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Vincent Purcell - HW 4 - ECE487

```
clear; clc; close all;
```

Problem 4.6

Problem 4.6 from the Text on page 248.

```
% Data generation based on inputs from text book
rng('default')
rng(1)
m = [-5 5 5 -5; 5 -5 5 -5];
s = 2;
N = 100;
[x1,y1] = data_generator(m,s,N);
x1 = x1';
y1 = y1';
rng(10);
[x2,y2] = data_generator(m,s,N);
x2 = x2';
y2 = y2';

C_vec = [1,100,1000]';
sigma_vec = [0.5,1,2,4]';
tol = 0.001;

% Create 12 models and plot them based on all combinations of sigma and C
for i=1:size(C_vec)
    for j=1:size(sigma_vec)
        plotSVM(x1,y1,x2,y2,tol,C_vec(i),sigma_vec(j));
    end
end

% Call Decision Tree Function
decisionTree(x1,y1,x2,y2);
```

SVM Classification

Classification and Plot Function

```
function plotSVM(x1,y1,x2,y2,tol,C,sigma)

%Get classifier model and errors
[model, test_err, train_err] = SVM_clas(x1,y1,x2,y2,tol,C,sigma);
svInd = model.IsSupportVector;
%Below plotting methods adapted from fitsvm MATLAB documentation
h = 0.02;
[X1,X2] = meshgrid(min(x1(:,1)):h:max(x1(:,1)),...
    min(x1(:,2)):h:max(x1(:,2)));
[~,score] = predict(model,[X1(:),X2(:)]);
scoreGrid = reshape(score(:,1),835,916);

figure
plot(x1(:,1),x1(:,2),'k.')
hold on
plot(x1(svInd,1),x1(svInd,2),'ko','MarkerSize',10)
```

```

contour(X1,X2,scoreGrid)
colorbar;
title_str = "SVM Classification C=" + num2str(C) + " \sigma=" + num2str(sigma);
title(title_str)
xlabel('X Axis')
ylabel('Y Axis')
legend('Observation','Support Vector')
a = gca; % get the current axis;
% set the width of the axis (the third value in Position)
% to be 60% of the Figure's width
a.Position(3) = 0.6;
text1 = {"Error","Train = " + num2str(train_err) ...
        ,"Test = " + num2str(test_err)};
annotation('textbox',[0.83 0 0 .5],'String',text1,'FitBoxToText','on')
hold off
snapnow
end

```

SVM Classifier

Function adapted from function on page 247 of the text

```

function [model,test_err,train_err]=SVM_clas(x1,y1,x2,y2,tol,C,sigma)

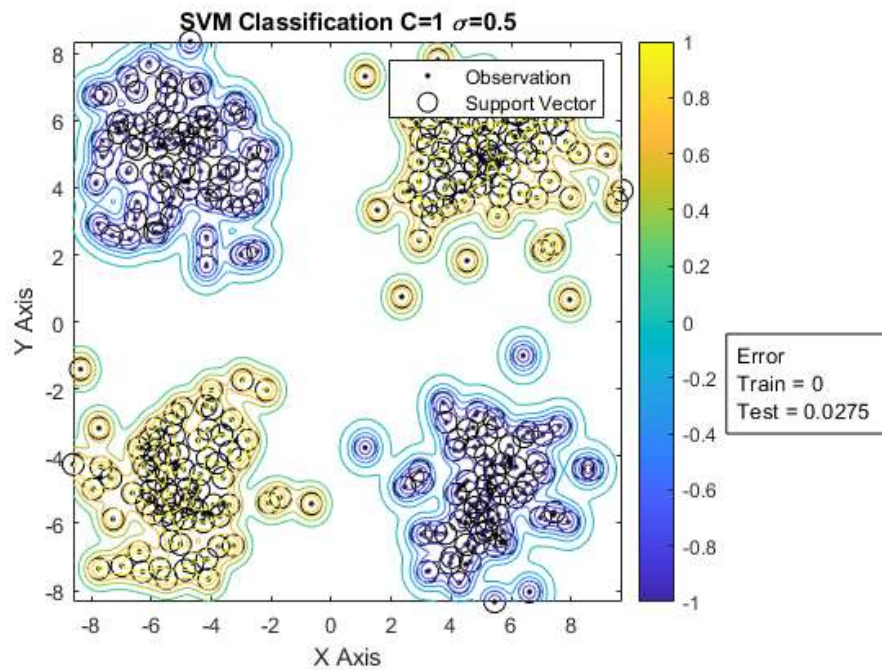
% The following options are from the function in the textbook, it
% required simple adaptation to the new function fitcsvm:
% DeltaGradientTolerance = tol
% Solver = SMO
% Verbose = 1
% IterationLimit = 20000
% CacheSize = 10000
% KernelFunction = RBF
% KernelScale = sigma
% BoxConstraint = C
model = fitcsvm(x1,y1, ...
    'DeltaGradientTolerance',tol,...
    'Solver','SMO',...
    'Verbose',1,...
    'IterationLimit',20000,...
    'CacheSize',10000,...
    'KernelFunction','RBF',...
    'KernelScale',sigma,...
    'BoxConstraint',C);

%Computation of the error probability
test_err = loss(model,x2,y2);
train_err = loss(model,x1,y1);
end

