# Homework 2

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3/6/19

### 1 Instructions

This assignment could be written in LaTeX, just as the last homework assignment could have been. Write in understandable, easy to follow English. Make sure you provide good illustrations and figures. Remember to include your Python programs in your assignment.

Your assignment should be submitted in two ways: through GitHub, and in hardcopy (in class). Use the **same** repository you have been using and submit your work in a folder named "lastname-xx", where lastname is your last name xx is the number of the assignment.

# 2 Problem Set

The following is a list of problems you will solve. These are from your textbook. When providing your solutions (hopefully using IATEX), do not simply give the final answer, show how you arrived to the solution, justify your assumptions, and explain your results clearly.

- **Problem 2.1**. *Hint:* it suffices to make  $\epsilon(M, N, \delta) = \sqrt{\frac{1}{2N} \ln \frac{2M}{\delta}} \le \epsilon$  and solve for N.
- **Problem 2.11**. *Hint:* you should use Equation (2.12) and remember that a confidence of 90% means a  $\delta = 0.1$ .
- **Problem 2.12**. *Hint:* use Equation (2.13).

For the following two problems you may use the dataset generator downloadable on iLearn: Resources/Homework/2/makeSemiCircles.py

- Problem 3.1
- (extra credit) Problem 3.2

#### 2.1 Problem 2.1

In Equation (2.1), set  $\delta = 0.03$  and let

$$\epsilon(M, N, \delta) = \sqrt{\frac{1}{2N} \ln \frac{2M}{\delta}}.$$

(a) For M=1, how many examples do we need to make  $\epsilon \leq 0.05$ ? We have that  $\epsilon(M,N,\delta) = \sqrt{\frac{1}{2N} \ln \frac{2M}{\delta}} \leq \epsilon$ . We also have that M=1 and  $\delta=0.03$ . We need to find

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N when  $\epsilon \le 0.05$ . Thus we have  $\sqrt{\frac{1}{2N} \ln \frac{2M}{\delta}} \le \epsilon \le 0.05$  or  $\sqrt{\frac{1}{2N} \ln \frac{2M}{\delta}} \le 0.05$ . Solving for N we have:

$$\sqrt{\frac{1}{2N} \ln \frac{2M}{\delta}} \le 0.05$$

$$\sqrt{\frac{1}{2N} \ln \frac{2(1)}{0.03}} \le 0.05$$

$$\frac{1}{2N} \ln \frac{2}{0.03} \le 0.0025$$

$$\frac{1}{2} \ln \frac{2}{0.03} \le 0.0025N$$

$$\frac{1}{2(0.0025)} \ln \frac{2}{0.03} \le N$$

$$839.9410 < N.$$

Thus, N = 840.

(b) For M=100, how many examples do we need to make  $\epsilon \leq 0.05$ ? We have that  $\epsilon(M,N,\delta) = \sqrt{\frac{1}{2N} \ln \frac{2M}{\delta}} \leq \epsilon$ . We also have that M=100 and  $\delta=0.03$ . We need to find N when  $\epsilon \leq 0.05$ . Thus we have  $\sqrt{\frac{1}{2N} \ln \frac{2M}{\delta}} \leq \epsilon \leq 0.05$  or  $\sqrt{\frac{1}{2N} \ln \frac{2M}{\delta}} \leq 0.05$ . Solving for N we have:

$$\sqrt{\frac{1}{2N} \ln \frac{2M}{\delta}} \le 0.05$$

$$\sqrt{\frac{1}{2N} \ln \frac{2(100)}{0.03}} \le 0.05$$

$$\frac{1}{2N} \ln \frac{200}{0.03} \le 0.0025$$

$$\frac{1}{2} \ln \frac{200}{0.03} \le 0.0025N$$

$$\frac{1}{2(0.0025)} \ln \frac{200}{0.03} \le N$$

$$1760.9751 < N.$$

Thus, N = 1761.

