## CMSC5724: Exercise List 3

**Problem 1.** Consider d boolean variables  $x_1, x_2, ..., x_d$ . An expression is a conjunction of d literals where the i-th literal ( $i \in [1, d]$ ) is either  $x_i$  or its negation  $\overline{x_i}$ . For example, when d = 3,  $x_1 \wedge x_2 \wedge \overline{x_3}$  and  $\overline{x_1} \wedge x_2 \wedge \overline{x_3}$  are both expressions. Let  $\mathcal{H}$  be the set of all expressions. Define the instance space  $\mathcal{X} = \{0, 1\}^d$  and label space  $\mathcal{Y} = \{-1, 1\}$ . Note that each  $x \in X$  is a d-dimensional boolean vector  $x = (x_1, ..., x_d)$ . For each expression (i.e., classifier)  $h \in \mathcal{H}$ , h(x) equals 1 if evaluating the expression h on x gives 1, or -1 if the evaluation gives 0. Consider d = 10. Denote by  $\mathcal{D}$  a distribution over  $\mathcal{X} \times \mathcal{Y}$ . Let S be a training set of objects drawn independently from  $\mathcal{D}$ . Prove: when  $|S| \geq 50000$ , with probability at least 0.9, the error of h on  $\mathcal{D}$  is higher than the error of h on S by at most 0.01, for every  $h \in \mathcal{H}$ .

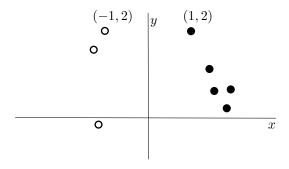
**Problem 2.** Let P be a set of 4 points: A = (1, 2, 1), B = (2, 1, 1), C = (0, 1, 1) and D = (1, 0, 1). A and B have label 1, while C and D have label -1. Execute Perceptron on P. Give the weight vector  $\boldsymbol{w}$  maintained by the algorithm after each iteration.

**Problem 3.** Let P be a set of multidimensional points where each point has a label equal to 1 or -1. We want to design an algorithm to achieve the following purpose:

- Either return a separation plane (see the lecture notes for the definition of separation plane);
- Or declare that P has no separation planes with a margin at least  $\gamma$ .

Your algorithm must still work even if no separation planes exist.

**Problem 4.** Consider the set of points below where points of different colors carry different labels. Only two points have their coordinates shown. Apply Perceptron to find a separation plane on the set. Prove: Perceptron finishes after at most 5 iterations.



**Problem 5.** Some people prefer the following variant of the Perceptron algorithm:

- 1. w = 0
- 2. while there is a violating point p
- 3. **if** p has label 1
- $4. w = w + \lambda \cdot p$

else

5. 
$$\mathbf{w} = \mathbf{w} - \lambda \cdot \mathbf{p}$$

where  $\lambda$  is a positive real-value constant. In the version we discussed in the lecture,  $\lambda = 1$ . Prove: regardless of  $\lambda$ , Perceptron always terminates in  $R^2/\gamma^2$  iterations, where R is the maximum distance of the points to the origin and  $\gamma$  the largest margin of all separation planes.

**Problem 6.** Let P be a set of points in  $\mathbb{R}^d$ , where each point is labeled 1 or -1. A d-dimensional plane  $\pi$  is a separation plane of P if

- $\pi$  does not pass any point in P;
- the points of the two labels in P fall on different sides of  $\pi$ .

Note that we do not require  $\pi$  to pass the origin.

Construct a (d+1)-dimensional point set P' as follows: given each  $p \in P$ , add to P' the point (p[1], p[2], ..., p[d], 1) (i.e., adding a new coordinate 1), carrying the same label as p. Prove: P has a separation plane if and only if P' has a separation plane passing the origin of  $\mathbb{R}^{d+1}$ .