

ELFVING'S THEOREM AND OPTIMAL DESIGNS FOR QUADRATIC LOSS¹

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1. Introduction. The purpose of this paper is to give a matrix analog of a geometric result of Elfving in the theory of optimal design of experiments. The connection with quadratic loss is indicated below.

Let $f = (f_1, \dots, f_m)$ denote m linearly independent continuous functions on a compact set X and let $\theta = (\theta_1, \dots, \theta_m)$ denote a vector of parameters. For each $x \in X$ an experiment can be performed. The outcome is a random variable $y(x)$ with mean value $\theta f'(x) = \sum_i \theta_i f_i(x)$ and a variance σ^2 independent of x . (Primes will denote transposes.) The functions f_1, \dots, f_m , called the regression functions, are assumed known while $\theta = (\theta_1, \dots, \theta_m)$ and σ^2 are unknown. An experimental design is a probability measure μ defined on a fixed σ -field of sets of X which include the one point sets. In practice, the experimenter is allowed N uncorrelated observations and the number of observations that he takes at each $x \in X$ is "proportional" to the measure μ . For a given μ let

$$m_{ij} = m_{ij}(\mu) = \int f_i f_j d\mu \text{ and } M(\mu) = \|m_{ij}\|_{i,j=1}^m.$$

The matrix $M(\mu)$ is called the information matrix of the design.

Suppose μ concentrates mass μ_i at the points x_i , $i = 1, \dots, r$ and $N\mu_i = n_i$ are integers. If N uncorrelated observations are made, taking n_i observations at x_i , then the variance of the best linear unbiased estimate of $a\theta' = \sum_i a_i \theta_i$ is given by $\sigma^2 N^{-1} a M^{-1}(\mu) a'$. The inverse $M^{-1}(\mu)$ must be suitably defined if $M(\mu)$ is singular. A design μ is called a -optimal if μ minimizes $V(a, \mu) = a M^{-1}(\mu) a'$. The following geometric result was given by Elfving (1952); see also Karlin and Studden (1966).

THEOREM (Elfving). *Let R denote the smallest convex set in Euclidean m -space which is symmetric with respect to the origin and contains all of the vectors $f(x) = (f_1(x), \dots, f_m(x))$, $x \in X$. A design μ_0 is a -optimal if and only if there exists a scalar valued function $\phi(x)$ satisfying $|\phi(x)| \equiv 1$ such that (i) $\int \phi(x) f(x) d\mu_0(x) = \beta a$ for some β and (ii) βa is a boundary point of R . Moreover βa lies on the boundary of R if and only if $\beta^2 = v^{-1}$ where $v = \min_{\mu} V(a, \mu)$.*

The quantity, analogous to $V(a, \mu)$, that we wish to consider is

$$(1.1) \quad V(A, \mu) = \text{tr } A' M^{-1}(\mu) A = \text{tr } M^{-1}(\mu) A A'$$

where A is an $m \times k$ matrix and tr denotes the trace. We thus wish to minimize the sum of quantities $V(a, \mu)$ where the a 's are given by the columns of A .

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