```

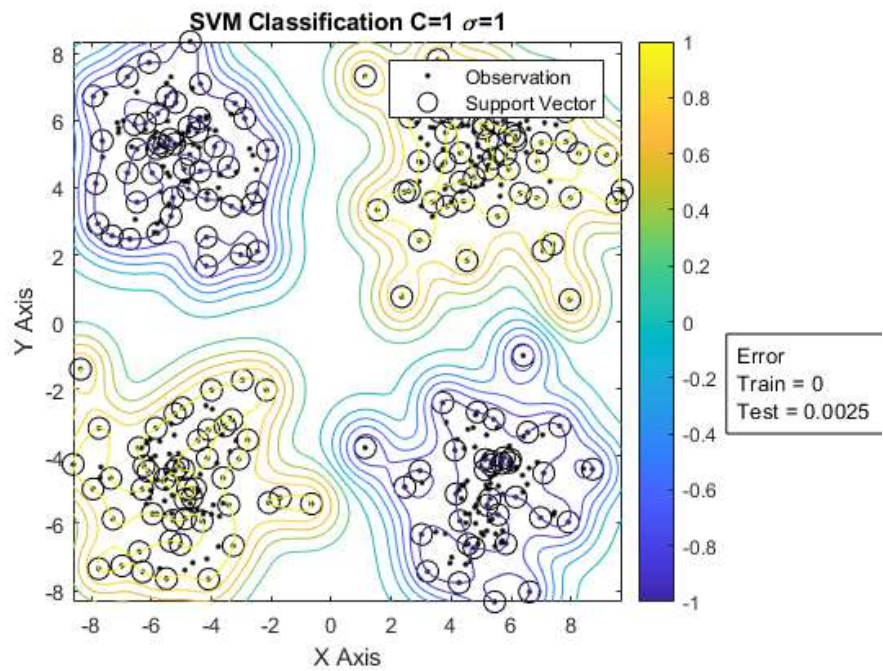
Iteration	Set	Set Size	Feasibility Gap	Delta Gradient	KKT Violation	Number of Supp. Vec.	Objective	Constraint Violation
0	active	400	9.975062e-01	2.000000e+00	1.000000e+00	0	0.000000e+00	0.000000e+00
1000	active	400	8.407885e-04	1.983371e-03	1.018865e-03	297	-7.239337e+01	3.339343e-16
1151	active	400	3.855007e-04	9.980384e-04	5.010877e-04	297	-7.239356e+01	6.071532e-16

Exiting Active Set upon convergence due to DeltaGradient.



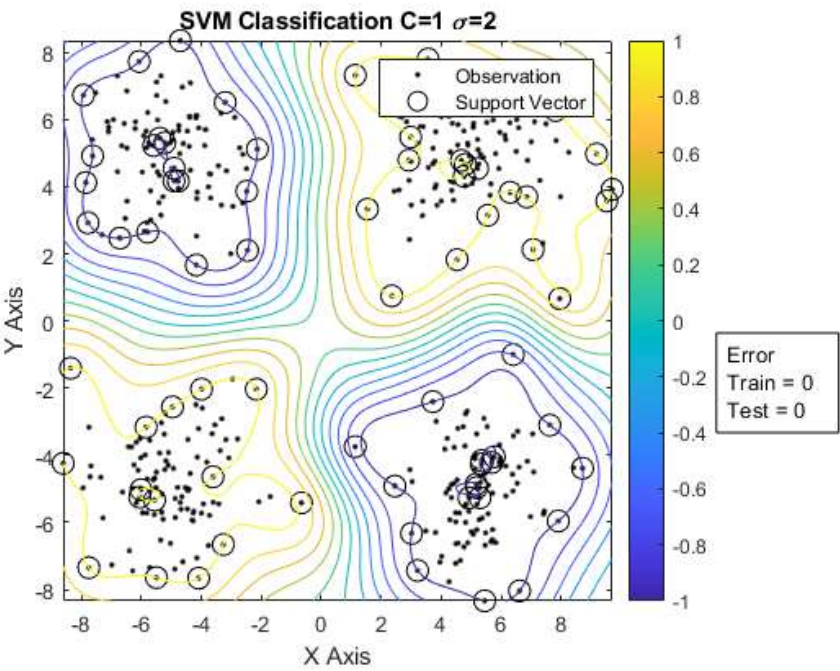
Iteration	Set	Set Size	Feasibility Gap	Delta Gradient	KKT Violation	Number of Supp. Vec.	Objective	Constraint Violation
0	active	400	9.975062e-01	2.000000e+00	1.000000e+00	0	0.000000e+00	0.000000e+00
1000	active	400	1.165708e-03	2.337429e-03	1.254457e-03	188	-3.195291e+01	8.413409e-17
1167	active	400	4.927793e-04	9.834192e-04	5.021250e-04	184	-3.195314e+01	2.645453e-17

Exiting Active Set upon convergence due to DeltaGradient.