(c) For M=10,000, how many examples do we need to make  $\epsilon \leq 0.05$ ? We have that  $\epsilon(M,N,\delta) = \sqrt{\frac{1}{2N} \ln \frac{2M}{\delta}} \leq \epsilon$ . We also have that M=10,000 and  $\delta=0.03$ . We need to find N when  $\epsilon \leq 0.05$ . Thus we have  $\sqrt{\frac{1}{2N} \ln \frac{2M}{\delta}} \leq \epsilon \leq 0.05$  or  $\sqrt{\frac{1}{2N} \ln \frac{2M}{\delta}} \leq 0.05$ . Solving for N we have:

$$\begin{split} \sqrt{\frac{1}{2N}\ln\frac{2M}{\delta}} &\leq 0.05\\ \sqrt{\frac{1}{2N}\ln\frac{2(10000)}{0.03}} &\leq 0.05\\ \frac{1}{2N}\ln\frac{20000}{0.03} &\leq 0.0025\\ \frac{1}{2}\ln\frac{20000}{0.03} &\leq 0.0025N\\ \frac{1}{2(0.0025)}\ln\frac{20000}{0.03} &\leq N\\ 2682.0091 &\leq N. \end{split}$$

Thus, N = 2683.

#### 2.2 Problem 2.11

Suppose  $m_{\mathcal{H}}(N) = N + 1$ , so  $d_{VC} = 1$ . You have 100 training examples. Use the generalization bound to give a bound for  $E_{\text{out}}$  with confidence 90%. Repeat for N = 10,000.

Equation (2.12) states that  $E_{\text{out}}(g) \leq E_{\text{in}}(g) + \sqrt{\frac{8}{N} \ln \frac{4m_{\mathcal{H}}(2N)}{\delta}}$ . Here we have N = 100,  $m_{\mathcal{H}} = 100 + 1 = 101$ , and  $\delta = 0.1$ . Plugging these values in, we can use the generalization bound to give a bound for  $E_{\text{out}}$  with confidence 90%. We have:

$$E_{\text{out}}(g) \leq E_{\text{in}}(g) + \sqrt{\frac{8}{N} \ln \frac{4m_{\mathcal{H}}(2N)}{\delta}}$$

$$E_{\text{out}}(g) \leq E_{\text{in}}(g) + \sqrt{\frac{8}{100} \ln \frac{4(101)(2(100))}{0.1}}$$

$$E_{\text{out}}(g) \leq E_{\text{in}}(g) + \sqrt{\frac{8}{100} \ln \frac{80800}{0.1}}$$

$$E_{\text{out}}(g) \leq E_{\text{in}}(g) + 1.0432.$$

Thus the bound for  $E_{\text{out}}$  with confidence 90% when N=100 can be represented as  $E_{\text{out}}(g) \leq E_{\text{in}}(g) + 1.0432$ . Now we can repeat with N=10,000. When N=10,000 we have  $m_{\mathcal{H}}=10000+1=10001$  and  $\delta=0.1$ . Plugging these values in we have:

$$E_{\text{out}}(g) \leq E_{\text{in}}(g) + \sqrt{\frac{8}{N} \ln \frac{4m_{\mathcal{H}}(2N)}{\delta}}$$

$$E_{\text{out}}(g) \leq E_{\text{in}}(g) + \sqrt{\frac{8}{10000} \ln \frac{4(10001)(2(10000))}{0.1}}$$

$$E_{\text{out}}(g) \leq E_{\text{in}}(g) + \sqrt{\frac{8}{10000} \ln \frac{800080000}{0.1}}$$

$$E_{\text{out}}(g) \leq E_{\text{in}}(g) + 0.1351.$$

Thus the bound for  $E_{\rm out}$  with confidence 90% when N=10,000 can be represented as  $E_{\rm out}(g) \leq E_{\rm in}(g) + 0.1351$ .

#### 2.3 Problem 2.12

For an  $\mathcal{H}$  with  $d_{VC} = 10$ , what sample size do you need (as prescribed by the generalization bound) to have 95% confidence that your generalization error is at most 0.05?

