

# Proofs and correction exercises on *The Elements of Statistical Learning*

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## Chapter 2 : Overview of Supervised Learning

### Exercise 1

In our context, each of our output variables  $(Y_1, \dots, Y_n)$  is categorical and belongs to some set of cardinal  $K$ . Let denote them as follows :

For  $i = 1, \dots, n$ ,  $Y_i \in \{0, 1\}^K$  and  $\sum_j Y_i(j) = 1$ . Given some new input  $x \in \mathbb{R}^p$ , we want to predict the group it belongs to. Therefore, we estimate the vector of probabilities  $\hat{p} := (\hat{p}_1, \dots, \hat{p}_K)$ , where  $\hat{p}_k$  denotes the (estimated) probability that  $x$  belongs to group  $k$ , with  $k = 1, \dots, K$ . The predicted class is the integer  $\bar{k}$  that maximizes the  $\hat{p}_k$ -s. Furthermore, for all integer  $k = 1, \dots, K$ , we denote by  $t_k$  a vector of all zeros except a 1 at  $k$ -th position.

Show that  $\bar{k} = \arg \max_k \hat{p}_k = \arg \min_k \|t_k - \hat{p}\|$ .

### Exercise 1 - Solution

$$\|t_k - \hat{p}\|^2 = \sum_{j=1}^K (t_k(j) - \hat{p}_j)^2 = (t_k(k) - \hat{p}_k)^2 = (1 - \hat{p}_k)^2$$

is a 2-d degree polynomial defined on  $[0, 1]$  and decreasing on that set. Therefore,  $\|t_k - \hat{p}\|$  is minimal when  $\hat{p}_k$  is maximal.

### Exercise 3 - Solution

For all generated points  $x_1, \dots, x_N$ , define  $r_i := \|x_i\|$  their distance to the origin. By definition, the median distance  $m$  from the closest point to the origin must satisfy  $\mathbb{P}(\min_i r_i \geq m) = \frac{1}{2}$ . By independence of the generation of points,

$$\begin{aligned} \mathbb{P}(\min_i r_i \geq m) &= \prod_{i=1}^N \mathbb{P}(r_i \geq m) = \prod_{i=1}^N (1 - \mathbb{P}(r_i \leq m)) = \prod_{i=1}^N \left(1 - \frac{\lambda_p(\mathbf{B}(0, m))}{\lambda_p(\mathbf{B}(0, 1))}\right) \\ &= \left(1 - \frac{\pi^{p/2} m^p}{\Gamma(p/2 + 1)} \frac{\Gamma(p/2 + 1)}{\pi^{p/2} 1^p}\right)^N = (1 - m^p)^N \end{aligned}$$

By solving  $(1 - m^p)^N = 1/2$  we find the solution  $m = (1 - 2^{-1/N})^{1/p}$ .