

## P9120 - Homework # 3

Assigned: October 27, 2022

Due at 1pm on November 10, 2022

**Maximum points that you can score in this Homework is 20.**

1. Suppose  $X \in \mathbb{R}^p$  and  $Y \in \{-1, 1\}$ . For any real-valued function  $f$  on  $\mathbb{R}^p$ , let  $L(Y, f(X))$  denote the loss function for measuring errors between  $Y$  and  $f(X)$ . Let  $f^* = \arg \min_f EL(Y, f(X))$ , where the expectation is taken over the joint distribution of  $X$  and  $Y$ . Show that
  - (a) (Logistic Regression) If  $L(y, f(\mathbf{x})) = \log[1 + \exp(-yf(\mathbf{x}))]$ , then  $f^*(\mathbf{x}) = \log \frac{Pr(Y=1|X=\mathbf{x})}{Pr(Y=-1|X=\mathbf{x})}$ .
  - (b) (SVM) If  $L(y, f(\mathbf{x})) = [1 - yf(\mathbf{x})]_+$ , then  $f^*(\mathbf{x}) = \text{sign}[Pr(Y = 1|X = \mathbf{x}) - \frac{1}{2}]$ .
  - (c) (Regression) If  $L(y, f(\mathbf{x})) = [y - f(\mathbf{x})]^2$ , then  $f^*(\mathbf{x}) = 2Pr(Y = 1|X = \mathbf{x}) - 1$ .
  - (d) (AdaBoost) If  $L(y, f(\mathbf{x})) = \exp[-yf(\mathbf{x})]$ , then  $f^*(\mathbf{x}) = \frac{1}{2} \log \frac{Pr(Y=1|X=\mathbf{x})}{Pr(Y=-1|X=\mathbf{x})}$ .
2. Ex 10.4, parts (b), (c), (d) of [ESL] (page 385). You can use existing AdaBoost package or write your own program to implement AdaBoost.
3. The “spam” data” ( <https://web.stanford.edu/hastie/ElemStatLearn/data>) has been divided into a training set and a test set. Fit a neural network to the training set, and calculate its classification error on the test set. Compare your results to the classification tree results presented in Section 9.2.5 of [ESL] on both the classification performance and interpretability of the final model.