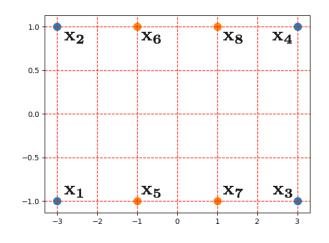
Consider the following 2-class binary problem. We will run Adaboost for two iterations.

Class 0 (y = -1): $\mathbf{x_1} = [-3, -1], \mathbf{x_2} = [-3, 1], \mathbf{x_3} = [3, -1], \mathbf{x_4} = [3, 1]$ Class 1 (y = 1): $\mathbf{x_5} = [-1, -1], \mathbf{x_6} = [-1, 1], \mathbf{x_7} = [1, -1], \mathbf{x_8} = [1, 1]$



(a) Assuming that the first classifier is $h_1(\mathbf{x}) = \begin{cases} -1 & \text{if } \mathbf{x} < -2 \\ +1 & \text{if } \mathbf{x} > -2 \end{cases}$, compute the classification error ϵ_1 and new weights $w_2(n)$ for every sample $n = 1, \ldots, 8$ that result from the first iteration (t = 1) of Adaboost.

 $w_1(n) = \frac{1}{8}, n = 1, \dots, 8$ $\epsilon_1 = w_1(3) + w_1(4) = \frac{2}{8} = \frac{1}{4}$ (samples $\mathbf{x_3}$ and $\mathbf{x_4}$ are incorrectly classified based on h_1) $\beta_1 = \log \frac{1-\epsilon_1}{\epsilon_1} = \log \frac{1-\frac{1}{4}}{\frac{1}{4}} = \log 3$

Using the weight update equation we have: $w_2(n) = \frac{1}{8}$ for n = 1, 2, 5, 6, 7, 8 (since those samples were classified correctly) $w_2(n) = \frac{1}{8}exp(\beta_1) = \frac{3}{8}$ for n = 3, 4 (since those samples were classified incorrectly)

After normalizing the weights (i.e., dividing by their total sum $\sum_{n=1}^{8} w_2(n) = \frac{3}{2}$), we get:

$$w_2(n) = \begin{cases} \frac{\frac{1}{8}}{\frac{3}{2}} = \frac{1}{12} & \text{if } n = 1, 2, 5, 6, 7, 8\\ \frac{3}{\frac{8}{3}} = \frac{1}{4} & \text{if } n = 3, 4 \end{cases}$$

We observe that the weights of the correctly classified samples decreased, while the weights of the incorrectly classified samples have increased.

(b) Based on the error that you computed in (a), what would be a reasonable classification boundary $h_2(\mathbf{x})$ for the second iteration of Adaboost?

A reasonable classifier would make sure that the samples with larger weight (i.e., $\mathbf{x_3}$ and $\mathbf{x_4}$) are correctly classified, therefore a reasonable choice would be: $h_2(\mathbf{x}) = \begin{cases} -1 & \text{if } \mathbf{x} > 2 \\ +1 & \text{if } \mathbf{x} < 2 \end{cases}$

(c) Compute the classification error ϵ_2 and new weights $w_3(n)$ for every sample $n = 1, \ldots, 8$ that result from the second iteration (t = 2) of Adaboost. $\epsilon_2 = w_2(2) + w_2(2) = \frac{2}{12} = \frac{1}{6}$ (samples $\mathbf{x_1}$ and $\mathbf{x_2}$ are incorrectly classified based on h_2)

$$\beta_2 = \log \frac{1-\epsilon_2}{\epsilon_2} = \log \frac{1-\frac{1}{6}}{\frac{1}{6}} = \log 6$$

Using the weight update equation we have: $w_3(n) = \frac{1}{12}$ for n = 5, 6, 7, 8 (since those samples were classified correctly) $w_3(n) = \frac{1}{4}$ for n = 3, 4 (since those samples were classified correctly) $w_3(n) = \frac{1}{12} exp(\beta_2) = \frac{5}{12}$ for n = 1, 2 (since those samples were classified incorrectly) After normalizing the weights (i.e., dividing by their total sum $\sum_{n=1}^{8} w_3(n) = \frac{5}{3}$), we get:

$$w_3(n) = \begin{cases} \frac{1}{12} = \frac{1}{20} & \text{if } n = 5, 6, 7, 8\\ \frac{1}{4} = \frac{3}{20} & \text{if } n = 3, 4\\ \frac{5}{12} = \frac{1}{4} & \text{if } n = 1, 2 \end{cases}$$

(d) What would be the final decision rule based on the first two iterations of Adaboost and the computed variables h_1 , h_2 , β_1 , and β_2 ? Would this result in all samples being correctly classified? What would you do if not?

$$\begin{aligned} h(\mathbf{x}) &= sign \left(\beta_1 h_1(\mathbf{x}) + \beta_2 h_2(\mathbf{x})\right) = sign \left(log_3 \cdot h_1(\mathbf{x}) + log_5 \cdot h_2(\mathbf{x})\right) \\ \mathbf{x_1} &= [-1,3]^T \colon h(\mathbf{x_1}) = -log_3 + log_5 > 0 \rightarrow \text{Class 1 (incorrect)} \\ \mathbf{x_2} &= [-3,1]^T \colon h(\mathbf{x_2}) = -log_3 + log_5 > 0 \rightarrow \text{Class 1 (incorrect)} \\ \mathbf{x_3} &= [3,-1]^T \colon h(\mathbf{x_2}) = log_3 - log_5 < 0 \rightarrow \text{Class -1 (correct)} \\ \mathbf{x_4} &= [3,1]^T \coloneqq h(\mathbf{x_4}) = log_3 - log_5 < 0 \rightarrow \text{Class -1 (correct)} \\ \mathbf{x_5} &= [-1,-1]^T \coloneqq h(\mathbf{x_5}) = -log_3 + log_5 > 0 \rightarrow \text{Class 1 (correct)} \\ \mathbf{x_6} &= [-1,1]^T \coloneqq h(\mathbf{x_6}) = -log_3 + log_5 > 0 \rightarrow \text{Class 1 (correct)} \\ \mathbf{x_7} &= [1,-1]^T \coloneqq h(\mathbf{x_7}) = log_3 - log_5 < 0 \rightarrow \text{Class -1 (incorrect)} \\ \mathbf{x_8} &= [1,-1]^T \coloneqq h(\mathbf{x_8}) = log_3 - log_5 < 0 \rightarrow \text{Class -1 (incorrect)} \\ \end{aligned}$$
We will need to perform more iterations in order to classify correctly all samples.

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