# **SVM** and Perceptron HW:

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```

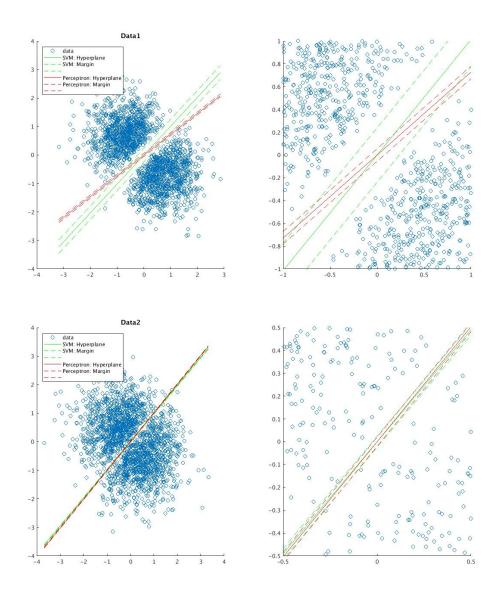
#### q3 - svm\_test and svm\_train:

### q4-5 - Visualization of HyperPlane and Margin:

For both Perceptron and SVM.

subplot(1,2,1)

```
title('Data1'); hold on;
plot_data_svm_percep(normX, svm_theta, perc_theta);
subplot(1,2,2)
plot_data_svm_percep(normX, svm_theta, perc_theta);
legend('off');
xlim([-1 1]); ylim([-1 1]);
load data2.mat
[normX, sDev, means] = data_normalization(X);
[perc_theta, k] = my_perceptron_train(normX, y);
svm theta = svm train(normX, y);
figure('Name', 'Data2')
set(gcf, 'Position', [300, 300, 1300, 650]);
subplot(1,2,1)
title('Data2'); hold on;
plot_data_svm_percep(normX, svm_theta, perc_theta);
subplot(1,2,2)
plot_data_svm_percep(normX, svm_theta, perc_theta);
legend('off');
xlim([-0.5 \ 0.5]); ylim([-0.5 \ 0.5]);
Minimum found that satisfies the constraints.
Optimization completed because the objective function is non-
decreasing in
feasible directions, to within the default value of the optimality
 tolerance,
and constraints are satisfied to within the default value of the
 constraint tolerance.
Minimum found that satisfies the constraints.
Optimization completed because the objective function is non-
decreasing in
feasible directions, to within the default value of the optimality
 tolerance,
and constraints are satisfied to within the default value of the
 constraint tolerance.
```



## q6

```
img2 = imread('test2.jpg');
digit2 = image2blocks(img2);
img3 = imread('test3.jpg');
digit3 = image2blocks(img3);

X = [digit2; digit3];
y = [ones(size(digit2,1),1) ; -1 * ones(size(digit3,1),1)];

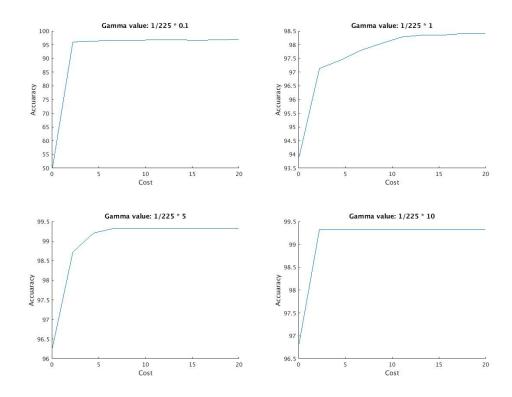
p = cvpartition( y, 'Holdout', 0.20);
X_train = X(p.training, :);
y_train = y(p.training);
X_test = X(p.test, :);
y_test = y(p.test);
```

```
[X_train_n, X_test_n] = normalizeTrainAndTest(X_train, X_test);
svmStruct = svmtrain(y_train, X_train_n, '-q');
[predicted labels, accuracy, dec values] = sympredict(y train,
X_train_n, svmStruct,'-q');
fprintf('\nThe accuracy for the training data is: %.4g%%\n',
accuracy(1))
fprintf('Its not 100%% because data is not linear seprable.\n')
[predicted_labels, accuracy, dec_values] = sympredict(y_test,
X_test_n, svmStruct,'-q');
fprintf('The accuracy for the test data is: %.4g%%\n', accuracy(1))
fprintf('\n**** GAMMA C PLAYGROUND ******\n')
def_c = 1;
def qamma = 1/size(X,2);
gamma coef = [1/10 1 5 10];
c_vals = linspace(def_c/20, def_c*20, 10);
value]
ACC = zeros(size(c vals));
figure('Name', 'SVM Accuracy to cost');
set(gcf, 'Position', [200, 200, 1300, 1300]);
for j = 1:length(gamma_coef)
    for i = 1:length(c vals)
        symoptions = ['-c', num2str(c_vals(i)), '-g',
num2str(gamma\_coef(j) * def\_gamma) , ' -q'];
       svmStruct = svmtrain(y_train, X_train_n, svmoptions);
        [predicted_labels, accuracy, dec_values] = sympredict(y_train,
X train n, svmStruct, '-q');
       ACC(i) = accuracy(1);
       if accuracy(1) > min error(1)
           min_error = [ accuracy(1) gamma_coef(j) * def_gamma
 c_vals(i) ];
       end
    end
    subplot(2,2,j); hold on;
    title(sprintf('Gamma value: 1/%g * %.4g', size(X,2),
gamma_coef(j)))
   plot(c_vals, ACC); xlabel('Cost'), ylabel('Accuaracy')
end
fprintf('Minimal Error Parameters:\n')
fprintf('Gamma = %g | Cost = %g:\n', min_error(2:3))
fprintf('We achived training accuaracy of: %g\n', min_error(1))
fprintf('\nWe can use the Gradiant Decent method to optimize the
parameters\n')
```

```
The accuracy for the training data is: 96.64% Its not 100% because data is not linear seprable. The accuracy for the test data is: 97.07%
```

```
***** GAMMA C PLAYGROUND ******
Minimal Error Parameters:
Gamma = 0.0222222 | Cost = 6.7:
We achived training accuaracy of: 99.3289
```

We can use the Gradiant Decent method to optimize the parameters



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