

## A COMPARISON OF THE MOST STRINGENT AND THE MOST STRINGENT SOMEWHERE MOST POWERFUL TEST FOR CERTAIN PROBLEMS WITH RESTRICTED ALTERNATIVE

BY W. SCHAAFSMA

*University of Groningen*

**0. Summary.** The paper studies hypothesis testing problems  $(H, K_1)$  for the mean of a vector variate having a multivariate normal distribution with known covariance matrix, in cases where the alternative  $K_1$  is restricted by a number of linear inequalities (Section 2). We describe a method for obtaining the *most stringent size- $\alpha$  test*  $\varphi^*$  for Problem  $(H, K_1)$  (Section 3). This method is used for constructing  $\varphi^*$  for a number of special problems  $(H, K_1)$  (Sections 5, 6 and 7) where (i)  $K_1'$  is symmetrical and (ii)  $\Psi_0$  is sufficiently small. Thus for these special problems, we can compare  $\varphi^*$  with the *most stringent SMP size- $\alpha$  test*  $\varphi_0$  that can be obtained by applying the general methods of [9] and [10] (described in Section 2). It turns out (for the special problems with  $\alpha = .05$  or  $.01$ ) that  $\varphi^*$  does not provide a worth-while improvement upon  $\varphi_0$ :  $\varphi^*$  has a smaller maximum shortcoming but  $\varphi_0$  is better than  $\varphi^*$  from other over-all points of view. This supports the opinion that generally no serious objections can be made to the use of the MSSMP tests constructed in [10] Part 2 for problems from actual practice. This is a fortunate circumstance, for these MSSMP tests require only *simple calculations*.

By the way we prove a theorem characterizing the MS size- $\alpha$  test for a very general problem  $(H, K_{(m)})$  where the hypothesis  $H$  may be composite while  $K_{(m)}$  consists of a *finite* number ( $m$ ) of parameters  $\theta_i (i = 1, \dots, m)$  (Section 4).

**1. Introduction.** Let  $(H, K)$  be a testing problem with hypothesis  $H$  and alternative  $K$ . The *shortcoming*  $\gamma_{\varphi, D}(\theta)$  in  $\theta \in K$  of test  $\varphi$  with respect to the class  $D$  of tests (test functions) is defined by

$$(1) \quad \gamma_{\varphi, D}(\theta) = \beta_D^*(\theta) - \beta_{\varphi}(\theta)$$

where  $\beta_D^*(\theta) = \sup_{\varphi \in D} \beta_{\varphi}(\theta)$  denotes the envelope power in  $\theta$  with respect to  $D$  and  $\beta_{\varphi}(\theta) = E_{\theta}\{\varphi(X)\}$  is the *power* in  $\theta$  of test (function)  $\varphi$ . A test  $\varphi^*$  is said to be MS( $D$ ): (*most stringent* with respect to the class  $D$ ) for Problem  $(H, K)$  if (i)  $\varphi^* \in D$  and (ii)

$$(2) \quad \sup_{\theta \in K} \gamma_{\varphi^*, D}(\theta) = \inf_{\varphi \in D} \sup_{\theta \in K} \gamma_{\varphi, D}(\theta).$$

In case  $D$  is the class of size- $\alpha$  tests,  $\varphi^*$  is said to be MS size- $\alpha$ .

A test  $\varphi$  is called SMP( $D$ ) (*somewhere most powerful* with respect to  $D$ ) if (i)  $\varphi \in D$  and (ii)  $\gamma_{\varphi, D}(\theta) = 0$  for some  $\theta \in K$ . Let  $C$  denote the class of all SMP( $D$ ) tests for Problem  $(H, K)$ . A test  $\varphi_0$  is said to be MSSMP( $D$ )

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(most stringent somewhere most powerful with respect to  $D$ ) if (i)  $\varphi_0 \in C$  and (ii)

$$(3) \quad \sup_{\theta \in K} \gamma_{\varphi_0, D}(\theta) = \inf_{\varphi \in C} \sup_{\theta \in K} \gamma_{\varphi, D}(\theta).$$

In case  $D$  is the class of size- $\alpha$  tests,  $\varphi_0$  is said to be MSSMP size- $\alpha$ .

In [9] and [10] Chapter 2 we described methods for obtaining MSSMP( $D$ ) tests for large classes of testing problems with a restricted alternative. Applications to problems from actual practice were given in [10] Part 2.

Though MSSMP( $D$ ) tests are admissible (at least if  $D$  is sufficiently large) they are not satisfactory for each problem with a restricted alternative: in [10] Section 2.15 we described a family of testing problems ( $H, K$ ) where  $K$  is defined by one inequality and where a worth-while improvement upon the MSSMP( $D$ ) tests turned out to be possible.

Obviously we are in want of a criterion characterizing the problems for which the MSSMP( $D$ ) tests cannot be improved upon to a worth-while extent (a provisional criterion has been described in [10] Section 2.13). For that purpose the MSSMP( $D$ ) test should be compared with all other (admissible) size- $\alpha$  tests, for a large number of problems.

In this paper we confine our attention to Problems ( $H, K_1$ ) (to be formulated in Section 2) which are of the form  $\{(H, K_1), \sigma^2 = 1\}$  described in [9] Section 3 and [10] Section 2.4. For such problems we try to obtain the MS size- $\alpha$  test  $\varphi^*$  by means of a method (Section 3) which resembles the method of *van Zwet and Oosterhoff* (1967) for obtaining MS size- $\alpha$  tests for combination of tests problems ( $H, K_1$ ) where  $K_1$  is a positive orthant with vertex  $H$  (in our notation  $H$  and  $K$  denote also the corresponding sets of parameters  $\Omega_H$  and  $\Omega_K$ ). Next we formulate a rule of thumb characterizing the problems for which the MS size- $\alpha$  test  $\varphi^*$  does not provide a worth-while improvement upon the MSSMP size- $\alpha$  test  $\varphi_0$ . Assuming that for such problems no other size- $\alpha$  test will improve upon  $\varphi_0$  to a worth-while extent, we may postulate the rule of thumb as the criterion wanted. Thus we obtained the opinion mentioned at the end of the Summary.

REMARK. Regarded as test-functions over the sample space,  $\varphi_0$  and  $\varphi^*$  are obviously not uniquely determined. In this paper we "identify" tests when their power-functions over  $K_1$  are identical.

**2. The formulation of problem ( $H, K_1$ ).** In accordance with [9] Section 3 and [10] Section 2.4, let  $X = (X_1, \dots, X_n)$  have the multivariate normal  $N(\xi, \Sigma)$  distribution with known nonsingular covariance matrix  $\Sigma = A^{-1}$  where  $A$  is the  $n \times n$  matrix  $(a^{ij})$ . The outcomes  $x$  of  $X$  and the admitted vectors  $\xi$  of means  $(\xi_1, \dots, \xi_n)$  are regarded as points in the same  $n$  dimensional space  $R^n$ . The vector  $\xi$  of means is known to lie in the parameter space  $\Omega = H \cup K_1$  defined by the  $r$  inequalities

$$(4) \quad \sum_{i=1}^n b^{ih} \xi_i \geq 0, \quad (h = n - s + 1, \dots, n - s + r)$$

in the subspace  $R^s$  defined by the  $n - s$  equalities,

$$(5) \quad \sum_{i=1}^n b^{ih} \xi_i = 0, \quad (h = 1, \dots, n - s);$$

$2 \leq r \leq s \leq n$ ;  $(b^{ih})$  ( $i = 1, \dots, n; h = 1, \dots, n - s + r$ ) is supposed to be a matrix of rank  $n - s + r$ .

The hypothesis  $H$  is to be tested that  $\xi$  lies in the subspace  $R^{s-r}$  in  $R^s$ , which is obtained when all  $\geq$  signs in (4) are replaced by  $=$  signs. The alternative  $K_1$  corresponds with the subset of  $R^s$  which is defined by (4) where at least one inequality is strong.

In  $R^n$  an inner product is defined by means of the bilinear form

$$(x, y) = \sum_{i=1}^n \sum_{j=1}^n a^{ij} x_i y_j$$

and accordingly (i) orthogonality  $x \perp y$  is defined by  $(x, y) = 0$ , (ii) the norm  $\|x\|$  by  $\|x\|^2 = (x, x)$  and (iii) if  $x$  spans the half-line  $l$  ( $l$  is the set of all points  $\theta x$  with  $\theta > 0$ ) and  $y$  spans the half-line  $m$  then  $\cos \{\Psi(l, m)\} = \|x\|^{-1} \|y\|^{-1} (x, y)$ , where  $\Psi(l, m)$  is the angle between  $l$  and  $m$  ( $0 \leq \Psi(l, m) \leq \pi$ ).