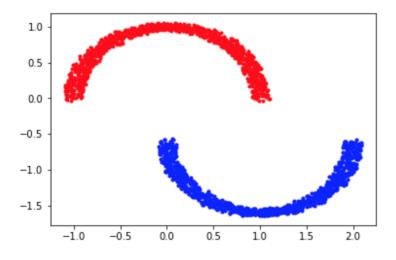
95% confidence that your generalization error is at most 0.05? Equation (2.13) states that  $N \geq \frac{8}{\epsilon^2} \ln \left( \frac{4((2N)^d V^C + 1)}{\delta} \right)$ . Here we have  $d_{VC} = 10$ ,  $\delta = 0.05$ , and  $\epsilon = 0.05$ . Plugging these values in we have  $N \geq \frac{8}{(0.05)^2} \ln \left( \frac{4((2N)^{10} + 1)}{0.05} \right)$ . To calculate N we will use an initial guess of N = 1000. Going through the calculations, we obtain  $N \geq 257251.3639$ . Then we take this new value and plug it back into the equation and obtain  $N \geq 434853.0816$ . We continue to plug in the value obtained into the equation until we see that it is converging to approximately 452956.

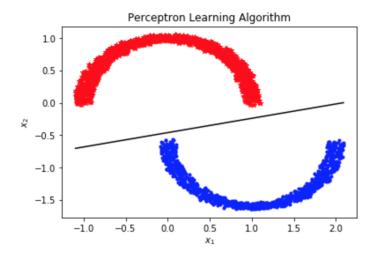
#### 2.4 Problem 3.1

Consider the double semi-circle "toy" learning task below.

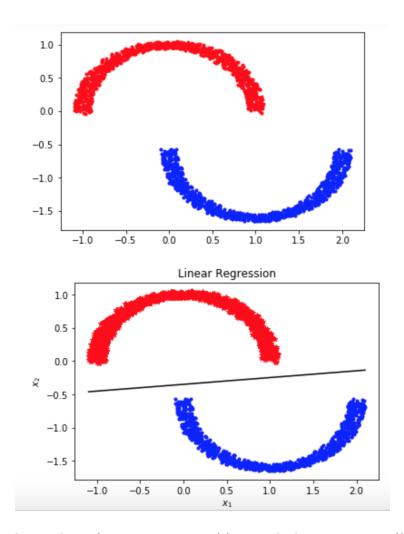
There are two semi-circles of width thk with inner radius, rad, separated by sep as shown (red is -1 and blue is +1). The center of the top semi-circle is aligned with the middle of the edge of the bottom semi-circle. This task is linearly separable when  $sep \geq 0$ , and not so for sep < 0. Set rad = 10, thk = 5 and sep = 5. Then, generate 2,000 examples uniformly, which means you will have approximately 1,000 examples for each class.

(a) Run the PLA starting from  $\mathbf{w} = \mathbf{0}$  until it converges. Plot the data and the final hypothesis.





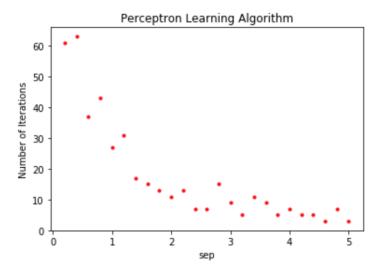
(b) Repeat part (a) using the linear regression (for classification) to obtain w. Explain your observations.



These are the results we obtain from repeating part (a) using the linear regression (for classification) to obtain  $\mathbf{w}$ . As we can see there are a few slight differences between the results but it shows us that we can use linear regression for this classification. These new values of  $\mathbf{w}$  still allow us to obtain an approximate solution for this model.

### 2.5 Problem 3.2 (extra credit)

For the double-semi-circle task in Problem 3.1, vary sep in the range  $\{0.2, 0.4, \ldots, 5\}$ . Generate 2,000 examples and run the PLA starting with  $\mathbf{w} = \mathbf{0}$ . Record the number of iterations PLA takes to converge. Plot sep versus the number of iterations taken for PLA to converge. Explain your observations. [Hint: Problem 1.3]



After plotting sep versus the number of iterations taken for PLA to converge, we can see that it is decreasing as the sep is increasing. In other words, the more separated the semi-circles are, the faster the PLA can converge to a solution.