## Parametric estimation using nearest neighbours

Parametric family  $\mathcal{F} = \{f_{\theta} \mid \theta \in \Theta\}$ . We are interested in finding  $\theta$  given a sample  $\mathcal{X} = \{X_1, \dots, X_n\}$  distributed according to  $f_{\theta}$ .

Given a proposed parameter  $\theta'$ , the Hellinger distance between  $f_{\theta'}$  and  $f_{\theta}$  ( $\theta$  fixed) is

$$\mathcal{H}^{2}(\theta') = \frac{1}{2} \int \left( \sqrt{f_{\theta'}}(x) - \sqrt{f_{\theta}}(x) \right)^{2} dx. \tag{1}$$

Expanding the square, we obtain

$$\mathcal{H}^{2}(\theta') = 1 - \int \sqrt{f_{\theta'}}(x)\sqrt{f_{\theta}}(x) dx \doteq 1 - \mathcal{B}(\theta'), \tag{2}$$

owing to  $\int f_{\theta'} = \int f_{\theta} = 1$ . The integral  $\mathcal{B}(\theta')$  is the 'Bhattacharyya affinity coefficient' [2] between the distribution  $f_{\theta'}$  and ground truth  $f_{\theta}$ . Minimising  $\mathcal{H}^2(\theta')$  is obviously equivalent to maximising  $\mathcal{B}(\theta')$  – so we are looking for making  $f_{\theta'}$  as similar as possible to  $f_{\theta}$  in the Bhattacharyya sense.

As  $\mathcal{B}(\theta')$  involves the unknown  $f_{\theta}$ , it needs to be estimated.

Here we propose to use [1, Lemma 1], allowing consistent estimation of quantities of type  $\int \phi \sqrt{f}$ . Specifically, we estimate  $\mathcal{B}(\theta')$  by

$$\widehat{\mathcal{B}}(\theta') \doteq \frac{2}{\sqrt{\pi}} \frac{1}{\sqrt{n}} \sum_{i=1}^{n} \sqrt{f_{\theta'}}(X_i) \sqrt{V_i}, \tag{3}$$

where,  $V_i$  is the volume of the ball centred at  $X_i$  whose radius is the distance from  $X_i$  to its closest neighbour in the sample.

We choose for parameter  $\theta'$  the maximiser:

$$\widehat{\theta} \doteq \arg\max_{\theta' \in \Theta} \widehat{\mathcal{B}}(\theta'). \tag{4}$$

For a regression setting with normally distributed iid errors  $\epsilon_i$ , we have

$$y_i = \alpha \, x_i + \epsilon_i \tag{5}$$

$$\widehat{\mathcal{B}}(\alpha) \doteq \frac{2}{\sqrt{\pi}} \frac{1}{\sqrt{n}} \sum_{i=1}^{n} \sqrt{N(y_i - \alpha x_i, \sigma)} \sqrt{E_i}, \tag{6}$$

## References

- [1] Carlos Aya-Moreno, Gery Geenens, and Spiridon Penev. "Shape-preserving wavelet-based multi-variate density estimation". In: *Journal of Multivariate Analysis* 168 (2018), pp. 30–47.
- [2] Anil Bhattacharyya. "On a measure of divergence between two multinomial populations". In:  $Sankhy\bar{a}$ : the indian journal of statistics (1946), pp. 401–406.