In [9] Section 4 and [10] Section 2.6, we constructed the MSSMP size- $\alpha$  test  $\varphi_0$  for Problem  $(H, K_1)$ . Let  $R^r$  be the linear subspace in  $R^s$  perpendicular to the  $R^{s-r}$  defined by  $H$  and let  $K_1'$  denote the intersection of  $K_1$  and  $R^r$ .  $K_1'$  is a polyhedral angle with edges  $e_1, \dots, e_r$  which are obtained when all inequalities (4) but one are replaced by the corresponding equalities. Next let the angle  $\Psi_0$  and the half-line  $l_0$  in  $K_1'$  be defined by

$$(6) \quad \Psi_0 = \inf_{l \subset K_1'} \sup_{m \subset K_1'} \Psi(l, m) = \sup_{m \subset K_1'} \Psi(l_0, m),$$

and let  $X^I$  be the projection of the sample point  $X$  on the  $R^1$  spanned by  $l_0$ ; further  $0X^I = \|X^I\|$  (or  $-\|X^I\|$ ) in case  $X^I \in$  (or  $\notin$ )  $l_0$ . The MSSMP size- $\alpha$  test  $\varphi_0$  for problem  $(H, K_1)$  rejects if and only if  $0X^I \geq u_\alpha$  where  $u_\alpha = (\Phi^\times)^{-1}(\alpha)$  and  $\Phi^\times(x) = P(U \geq x)$  when  $U$  has the normal  $N(0, 1)$  distribution.

$D$  being the class of size- $\alpha$  tests for Problem  $(H, K_1)$ , the envelope power in a point  $\xi \in K_1'$  is equal to  $\beta_D^*(\xi) = \Phi^\times(u_\alpha - \|\xi\|)$ . The maximum of the shortcoming of  $\varphi_0$  over an arbitrary half-line  $l$  in  $K_1'$  is a strictly increasing function of the angle  $\Psi(l, l_0)$ :

$$(7) \quad \gamma_{\varphi_0, D}(l) = \sup_{\|\xi\| > 0} [\Phi^\times(u_\alpha - \|\xi\|) - \Phi^\times[u_\alpha - \|\xi\| \cos \{\Psi(l, l_0)\}]].$$

Tables of  $\gamma_{\varphi_0, D}(l)$  as a function of  $\Psi = \Psi(l, l_0)$  and as a function of  $\cos \Psi$  will be given in [8]. We shall compare the MSSMP size- $\alpha$  test  $\varphi_0$  with the MS size- $\alpha$  test  $\varphi^*$  by considering graphs of  $\gamma_{\varphi, D}(l)$  for  $\varphi = \varphi_0$  and  $\varphi = \varphi^*$  and  $l$  varying over (subsets of)  $K_1'$  (see the Figures 3, 4, 5, 8, 9 and 10).

REMARK. It may be proved easily that the (power function of the) MSSMP size- $\alpha$  test  $\varphi_0$  for Problem  $(H, K_1)$  is uniquely determined.

**3. A general method for obtaining the MS size- $\alpha$  test  $\varphi^*$  for Problem  $(H, K_1)$ .**

Our method is based on the following two lemmas:

First let  $(H, K_{(m)})$  be a testing problem for the variate  $X$  over the sample space  $\mathfrak{X}$ , such that  $K_{(m)}$  consists of exactly  $m$  parameterpoints  $\theta_i$  defining pdf's  $f_{\theta_i}(x)$  over  $\mathfrak{X}$  ( $i = 1, \dots, m$ ). Let  $\varphi$  be MP size- $\alpha$  for testing  $H$  against the simple alternative that  $X$  has the pdf  $\sum p_i f_{\theta_i}(x)$  over  $\mathfrak{X}$  where  $p_1, \dots, p_m$  are certain probabilities ( $p_i \geq 0; \sum p_i = 1$ ).

LEMMA 1. *If these probabilities  $p_1, \dots, p_m$  are such that*

$$(8) \quad \gamma_{\varphi,D}(\theta_1) = \gamma_{\varphi,D}(\theta_2) = \dots = \gamma_{\varphi,D}(\theta_m)$$

*holds, then  $\varphi$  is MS size- $\alpha$  for problem  $(H, K_{(m)})$ .*

PROOF. Suppose  $\varphi$  does not minimize the maximum shortcoming over  $K_{(m)}$ . Then there exists a test  $\psi \in D$  such that  $\gamma_{\psi,D}(\theta_i) < \gamma_{\varphi,D}(\theta_i)$  and consequently  $\beta_{\psi}(\theta_i) > \beta_{\varphi}(\theta_i)$  ( $i = 1, \dots, m$ ). But the power of  $\psi$  against the above-described simple alternative is  $\sum p_i \beta_{\psi}(\theta_i)$ . Consequently  $\psi$  would be more powerful than  $\varphi$ . Thus we obtain a contradiction.

A complement of Lemma 1 will be given in the characterization Theorem 2 (Section 4).

Next let  $(H, K^{(1)})$  and  $(H, K^{(2)})$  be two problems for the variate  $X$  over  $\mathfrak{X}$ , such that  $K^{(1)} \subset K^{(2)}$ .

LEMMA 2. *If the test  $\varphi$  is MS size- $\alpha$  for Problem  $(H, K^{(1)})$  and*

$$(9) \quad \sup_{\theta \in K^{(1)}} \gamma_{\varphi,D}(\theta) = \sup_{\theta \in K^{(2)}} \gamma_{\varphi,D}(\theta)$$

*then  $\varphi$  is MS size- $\alpha$  for Problem  $(H, K^{(2)})$ .*

PROOF. For an arbitrary testing problem  $\Pi = (H, K)$  we define

$$(10) \quad b_D(\Pi) = \inf_{\varphi \in D} \sup_{\theta \in K} \gamma_{\varphi,D}(\theta),$$

where  $D$  is the class of all size- $\alpha$  tests;  $b_D(\Pi)$  is the maximum shortcoming on  $K$  of the MS size- $\alpha$  test for Problem  $\Pi$  if this test exists. Next let  $\Pi^{(i)}$  denote problem  $(H, K^{(i)})$  ( $i = 1, 2$ ). Then it follows from  $K^{(1)} \subset K^{(2)}$  that  $b_D(\Pi^{(1)}) \leq b_D(\Pi^{(2)})$  (see Lemma 3 in Section 7).

Moreover the left-hand side of (9) is equal to  $b_D(\Pi^{(1)})$ , for  $\varphi$  is MS size- $\alpha$  for Problem  $\Pi^{(1)}$ . Thus it follows from (9) that  $b_D(\Pi^{(1)}) \geq b_D(\Pi^{(2)})$ . Hence  $b_D(\Pi^{(1)}) = b_D(\Pi^{(2)})$  and (9) shows that  $\varphi$  is MS size- $\alpha$  for Problem  $\Pi^{(2)}$ .

In order to obtain the MS size- $\alpha$  test  $\varphi^*$  for Problem  $(H, K_1)$ , we shall try to find  $m$  parameterpoints  $\theta_1, \dots, \theta_m$  in  $K_1$ , defining pdf's  $f_{\theta_i}(x)$  over  $R^n$ , with probabilities  $p_1, \dots, p_m$  ( $\sum p_i = 1$ ) such that the MP size- $\alpha$  test  $\varphi$  for  $H$  against the simple alternative that  $X$  has the pdf  $\sum p_i f_{\theta_i}(x)$  satisfies (8), and (9) with  $K_{(m)}$  and  $K_1$  substituted for  $K^{(1)}$  and  $K^{(2)}$ . In that case the lemmas show that  $\varphi$  is the MS size- $\alpha$  test  $\varphi^*$  for Problem  $(H, K_1)$  and we write  $\theta_i^*$  and  $p_i^*$  instead of  $\theta_i$  and  $p_i$ .

Let  $f_1, \dots, f_n$  be an orthonormal basis of  $R^n$  such that  $f_{n-s+r+1}, \dots, f_n$  span  $R^{s-r}$  and  $f_{n-s+1}, \dots, f_n$  span  $R^s$ ; consequently  $f_{n-s+1}, \dots, f_{n-s+r}$  span  $R^r$  (see Section 2, [9] Section 3, or [10] Section 2.5). We shall only look for  $\theta_1^*, \dots, \theta_m^*$  among the points of  $K_1'$ . Suppose that  $\theta_1, \dots, \theta_m$  are  $m$  parameterpoints in  $K_1'$ . With respect to the basis  $f_1, \dots, f_n$  the coordinates of  $\theta_i$  are  $\eta_{i1}, \dots, \eta_{in}$  where  $\eta_{ij} = 0$  ( $j = 1, \dots, n - s; n - s + r + 1, \dots, n$ ) and accordingly  $\theta_i$  defines the pdf

$$f_{\theta_i}(x) = (2\pi)^{-\frac{1}{2}n} \exp \left\{ -\frac{1}{2} \sum_{j=1}^n (y_j - \eta_{ij})^2 \right\}$$

in the point  $x$  of  $R^n$  with coordinates  $y_1, \dots, y_n$  with respect to  $f_1, \dots, f_n$ .

Next suppose that  $p_1, \dots, p_m$  are  $m$  probabilities defining the pdf  $\sum p_i f_{\theta_i}(x)$  with respect to  $f_1, \dots, f_n$ , over  $R^n$ . The MP size- $\alpha$  test  $\varphi$  for the simple hypothesis:  $\eta_i = 0$  ( $i = 1, \dots, n$ ), against the simple alternative that  $X$  has the pdf  $\sum p_i f_{\theta_i}(x)$  is obtained by applying the Neyman-Pearson fundamental lemma. This test rejects when

$$(11) \quad \sum_{i=1}^m p_i \exp \left\{ \sum_{j=n-s+1}^{n-s+r} \eta_{ij} Y_j \right\} \geq c_\alpha,$$

where  $c_\alpha$  is determined such that the probability of (11) is equal to  $\alpha$  when  $Y_1, \dots, Y_n$  are independent  $N(0, 1)$  variates. Obviously this test  $\varphi$  is of size- $\alpha$  for testing the whole hypothesis  $H$  and consequently  $\varphi$  is MP size- $\alpha$  for testing  $H$  against the simple alternative that  $X$  has the pdf  $\sum p_i f_{\theta_i}(x)$ .

