## One-against-all

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**Theorem**. Let  $\hat{\eta}_1, \dots, \hat{\eta}_K \colon \mathcal{X} \to [0, 1]$  be estimates of conditional probability functions  $x \mapsto \mathbb{P}(Y = k \mid X = x)$ for  $k = 1, \ldots, K$ , and let

$$\epsilon := \mathbb{E}\left[\max_{k=1,\dots,K} \left| \hat{\eta}_k(X) - \mathbb{P}(Y = k \mid X) \right| \right].$$

Let  $\hat{f}: \mathcal{X} \to \{1, \dots, K\}$  be the one-against-all classifier based on  $\hat{\eta}_1, \dots, \hat{\eta}_K$ , i.e.,

$$\hat{f}(x) = \underset{k=1,\dots,K}{\operatorname{arg\,max}} \hat{\eta}_k(x), \quad x \in \mathcal{X},$$

(with ties broken arbitrarily), and let  $f^* : \mathcal{X} \to \{1, \dots, K\}$  be the Bayes optimal classifier. Then

## Assignment Project Exam Help Proof. Fix $x \in \mathcal{X}$ , $y^* := f^*(x)$ , and $\hat{y} := \hat{f}(x)$ . Let $\eta_k(x) := \mathbb{P}(Y = k \mid X = x)$ for all $k = 1, \dots, K$ . Then

## $\mathbb{P}^{(\hat{f}(X) \neq Y \mid X =} \textbf{https://eduassistpro.github.io/}$

$$\begin{array}{l} \textbf{Add WeChatedu}\_\underset{k=1,\ldots,K}{\overset{\hat{\eta}_{\hat{y}^{\star}}()}{\text{--}}} (\textbf{assist}\_\underset{k}{\overset{\hat{\eta}_{\hat{y}^{\star}}(x) - \eta_{\hat{y}}(x)}{\text{--}}} (\textbf{assist}\_\underset{k}{\overset{\hat{\eta}_{\hat{y}^{\star}}(x) - \eta_{\hat{y}}(x)}{\text{--}}} (\textbf{assist})) \\ & \leq 2 \max_{k=1,\ldots,K} \sum_{k=1,\ldots,K} (\textbf{assist}) (\textbf{as$$

Therefore, taking expectations with respect to X,

$$\mathbb{P}(\hat{f}(X) \neq Y) - \mathbb{P}(f^{\star}(X) \neq Y) \leq 2 \cdot \mathbb{E}\left[\max_{k=1,\dots,K} \left| \hat{\eta}_k(X) - \eta_k(X) \right| \right].$$

The bound on the excess risk is tight. To see this, suppose for a given  $x \in \mathcal{X}$  (with  $y^* = f^*(x)$  and  $\hat{y} = \hat{f}(x)$ ), we have  $\hat{\eta}_{y^*}(x) = \hat{\eta}_{\hat{y}}(x) - \delta$ , but  $\eta_{y^*}(x) = \hat{\eta}_{y^*}(x) + \epsilon$  and  $\eta_{\hat{y}}(x) = \hat{\eta}_{\hat{y}}(x) - \epsilon$ . Then

$$\eta_{y^{\star}}(x) - \eta_{\hat{y}}(x) = (\hat{\eta}_{y^{\star}(x)} + \epsilon) - (\hat{\eta}_{\hat{y}}(x) - \epsilon)$$
$$= 2\epsilon - \delta$$

which tends to  $2\epsilon$  as  $\delta \to 0$ .