

## $\phi$ -CORRECT DECISION FOR SELECTION AND ELIMINATION

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(Received Oct. 8, 1984; revised Feb. 19, 1985)

### Summary

The selection of  $t$  out of  $k$  populations with parameters  $\theta_i$  ( $i=1, \dots, k$ ) is said to result in an  $\phi$ -correct decision provided

$$\phi(\text{minimum selected } \theta) > \text{maximum non-selected } \theta$$

where  $\phi(\theta)$  ( $>\theta$ ) is an increasing function. For the cases of location or scale parameters the minimum probability of  $\phi$ -correct decision over the entire parameter space is shown to be no less than the minimum probability of correct selection over a preference zone determined by  $\phi(\theta)$ . For other types of parameters this result is shown to be true under certain conditions linking the distribution function and the  $\phi$  function.

### 1. Introduction

Consider  $k$  independent random variables  $X_i$  ( $i=1, \dots, k$ ) from populations  $\pi_i$  with continuous distribution functions  $F(\cdot; \theta_i)$  which are stochastically increasing in the parameters  $\theta_i$ . We suppose that the parameter space is given by

$$(1.1) \quad \Omega = \{\theta = (\theta_1, \dots, \theta_k) : \theta_i \in \Theta\}$$

where  $\Theta$  is a subset of the real line.

The ordered  $\theta_i$  are given by

$$(1.2) \quad \theta_{\rho(1)} \leq \theta_{\rho(2)} \leq \dots \leq \theta_{\rho(k-t)} < \theta_{\rho(k-t+1)} \leq \dots \leq \theta_{\rho(k)}$$

where  $\rho(\cdot)$  is an unknown parametric function, and the ordered  $X_i$  are given by

$$(1.3) \quad X_{R(1)} < X_{R(2)} < \dots < X_{R(k)}$$

where  $R(\cdot)$  is a random function.

Key words: Correct selection, correct decision, preference zone.

Bechhofer [2] considers the decision rule :

$R$ : Select populations  $\pi_i$  for all  $i \in G$  where

$$(1.4) \quad G = \{R(k), R(k-1), \dots, R(k-t+1)\}.$$

Notice that  $R$  selects the populations with the  $t$  largest values of  $X_i$  and eliminates those with the  $k-t$  smallest values. Since the distributions of the  $X_i$  are stochastically increasing in the  $\theta_i$ , we expect, with this decision rule, to select populations with large  $\theta_i$  values and to eliminate those with small  $\theta_i$ .

Let the set of indices of populations with the  $t$  largest  $\theta_i$  values be denoted by

$$(1.5) \quad \gamma = \{\rho(k), \rho(k-1), \dots, \rho(k-t+1)\}.$$

Correct selection, CS, is defined as the event

$$(1.6) \quad \text{CS} = \{G = \gamma\}.$$

In order to put a lower bound on the probability of correct selection, P(CS), Barr and Rizvi [1] consider a preference zone,  $\Omega_\phi$ , taken to be a non-empty subset of  $\Omega$  and given by

$$(1.7) \quad \Omega_\phi = \{\theta: \phi(\theta_j) \leq \theta_i \quad \forall i \in \gamma, j \in \gamma^c\}$$

where  $c$  indicates the complementary set and  $\phi(\cdot)$  is an increasing function with

$$\phi(\theta) > \theta \quad \forall \theta \in \Theta.$$

It is usual to take  $\phi(\theta) = \theta + \Delta_L$  when the  $\theta_i$  are location parameters, with  $\Delta_L$  non-negative, and to take  $\phi(\theta) = \theta/\Delta_s$  when the  $\theta_i$  are scale parameters, with  $\Delta_s \in (0, 1)$ .