In the rest of this section and in the Sections 5, 6 and 7 we confine our attention to Problems  $(H, K_1')$  where  $K_1'$  is a *symmetrical* polyhedral angle (see Sections 6 and 7 for the definition of symmetry in case  $r = 3$  or 4);  $\bar{e}_i$  will denote the unit vector along the  $i$ th edge  $e_i$  of  $K_1'$  ( $i = 1, \dots, r$ ). For these problems we look for  $\theta_1^*, \dots, \theta_m^*$  only among the  $r$ -tuples  $(\theta_1, \dots, \theta_r)$  where  $\theta_i = \nu \bar{e}_i$  ( $i = 1, \dots, r$ ). On account of the symmetry of  $K_1'$ , the MP size- $\alpha$  test  $\varphi_\nu$  for  $H$  against the simple alternative that  $X$  has the pdf  $\sum p_i f_{\theta_i}(x)$  where  $\theta_i = \nu \bar{e}_i$ , will satisfy (8) if  $p_i = r^{-1}$  ( $i = 1, \dots, r$ ). So we look for  $\varphi^*$  only among these tests  $\varphi_\nu$  ( $\nu > 0$ ). The shortcoming of  $\varphi_\nu$  in all  $r$  points  $\theta_i = \nu \bar{e}_i$  is equal to the following function of  $\nu$ :

$$(12) \quad \gamma_{\varphi_\nu, D} \{ \theta(\nu) \} = \Phi^X(u_\alpha - \nu) - \beta_{\varphi_\nu}(\nu \bar{e}_i)$$

(the notation  $\theta(\nu)$  is introduced in order to indicate that this shortcoming does not depend on  $i$ ). Next we define  $\nu^*$  by the requirement that

$$(13) \quad \gamma_{\varphi_\nu^*, D} \{ \theta(\nu^*) \} = \sup_{\nu > 0} \gamma_{\varphi_\nu, D} \{ \theta(\nu) \}$$

stating that  $\varphi_{\nu^*}$  is a test of the form  $\varphi_\nu$  such that the shortcoming in  $\theta(\nu)$  (or equivalently in  $\nu \bar{e}_i$ ) is *maximized*.

The maximum shortcoming of  $\varphi_{\nu^*}$  over the half-line  $l$  in  $K_1'$  is determined by (see Formula (7))

$$(14) \quad \gamma_{\varphi_{\nu^*}, D}(l) = \sup_{\kappa > 0} \{ \Phi^X(u_\alpha - \kappa) - \beta_{\varphi_{\nu^*}}(\kappa \bar{l}) \}$$

where  $\bar{l}$  is the unit vector spanning  $l$ .

**THEOREM 1.** *The test  $\varphi_{\nu^*}$  is the MS size- $\alpha$  test  $\varphi^*$  for Problem  $(H, K_1')$  if*

$$(15) \quad \gamma_{\varphi_{\nu^*}, D}(l) \leq \gamma_{\varphi_{\nu^*}, D} \{ \theta(\nu^*) \}$$

*holds for each half-line  $l$  in  $K_1'$ .*

**PROOF.**  $\varphi_{\nu^*}$  is of the form  $\varphi_\nu$ . Consequently (Lemma 1)  $\varphi_{\nu^*}$  is MS size- $\alpha$  for Problem  $(H, K_{(r)}(\nu^*))$  where  $K_{(r)}(\nu^*)$  consists of the  $r$  points  $\nu^* \bar{e}_i$  ( $i = 1, \dots, r$ ) in  $K_1'$ . We have

$$(16) \quad \sup_{\theta \in K_{(r)}(\nu^*)} \gamma_{\varphi_{\nu^*}, D}(\theta) = \gamma_{\varphi_{\nu^*}, D} \{ \theta(\nu^*) \}.$$

On account of (15) we have

$$(17) \quad \gamma_{\varphi_{\nu^*}, D}(\theta) \leq \gamma_{\varphi_{\nu^*}, D}\{\theta(\nu^*)\},$$

for all  $\theta \in K_1'$  with equality for  $\theta = \nu^* \bar{e}_i \in K_1'$ . But  $\varphi_{\nu^*}$  and consequently  $\gamma_{\varphi_{\nu^*}, D}(\theta)$  are invariant under all translations parallel to  $R^{s-r}$  (see Formula (11)). Hence (17) holds for all  $\theta \in K_1$  with equality for  $\theta = \nu^* \bar{e}_i$ . This and (16) gives

$$(18) \quad \sup_{\theta \in K_1} \gamma_{\varphi_{\nu^*}, D}(\theta) = \sup_{\theta \in K_{(r)}(\nu^*)} \gamma_{\varphi_{\nu^*}, D}(\theta).$$

By applying Lemma 2 we complete the proof of the theorem.

REMARK. *The verification of Condition (15) is based on the results of the computations. For all problems which will be considered in the Sections 5, 6 and 7 we found*

$$(19) \quad \gamma_{\varphi_{\nu^*}, D}\{\theta(\nu^*)\} = \gamma_{\varphi_{\nu^*}, D}(e_i) \quad (i = 1, \dots, r)$$

showing that we indeed *minimize* the *maximum* shortcoming  $\gamma_{\varphi_{\nu}, D}(e_i)$  over  $e_i$  ( $i = 1, \dots, r$ ) when we *maximize* the shortcoming  $\gamma_{\varphi_{\nu}, D}\{\theta(\nu)\}$  of  $\varphi_{\nu}$  in  $\nu \bar{e}_i$  as a function of  $\nu$ . We shall prove (19) assuming that (i) there exists a  $\nu^*$  ( $0 < \nu^* < \infty$ ) satisfying (13) and (ii) the corresponding shortcoming  $f(\kappa) = \gamma_{\varphi_{\nu^*}, D}(\kappa \bar{e}_i)$  over  $e_i$  has exactly one maximum whereas  $f'(\kappa) \neq 0$  everywhere else ( $\kappa > 0$ ). Let the function  $g(\nu)$  denote the left-hand side of (12). Then (i)  $g(\nu)$  has a maximum for  $\nu = \nu^*$ , (ii)  $f(\nu^*) = g(\nu^*)$ , (iii)  $f(\kappa) \geq g(\kappa)$  for all  $\kappa > 0$  (otherwise for a certain  $\kappa$ ,  $\varphi_{\nu^*}$  would be more stringent than  $\varphi_{\kappa}$  for testing  $H$  against  $K_{(r)}(\kappa)$ ). Hence  $f'(\nu^*) = 0$  and on account of the assumptions,  $g(\nu^*) = f(\nu^*) = \sup_{\kappa} f(\kappa)$  what proves (19).

Consequently we may verify that

$$(20) \quad \gamma_{\varphi_{\nu^*}, D}(l) \leq \gamma_{\varphi_{\nu^*}, D}(e_i)$$

holds for each half-line  $l$  in  $K_1'$ , instead of Condition (15). Of course the right-hand side in (20) does not depend on  $i$  ( $i = 1, \dots, r$ ). It will turn out in the Sections 5, 6 and 7 that Condition (20) is satisfied provided that  $\Psi_0$  (see Section 2) is *not too large*. If  $\Psi_0$  is so large that (20) is not satisfied, then  $\varphi_{\nu^*}$  will be not MS size- $\alpha$  for Problem  $(H, K_1)$  (this follows from Theorem 4 under the assumptions of Theorem 3).

**4. Back-ground material for Section 3.** The Lemmas 1 and 2 and Theorem 1 give *sufficient* conditions for a test to be MS size- $\alpha$ . This section is devoted to some complementary results. It is not necessary to read this section before the Sections 5, 6 and 7 where only Theorem 1 is needed in order to prove that certain tests are MS size- $\alpha$ .

Condition (8) is sufficient for  $\varphi$  to be MS size- $\alpha$  (Lemma 1) for the general problem  $(H, K_{(m)})$  described at the beginning of Section 3. In case  $m = 2$ , Condition (8) is also necessary for  $\varphi$  to be MS size- $\alpha$ . For suppose  $\gamma_{\varphi, D}(\theta_1) < \gamma_{\varphi, D}(\theta_2)$  and let  $\varphi_2$  be the MP size- $\alpha$  test for  $H$  against the simple alternative  $\theta_2$ , then (for sufficiently small  $\rho > 0$ )  $(1 - \rho)\varphi + \rho\varphi_2 \in D$  will be more stringent than  $\varphi$  for

Problem  $(H, K_{(2)})$ . Moreover in case  $m = 2$  it can be shown easily that there exist probabilities  $p_1, p_2 (p_1 + p_2 = 1)$  such that the corresponding MP test  $\varphi$  satisfies (8). In case  $m > 2$  there will not always exist probabilities  $p_1, \dots, p_m$  such that (8) holds. For let  $\varphi$  be MS size- $\alpha$  for Problem  $(H, K_{(2)})$  where  $K_{(2)}$  consists of  $\theta_1, \theta_2 \in K_1'$  ( $\theta_1$  and  $\theta_2$  not on the same half-line  $l$ ). Let  $\theta_3, \dots, \theta_m$  be parameter-points in  $K_1'$  such that  $\gamma_{\varphi, D}(\theta_i) < \gamma_{\varphi, D}(\theta_1) = \gamma_{\varphi, D}(\theta_2)$  ( $i = 3, \dots, m$ ). Then  $\varphi$  is MS size- $\alpha$  for Problem  $(H, K_{(m)})$  (Lemma 2) but  $\varphi$  does not satisfy (8).