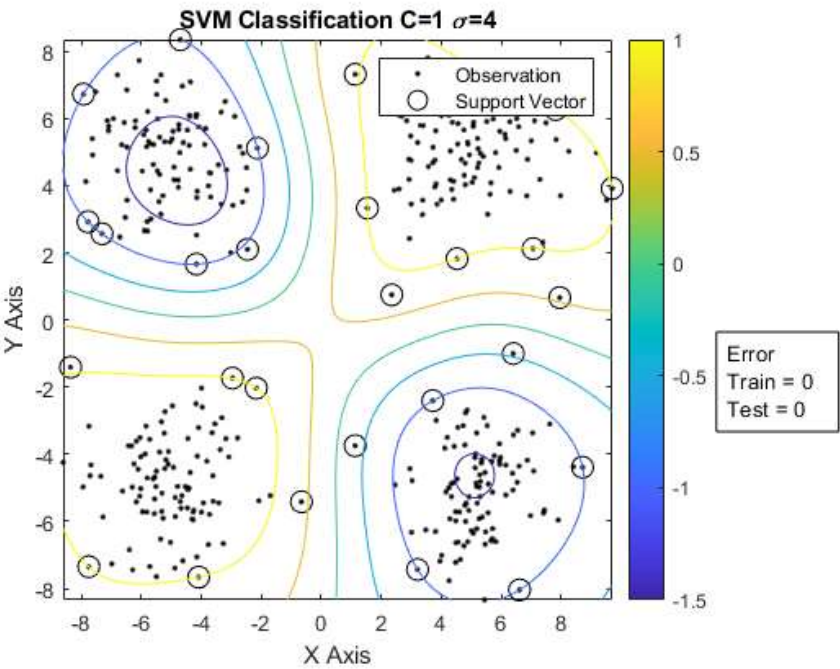
Iteration	Set	Set Size	Feasibility Gap	Delta Gradient	KKT Violation	Number of Supp. Vec.	Objective	Constraint Violation
0	active	400	9.975062e-01	2.000000e+00	1.000000e+00	0	0.000000e+00	0.000000e+00
419	active	400	5.294154e-04	9.618164e-04	4.838165e-04	75	-1.382540e+01	1.953732e-16

Exiting Active Set upon convergence due to DeltaGradient.



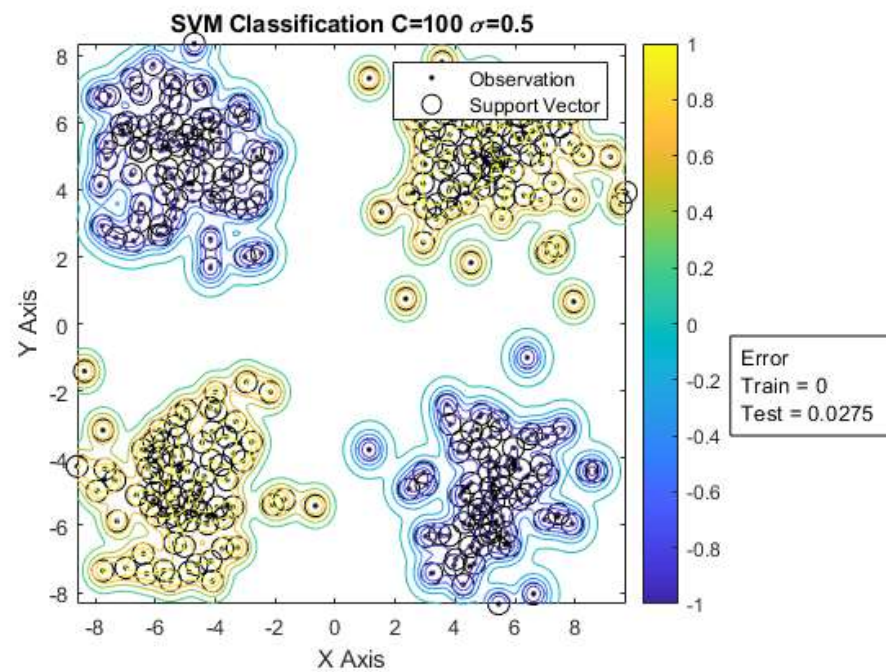
Iteration	Set	Set Size	Feasibility Gap	Delta Gradient	KKT Violation	Number of Supp. Vec.	Objective	Constraint Violation
0	active	400	9.975062e-01	2.000000e+00	1.000000e+00	0	0.000000e+00	0.000000e+00
88	active	400	1.282562e-04	6.374282e-04	3.575460e-04	28	-9.747927e+00	2.775558e-17

Exiting Active Set upon convergence due to DeltaGradient.



