
```

function [trainedClassifier, validationAccuracy] =
    gaussianSVMTrain(trainingData)
% trainClassifier(trainingData)
% returns a trained classifier and its validation accuracy.
% This code recreates the classification model trained in
% Classification Learner app.
%
% Input:
%     trainingData: the training data of same data type as imported
%                   in the app (table or matrix).
%
% Output:
%     trainedClassifier: a struct containing the trained classifier.
%                       The struct contains various fields with information about the
%                       trained classifier.
%
%     trainedClassifier.predictFcn: a function to make predictions
%                                   on new data. It takes an input of the same form as this
%                                   training
%                                   code (table or matrix) and returns predictions for the
%                                   response.
%     If you supply a matrix, include only the predictors columns
%     (or
%     rows).
%     validationAccuracy: a double containing the validation
%     accuracy
%     score in percent. In the app, the History list displays this
%     overall accuracy score for each model.
% Use the code to train the model with new data.
% To retrain your classifier, call the function from the command line
% with your original data or new data as the input argument
% trainingData.
%
% For example, to retrain a classifier trained with the original data
% set
% T, enter:
%     [trainedClassifier, validationAccuracy] = trainClassifier(T)
%
% To make predictions with the returned 'trainedClassifier' on new
% data T,
% use
%     yfit = trainedClassifier.predictFcn(T)
%
% To automate training the same classifier with new data, or to learn
% how
% to programmatically train classifiers, examine the generated code.

% Auto-generated by MATLAB on 14-Feb-2016 19:34:18

% Convert input to table
inputTable = table(trainingData);

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{'column'}}), 'VariableNames',
{'column_1', 'column_2', 'column_3', 'column_4', 'column_5', 'column_6', 'column_7', 'column_8', 'column_9', 'column_10', 'column_11', 'column_12', 'column_13', 'column_14', 'column_15', 'column_16', 'column_17', 'column_18', 'column_19', 'column_20', 'column_21', 'column_22', 'column_23', 'column_24', 'column_25', 'column_26', 'column_27', 'column_28', 'column_29', 'column_30', 'column_31', 'column_32', 'column_33', 'column_34', 'column_35', 'column_36', 'column_37', 'column_38', 'column_39', 'column_40', 'column_41', 'column_42', 'column_43', 'column_44', 'column_45', 'column_46', 'column_47', 'column_48', 'column_49', 'column_50', 'column_51', 'column_52', 'column_53', 'column_54', 'column_55', 'column_56', 'column_57', 'column_58', 'column_59', 'column_60', 'column_61', 'column_62', 'column_63', 'column_64', 'column_65', 'column_66', 'column_67', 'column_68', 'column_69', 'column_70', 'column_71', 'column_72', 'column_73', 'column_74', 'column_75', 'column_76', 'column_77', 'column_78', 'column_79', 'column_80', 'column_81', 'column_82', 'column_83', 'column_84', 'column_85', 'column_86', 'column_87', 'column_88', 'column_89', 'column_90', 'column_91', 'column_92', 'column_93', 'column_94', 'column_95', 'column_96', 'column_97', 'column_98', 'column_99', 'column_100'}

% Extract predictors and response
% This code processes the data into the right shape for training the
% classifier.
predictorNames =
{'column_1', 'column_2', 'column_3', 'column_4', 'column_5', 'column_6', 'column_7', 'column_8', 'column_9', 'column_10', 'column_11', 'column_12', 'column_13', 'column_14', 'column_15', 'column_16', 'column_17', 'column_18', 'column_19', 'column_20', 'column_21', 'column_22', 'column_23', 'column_24', 'column_25', 'column_26', 'column_27', 'column_28', 'column_29', 'column_30', 'column_31', 'column_32', 'column_33', 'column_34', 'column_35', 'column_36', 'column_37', 'column_38', 'column_39', 'column_40', 'column_41', 'column_42', 'column_43', 'column_44', 'column_45', 'column_46', 'column_47', 'column_48', 'column_49', 'column_50', 'column_51', 'column_52', 'column_53', 'column_54', 'column_55', 'column_56', 'column_57', 'column_58', 'column_59', 'column_60', 'column_61', 'column_62', 'column_63', 'column_64', 'column_65', 'column_66', 'column_67', 'column_68', 'column_69', 'column_70', 'column_71', 'column_72', 'column_73', 'column_74', 'column_75', 'column_76', 'column_77', 'column_78', 'column_79', 'column_80', 'column_81', 'column_82', 'column_83', 'column_84', 'column_85', 'column_86', 'column_87', 'column_88', 'column_89', 'column_90', 'column_91', 'column_92', 'column_93', 'column_94', 'column_95', 'column_96', 'column_97', 'column_98', 'column_99', 'column_100'}
predictors = inputTable(:, predictorNames);
response = inputTable.column_37;

% Set up holdout validation
cvp = cvpartition(response, 'Holdout', 0.31);
trainingPredictors = predictors(cvp.training,:);
trainingResponse = response(cvp.training,:);

% Train a classifier
% This code specifies all the classifier options and trains the
% classifier.
template = templateSVM(...
    'KernelFunction', 'gaussian', ...
    'PolynomialOrder', [], ...
    'KernelScale', 'auto', ...
    'BoxConstraint', 3, ...
    'Standardize', true);
classificationSVM = fitcecoc(...
    trainingPredictors, ...
    trainingResponse, ...
    'Learners', template, ...
    'Coding', 'onevsall', ...
    'ClassNames', [1; 2; 3; 4; 5; 6]);

svmPredictFcn = @(x) predict(classificationSVM, x);
validationPredictFcn = @(x) svmPredictFcn(x);

% Compute validation accuracy
validationPredictors = predictors(cvp.test,:);
validationResponse = response(cvp.test,:);

[validationPredictions, validationScores] =
    validationPredictFcn(validationPredictors);
correctPredictions = (validationPredictions == validationResponse);
validationAccuracy = sum(correctPredictions)/
    length(correctPredictions);

Not enough input arguments.

Error in gaussianSVMTrain (line 43)
inputTable = table(trainingData);

```

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