**CHARACTERIZATION THEOREM 2.** *If the sample space  $\mathfrak{X}$  satisfies a regularity condition ([7] p. 354) then there exists a number  $k (2 \leq k \leq m)$ , indices  $i_j (j = 1, \dots, k; i_j = 1, \dots, m)$  and probabilities  $p(i_j) (\sum p(i_j) = 1)$  such that the MP size- $\alpha$  test  $\varphi$  for  $H$  against the simple alternative that  $X$  has the pdf  $\sum p(i_j) f_{\theta(i_j)}(x)$  satisfies*

$$\gamma_{\varphi, D}(\theta_{i_1}) = \dots = \gamma_{\varphi, D}(\theta_{i_k}) = \gamma^* = \max_{i=1, \dots, m} \gamma_{\varphi, D}(\theta_i)$$

so that  $\varphi$  is MS size- $\alpha$  for Problem  $(H, K_{(m)})$  on account of the Lemmas 1 and 2.

**PROOF.** This is the situation of a *S-game* ([2] p. 47) where Player I (Nature) has a finite number of pure strategies  $\theta_i (i = 1, \dots, m)$  while Player II (Statistician) selects his strategy (test function)  $\varphi$  from the class  $D$  of all size- $\alpha$  tests. In the  $R^m$  of points  $g = (g_1, \dots, g_m)$  we define the subset  $S$  of all points  $g$  with coordinates  $g_i = \gamma_{\varphi, D}(\theta_i) (i = 1, \dots, m) (\varphi \in D)$ . Obviously equivalent tests (with the same power over  $K_{(m)}$ ) are represented by the same point in  $S$ . Let  $K$  denote the unit hypercube of all points  $g$  with  $0 \leq g_i \leq 1 (i = 1, \dots, m)$ .  $S$  is a closed, convex subset of  $K$  and  $S$  has a point in common with each of the  $m$  faces  $g_i = 0 (i = 1, \dots, m)$  of  $K$ . The convexity of  $S$  can be proved by remarking that when  $\varphi, \varphi' \in D$  are characterized by the points  $g, g' \in S$ , then for  $0 \leq \rho \leq 1$ , the test  $\rho\varphi + (1 - \rho)\varphi' \in D$  and this test is characterized by the point  $\rho g + (1 - \rho)g'$ . It follows from the Weak Compactness Theorem ([7] p. 354) provided that  $\mathfrak{X}$  satisfies a regularity condition, that (i) there exists a MP size- $\alpha$  test for  $H$  against the simple alternative  $\theta_i$  ( $S$  has a point in common with  $g_i = 0$ ) and (ii)  $S$  is closed.

Next ([2] p. 49) let  $\Gamma_\gamma = \{g; g \in R^m, g_i < \gamma (i = 1, \dots, m)\}$  denote the "negative orthant" in  $R^m$  shifted so that the vertex  $(0, \dots, 0)$  is at the point  $\gamma = (\gamma, \dots, \gamma)$ .  $\Gamma_\gamma$  is obviously open and convex. Let  $\gamma^* = \sup_\gamma \{\gamma; \Gamma_\gamma \cap S = \emptyset\}$ . Then  $\Gamma_{\gamma^*} \cap S = \emptyset$ ; hence there exists a hyperplane  $\sum p_i g_i = c$  separating  $\Gamma_{\gamma^*}$  and  $S$ . Without loss of generality we can assume  $p_i \geq 0 (i = 1, \dots, m), \sum p_i = 1$ ; moreover the hyperplane contains the vertex  $\gamma^*$  of  $\Gamma_{\gamma^*}$  ([2] page 43, [5] pages 11 and 12).  $[\Gamma_{\gamma^*}] \cap S \neq \emptyset$ . Let  $g^* \in S$  be such that  $g^* \in [\Gamma_{\gamma^*}]$ .  $S$  is convex and has a point in common with each of the  $m$  faces  $g_i = 0$  of  $K$ . Consequently at least two coordinates  $g_i^*$  of  $g^*$  are equal to  $\gamma^*$ . So there exists a number  $k (2 \leq k \leq m)$  and indices  $i_j (j = 1, \dots, k; i_j = 1, \dots, m)$  such that the coordinates  $g^*(i_j)$  are equal to  $\gamma^* (j = 1, \dots, k)$  whereas the other coordinates of  $g^*$  are smaller. The separating hyperplane  $\sum p_i g_i = c$  contains  $g^*$  and the vertex  $\gamma^*$ . Consequently  $p_i = 0$  for  $i \neq i_j (j = 1, \dots, k)$ . We complete the proof by remarking that the test  $\varphi$  which is represented by  $g^* \in S$  is MP size- $\alpha$  for testing  $H$  against the simple

alternative that  $X$  has the pdf  $\sum p_i f_{\theta(i)}(x)$ ; for the power  $\sum p_i \beta_\varphi(\theta_i)$  against this simple alternative becomes as large as possible when  $\sum p_i \gamma_{\varphi,D}(\theta_i)$  is minimized; the latter expression is  $\geq c$  for all points  $g \in S$  and  $=c$  for the point  $g^*$ .

Theorem 2 characterizes the parameterpoints  $\theta(i_j)$  ( $j = 1, \dots, k$ ) as those points where the shortcoming  $\gamma_{\varphi,D}(\theta)$  of the MS size- $\alpha$  test  $\varphi$  for Problem  $(H, K_{(m)})$  is as large as possible. We remark that each MS size- $\alpha$  test  $\psi$  for Problem  $(H, K_{(m)})$  is MP size- $\alpha$  for  $H$  against the simple alternative that  $X$  has the pdf  $\sum p(i_j) f_{\theta(i_j)}(x)$  for otherwise  $\sum p(i_j) \beta_\psi\{\theta(i_j)\} < \sum p(i_j) \beta_\varphi\{\theta(i_j)\}$  and hence  $\sum p(i_j) \gamma_{\psi,D}\{\theta(i_j)\} > \gamma^*$  and  $\gamma_{\psi,D}\{\theta(i_j)\} > \gamma^*$  for some index  $j$ . This remark will be useful for proving that the MS size- $\alpha$  test for Problem  $(H, K_{(m)})$  is *uniquely determined* under certain assumptions (see Theorem 4).

It can be proved that there exists a MS size- $\alpha$  test  $\psi$  for Problem  $(H, K_1)$  which is "invariant" or in other words, which only depends on the coordinates  $y_i$  ( $i = n - s + 1, \dots, n - s + r$ ) of the sample point  $x$  (see Section 3). For such tests we can restrict our attention to the power over  $K_1'$ .

**CHARACTERIZATION THEOREM 3.** *If there exists an invariant MS size- $\alpha$  test  $\psi$  for Problem  $(H, K_1)$ , such that a (finite) number  $m$  and parameterpoints  $\theta_i \in K_1'$  ( $i = 1, \dots, m$ ) exist with the property that*

$$\gamma_{\psi,D}(\theta_i) = \sup_{\theta \in K_1'} \gamma_{\psi,D}(\theta) = \gamma^*, \quad (i = 1, \dots, m),$$

*while for all other  $\theta \in K_1'$  we have  $\gamma_{\psi,D}(\theta) < \gamma^*$  (and even  $\gamma_{\psi,D}(\theta) < \gamma^* - \epsilon$  for some  $\epsilon > 0$ , provided that  $\|\theta\|$  is sufficiently large) then  $\psi$  is MS size- $\alpha$  for Problem  $(H, K_{(m)})$ .*

**PROOF.** Suppose  $\psi$  is not MS size- $\alpha$  for Problem  $(H, K_{(m)})$ . Let  $\varphi$  be a MS size- $\alpha$  test of the form (11), for Problem  $(H, K_{(m)})$  (see Theorem 2). Then  $\gamma_{\varphi,D}(\theta_i) < \gamma_{\psi,D}(\theta_i)$  ( $i = 1, \dots, m$ ) and we can construct  $\rho$  ( $0 < \rho < 1$ ) such that the test  $\rho\varphi + (1 - \rho)\psi \in D$  is more stringent for Problem  $(H, K_1)$  than  $\psi$ . Thus we obtain a contradiction.

**UNIQUENESS THEOREM 4.** *If  $(H, K_1)$  satisfies the condition of Theorem 3, then each MS size- $\alpha$  test  $\varphi^*$  for Problem  $(H, K_1)$  satisfies  $\varphi^* = \psi$  a.e. ( $\mu$ ).*

**PROOF.**  $\psi$  is MS size- $\alpha$  for Problem  $(H, K_{(m)})$  and  $\gamma_{\psi,D}(\theta_i) = \gamma^*$  ( $i = 1, \dots, m$ ) (Theorem 3). Hence  $\varphi^*$  is MS size- $\alpha$  for Problem  $(H, K_{(m)})$ . Consequently both  $\psi$  and  $\varphi^*$  are MP size- $\alpha$  for testing  $H$  against the simple alternative that  $X$  has a pdf  $\sum p(i) f_{\theta(i)}(x)$  where  $\theta(i) \in K_1'$ . Hence  $\varphi^* = \psi$  a.e. ( $\mu$ ) where  $\mu$  is the Lebesgue measure over  $R^n$ .