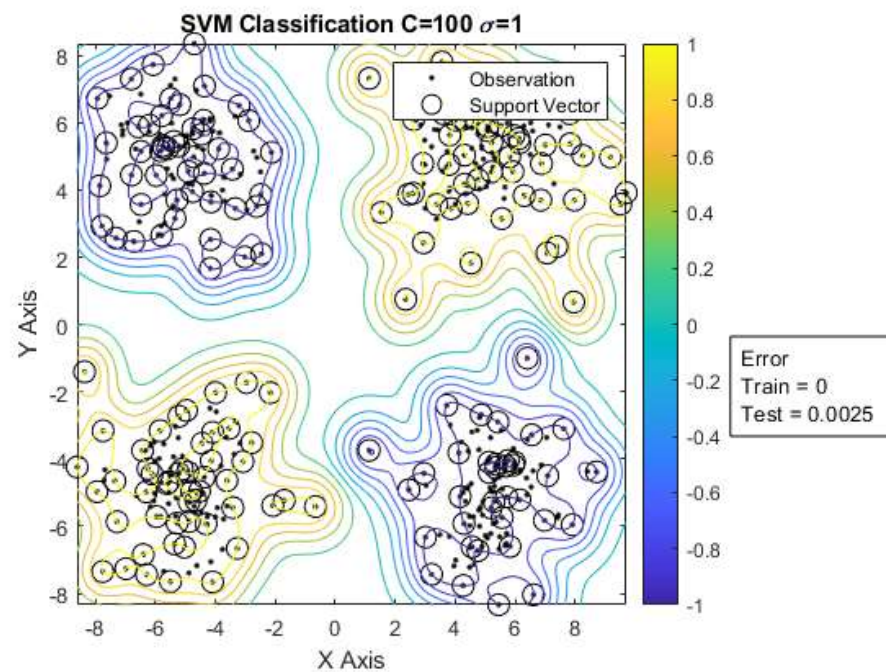
Iteration	Set	Set Size	Feasibility Gap	Delta Gradient	KKT Violation	Number of Supp. Vec.	Objective	Constraint Violation
0	active	400	9.999750e-01	2.000000e+00	1.000000e+00	0	0.000000e+00	0.000000e+00
1000	active	400	7.582241e-02	2.810799e-03	1.665751e-03	296	-7.240008e+01	5.585810e-16
1160	active	400	4.109412e-02	9.976687e-04	5.177254e-04	297	-7.240025e+01	7.754214e-16

Exiting Active Set upon convergence due to DeltaGradient.



Iteration	Set	Set Size	Feasibility Gap	Delta Gradient	KKT Violation	Number of Supp. Vec.	Objective	Constraint Violation
0	active	400	9.999750e-01	2.000000e+00	1.000000e+00	0	0.000000e+00	0.000000e+00
1000	active	400	7.382879e-02	2.080937e-03	1.355332e-03	187	-3.195580e+01	3.068292e-16
1063	active	400	5.763212e-02	9.999235e-04	5.191424e-04	186	-3.195587e+01	1.040834e-16

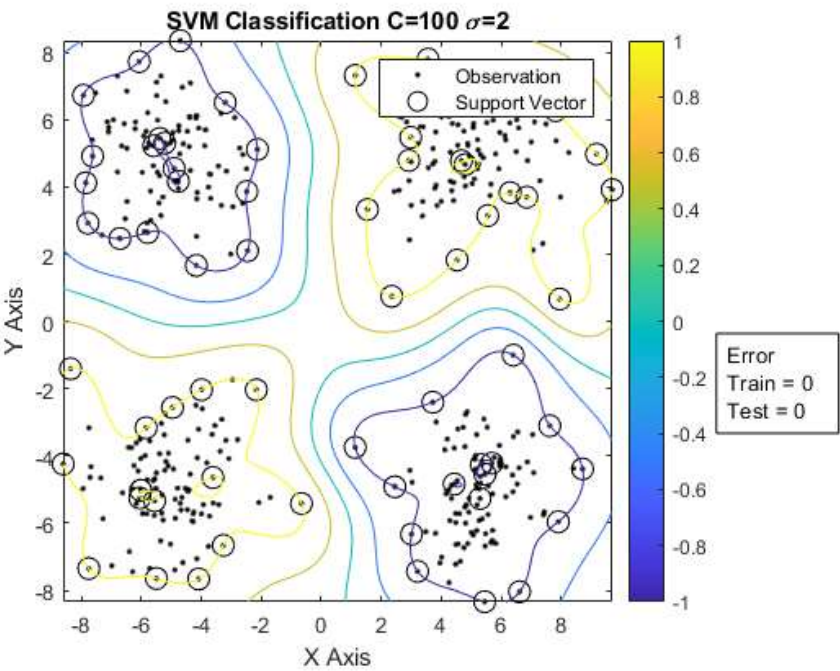
Exiting Active Set upon convergence due to DeltaGradient.



