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## ON MINIMUM VARIANCE AMONG CERTAIN LINEAR FUNCTIONS OF ORDER STATISTICS

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1. Summary. Suppose there are n normal populations  $N(\mu_i, 1)$ ,  $i = 1, \dots, n$  and that one random observation from each of these n populations is given. Let  $x_1 \leq x_2 \leq \dots \leq x_n$  be the observations when arranged in order of magnitude and let the corresponding n random variables be denoted by  $X_i$ ,  $i = 1, \dots, n$ .

The following theorem is proved:

THEOREM.

(1) 
$$\operatorname{Var}\left(\sum_{i=1}^{n}c_{i}X_{i}\right), \text{ where}$$

$$\sum_{i=1}^{n}c_{i}=1,$$

is minimum when  $c_i = 1/n$ ,  $i = 1, \dots, n$ .

The above theorem may be applied to provide a direct proof of the result that  $\sum_{i=1}^{n} X_i$  is the best unbiased linear function of order statistics for estimating the sum  $\sum_{i=1}^{n} \mu_i$ .

**2. Proof.** Let  $(\sigma_{ij})$  be the variance-covariance matrix of  $X_i$  and  $X_j$ , i=1,  $\cdots$ , n; j=1,  $\cdots$ , n. The above theorem will follow from the following lemma. Lemma 1.

(2) 
$$\sum_{i=1}^{n} \sigma_{ij} = 1, \qquad j = 1, \cdots, n.$$

PROOF. The joint probability density function (pdf) of  $X_1, \dots, X_n$  can be easily shown (see [2], pp. 12-17) to be given by

(3) 
$$(2\pi)^{-n/2} \sum_{\tau} \exp\left\{-\frac{1}{2} \sum_{i=1}^{n} (x_i - \mu_{t_i})^2\right\} d\xi,$$

$$x_1 \leq x_2 \leq \cdots \leq x_n,$$

where  $\tau = (t_1, \dots, t_n)$  is a permutation of  $(1, 2, \dots, n)$ ,  $\Sigma_{\tau}$  denotes the summation over n! such permutations and  $\xi$  represents the row vector  $(x_1, \dots, x_n)$ .

Let g be any differentiable function such that the integrals involved exist and we have identically in u,

$$Eg(X_{j} + u) = \int_{\substack{x_{1} \leq \dots \leq x_{n} \\ x_{1} \leq \dots \leq x_{n}}} \dots \int_{\substack{x_{1} \leq \dots \leq x_{n} \\ x_{1} \leq \dots \leq x_{n}}} g(x_{j} + u)(2\pi)^{-n/2} \sum_{\tau} \exp\left\{-\frac{1}{2} \sum_{i=1}^{n} (x_{i} - \mu_{t_{i}})^{2}\right\} d\xi$$

$$= \int_{\substack{x_{1} \leq \dots \leq x_{n} \\ x_{1} \leq \dots \leq x_{n}}} g(x_{j})(2\pi)^{-n/2} \sum_{\tau} \exp\left\{-\frac{1}{2} \sum_{i=1}^{n} (x_{i} - u - \mu_{t_{i}})^{2}\right\} d\xi.$$

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Differentiating both sides of (4) with respect to u and setting u = 0, we obtain  $Eq'(X_i)$ 

$$= \int_{x_{1} \leq \cdots \leq x_{n}} g(x_{j})(2\pi)^{-n/2} \sum_{\tau} \left[ \sum_{i=1}^{n} (x_{i} - \mu_{t_{i}}) \exp \left\{ -\frac{1}{2} \sum_{i=1}^{n} (x_{i} - \mu_{t_{i}})^{2} \right\} \right] d\xi$$

$$= \int_{x_{1} \leq \cdots \leq x_{n}} g(x_{j}) \sum_{i=1}^{n} (x_{i} - \mu_{i})(2\pi)^{-n/2} \sum_{\tau} \exp \left\{ -\frac{1}{2} \sum_{i=1}^{n} (x_{i} - \mu_{t_{i}})^{2} \right\} d\xi$$

$$= E \left[ g(x_{j}) \sum_{i=1}^{n} (x_{i} - \mu_{i}) \right].$$

With g(x) = x, equation (5) gives the required lemma

$$1 = E\left[X_j \sum_{\gamma}^n (X_i - \mu_i)\right] = \sum_{i=1}^n \sigma_{ij}.$$

PROOF OF THE THEOREM.

(6) 
$$\operatorname{Var}\left(\sum_{i=1}^{n} c_{i} X_{i}\right) = \sum_{i=1}^{n} \sum_{j=1}^{n} c_{i} c_{j} \sigma_{ij}.$$

Hence, to minimize (6) subject to the condition (1), we get the following equations to be satisfied by  $c_i$ 's,  $i = 1, \dots, n$ ,

(7) 
$$\sum_{i=1}^{n} c_i \sigma_{ij} = \lambda, \qquad j = 1, \dots, n,$$

where  $2 \lambda$  is used as Lagrangian undetermined multiplier.

From (2) and (7) it follows, on summing over the n equations, that  $\lambda = 1/n$ , so that the desired values of  $c_i$ 's,  $i = 1, \dots, n$ , should satisfy

(8) 
$$\sum_{i=1}^{n} c_i \sigma_{ij} = 1/n, \qquad j = 1, \dots, n.$$

Comparing the equations (2) with (8) and noting that the matrix  $(\sigma_{ij})$  is non-singular, it follows that the solution of equation (8) is  $c_i = 1/n$ ,  $i = 1, \dots, n$ .

This proves the theorem.

In the above theorem, when

$$\mu_1 = \mu_2 = \cdots = \mu_n$$

Lemma 1 was derived by Lloyd [1]. Also we get in this special case the known result that Var ( $\sum_{i=1}^{n} c_i U_i$ ), where  $\sum_{i=1}^{n} c_i = 1$ , and  $u_1 \leq u_2 \leq \cdots \leq u_n$  are n ordered values from  $N(\mu, 1)$ , is minimum when  $c_i = 1/n$ ,  $i = 1, \dots, n$ .

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## REFERENCES

- E. H. Lloyd, "Least squares estimation of location and scale parameters using order statistics," Biometrika, Vol. 39 (1952), pp. 88-95.
- [2] K. C. Seal, "On a class of decision procedures for ranking means," Unpublished Ph.D. Thesis (1954), University of North Carolina, Chapel Hill.