

# ON THE EFFICIENCY OF EXPERIMENTAL DESIGNS<sup>1</sup>

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**1. Introduction.** Many models used in statistical investigations can be formulated in terms of least-square theory. The models that will be discussed can be stated as follows. Let  $y_1, \dots, y_N$  be  $N$  independently and normally distributed random variables with common variance  $\sigma^2$ . It is assumed that the expected value of  $y_\alpha$  is given by

$$(1.1) \quad E(y_\alpha) = \beta_1 x_{\alpha 1} + \beta_2 x_{\alpha 2} + \dots + \beta_p x_{\alpha p}, \quad \alpha = 1, \dots, N,$$

where quantities  $x_{\alpha i}$  for  $i = 1, \dots, p; \alpha = 1, \dots, N$  are known constants and  $\beta_1, \dots, \beta_p$  are unknown constants. The coefficients  $\beta_1, \dots, \beta_p$  are the population regression coefficients of  $y$  on  $x_1, x_2, \dots, x_p$  respectively. In matrix notation the above model can be expressed as

$$(1.2) \quad E(y) = X\beta$$

$$\beta = \begin{bmatrix} \beta_1 \\ \cdot \\ \cdot \\ \cdot \\ \beta_p \end{bmatrix}, \quad X = \begin{bmatrix} x_{11} & \dots & x_{1p} \\ \cdot & & \cdot \\ \cdot & & \cdot \\ \cdot & & \cdot \\ x_{N1} & \dots & x_{Np} \end{bmatrix}, \quad y = \begin{bmatrix} y_1 \\ \cdot \\ \cdot \\ \cdot \\ y_N \end{bmatrix},$$

The  $p$  column vectors in  $X$  will be denoted by  $x_1, x_2, \dots, x_p$  where  $x'_j = (x_{1j}, x_{2j}, \dots, x_{Nj})$ . In some cases the experimenter has some amount of freedom in the choice of the  $p$  vectors  $x_i$ . The efficiency and sensitivity of the design may be very much affected by the choice of the design matrix  $X$ . The choice of this matrix is equivalent to that of  $p$  vectors in  $N$ -dimensional Euclidean space.

A simple illustration is furnished by the following example. Suppose  $y_\alpha$  are independent random variables with equal variance  $\sigma^2$ , where  $E(y_\alpha) = \beta_1(x_\alpha - \bar{x}) + \beta_2$ . The  $x$ 's are assumed to be fixed constants. Suppose, furthermore, that we have  $N$  pairs of observations  $(x_1, y_1), \dots, (x_N, y_N)$  and we want to estimate  $\beta_1$ . It is known that the variance of the least square estimate of  $\beta_1$  is inversely proportional to  $\sum_\alpha (x_\alpha - \bar{x})^2$ . Hence, if we could choose values  $x_1, \dots, x_N$  in a domain  $T$ , we would choose them such that  $\sum_\alpha (x_\alpha - \bar{x})$  is as large as possible.

In Section 2 we will prove a theorem about quadratic forms which, with the aid of other considerations, will motivate a criterion for the efficiency of a design matrix  $X$ . In Section 3 two theorems will be proved to aid in the applica-

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