In the following sections, the condition of Theorem 3 turns out to hold for certain problems. For problems  $(H, K_1)$  satisfying the condition (we conjecture that each problem  $(H, K_1)$  will do that) there obviously exist probabilities  $p_i$  ( $i = 1, \dots, m$ ) such that the MS size- $\alpha$  test  $\varphi^*$  for Problem  $(H, K_1)$  can be obtained by applying Lemma 1 (Condition (8) is satisfied) and Lemma 2. This shows that Condition (20) (or (15)) is necessary for  $\varphi_r$  to be MS size- $\alpha$  for Problem  $(H, K_1)$  where  $K_1'$  is symmetrical (Remark in Section 3). By introducing the condition of Theorem 3, we avoided the difficulties that emerge when the alternative consists of an infinite number of points



Bartholomew [1], Kudô [6] a.o. applied the L.R. (*likelihood-ratio*) criterion to problems of the form  $(H, K_1)$ . Their L.R. size- $\alpha$  test  $\varphi'$  is not of the form (11). Consequently  $\varphi'$  is not MS size- $\alpha$  (if at least the condition of Theorem 3 is satisfied). Theoretically, the L.R. criterion has the disadvantage that it is not formulated as an optimum property in terms of the power ([7], page 15). From a practical point of view,  $\varphi'$  seems to stand somewhere midway between  $\varphi_0$  and  $\varphi^*$ : the maximum shortcoming will lie between that of  $\varphi_0$  and  $\varphi^*$ ; the computations involved are more complicated than for  $\varphi_0$  but much less difficult than for  $\varphi^*$ .

Doornbos and Prins [3], Kander and Zacks [4], a.o. considered testing problems which can be regarded as (generalizations of) special cases of Problem  $(H, S_1)$  where on the basis of an outcome of the random variable  $X$  described in Section 2, the hypothesis  $H$  has to be tested against the alternative  $S_1$  corresponding with the subset of  $R^r$  which is defined by the existence of an index  $h$  ( $h = n - s + 1, \dots, n - s + r$ ) such that strict inequality holds in (4) for this index whereas equality holds for all other indices.

EXAMPLE. Let  $X_1, \dots, X_r$  have independent normal  $N(\mu_i, 1)$  distributions ( $i = 1, \dots, r$ ). The hypothesis  $H: \mu_i = 0$  ( $i = 1, \dots, r$ ) has to be tested against the alternative  $S_1$ : for some index  $h$  ( $h = 1, \dots, r$ ) we have  $\mu_h > 0$  whereas  $\mu_i = 0$  ( $i = 1, \dots, r; i \neq h$ ), or in other words that the vector of means  $(\mu_1, \dots, \mu_r)$  is situated on one of the edges of the positive orthant.

Obviously for the general problem  $(H, S_1)$ , the intersection  $S_1'$  of  $S_1$  and  $R^r$  (see Section 2) consists of the  $r$  edges  $e_i$  ( $i = 1, \dots, r$ ) of  $K_1'$ . The class of all SMP size- $\alpha$  tests for Problem  $(H, S_1)$  consists of the  $r$  tests  $0X^I \geq u_\alpha$  where  $X^I$  is the projection of  $X$  on the  $R^1$  spanned by the half-line  $l$  and  $l = e_i$  ( $i = 1, \dots, r$ ). Consequently the criterion MSSMP is inappropriate for Problem  $(H, S_1)$ . It is seen easily that the MSSMP size- $\alpha$  test  $\varphi_0$  for Problem  $(H, K_1)$  is MS ( $C$ ) for Problem  $(H, S_1)$  when  $C$  is the class of all tests of the form  $0X^I \geq u_\alpha$ , which are based on a *linear* combination of the coordinates  $X_1, \dots, X_n$  (Kander and Zacks applied a Bayesian approach with a limiting argument and thus obtained a test of the form  $0X^I \geq u_\alpha$  which is not MS ( $C$ )). The restriction to the above-mentioned class  $C$  seems to be rather undesirable for problems of the form  $(H, S_1)$ . The method described after the proof of Lemma 2 where  $m = r$  and  $\theta_i = \nu_i \bar{e}_i$  ( $i = 1, \dots, r$ ) will provide the MS size- $\alpha$  test for Problem  $(H, S_1)$ . The computations are feasible if we restrict our attention to Problems  $(H, S_1)$  where the corresponding polyhedral angle  $K_1'$  is *symmetrical* (see Example). For such cases it follows from (19) that  $\varphi_{\nu^*}$  is the MS size- $\alpha$  test for Problem  $(H, S_1)$  (Condition (20) needs not to be satisfied).

The above-mentioned Example provides an interesting test-case. The corresponding tests  $\varphi_0$ ,  $\varphi_\nu$  and the L.R. size- $\alpha$  test  $\varphi'$  respectively reject when  $\sum X_i$ ,  $\sum \exp(\nu X_i)$  and  $\max X_i$  are sufficiently large.  $\varphi_0$  and  $\varphi'$  may be regarded as limiting cases ( $\nu \rightarrow 0$  and  $\nu \rightarrow \infty$  respectively) of  $\varphi_\nu$ ; consequently the MS size- $\alpha$  test  $\varphi_{\nu^*}$  has an intermediate position between  $\varphi_0$  and  $\varphi'$ . The method of Kander and Zacks inheres in the construction of  $\varphi_0$  (their limiting argument corresponds with  $\nu \rightarrow 0$  in  $\varphi_\nu$ ) whereas the method of Doornbos and Prins provides

$\varphi'$ . The tests  $\varphi_0$ ,  $\varphi_{\nu^*}$  and  $\varphi'$  respectively have the maximum shortcomings .26, .11, .11 ( $r = 2$ ); .35, .17, .17 ( $r = 3$ ) and .47, .21, .21 ( $r = 4$ ) over the alternative  $S_1$  in case  $\alpha = .05$ . This shows that for these cases  $\varphi'$  does not differ much from the MS size- $\alpha$  test  $\varphi_{\nu^*}$ . The maximum shortcoming of  $\varphi_0$  is unsatisfactory large; the corresponding advantages of  $\varphi_{\nu^*}$  (and  $\varphi'$ ) over  $\varphi_0$  are not neutralized by advantages of  $\varphi_0$  over  $\varphi_{\nu^*}$  from other reasonable over-all points of view (see [11] Figure 5.2 where the shortcoming of  $\varphi_0$  and  $\varphi_{\nu^*}$  is plotted for the  $r = 2$  case).

**5. Comparing  $\varphi_0$  and  $\varphi^*$  for Problems  $(H, K_1)$  with  $r = 2$ .** In the case  $r = 2$ ,  $K_1'$  is an angle with vertex 0 in  $R^r = R^2$  and  $K_1'$  is obviously symmetrical. We introduce a new orthonormal basis  $g_1, \dots, g_n$  for  $R^n$  with  $g_1 = \|\bar{e}_1 - \bar{e}_2\|^{-1}(\bar{e}_1 - \bar{e}_2)$  and  $g_2 = \|\bar{e}_1 + \bar{e}_2\|^{-1}(\bar{e}_1 + \bar{e}_2)$ . The half-line  $l_0$  satisfying (6) is spanned by  $g_2$ . Further  $\Psi_0 = \Psi(g_2, e_1)$ . The coordinates of the sample point  $X$  with respect to the basis  $g_1, \dots, g_n$  are denoted by  $Z_1, \dots, Z_n$ . Obviously  $\varphi_0$  rejects when  $Z_2 \geq u_\alpha$ . Graphs of  $\gamma_{\varphi_0, D}(l)$  as a function of  $\Psi = \Psi(l, l_0)$  (see Formula (7)) are given in the Figures 3, 4 and 5.

Next we try to obtain  $\varphi^*$  by determining  $\varphi_{\nu^*}$  (see Section 3). Using the basis  $g_1, \dots, g_n$  we have  $\bar{e}_i = ((-1)^{i-1} \sin \Psi_0, \cos \Psi_0, 0 \dots)$  ( $i = 1, 2$ );  $\theta_i = \nu \bar{e}_i$  defines the  $n$ -variate normal  $N(\theta_i, I)$  distribution with pdf  $f_{\theta_i}(x)$  ( $i = 1, 2$ ). The MP size- $\alpha$  test  $\varphi_\nu$  for  $H$  against the simple alternative that  $X$  has the pdf  $\frac{1}{2}f_{\theta_1}(x) + \frac{1}{2}f_{\theta_2}(x)$  rejects when

$$\{\exp(-\frac{1}{2}Z_1^2 - \frac{1}{2}Z_2^2)\}^{-1} [\exp[-\frac{1}{2}\{(Z_1 - \nu \sin \Psi_0)^2 + (Z_2 - \nu \cos \Psi_0)^2\}] + \exp[-\frac{1}{2}\{(Z_1 + \nu \sin \Psi_0)^2 + (Z_2 - \nu \cos \Psi_0)^2\}]]$$

is sufficiently large, or equivalently when

$$(21) \quad Z_2 \geq f_{\nu, \alpha, \Psi_0}(Z_1),$$

where

$$(22) \quad f_{\nu, \alpha, \Psi_0}(z_1) = (\nu \cos \Psi_0)^{-1}$$

$$\log_e \{c_\alpha(\nu, \Psi_0) / [\exp(z_1 \nu \sin \Psi_0) + \exp(-z_1 \nu \sin \Psi_0)]\}$$

with  $c_\alpha(\nu, \Psi_0)$  such that Test (21) is of size- $\alpha$  for testing  $H$  or equivalently for testing the hypothesis that  $Z_1$  and  $Z_2$  have independent normal  $N(0, 1)$  distributions.

