## Introduction to Machine Learning Homework 6: Support Vector Machines

Prof. Sundeep Rangan

1. Consider the data set for four points with features  $\mathbf{x}_i = (x_{i1}, x_{i2})$  and binary class labels  $y_i = \pm 1$ .

ſ	$x_{i1}$	0	1	1	2
ĺ	$x_{i2}$	0	0.3	0.7	1
	$y_i$	-1	-1	1	1

(a) Find a linear classifier that separates the two classes. Your classifier should be of the form

$$\hat{y} = \begin{cases} 1 & \text{if } b + w_1 x_1 + w_2 x_2 > 0 \\ -1 & \text{if } b + w_1 x_1 + w_2 x_2 < 0 \end{cases}$$

State the intercept b and weights  $w_1$  and  $w_2$  for your classifier. Note there is no unique answer as there are multiple linear classifiers that could separate the classes.

(b) Find the maximum  $\gamma$  such that

$$y_i(b + w_1x_{i1} + w_{i2}x_{i2}) \ge \gamma$$
, for all i,

for the classifier in part (a)?

(c) Compute the margin of the classifier

$$m = \frac{\gamma}{\|\mathbf{w}\|}, \quad \|\mathbf{w}\| = \sqrt{w_1^2 + w_2^2}.$$

- (d) Which samples i are on the margin for your classifier?
- 2. Consider the data set with scalar features  $x_i$  and binary class labels  $y_i = \pm 1$ .

$x_i$	0	1.3	2.1	2.8	4.2	5.7
$y_i$	-1	-1	-1	1	-1	1

Consider a linear classifier for this data of the form,

$$\hat{y} = \begin{cases} 1 & z > 0 \\ -1 & z < 0, \end{cases} \quad z = x - t,$$

where t is a threshold. For each threshold t, let J(t) denote the sum hinge loss,

$$J(t) = \sum_{i} \epsilon_i, \quad \epsilon_i = \max(0, 1 - y_i z_i).$$

- (a) Write a short python program to plot J(t) vs. t for 100 values of t in the interval  $t \in [0, 5]$ .
- (b) Based on the plot, what is one value of t that minimizes J(t).
- (c) For the value of t in part (b), find the corresponding slack variables  $\epsilon_i$ .
- (d) Which samples i violate the margin  $(\epsilon_i > 0)$  and which samples i are misclassified  $(\epsilon_i > 1)$ .
- 3. Consider an image recognition problem, where an image **X** and filter **W** are  $4 \times 4$  matrices:

$$\mathbf{X} = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix}, \quad \mathbf{W} = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}.$$

- (a) Recall that in linear classification, the  $4 \times 4$  image matrices  $\mathbf{X}$  and  $\mathbf{W}$  can be represented as 16-dimensional vectors,  $\mathbf{x} = \text{vec}(\mathbf{X})$  and  $\mathbf{w} = \text{vec}(\mathbf{W})$  by stacking the columns of the matrices vertically. What are  $\mathbf{x}$  and  $\mathbf{w}$  for the matrices above.
- (b) What is the inner product  $z = \mathbf{w}^{\mathsf{T}} \mathbf{x}$ .
- (c) What is the inner product  $z = \mathbf{w}^\mathsf{T} \mathbf{x}_{\text{right}}$  where  $\mathbf{x}_{\text{right}}$  is the vector corresponding to the matrix  $\mathbf{X}$  right shifted by one pixel with the left column filled with zeros.
- (d) What is the inner product  $z = \mathbf{w}^\mathsf{T} \mathbf{x}_{\text{left}}$  where  $\mathbf{x}_{\text{left}}$  is the vector corresponding to the matrix  $\mathbf{X}$  left shifted by one pixel with the right column filled with zeros.
- (e) Write the python command that can covert a  $4 \times 4$  image matrix, Xmat to the 16-dimensional vector, x. What is the python command to go from x to Xmat.
- 4. Consider the data set with scalar features  $x_i$  and binary class labels  $y_i = \pm 1$ .

I	$x_i$	0	1	2	3
	$y_i$	1	-1	1	-1

A support vector classifier is of the form

$$\hat{y} = \begin{cases} 1 & z > 0 \\ -1 & z < 0, \end{cases} \quad z = \sum_{i} \alpha_i y_i K(x_i, x),$$

where K(x, x') is the radial basis function,  $K(x, x') = e^{-\gamma(x-x')^2}$ , and  $\gamma > 0$  and  $\alpha = [\alpha_1, \ldots, \alpha_4]$  are parameters of the classifier.

- (a) Use python to plot z vs. x and  $\hat{y}$  vs. x when  $\gamma = 3$  and  $\alpha = [0, 0, 1, 1]$ .
- (b) Repeat (a) with  $\gamma = 0.3$  and  $\alpha = [1, 1, 1, 1]$ .
- (c) Which classifier makes more errors on the training data.