Iteration	Set	Set Size	Feasibility Gap	Delta Gradient	KKT Violation	Number of Supp. Vec.	Objective	Constraint Violation
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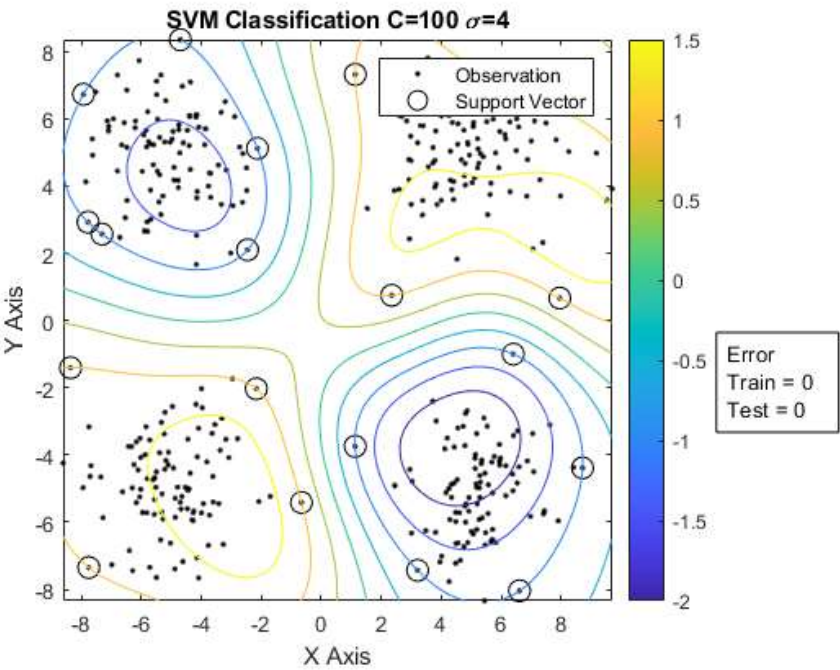
	0	active	400	9.999750e-01	2.000000e+00	1.000000e+00	0	0.000000e+00	0.000000e+00
	386	active	400	3.377326e-02	8.624039e-04	4.503157e-04	68	-1.387275e+01	6.982262e-17

Exiting Active Set upon convergence due to DeltaGradient.



Iteration	Set	Set Size	Feasibility Gap	Delta Gradient	KKT Violation	Number of Supp. Vec.	Objective	Constraint Violation
0	active	400	9.999750e-01	2.000000e+00	1.000000e+00	0	0.000000e+00	0.000000e+00
61	active	400	1.694130e-02	9.455347e-04	5.715393e-04	19	-1.299826e+01	4.996004e-16

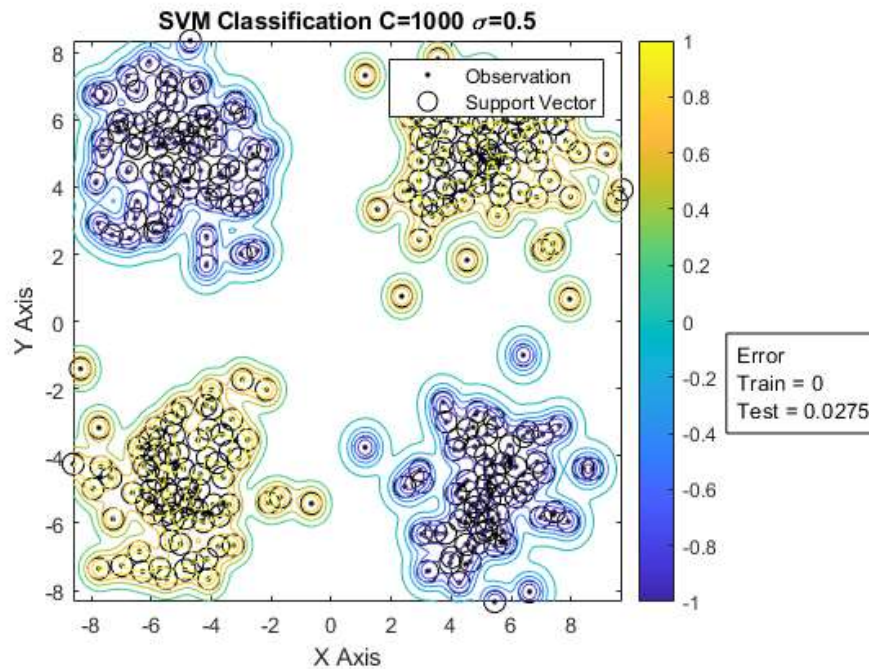
Exiting Active Set upon convergence due to DeltaGradient.