Let  $l$  denote the half-line in  $K_1'$  spanned by the vector  $\bar{l} = (\sin \Psi, \cos \Psi, 0 \dots)$  where  $|\Psi| \leq \Psi_0$ . The power of  $\varphi_\nu$  in the arbitrary point  $\kappa \bar{l}$  of  $l$  ( $\kappa > 0$ ) is determined by

$$(23) \quad \beta_{\varphi_\nu}(\kappa \bar{l}) = P\{Z_2 \geq f_{\nu, \alpha, \Psi_0}(Z_1)\},$$

where in the right-hand side  $Z_1$  and  $Z_2$  have independent normal  $N(\kappa \sin \Psi, 1)$  and  $N(\kappa \cos \Psi, 1)$  distributions respectively.

Next we compute  $\nu^*$  satisfying (13) (the computations are based on the assumption that the function  $g(\nu)$  that is defined by the left-hand side of (12) has exactly one summit for  $\nu > 0$ , cf [11] Theorem 4.1). Thus we obtain the test  $\varphi_{\nu^*}$

(that is MS size- $\alpha$  for the allied Problem  $(H, S_1)$  described in Section 4) and the problem arises whether  $\varphi_{\nu^*}$  is the MS size- $\alpha$  test  $\varphi^*$  for Problem  $(H, K_1)$  or equivalently whether Condition (20) is satisfied for each half-line  $l$  in  $K_1'$  (see also (14) and (23)). In the limiting case  $\Psi_0 = \frac{1}{2}\pi$  the tests  $\varphi_{\nu}$  ( $\nu > 0$ ) degenerate into the same test  $\varphi'$  which rejects when  $|Z_1| \geq u_{\frac{1}{2}\alpha}$ . We have  $\gamma_{\varphi', D}(l_0) = 1 - \alpha$  and (20) is not satisfied. There will exist a critical angle  $\Psi_0^{(cr)}$  ( $\alpha$ ) such that (20) is satisfied and  $\varphi_{\nu^*}$  is the MS size- $\alpha$  test  $\varphi^*$  for Problem  $(H, K_1)$ , if and only if

$$(24) \quad 0 \leq \Psi_0 \leq \Psi_0^{(cr)}(\alpha).$$

In order to determine this critical angle we examined Condition (20) by considering graphs of  $\gamma_{\varphi_{\nu^*}, D}(l)$  as a function of  $\Psi = \Psi(l, l_0)$  (see Figures 3, 4 and 5). Indeed an angle  $\Psi_0^{(cr)}$  ( $\alpha$ ) turned out to exist such that  $\varphi_{\nu^*} = \varphi^*$  if (24) holds, whereas otherwise  $\gamma_{\varphi_{\nu^*}, D}(l)$  obtains an absolute maximum for  $l = l_0$ . In this connection we refer to [11] where *van Zwet and Oosterhoff* proved analytically for the  $\Psi_0 = 45^\circ$  case that  $\gamma_{\varphi_{\nu^*}, D}(l)$  as function of  $\Psi = \Psi(l, l_0)$  can only have maxima for  $\Psi = 0$  and  $\Psi = \Psi_0$ .

The test  $\varphi_{\nu^*}$  is of the form (21), (22) and is completely determined if we know  $\nu^*$  and  $c_\alpha(\nu^*, \Psi_0)$ . Writing  $\nu^* = \nu_\alpha^*(\Psi_0)$  and  $c_\alpha(\nu_\alpha^*(\Psi_0), \Psi_0) = c_\alpha^*(\Psi_0)$  we indicate that  $\nu^*$  and  $c^*$  are completely determined by  $\alpha$  and  $\Psi_0$ . Corresponding graphs for all values  $\Psi_0$  satisfying (24) are given in the Figures 1 and 2. These figures determine  $\varphi_{\nu^*}$  completely, exactly in those cases where  $\varphi_{\nu^*}$  is the MS size- $\alpha$  test  $\varphi^*$  for Problem  $(H, K_1)$ . Moreover these figures determine  $\Psi_0^{(cr)}$  ( $\alpha$ ).

We proposed to compare the MSSMP size- $\alpha$  test  $\varphi_0$  and the MS size- $\alpha$  test  $\varphi^*$  for Problem  $(H, K_1)$  with  $r = 2$  by means of graphs of  $\gamma_{\varphi, D}(l)$  as function of  $\Psi = \Psi(l, l_0)$  for  $\varphi = \varphi_0$  and  $\varphi = \varphi^*$ . Such graphs are given in the Figures 3, 4 and 5 ( $\alpha = .05$ ).

By definition, the maximum shortcoming of  $\varphi^*$  is smaller than that of  $\varphi_0$ . But from other points of view  $\varphi_0$  may be better. In Figure 6 for example we give a graph of (see also [10] Section 2.15)

$$\lambda_\alpha(\Psi_0) = \Psi_0^{-1} \Psi_1(\Psi_0, \alpha),$$

where  $\Psi_1(\Psi_0, \alpha)$  is by definition such that the condition  $\Psi(l, l_0) \leq \Psi_1(\Psi_0, \alpha)$  is necessary and sufficient for  $\gamma_{\varphi_0, D}(l) \leq \gamma_{\varphi^*, D}(l)$ . So  $\varphi_0$  is regarded as better than  $\varphi^*$  from this point of view if  $\lambda_\alpha(\Psi_0) > \frac{1}{2}$ .

We can also compute (cf [10] Sections 2.15 and 4.3)

$$A_\varphi(\Psi_0) = \frac{1}{2} \Psi_0^{-1} \int_{-\Psi_0}^{\Psi_0} \gamma_{\varphi, D}(l) d\Psi$$

where  $\gamma_{\varphi, D}(l)$  is regarded as a function of  $\Psi = \Psi(l, l_0)$ . Corresponding graphs for  $\varphi_0$  and  $\varphi^*$  ( $0 \leq \Psi_0 \leq \Psi_0^{(cr)}(\alpha)$ ) are given in Figure 7.

Interpreting the Figures 3, . . . , 7 and the theory of [10] Section 2.15 we arrive at the following *rule of thumb*: for Problems  $(H, K_1)$  with  $r = 2$  and  $\Psi_0 < 60^\circ$ ,  $\varphi^*$  does not provide a worth-while improvement upon  $\varphi_0$ .

REMARK. In [9] Section 2 and [10] Section 2.3 we made objections to the criterion MS size- $\alpha$ . Supposing that  $\gamma_{\varphi^*, D}(l)$  was almost constant over  $|\Psi| \leq \Psi_0$  for

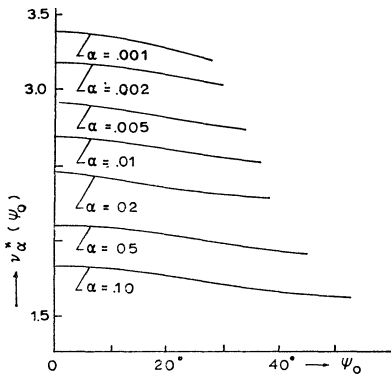


FIG. 1.

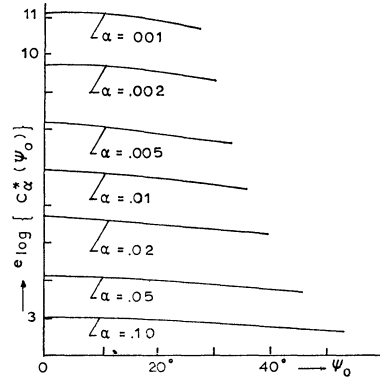


FIG. 2.

