

# QUASI-LINEARLY INVARIANT PREDICTION

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**1. Introduction, the geometry of the problem.** Consider a sequence of random variables  $X_1, \dots, X_n, X_{n+1}, \dots, X_{n+m}$  from which we have observed the first  $n$  and we want to predict the value of some known function of the next  $m$  random variables.

Let us suppose that the distribution of the random sequence is known except for parameters of location and dispersion  $(\lambda, \delta)$  ( $-\infty < \lambda < +\infty, 0 < \delta < +\infty$ ); this is a restricted but important case, which, in some applications, may be very useful. We will also suppose that the known function  $\varphi(X_{n+1}, \dots, X_{n+m})$  of the next  $m$  random variables about which we want to predict is *quasi-linearly invariant*, that is,

$$\varphi(\lambda + \delta X_{n+1}, \dots, \lambda + \delta X_{n+m}) = \lambda + \delta \varphi(X_{n+1}, \dots, X_{n+m}).$$

Examples of such functions are the order statistics, and linear combinations of the order statistics.

Let us suppose that the process has a density and the function  $\varphi$  is continuous. Denote by  $(1/\delta^{n+1})\mathcal{L}((x_1 - \lambda)/\delta, \dots, (x_n - \lambda)/\delta; (z - \lambda)/\delta)$  the likelihood of the random vector  $(X_1, \dots, X_n; Z)$ , where  $Z = \varphi(X_{n+1}, \dots, X_{n+m})$ . Our purpose is to obtain mean-square predictors and prediction regions for  $Z$ , which will be quasi-linearly invariant.

Let  $R^{n+1}$  denote  $(n+1)$ -dimensional real space. As  $Z$  must be quasi-linearly invariant the points  $(x_1, \dots, x_n; z)$  and  $(\lambda + \delta x_1, \dots, \lambda + \delta x_n; \lambda + \delta z)$  are in correspondence by the quasi-linear group of transformations  $x \rightarrow \lambda + \delta x$  acting in  $R^{n+1}$ .

The equivalence relation introduced by this group splits  $R^{n+1}$  into equivalence classes which are the half-planes of a bundle, whose axis is the line  $x_1 = \dots = x_n = z$ . Each of the half-planes can be described by a system of quantities

$$\xi_3 = (x_3 - x_1)/(x_2 - x_1), \dots, \xi_n = (x_n - x_1)/(x_2 - x_1), \quad \zeta = (z - x_1)/(x_2 - x_1)$$

invariant under the quasi-linear group acting on  $R^{n+1}$ .

**2. Best predictors.** We now consider the problem of finding a best quasi-linearly invariant predictor,  $p(x_1, \dots, x_n)$ , where the loss function is given by  $(z - p(x_1, \dots, x_n))^2$ .

If  $p(x_1, \dots, x_n)$  is an invariant predictor, write  $h(\xi_3, \dots, \xi_n) = p(0, 1, \xi_3, \dots, \xi_n)$ . Also, define  $\mu(\xi_3, \dots, \xi_n)$  by

$$\mu(\xi_3, \dots, \xi_n) = \int_{-\infty}^{+\infty} d\zeta \zeta \mathcal{L}^*(\xi_3, \dots, \xi_n; \zeta) / \int_{-\infty}^{+\infty} d\zeta \mathcal{L}^*(\xi_3, \dots, \xi_n; \zeta)$$

where  $\mathcal{L}^*(\xi_3, \dots, \xi_n; \zeta) = \int_0^{+\infty} db b^{n+1} \int_{-\infty}^{+\infty} da \mathcal{L}(a, a+b, a+b\xi_3, \dots, a+b\xi_n;$

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