Feigin and Weissman [5] define  $\phi$ -correct selection,  $\phi$ -CS, as

$$(1.8) \quad \phi\text{-CS} = \{\min_{i \in G} \phi(\theta_i) > \theta_{\rho(k-t+1)}\}$$

and show that

$$(1.9) \quad \inf_{\theta \in \Omega} P\{\phi\text{-CS}\} = \inf_{\theta \in \Omega_\phi} P(\text{CS}).$$

Let us define  $\phi$ -correct decision,  $\phi$ -CD, as

$$(1.10) \quad \phi\text{-CD} = \{\min_{i \in G} \phi(\theta_i) > \max_{j \in G^c} \theta_j\}.$$

This is considered also by Feigin and Weissman who refer to it as F-CS since it is a particular case of correct selection as studied by Fabian [4]. We prefer the term  $\phi$ -CD since there is a joint state-

ment about the eliminated as well as the selected populations.

Feigin and Weissman remark that it is an open question whether or not

$$(1.11) \quad \inf_{\theta \in \Omega} P \{ \phi - \text{CD} \} = \inf_{\theta \in \Omega_\phi} P \{ \text{CS} \}.$$

In Section 2 it will be shown that (1.11) holds under certain conditions on the distribution functions and the  $\phi$  function. This generalizes the work on location and scale parameters in Bofinger [3].

## 2. $\phi$ -correct decision

Let

$$T = \{(i, j) : i \in \gamma, j \in \gamma^c \text{ and } \phi(\theta_j) \leq \theta_i\}$$

$$I = \{i : (i, j) \in T\}$$

$$J = \{j : (i, j) \in T\}$$

and

$$D(\theta) = \{X_j < X_i \quad \forall (i, j) \in T | \theta\}.$$

Before proving the main result we indicate the connection between  $D(\theta)$  and  $\phi$ -CD with the following lemma :

LEMMA 2.1.

$$D(\theta) \Rightarrow \phi - \text{CD}$$

where the symbol " $\Rightarrow$ " is to be interpreted as "implies".

PROOF.

$$\begin{aligned} \overline{\phi - \text{CD}} &\Rightarrow \{i \in G^c \text{ and } j \in G \text{ for some } (i, j) \in T\} \\ &\Rightarrow \{X_j > X_i \text{ for some } (i, j) \in T\} \\ &\Rightarrow \overline{D(\theta)}. \end{aligned}$$

Hence the result follows.

THEOREM 2.1.

$$\inf_{\theta \in \Omega} P \{ \phi - \text{CD} \} = \inf_{\theta \in \Omega_\phi} P \{ \text{CS} \}$$

provided that there exists a  $\theta^* \in \Theta$  : either

$$(2.1) \quad (a) \quad \forall j \in J, \quad F(X_j; \phi(\theta_j)) \leq_{st} F(Y; \phi(\theta^*))$$

or

$$(2.2) \quad (b) \quad \forall i \in I, \quad F(X_i; \phi^{-1}(\theta_i)) \geq_{st} F(Y; \phi^{-1}(\theta^*))$$

where  $Y$  has distribution function  $F(\cdot; \theta^*)$ ,  $\phi^{-1}(\cdot)$  is the function inverse to  $\phi(\cdot)$  and  $\leq_{st}$  or  $\geq_{st}$  indicates stochastic ordering.

PROOF. Since  $F(x; \theta_i)$  is increasing in  $x$

$$(2.3) \quad \begin{aligned} D(\theta) &= \{F(X_j; \theta_i) < F(X_i; \theta_i) \quad \forall (i, j) \in T\} \\ &\Leftarrow \{F(X_j; \phi(\theta_j)) < F(X_i; \theta_i) \quad \forall (i, j) \in T\} \end{aligned}$$

since  $F(x; \theta)$  is decreasing in  $\theta$  and

$$\theta_i \geq \phi(\theta_j) \quad \forall (i, j) \in T.$$

If condition (a) holds we may replace the LHS of the inequality in (2.3) by

$$F(Y_j; \phi(\theta^*))$$

where, for all  $j \in J$ , the  $Y_j$  are i.i.d. with distribution function  $F(\cdot; \theta^*)$ .

Also, since the  $F(X_i; \theta_i)$  are uniformly distributed on  $(0, 1)$  we may replace the RHS of the inequality by

$$F(Z_i; \phi(\theta^*))$$

where, for all  $i \in I$ , the  $Z_i$  are i.i.d. with distribution function  $F(\cdot; \phi(\theta^*))$ .

