## Problem 2.1: Expected Zero-One-Loss

The expected zero-one-loss is given by:

$$E(x,y) \sim P[I[f_0(x) \neq y]]$$

$$= P[Y = +1] \cdot P[f_0(x) \neq +1 | Y = +1]$$

$$+ P[Y = -1] \cdot P[f_0(x) \neq -1 | Y = -1]$$

$$= p^+ \cdot 0.5 + (1 - p^+) \cdot 0.5$$

$$= 0.5$$

So, the expected zero-one-loss for  $f_0(x)$  is 0.5.

## Problem 2.2: Probability of Zero Training Loss

The probability of zero training loss with N samples is given by:

$$\begin{split} P(\text{Zero training loss}) &= \left[P(X(1) < 0 | Y = +1)\right]^{N/2} \cdot \left[P(X(1) \ge 0 | Y = +1)\right]^{N/2} \\ & \cdot \left[P(X(1) < 0 | Y = -1)\right]^{N/2} \cdot \left[P(X(1) \ge 0 | Y = -1)\right]^{N/2} \\ &= (0.5^{N/2}) \cdot (0.5^{N/2}) \cdot (0.5^{N/2}) \cdot (0.5^{N/2}) \\ &= 0.25^{N} \end{split}$$

So, the probability of zero training loss with N samples is  $0.25^N$ .

## Problem 2.3: Evolution of Probability with Increasing N

As N increases to infinity, the probability of zero training loss, P(Zero training loss), approaches zero. This is because as N becomes very large, the probability of all the samples satisfying the conditions for zero training loss becomes extremely low, and it becomes increasingly unlikely to obtain zero training loss as N goes to infinity.