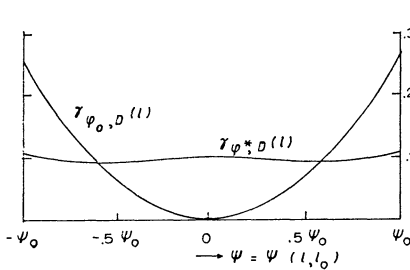


FIG. 3. ( $\alpha = .05; \psi_0 = 45^\circ$ )

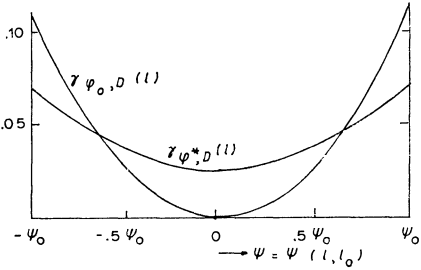


FIG. 4. ( $\alpha = .05; \psi_0 = 30^\circ$ )

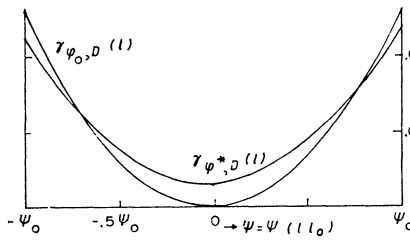


FIG. 5. ( $\alpha = .05; \psi_0 = 15^\circ$ )

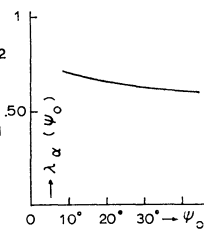


FIG. 6. ( $\alpha = .05$ );

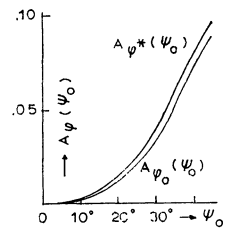


FIG. 7. ( $\alpha = .05$ )

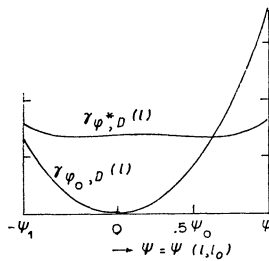


FIG. 8. ( $\alpha = .05; \psi = 52.5^\circ$ );

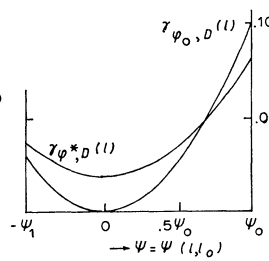


FIG. 9. ( $\alpha = .05; \psi = 30^\circ$ )

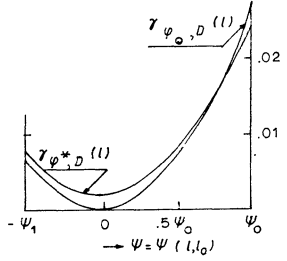


FIG. 10. ( $\alpha = .05; \psi = 15^\circ$ )

$\Psi = \Psi(l, l_0)$ , we expected that  $\varphi_0$  would provide a worth-while improvement upon  $\varphi^*$  for Problems  $(H, K_1)$  with  $\Psi_0$  small. The Figures 4 and 5 show that these objections do not apply to Problems  $(H, K_1)$  in case  $r = 2$ . For such problems both  $\varphi_0$  and  $\varphi^*$  have similar power properties if  $\Psi_0$  is small.

**6. Comparing  $\varphi_0$  and  $\varphi^*$  for Problem  $(H, K_1)$  with  $r = 3$  and  $K_1'$  symmetrical.** In this case,  $K_1'$  is a symmetrical polyhedral angle in  $R^r = R^3$ . The symmetry may be defined by  $\Psi(e_1, e_2) = \Psi(e_1, e_3) = \Psi(e_2, e_3)$ . We can introduce an orthonormal basis  $g_1, \dots, g_n$  for  $R^n$  such that the unit vectors  $\bar{e}_i$  along the edges  $e_i$  of  $K_1'$  have the coordinates

$$\begin{aligned} \bar{e}_1 &= (\sin \Psi_0, 0, \cos \Psi_0, 0 \dots) \\ \bar{e}_2 &= (-\frac{1}{2} \sin \Psi_0, \frac{1}{2} 3^{\frac{1}{2}} \sin \Psi_0, \cos \Psi_0, 0 \dots) \\ \bar{e}_3 &= (-\frac{1}{2} \sin \Psi_0, -\frac{1}{2} 3^{\frac{1}{2}} \sin \Psi_0, \cos \Psi_0, 0 \dots). \end{aligned}$$

The half-line  $l_0$  satisfying (6) is spanned by  $g_3$ ;  $\Psi(l_0, e_i) = \Psi_0$  ( $i = 1, 2, 3$ ). The coordinates of the sample point  $X$  with respect to the basis  $g_1, \dots, g_n$  are denoted by  $Z_1, \dots, Z_n$ . Obviously  $\varphi_0$  rejects when  $Z_3 \geq u_\alpha$ .

The test  $\varphi_\nu$  (see Sections 3 and 5) rejects when

$$(25) \quad Z_3 \geq f_{\nu, \alpha, \Psi_0}(Z_1, Z_2)$$

where  $f$  is defined by

$$(26) \quad (\nu \cos \Psi_0) f_{\nu, \alpha, \Psi_0}(z_1, z_2) = \log_e \{ c_\alpha(\nu, \Psi_0) [\exp(z_1 \nu \sin \Psi_0) + \exp(-\frac{1}{2} z_1 \nu \sin \Psi_0) \{ \exp(\frac{1}{2} 3^{\frac{1}{2}} z_2 \nu \sin \Psi_0) + \exp(-\frac{1}{2} 3^{\frac{1}{2}} z_2 \nu \sin \Psi_0) \}]^{-1} \},$$

with  $c_\alpha(\nu, \Psi_0)$  such that Test (25) is of size- $\alpha$  for testing  $H$ .

The test  $\varphi_{\nu^*}$  is determined according to (12) and (13) (the computations are based on the assumption that (12) as function of  $\nu$  has exactly one summit). Condition (20) is examined by computing  $\gamma_{\varphi^*, D}(l)$  as a function of  $\Psi = \Psi(l, l_0)$  for  $l$  varying over the intersection of  $K_1'$  and the  $R^2$  spanned by  $e_1$  and  $g_3$ . So  $\Psi$  varies over  $[-\Psi_1, \Psi_0]$  where

$$(27) \quad \Psi_1 = \arccos \{ \Psi(\bar{e}_2 + \bar{e}_3, g_3) \} = \arccos \{ 2 \cos \Psi_0 / (1 + 3 \cos^2 \Psi_0)^{\frac{1}{2}} \}.$$

Corresponding graphs of  $\gamma_{\varphi, D}(l)$  for  $\varphi = \varphi_0$  and  $\varphi = \varphi_{\nu^*} = \varphi^*$  are given in the Figures 8, 9 and 10. Here  $\varphi_{\nu^*}$  is assumed to be the MS size- $\alpha$  test  $\varphi^*$  for Problem  $(H, K_1)$  because the function  $\gamma_{\varphi^*, D}(l)$  of  $\Psi = \Psi(l, l_0)$  over  $[-\Psi_1, \Psi_0]$  has an absolute maximum for  $\Psi = \Psi_0$  (see Formula (19)).

Interpreting these figures (and other results mentioned in [8] and in [10] Section 4.3) we obtain the opinion that the *rule of thumb* at the end of Section 5 may be modified for Problems  $(H, K_1)$  with  $r = 3$ , substituting  $70^\circ$  for  $60^\circ$  for example.

From certain over-all points of view,  $\varphi_0$  provides a slight improvement upon  $\varphi^*$  if  $40^\circ < \Psi_0 < 60^\circ$ . Further the power properties of  $\varphi_0$  and  $\varphi^*$  are quite similar for  $\Psi_0 < 30^\circ$ .

**7. Comparing  $\varphi_0$  and  $\varphi^*$  for Problems  $(H, K_1)$  with  $r > 3$ .** The case  $K_1'$  symmetrical and  $r = 4$  was treated also along the lines of Section 3. In this case we introduce an orthonormal basis  $g_1, \dots, g_n$  for  $R^n$  such that

$$\begin{aligned} \bar{e}_1 &= (\sin \Psi_0, 0, 0, \cos \Psi_0, 0 \dots), \\ \bar{e}_2 &= (-3^{-1} \sin \Psi_0, 3^{-1} 8^{\frac{1}{2}} \sin \Psi_0, 0, \cos \Psi_0, 0 \dots), \\ \bar{e}_3 &= (-3^{-1} \sin \Psi_0, -3^{-1} 2^{\frac{1}{2}} \sin \Psi_0, 3^{-\frac{1}{2}} 2^{\frac{1}{2}} \sin \Psi_0, \cos \Psi_0, 0 \dots), \\ \bar{e}_4 &= (-3^{-1} \sin \Psi_0, -3^{-1} 2^{\frac{1}{2}} \sin \Psi_0, -3^{-\frac{1}{2}} 2^{\frac{1}{2}} \sin \Psi_0, \cos \Psi_0, 0 \dots), \end{aligned}$$

and the MSSMP size- $\alpha$  test  $\varphi_0$  rejects when  $Z_4 \geq u_\alpha$  whereas the test  $\varphi_\nu$  (see Sections 3 and 6) rejects when

$$(28) \quad Z_4 \geq f_{\nu, \alpha, \Psi_0}(Z_1, Z_2, Z_3)$$

where  $f$  is defined by a formula similar to (26) but with a more intricate denominator in the right-hand side.

Again  $\varphi_{\nu^*}$  can be determined and Condition (20) is examined by computing  $\gamma_{\varphi_{\nu^*}, D}(l)$  as a function of  $\Psi = \Psi(l, l_0) = \Psi(l, g_4)$  for  $l$  varying over the intersection of  $K_1'$  and the  $R^2$  spanned by  $e_1$  and  $g_4$ . So  $\Psi$  varies over  $[-\Psi_1, \Psi_0]$  where

$$(29) \quad \Psi_1 = \Psi(\bar{e}_2 + \bar{e}_3 + \bar{e}_4, g_4) = \arccos \{3 \cos \Psi_0 / (1 + 8 \cos^2 \Psi_0)^{\frac{1}{2}}\}.$$

Interpreting the corresponding graphs of  $\gamma_{\varphi_{\nu^*}, D}(l)$  we obtain the opinion that the *rule of thumb* at the end of Section 5 may be modified for Problems  $(H, K_1)$  with  $r = 4$ , substituting  $75^\circ$  for  $60^\circ$  for example.

