

**Homework # 11: Support Vector Machines (Chap. 9)**

*Due April 24 by noon on Blackboard. Quiz #9 is on April 25.*

1. (**Chap. 9, # 1a, p. 368**) This problem involves a hyperplane in two dimensions. Sketch the hyperplane  $1 + 3X_1 - X_2 = 0$ . Indicate the set of points for which  $1 + 3X_1 - X_2 > 0$ , as well as the set of points for which  $1 + 3X_1 - X_2 < 0$ .
2. (**Chap. 9, # 2, p. 368**) We have seen that in  $p = 2$  dimensions, a linear decision boundary takes the form  $\beta_0 + \beta_1 X_1 + \beta_2 X_2 = 0$ . We now investigate a **non-linear** decision boundary.
  - (a) Sketch the curve  $(1 + X_1)^2 + (2 - X_2)^2 = 4$ .
  - (b) On your sketch, indicate the set of points for which  $(1 + X_1)^2 + (2 - X_2)^2 > 4$ , as well as the set of points for which  $(1 + X_1)^2 + (2 - X_2)^2 \leq 4$ .
  - (c) Suppose that a classifier assigns an observation to the blue class if  $(1 + X_1)^2 + (2 - X_2)^2 > 4$ , and to the red class otherwise. To what class is the observation  $(0, 0)$  classified?  $(-1, 1)$ ?  $(2, 2)$ ?  $(3, 8)$ ?
  - (d) Argue that while the decision boundary in (c) is not linear in terms of  $X_1$  and  $X_2$ , it is linear in terms of  $X_1$ ,  $X_1^2$ ,  $X_2$ , and  $X_2^2$ .
3. (**Chap. 9, # 3, p. 368**) Here we explore the maximal margin classifier on a toy data set.
  - (a) We are given  $n = 7$  observations in  $p = 2$  dimensions. For each observation, there is an associated class label.

Obs.	$X_1$	$X_2$	Y
1	3	4	Red
2	2	2	Red
3	4	4	Red
4	1	4	Red
5	2	1	Blue
6	4	3	Blue
7	4	1	Blue

Sketch the observations.

- (b) Sketch the optimal separating hyperplane, and provide the equation for this hyperplane such as in exercise #1.
- (c) Describe the classification rule for the maximal margin classifier. It should be something along the lines of “Classify to Red if  $\beta_0 + \beta_1 X_1 + \beta_2 X_2 > 0$ , and classify to Blue otherwise”. Provide the values for  $\beta_0$ ,  $\beta_1$ , and  $\beta_2$ .
- (d) On your sketch, indicate the margin for the maximal margin hyperplane.
- (e) Indicate the support vectors for the maximal margin classifier.
- (f) Argue that a slight movement of the seventh observation would not affect the maximal margin hyperplane.
- (g) Sketch a hyperplane that is not the optimal separating hyperplane, and provide the equation for this hyperplane.
- (h) Draw an additional observation on the plot so that the two classes are no longer separable by a hyperplane.

4. (**Chap. 9, # 7, p. 371**) In this problem, you will use support vector approaches in order to predict whether a given car gets high or low gas mileage based on the Auto data set (from ISLR).
- (a) Create a binary variable that takes on a 1 for cars with gas mileage above the median, and a 0 for cars with gas mileage below the median.
  - (b) Fit a support vector classifier to the data with various values of cost, in order to predict whether a car gets high or low gas mileage. Report the cross-validation errors associated with different values of this parameter. Comment on your results.
  - (c) Now repeat (b), this time using SVMs with radial and polynomial basis kernels. Comment on your results.
  - (d) Make some plots to back up your assertions in (b) and (c). When  $p > 2$ , you can use the `plot()` function to create plots displaying pairs of variables at a time. For example, to plot horsepower and year, the syntax is `plot(svm.result, data=Auto, horsepower~year)`.