

# Reading Summary 1–A-optimality

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## Problem discussed

- A-optimality for a linear regression model  $y = X\beta + \epsilon$ .
- Consider the problem of determining optimal block designs in which  $\nu$  treatments are to be applied to experimental units arranged in  $b$  blocks of size  $k$ . And A-optimality for such block design.

## A-optimality for linear regression model

Consider the linear regression model

$$y = \mathbf{X}\beta + \epsilon$$

where  $y$  is an  $m \times 1$  vector of observations,  $\mathbf{X}$  is an  $m \times n$  design matrix,  $\beta$  is an  $n \times 1$  vector of unknown parameters.  $\epsilon$  is a an  $m \times 1$  vector of random variables with mean the  $m \times 1$  zero vector and known covariance matrix  $\Lambda$ . Assume that  $m \geq n$  and denote the eigenvalues of  $\Lambda$  in increasing order:

$$\lambda_1 \leq \lambda_2 \leq \dots \leq \lambda_n \dots \leq \lambda_m$$

.

For a design matrix  $\mathbf{X}$  if rank  $n$ , based on observation  $y$ , an unbiased simple least squares estimation of parameter  $\beta$  is

$$(\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'y$$

and covariance matrix is given by

$$(\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'\Lambda\mathbf{X}(\mathbf{X}'\mathbf{X})^{-1}$$

To choose  $\mathbf{X}$  from a given experimental region, A-optimality try to find the minimum of the trace of covariance matrix.

The experimental region under consideration is taken to be the set  $H$  of all  $m \times n$  real matrices of rank  $n$  whose  $i$ th column has a Euclidean norm not exceeding given positive numbers  $c_i, i = 1, \dots, n$ .

## a basic inequality

For matrix  $\Lambda$  there exists an orthogonal matrix  $P$  and a diagonal matrix  $\Lambda_m$  such that

$$\Lambda = P' \Lambda_m P$$

and for any  $X$  in  $H$

$$tr\{(X'X)^{-1}X'\Lambda X(X'X)^{-1}\} \geq (\sum_{i=1}^n c_i^2)^{-1}(\sum_{i=1}^n \lambda_i^{1/2})^2$$

and by the definition of the set  $H$

$$tr\{X'X\} \leq \sum_{i=1}^n c_i^2$$

(Chan 1982)

## A-Optimal designs

So the main goal is to obtain the matrix in  $H$  such that the lower bound is attained. the article only gives the theorem of existence:

Suppose that the positive numbers  $c_i, i = 1, \dots, n$  are arranged in ascending order of magnitude and that the smallest eigenvalue  $\lambda_1$  of the covariance matrix  $\Lambda$  is positive. Then there is an  $X$  in  $H$  such that

$$tr\{(X'X)^{-1}X'\Lambda X(X'X)^{-1}\} = (\sum_{i=1}^n c_i^2)^{-1}(\sum_{i=1}^n \lambda_i^{1/2})^2$$

if and only if

$$(\sum_{i=1}^n c_i^2)^{-1} \sum_{i=1}^k c_i^2 \geq (\sum_{i=1}^n \lambda_i^2)^{-1} \sum_{i=1}^k \lambda_i^2, \quad k = 1, \dots, n-1$$

(Chan 1982)

# A-optimality for block design

Here size of blocks  $k = \nu - 1$  or  $k = \nu + 1$

## Some notatinos

Use  $d$  to denote some particular block design that can be used. Let  $N_d = (n_{dij})$  denote the treatment block incidence matrix of  $d$ ; and  $n_{dij}$  is the number of experimental units to which treatment  $i$  is applied in block  $j$ . The  $i$ th row sum of  $N_d$  is denoted by  $r_{di}$  and represents the number of times treatment  $i$  is replicated in  $d$ .  $N_d N_d'$  is called the concurrence matrix of  $d$ . (Jacroux 1989).

Now for observations  $Y_{ijp}$  obtained by applying treatment  $i$  to an experimental unit occurring in block  $j$ . Assumed model is:

$$Y_{ijp} = \alpha_i + \beta_j + E_{ijp}, \quad 0 \leq i \leq \nu, \quad 1 \leq j \leq b, \quad 0 \leq p \leq n_{dij}$$

where  $\alpha_i$  is the effect of the  $i$ th treatment,  $\beta_j$  in the effect of the  $j$ th block, and the  $E_{ijp}$  are independent random variables having expectation 0 and constant variance  $\sigma^2$

The coefficient matrix of the reduced normal equation for obtaining the least squares estimates of the treatment effects in  $d$  can be written as

$$C_d = \text{diag}(r_{d0}, r_{d1}, \dots, r_{d\nu}) - (1/k)N_d N_d'$$

where  $\text{diag}(r_{d0}, r_{d1}, \dots, r_{d\nu})$  denotes a  $(\nu + 1) \times (\nu + 1)$  diagonal matrix. The matrix  $C_d$  is called the  $C$ -matrix of  $d$  and is positive-semidefinite with zero row sums, having rank  $\nu - 1$

Use  $D(\nu, b, k)$  to denote the class of all connected block designs having  $\nu$  treatments arranged in  $b$  blocks of size  $k$ .

For  $d \in D(\nu, b, k)$  let  $0 = z_{d0} < z_{d1} \leq \dots \leq z_{d, \nu-1}$  denote the nonzero eigen values of  $C_d$  (Jacroux 1992)

## References

- Chan, NN. 1982. "A-Optimality for Regression Designs." *Journal of Mathematical Analysis and Applications* 87 (1): 45–50.
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- . 1992. "On the a-Optimality of Block Designs." *Journal of Statistical Planning and Inference* 32 (3): 401–15.