```
응 {
Support vector machine
응 }
clc
% ######## Set aside some of the data for testing #########
% We want to shuffle both inputs and outputs while preserving the
% correlation
p = randperm(length(final inputs svm));
random_final_inputs = final_inputs_svm(p, :);
random_final_targets = final_targets_svm(p, :);
% set some percentage of it aside for testing
test_percent = 85;
test element count =
uint32((test_percent/100)*length(random_final_inputs));
% Define which features to include in the input set.
train_inputs = random_final_inputs(1:end-test_element_count,:);
Take all the rows, and all the 10 features as inputs. Could also use:
inputs = dataSet(:,1:end-2).
test_inputs= random_final_inputs(end-test_element_count+1:end,:);
% Define the target set
train_targets = random_final_targets(1:end-test_element_count, :);
test_targets = random_final_targets(end-test_element_count+1:end,:);
%Create a SVM Model template to fit into fitcecoc()
t = templateSVM('Standardize',true);
SVMModel_Multi = fitcecoc(train_inputs,train_targets,'Learners',t);
SVMModel_Multi.ClassNames
CodingMat = SVMModel_Multi.CodingMatrix;
% predict using the test data extracted earlier
targets_from_test_prediction =
predict(SVMModel_Multi,test_inputs,'Verbose',1);
% print out the Binary Loss function has been applied to the model
 ('hinge' by default)
SVMModel_Multi.BinaryLoss
```

```
% Plot Confusion Matrix
confusionchart(test_targets, targets_from_test_prediction);
응 {
now I have created new method for labelling data for SVMs as it is a
different process
than that for ANNs. Created a multiclass SVM classifier for the data
currently receiving 100% classification accuracy using the hinge
binary loss
function. Will switch to assessing the classifier performance in a
more robust way
by obtaining the Posterior Probabilities.
Explained
openExample('stats/
EstimatePosteriorProbabilitiesUsingECOCClassifiersExample')
https://www.youtube.com/watch?v=FpKiHpYnY_I
응 }
```

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