

Problem 3

Let $p(y|x; \theta) = (1 - \delta)(h_\theta(x))^y(1 - h_\theta(x))^{1-y}$ since we have $p(\text{error}) = \delta$

Then the likelihood function is:

$$L(\theta) = \prod_{i=1}^m p(y^{(i)} | x^{(i)}; \theta) = \prod_{i=1}^m (1 - \delta)(h_\theta(x^{(i)}))^{y^{(i)}}(1 - h_\theta(x^{(i)}))^{1-y^{(i)}}$$

Thus the log-likelihood is $l(\theta) = \sum_{i=1}^m \log(1 - \delta) + y^{(i)} \log h_\theta(x^{(i)}) + (1 - y^{(i)}) \log(1 - h_\theta(x^{(i)}))$

Problem 4

The loss function is

$$L = \frac{1}{2} \sum_i (y - \hat{y})^2 = \frac{1}{2} (80(1 - \hat{y})^2 + 20(0 - \hat{y})^2)$$

So the derivative of L is $-80(1 - \hat{y}) + 20\hat{y} = 100\hat{y} - 80$

Let it be 0, then $\hat{y} = 0.8$

Therefore, there is 0.8 probability to output 1.

Problem 5&6

In the implementation part, I followed the pseudocode in R&N textbook. According to professor's word in piazza, I normalize the data by subtracting the column mean, and if column standard deviations is not zero, I divide the data by them. And then to initialize the weight, I used the recommended Xavier Initialization. And then the whole algorithm follows the textbook by first doing forward propagation, and then back propagation.

But my code does not have code accuracy, I am not sure