

## 1 Exercise 1

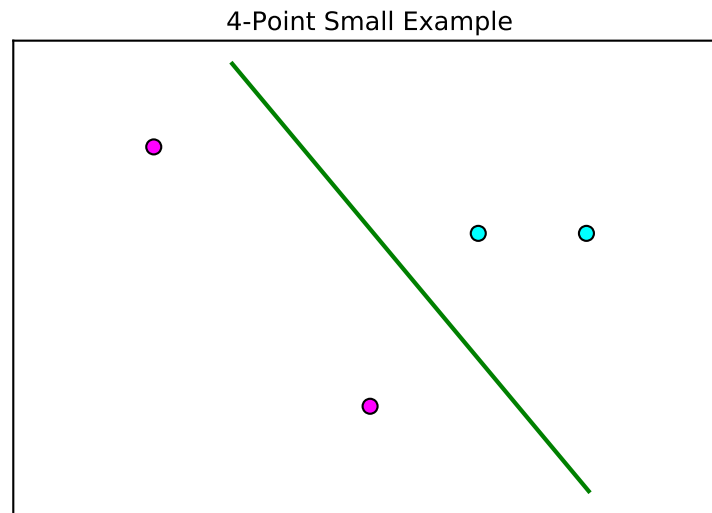


Figure 1: Something

1.2 - since smallOverlap was the same, took half the data for test and half for train

2.2: \* do training set over some values of lambda \* use that to pick lambda \* test on 'validation' data and report an error value

2.3: will need to plot lambda versus sparsity

2.4: \* do training set over some values of bandwidth with lambda = 0 \* do validation set over some values of lambda \* test on test and report results

2.5: \* will need to think about this a little bit more.

3.1: \* <https://piazza.com/class/hzdfawvtilo7hf?cid=434> \* L2 regularization, linear case \* <http://blog.datumbox.com/machine-learning-tutorial-the-multinomial-logistic-regression-softmax-regression/>

\* maybe need to save coefficients or predictions or something?

3.1 - LR: \* 2 features, 100 pts, 2 classes,  $\lambda = 0.01$  - error = .485 \* might need to randomly select subsets of data points

\* tips for getting numerics to work better: \* normalize data beforehand (then don't forget to re-normalize later) \*

3.2 - multiclass SVM: \* used <http://scikit-learn.org/stable/modules/generated/sklearn.svm.LinearSVC.html> \* can do this SO FAST it doesn't even make sense to try to do this by myself. \* tried an array of  $\lambda$  values with L1 loss (multiclasssvm.py), best  $\lambda$  with random partitioning of data into 3 sets. rigorous stopping criteria \* hinge loss, l2 regularization: \* validation error: .187 \* test error: .261 \*  $\lambda$ : 0.01 \* squared loss, l1

regularization: \* validation error: .143 \* test error: .143 \* l: 7e-7