Iteration	Set	Set Size	Feasibility Gap	Delta Gradient	KKT Violation	Number of Supp. Vec.	Objective	Constraint Violation
0	active	400	9.999750e-01	2.000000e+00	1.000000e+00	0	0.000000e+00	0.000000e+00
61	active	400	1.694130e-02	9.455347e-04	5.715393e-04	19	-1.299826e+01	4.996004e-16

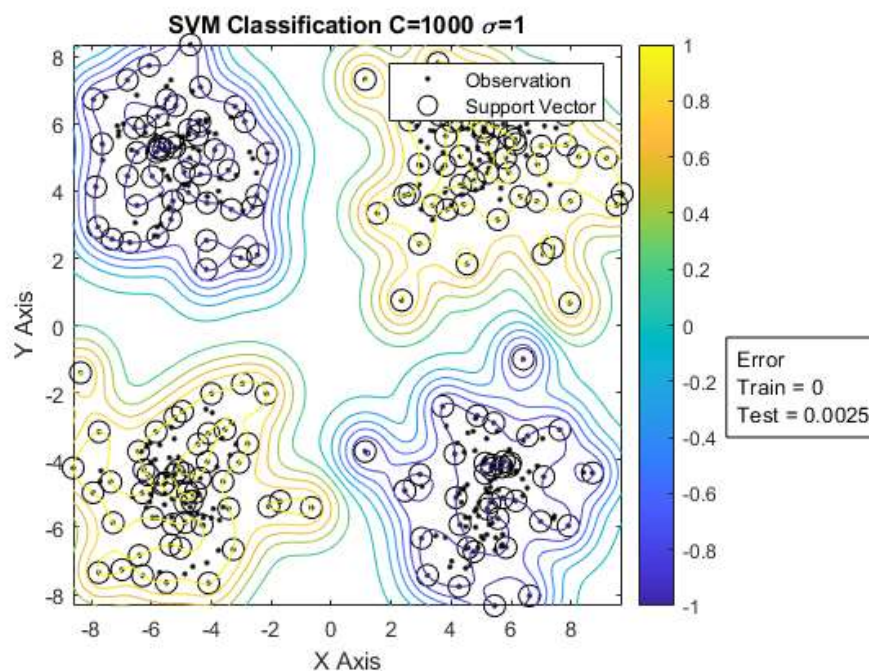
0	active	400	9.999975e-01	2.000000e+00	1.000000e+00	0	0.000000e+00	0.000000e+00
1000	active	400	4.503944e-01	2.810799e-03	1.665751e-03	296	-7.240008e+01	5.585810e-16
1160	active	400	2.999707e-01	9.976687e-04	5.177254e-04	297	-7.240025e+01	7.754214e-16

Exiting Active Set upon convergence due to DeltaGradient.



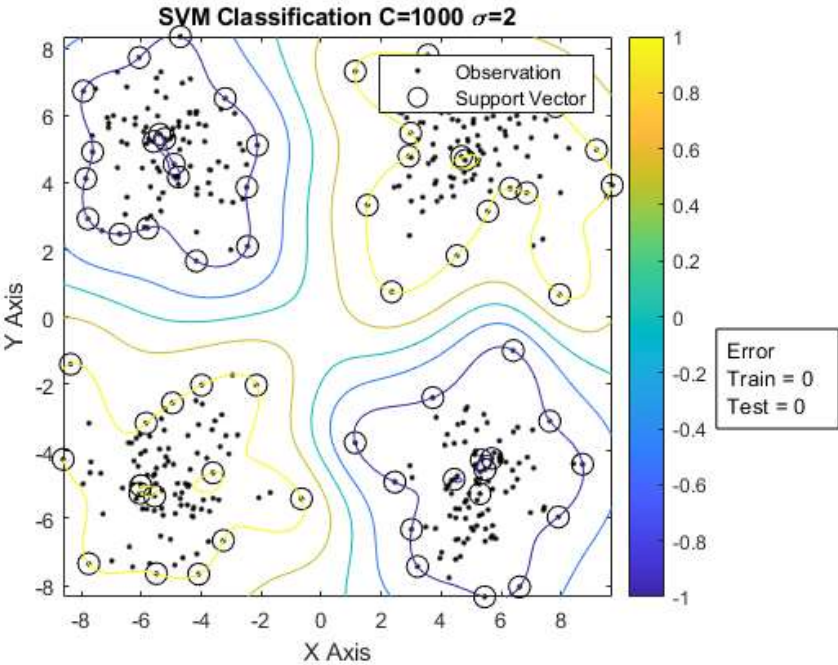
Iteration	Set	Set Size	Feasibility Gap	Delta Gradient	KKT Violation	Number of Supp. Vec.	Objective	Constraint Violation
0	active	400	9.999975e-01	2.000000e+00	1.000000e+00	0	0.000000e+00	0.000000e+00
1000	active	400	4.435076e-01	2.080937e-03	1.355332e-03	187	-3.195580e+01	3.068292e-16
1063	active	400	3.794840e-01	9.999235e-04	5.191424e-04	186	-3.195587e+01	1.040834e-16

Exiting Active Set upon convergence due to DeltaGradient.



