## CSE 250B: Section 5 - Sharad Vikram

1. Perceptron Practice

Recall in the perceptron algorithm, our classifier is of the form:

$$h_w(x) = \begin{cases} +1 & \text{if } w^T x > 0\\ -1 & \text{if } w^T x \le 0 \end{cases}$$

Let's try running through a few iterations of the perceptron algorithm! Let  $w_0 = [0,0]$  and recall that the update rule is if a training point (x,y) is misclassified, update  $w_{t+1} = w_t + yx$ . Please record the following: (a) whether or not the point is misclassified (b) the new weights, and (c) a plot of the decision boundary. The first two have been filled out for you.

- 1. Training point:  $x_1 = (1, 1), y_1 = +1$ 
  - (a) Point is misclassified.
  - (b)  $w_1 = w_0 + [1, 1] = [1, 1]$
  - (c) See board
- 2. Training point: x = (2, -1), y = -1
  - (a) Point is misclassified.
  - (b)  $w_2 = w_1 [2, -1] = [-1, 2]$
  - (c) See board
- 3. Training point: x = (0, 2), y = +1

4. Training point: x = (-3, -1), y = -1

5. Training point: x = (3, 1), y = 1

#### 2. Matrix calculus

Let A be a  $n \times n$  matrix, and let  $x \in \mathbb{R}^n$ . Show the following identities are true (the gradient will be typically be an  $n \times 1$  vector):

- (a)  $\nabla_x(x^Ty) = y$
- (b)  $\nabla_x(y^Tx) = y$
- (e)  $\nabla_x(x^Tx) = 2x$
- (c)  $\nabla_x(Ax) = A$
- (d)  $\nabla_x(x^T A) = A^T$
- (f)  $\nabla_x(x^TAx) = (A + A^T)x$  (hint: use product rule)

# 3. Linearly Separable Data with Logistic Regression

Show (or explain) that for a linearly separable data set, the maximum likelihood solution for the logistic regression model is obtained by finding a vector  $\beta$  whose decision boundary  $\beta^T x = 0$  separates the classes, and taking the magnitude of  $\beta$  to be infinity.

### 4. Quadratic Kernel

Find a feature mapping  $\Phi$  such that  $\Phi(x)^T \Phi(y) = K(x,y)$  where the kernel function is  $K(x,y) = (x^T y + 1)^2$ . For simplicity, you may assume that the data is 2-dimensional, i.e.  $x = [x_1, x_2]^T$ .

#### 5. Fun with Newton's method for root-finding

- (a) Write down the iterative update equation of Newton's method for finding a root x : f(x) = 0 for a real-valued function f.
- (b) Prove that if f(x) is a quadratic function  $(f(x) = ax^2 + bx + c)$ , then it only takes one iteration of Newton's Method to find the minimum/maximum.