that endpoints constraints are allowed in the test sequence.

**editDistance** Computes the Edit (Levenstein) distance between two sequences of characters.

**Auxiliary functions** *stEnergy*, *stZeroCrossingRate*, *IsoDigitRec*.

## Chapter 6

**BWDoHMMsc** Computes the recognition probability of an HMM, given a sequence of discrete observations, by means of the scaled version of the Baum-Welch (any-path) method.

**BWDoHMMst** Same as *BWDoHMMsc*, except that no scaling is employed.

**MultSeqTrainDoHMMBWsc** Baum-Welch training (scaled version) of a Discrete Observation HMM, given multiple training sequences.

**MultSeqTrainDoHMMVITsc** Viterbi training (scaled version) of a Discrete Observation HMM, given multiple training sequences.

**MultSeqTrainCoHMMBWsc** Baum-Welch training (scaled version) of a Continuous Observation HMM, given multiple training sequences.

**VitCoHMMsc** Computes the scaled Viterbi score of an HMM, given a sequence of  $l$ -dimensional vectors of continuous observations, under the assumption that the pdf of each state is a Gaussian mixture.

**VitCoHMMst** Same as *VitCoHMMsc* except that no scaling is employed.

**VitDoHMMsc** Computes the scaled Viterbi score of a Discrete Observation HMM, given a sequence of observations.

**VitDoHMMst** Same as *VitDoHMMsc*, except that no scaling is employed.

## Chapter 7

**agglom** Generalized Agglomerative Scheme (GAS) for data clustering. It runs, on demand, either the single-link or the complete-link algorithm.

**BSAS** Basic Sequential Algorithmic Scheme (BSAS algorithm) for data clustering.

**CL\_step** Performs a step of the complete-link algorithm.

**dendrogram\_cut** Determines the clusterings of a hierarchy that best fit the underlying clustering structure of the data set at hand.

**fuzzy\_c\_means** FCM algorithm for data clustering.

**GMDAS** Generalized Mixture Decomposition Algorithmic Scheme (GMDAS algorithm) for data clustering.

**k\_means** k-means clustering algorithm.

**k\_medoids** k-medoids clustering algorithm.

**LLA** Competitive leaky learning algorithm for data clustering.

**possibi** Possibilistic clustering algorithm, adopting the squared Euclidean distance.

**SL\_step** Performs a step of the single-link algorithm.

**spectral\_Ncut2** Spectral clustering based on the normalized cut criterion.

**valley\_seeking** Valley-seeking algorithm for data clustering.

**Auxiliary functions** cost\_comput, distan, distant\_init, rand\_data\_init, rand\_init, reassign.