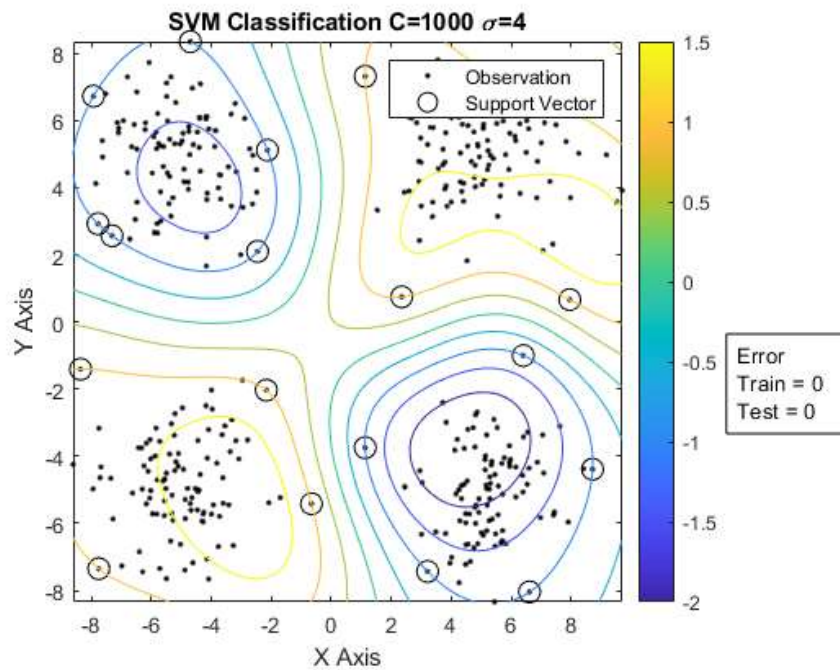
Iteration	Set	Set Size	Feasibility Gap	Delta Gradient	KKT Violation	Number of Supp. Vec.	Objective	Constraint Violation
0	active	400	9.999975e-01	2.000000e+00	1.000000e+00	0	0.000000e+00	0.000000e+00
386	active	400	2.581275e-01	8.624039e-04	4.503157e-04	68	-1.387275e+01	6.982262e-17

Exiting Active Set upon convergence due to DeltaGradient.



Iteration	Set	Set Size	Feasibility Gap	Delta Gradient	KKT Violation	Number of Supp. Vec.	Objective	Constraint Violation
0	active	400	9.999975e-01	2.000000e+00	1.000000e+00	0	0.000000e+00	0.000000e+00
61	active	400	1.475263e-01	9.455347e-04	5.715393e-04	19	-1.299826e+01	4.996004e-16

Exiting Active Set upon convergence due to DeltaGradient.