Hence

$$\begin{aligned} D(\theta) &\Leftarrow \{F(Y_j; \phi(\theta^*)) < F(Z_i; \phi(\theta^*)) \quad \forall (i, j) \in T\} \\ &\Leftarrow \{Y_j < Z_i \quad \forall i \in \gamma, j \in \gamma^c\}. \end{aligned}$$

Now

$$\inf_{\theta \in \Theta} P\{Y_j < Z_i \quad \forall i \in \gamma, j \in \gamma^c\} \geq \inf_{\theta \in \Theta_\phi} P(\text{CS})$$

which, using Lemma (2.1), shows that

$$\inf_{\theta \in \Theta} P(\phi - \text{CD}) \geq \inf_{\theta \in \Theta_\phi} P(\text{CS}).$$

However,

$$\inf_{\theta \in \Theta_\phi} P(\phi - \text{CD}) = \inf_{\theta \in \Theta_\phi} P(\text{CS})$$

which completes the theorem when condition (a) holds.

When condition (b) holds a similar argument is used, where the expression (2.3) is replaced by

$$\{F(X_j; \theta_j) < F(X_i; \phi^{-1}(\theta_i)) \quad \forall (i, j) \in T\}.$$

*Remarks.* Condition (a) is satisfied if there exists a  $\theta^* \in \Theta$  such

that, for  $X$  and  $Y$  random variables with distribution functions  $F(\cdot; \theta)$  and  $F(\cdot; \theta^*)$  respectively,

$$(2.4) \quad (a') \quad F(X; \phi(\theta)) \leq_s F(Y; \phi(\theta^*)) \quad \forall \theta < \phi^{-1}(U) \\ \text{where } U = \max_{\theta} \theta.$$

Similarly, condition (b) is satisfied if there exists a  $\theta^* \in \theta$  such that

$$(b') \quad F(X; \phi^{-1}(\theta)) \geq_s F(Y; \phi^{-1}(\theta^*)) \quad \forall \theta > \phi(L) \\ \text{where } L = \min_{\theta} \theta.$$

In Section 3 these conditions will be considered for location and scale parameters.

### 3. Location and scale parameters

It will be shown that, for location and scale parameters, the conditions of Theorem (2.1) are satisfied for any choice of the function  $\phi(\cdot)$ .

*Location.* Suppose

$$F(x; \theta) = H(x - \theta).$$

Then condition (a) of Theorem 2.1 may be written:

$$\forall j \in J \quad X_j - \phi(\theta_j) \leq_s Y - \phi(\theta^*).$$

Since  $X_j - \theta_j$  and  $Y - \theta^*$  have the same distribution we see that condition (a) becomes

$$(3.1) \quad \forall j \in J \quad \phi(\theta_j) - \theta_j \geq \phi(\theta^*) - \theta^*$$

and a possible value for  $\theta^*$  is that  $\theta_j$  which gives the minimum value of  $\phi(\theta_j) - \theta_j$ . Hence, although we may not be able to specify  $\theta^*$  (since the  $\theta_j$  are unknown), we know that an appropriate  $\theta^*$  exists and therefore the theorem is true for any choice of the function  $\phi(\cdot)$ .

In the particular case where

$$\phi(\theta) = \theta + \Delta_L$$

we see that any value of  $\theta^*$  satisfies (3.1).

*Scale.* Suppose

$$F(x; \theta) = H(x/\theta).$$

Then condition (a) of Theorem 2.1 may be written

$$\forall j \in J \quad X_j / \phi(\theta_j) \leq_{st} Y / \phi(\theta^*) .$$

Since  $X_j / \theta_j$  and  $Y / \theta^*$  have the same distribution the condition becomes

$$(3.2) \quad \forall j \in J \quad \phi(\theta_j) / \theta_j \geq \phi(\theta^*) / \theta^*$$

and a possible value for  $\theta^*$  is that  $\theta_j$  which gives the minimum value of  $\phi(\theta_j) / \theta_j$ . Hence condition (a) is satisfied for any choice of  $\phi(\cdot)$ .

In the particular case where

$$\phi(\theta) = \theta / \Delta,$$

any value of  $\theta^*$  satisfies (3.2).

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