From certain over-all points of view,  $\varphi_0$  provides a worth-while improvement upon  $\varphi^*$  if  $35^\circ < \Psi_0 < 65^\circ$ . The power properties of  $\varphi_0$  and  $\varphi^*$  are quite similar for  $\Psi_0 < 30^\circ$ .

The rest of this section is devoted to a comparison of  $\varphi_0$  and  $\varphi^*$  for Problems  $(H, K_1)$  where  $K_1'$  is an *orthant*:  $\Psi(e_i, e_j) = 90^\circ$  for  $i, j = 1, \dots, r; i \neq j$  (in [10] Section 4.3 we examined such problems as feasible test-cases for the provisional classification in [10] Section 2.13). For such problems we have  $\Psi_0 = \arccos(r^{\frac{1}{2}})$ . Let  $A_\varphi(\Psi_0)$  denote the averaged maximum shortcoming  $\gamma_{\varphi, D}(l)$  when  $\bar{l}$  has the uniform distribution over the intersection of  $K_1'$  and the surface of the unit sphere in  $R^r$ . For  $\alpha = .05$  we have

$$(30) \quad A_{\varphi_0}(\Psi_0) \approx (1 - \alpha) \{1 - \pi^{-\frac{1}{2}} [\Gamma\{\frac{1}{2}(r + 1)\}]^{-1} r^{\frac{1}{2}} \Gamma(\frac{1}{2}r)\}$$

(see [10] Section 4.3, the formula for  $A_\alpha(m)$ ). We shall use  $A_\varphi(\Psi_0)$  as a measure for comparing  $\varphi_0$  and  $\varphi^*$  for Problems  $(H, K_1)$  where  $K_1'$  is an orthant,  $r = 2, 3, 4, \dots$  and  $\alpha = .05$ .

Figure 3 applies to the case  $r = 2$ . We compute  $A_{\varphi^*}(\Psi_0) = .097$  and  $A_{\varphi_0}(\Psi_0) = .084$ ; further  $\Psi_0^{(cr)}(\alpha) \approx 46^\circ$ . Neither  $\varphi_0$  nor  $\varphi^*$  provides a worth-while improvement upon the other.

Condition (20) is not satisfied in the case  $r = 3$ :  $\Psi_0^{(cr)}(\alpha) \approx 54^\circ < \Psi_0 = 54.7^\circ$ .  $\gamma_{\varphi_{\nu^*}, D}(l)$  is almost constantly equal to .16 for  $l$  varying over the region studied

(see Section 6). We shall have  $A_{\varphi^*}(\Psi_0) \approx .16$  for the unknown MS size- $\alpha$  test  $\varphi^*$ . From (30) we obtain  $A_{\varphi_0}(\Psi_0) \approx .13$ . Neither  $\varphi_0$  nor  $\varphi^*$  provides a worth-while improvement upon the other;  $\varphi_0$  is slightly better from the  $A_{\varphi}(\Psi_0)$  point of view.

In the case  $r = 4$  we have  $A_{\varphi_0}(\Psi_0) \approx .15$  (see (30)). Condition (20) is not satisfied:  $\Psi_0^{(er)}(\alpha) \approx 59^\circ < \Psi_0 = 60^\circ$ . But  $\gamma_{\varphi^*,D}(l)$  is almost constantly equal to about .21 for  $l$  varying over the region studied. We shall have  $A_{\varphi^*}(\Psi_0) \approx .21$  for the unknown MS size- $\alpha$  test  $\varphi^*$ . So  $\varphi_0$  provides a worth-while improvement upon  $\varphi^*$  from the  $A_{\varphi}(\Psi_0)$  point of view whereas the maximum shortcoming of  $\varphi^*$  (about .21) is much smaller than the maximum shortcoming (.47) of  $\varphi_0$ .

Next, by applying two simple lemmas, we prove Theorem 5 elucidating the cases  $r > 4$ .

LEMMA 3. If  $\Pi^{(1)} = (H, K^{(1)})$  and  $\Pi^{(2)} = (H, K^{(2)})$  are two testing problems for the variate  $X$  over  $\mathfrak{X}$  such that  $K^{(1)} \subset K^{(2)}$ , then  $b_D(\Pi^{(1)}) \leq b_D(\Pi^{(2)})$  ( $b_D(\Pi)$  was defined in (10)).

Next let  $\Pi_r$  and  $\bar{\Pi}_r$  be two testing problems of the form  $(H, K_1)$  such that for both problems  $K_1'$  is an orthant in an  $R^r$  ( $n$  and  $s$  may be different).

LEMMA 4. For  $\Pi_r$  and  $\bar{\Pi}_r$  we have  $b_D(\Pi_r) = b_D(\bar{\Pi}_r)$ .

We shall elucidate the cases  $r > 4$  by considering a sequence  $\{\Pi_r\}$  ( $r = 2, 3, \dots$ ) of problems of the form  $(H, K_1)$  where  $K_1'$  is an orthant in  $R^r$ .

THEOREM 5. For each sequence  $\{\Pi_r\}$  ( $r = 2, 3, \dots$ ) we have (i)  $b_D(\Pi_r) \leq b_D(\Pi_{r+1})$  ( $r = 2, 3, \dots$ ) and (ii)  $\lim_{r \rightarrow \infty} b_D(\Pi_r) = 1 - \alpha$ .

PROOF. (i) may be proved by applying Lemma 3 and Lemma 4. (ii) will be proved by applying Lemma 1 and Lemma 3. First we remark that the trivial size- $\alpha$  test  $\varphi_t(x) \equiv \alpha$  belongs to  $D$ . Moreover  $\sup_{\theta \in K_1} \gamma_{\varphi_t, D}(\theta) = 1 - \alpha$ . Hence  $b_D(\Pi_r) \leq 1 - \alpha$  ( $r = 2, 3, \dots$ ).

We shall construct for  $\epsilon > 0$  an integer  $r_\epsilon$  such that  $b_D(\Pi_r) > 1 - \alpha - \epsilon$  for  $r > r_\epsilon$ . For that purpose we define  $\nu_\epsilon$  such that  $\Phi^X(u_\alpha - \nu_\epsilon) - \alpha = 1 - \alpha - \frac{1}{2}\epsilon$ . With respect to Problem  $\Pi_r$  we introduce an orthonormal basis  $g_1, \dots, g_n$  for  $R^n$  such that  $g_i = \bar{e}_i$  ( $i = 1, \dots, r$ ). This is possible because  $K_1'$  is an orthant. The coordinates of the sample point  $X$  are denoted by  $Z_1, \dots, Z_n$ . By applying Lemma 1 we obtain the MS size- $\alpha$  test  $\varphi_{\nu_\epsilon}$  for  $H$  against the alternative  $K_{(r)}(\nu_\epsilon) \subset K_1$  where  $K_{(r)}(\nu_\epsilon)$  consists of the parameterpoints  $\nu_\epsilon g_i$  ( $i = 1, \dots, r$ ).  $\varphi_{\nu_\epsilon}$  rejects when

$$\sum_{i=1}^r \exp(\nu_\epsilon Z_i) \geq c_{\alpha,r}(\nu_\epsilon)$$

where  $c_{\alpha,r}(\nu_\epsilon)$  is determined such that the test is of size- $\alpha$ . By applying the Central Limit Theorem for  $r \rightarrow \infty$  ( $\nu_\epsilon$  is fixed) it can be shown that the power of  $\varphi_{\nu_\epsilon}$  in  $\nu_\epsilon g_i$  tends to  $\alpha$  as  $r \rightarrow \infty$ . Thus we can choose  $r_\epsilon$  such that for  $r > r_\epsilon$  this power is smaller than  $\alpha + \frac{1}{2}\epsilon$ . Consequently the corresponding shortcoming is larger than  $\Phi^X(u_\alpha - \nu_\epsilon) - \alpha - \frac{1}{2}\epsilon$ . But this shortcoming is equal to  $b_D(H, K_{(r)}(\nu_\epsilon))$ . By applying Lemma 3 we obtain

$$b_D(\Pi_r) > \Phi^X(u_\alpha - \nu_\epsilon) - \alpha - \frac{1}{2}\epsilon \geq 1 - \alpha - \epsilon, \text{ for } r > r_\epsilon.$$

This completes the proof of the theorem.

By definition the MS size- $\alpha$  test  $\varphi^*$  has the advantage over the MSSMP size- $\alpha$  test  $\varphi_0$ , that the corresponding maximum shortcoming is smaller. Theorem 5 shows that this advantage becomes unimportant when  $r \rightarrow \infty$ , for the maximum shortcoming tends to  $1 - \alpha$  as  $r \rightarrow \infty$  both for  $\varphi^*$  and  $\varphi_0$ .

$\varphi_0$  has the important advantage that good power properties for large regions inside  $K_1$  are warranted: the expression in the right-hand side of (30) tends to  $(1 - \alpha)(1 - 2^{\frac{1}{2}}\pi^{-\frac{1}{2}}) \approx .19$  as  $r \rightarrow \infty$  ( $\alpha = .05$ ).

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