Decision Tree Classification

```
function decisionTree(x1,y1,x2,y2)
    rng(1);
    tree = fitctree(x1, y1, 'Prune', 'off', 'PruneCriterion', 'impurity');
    tree_pruned = prune(tree);

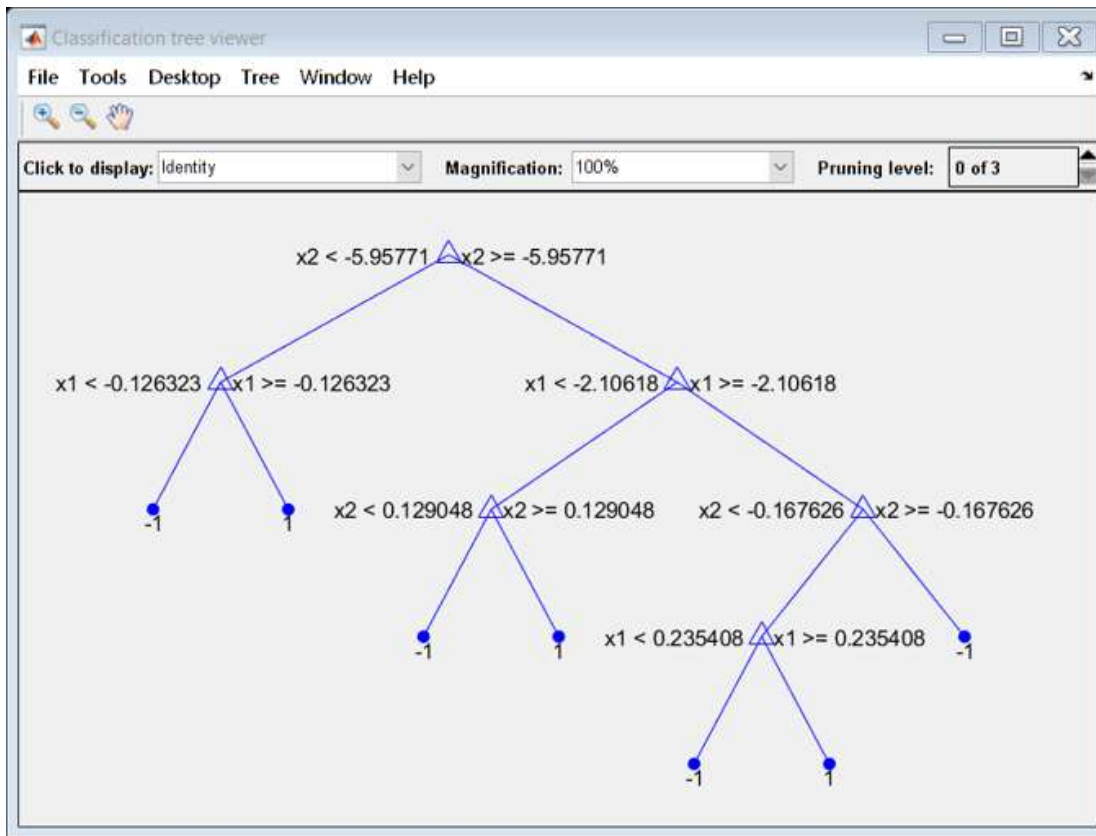
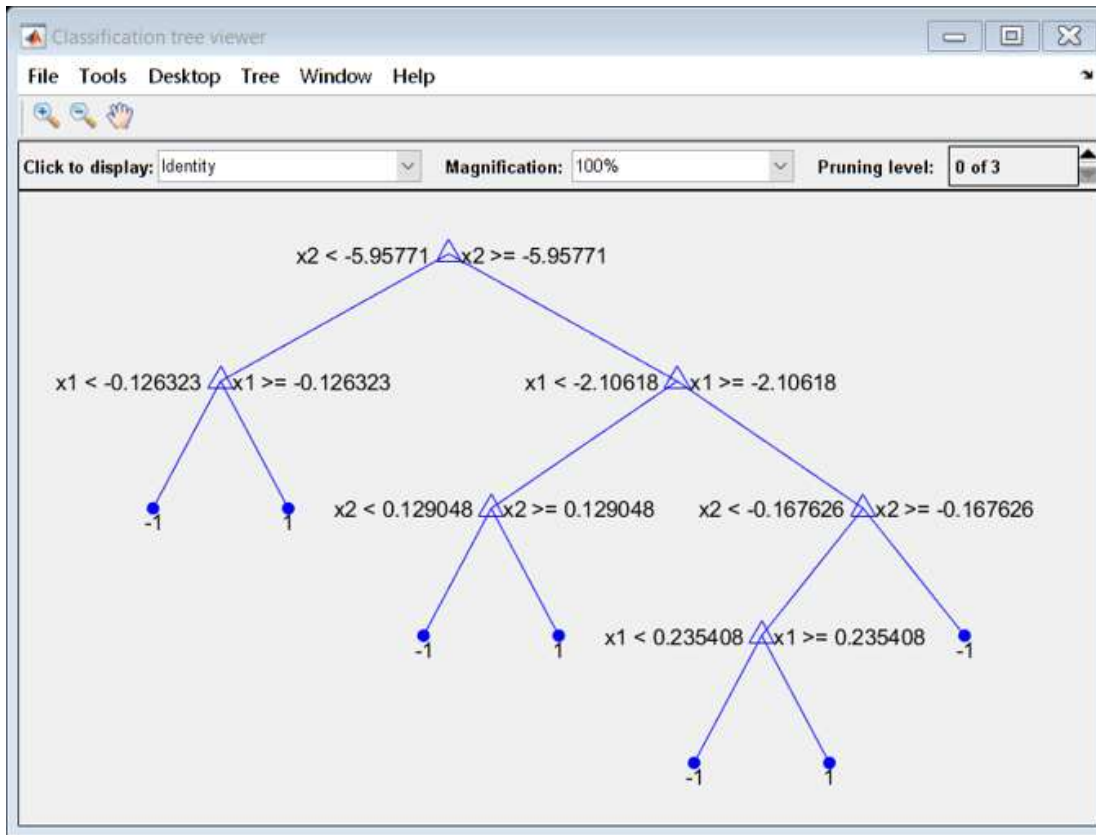
    view(tree, 'Mode', 'graph');
    view(tree_pruned, 'Mode', 'graph');

    test_err = loss(tree, x2, y2);
    train_err = loss(tree, x1, y1);

    test_err_p = loss(tree_pruned, x2, y2);
    train_err_p = loss(tree_pruned, x1, y1);

    fprintf('Testing Error without Pruning: %f\n', test_err);
    fprintf('Training Error without Pruning: %f\n', train_err);
    fprintf('Testing Error with Pruning: %f\n', test_err_p);
    fprintf('Training Error with Pruning: %f\n', train_err_p);
end
```

```
Testing Error without Pruning: 0.005000
Training Error without Pruning: 0.000000
Testing Error with Pruning: 0.005000
Training Error with Pruning: 0.000000
```



Functions Received From Textbook

The following functions were received from the Textbook
 Pattern Recognition - Theodoridis, Koutroumbas

Data Generation Class

Received from page 244 of the text

```
function [x,y]=data_generator(m,s,N)
    S = s*eye(2);
    [~,c] = size(m);
    x = []; % Creating the training set
    for i = 1:c
        x = [x mvnrnd(m(:,i)',S,N)'];
    end
    y=[ones(1,N) ones(1,N) -ones(1,N) -ones(1,N)];
end
```

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