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Optimal Two-Sided Test for the Location Parameter  
of the Uniform Distribution Based on Lagrange's Method

by

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Abstract.

In this paper we deal with the uniform distribution with the density

$$f(x|\theta) = \begin{cases} (\delta_2 - \delta_1)^{-1}, & \text{for } \delta_1 \leq x - \theta < \delta_2 \\ 0, & \text{otherwise} \end{cases}$$

where  $-\infty < \theta < \infty$  and  $\delta_i (i=1, 2)$  are real values such that  $\delta_1 < \delta_2$ . Based on a random sample  $X_1, \dots, X_n$  from  $f(x|\theta)$  we construct the two-sided test for testing the hypotheses  $H_0: \theta = \theta_0$  versus  $H_1: \theta \neq \theta_0$  for a constant  $\theta_0$ , inverting the shortest interval estimate for  $\theta_0$ . We show that the power function of this test is minimized at  $\theta_0$  and exhibit its exact form. We also prove that this test has the greatest power among the tests symmetric about  $\theta_0$ .

## §1. Introduction.

In this paper we deal with the uniform distribution over the interval  $[\theta + \delta_1, \theta + \delta_2)$  with the density

$$(1.1) \quad f(x|\theta) = \begin{cases} (\delta_2 - \delta_1)^{-1}, & \text{for } \delta_1 \leq x - \theta < \delta_2 \\ 0, & \text{otherwise} \end{cases}$$

where  $-\infty < \theta < \infty$  and  $\delta_i (i=1,2)$  are real values such that  $\delta_1 < \delta_2$ . Based on a random sample  $X_1, \dots, X_n$  from  $f(x|\theta)$  we test the hypothesis  $H_0: \theta = \theta_0$  versus the alternative hypothesis  $H_1: \theta \neq \theta_0$  for a constant  $\theta_0$ . Let  $X_{(i)}$  be the  $i$ -th smallest observation of  $X_1, \dots, X_n$ . We first estimate  $\theta$  by an unbiased estimate

$$(1.2) \quad Y = (X_{(1)} + X_{(n)} - \delta_0) / 2$$

with  $\delta_0 = \delta_1 + \delta_2$ .

The author needs to remark on the estimator  $Y$ . Although the author had been noticed in Y. Nogami(1992) (see p.1) that the unbiased estimator  $Y$  could be a good estimator, miscalculation in deriving the probability density function (p.d.f.) of  $Y$  had prevented her from reaching  $Y$ . Later, in Y. Nogami(1995), the author corrected this and used  $Y$  to obtain the two-sided test based on the shortest interval estimate. This paper is the generalization of Sections 1 and 2 of Y. Nogami(1995) to the density (1.1) and introduce in Section 4 extra property which is the generalization of Y. Nogami(1997).

Let  $0 < \alpha < 1$ . We call  $(U_1, U_2)$  a  $(1-\alpha)$  interval estimate for the parameter  $\nu$  if  $P_\nu [U_1 < \nu < U_2] = 1 - \alpha$ . We use Lagrange's method to get the shortest  $(1-\alpha)$  interval estimate for the parameter  $\theta$  based on  $Y$ . Inverting this interval for  $\theta_0$  we obtain the acceptance region of the two-sided test in Section 2.

As the bibliography the problem of testing hypotheses  $H_0: \theta = 0$  versus  $H_1: \theta \neq 0$  is treated by A. Birnbaum(1954). He treats the case of  $\delta_2 = -\delta_1 = 2^{-1}$  and states in his paper that there exists no uniformly most powerful unbiased test of  $H_0: \theta = 0$ . However, he had not mentioned the estimator  $Y$ . Also, the alternative hypothesis  $H_1: 0 < |\theta| < 2^{-1}(1-\alpha^{1/n})$  in his Example 2 is too close to the hypothesis  $H_0: \theta = 0$  as  $n \rightarrow \infty$  because  $\alpha^{1/n}$  becomes close to 1 when  $n$  is large. So, the author still believes goodness of our test introduced in Section 2. In fact, in Section 3 we see that the power function of this test is minimized at  $\theta_0$  with the minimum value  $\alpha$  and exhibit its exact form. (see (3.1).) Furthermore, in Section 4 we state the theorem that this test has the greatest power among size- $\alpha$  tests symmetric about  $\theta_0$  and prove it by usage of generalized Neyman-Pearson Lemma.

Hereafter, we let  $c = \delta_2 - \delta_1 (> 0)$ . Let  $\pm$  be the defining property.

§2. The two-sided test for  $\theta$ .

Let  $X_1, \dots, X_n$  be a random sample of size  $n$  taken from the density (1.1). We find the shortest  $(1-\alpha)$  interval estimate for  $\theta$  based on  $Y$  defined by (1.2) using Lagrange's method and invert this interval estimate for  $\theta_0$  to get the two-sided test for testing hypotheses  $H_0: \theta = \theta_0$  versus  $H_1: \theta \neq \theta_0$ .

We first find the probability density function (p.d.f.) of  $Y$ . Applying the variable transformation  $Y = (X_{(1)} + X_{(n)} - \theta_0)/2$  and  $Z = X_{(1)}$  to the joint p.d.f. of  $(X_{(1)}, X_{(n)})$  and taking the marginal p.d.f. we obtain the p.d.f. of  $Y$  as follows:

$$(2.1) \quad g_Y(y|\theta) = \begin{cases} nc^{-n}(c-2|y-\theta|)^{n-1}, & \text{for } -c/2 < y-\theta < c/2 \\ 0, & \text{otherwise.} \end{cases}$$

To get the shortest  $(1-\alpha)$  interval estimate for  $\theta$  we shall find real numbers  $r_1$  and  $r_2$  ( $r_1 < r_2$ ) which minimize  $r_2 - r_1$  subject to

$$(2.2) \quad P_\theta[r_1 < Y - \theta < r_2] = \int_{\theta+r_1}^{\theta+r_2} g_Y(y|\theta) dy = 1-\alpha.$$

Letting  $\lambda$  be a real number we define

$$L = L(r_1, r_2; \lambda) = r_2 - r_1 - \lambda \left\{ \int_{\theta+r_1}^{\theta+r_2} g_Y(y|\theta) dy - 1 + \alpha \right\}.$$

By Lagrange's method  $\partial L / \partial r_1 = 0 = \partial L / \partial r_2$ , which leads to

$$(2.3) \quad g_Y(\theta+r_1|\theta) = g_Y(\theta+r_2|\theta) = (\lambda^{-1}), \quad \forall \theta.$$

Since  $\partial L / \partial \lambda = 0$  is equivalent to (2.2), (2.2) and (2.3) lead to  $r_2 = -r_1$  ( $\stackrel{\text{e}}{=} r$ ). Substituting these into (2.2), making a variable change  $u = y - \theta$  and performing further calculations leads to

$$\text{the extreme left hand side of (2.2)} = 2 \int_0^r nc^{-n}(c-2u)^{n-1} du = 1 - (1 - (2r/c))^n.$$

Solving  $\alpha = (1 - (2r/c))^n$  we get

$$(2.4) \quad r = c(1 - \alpha^{1/n})/2.$$

Hence, the shortest  $(1-\alpha)$  interval estimate for  $\theta$  is

$$(2.5) \quad (Y-r, Y+r)$$

where  $r$  is given by (2.4).

Therefore, by inverting the shortest  $(1-\alpha)$  interval estimate (2.5) for  $\theta_0$  we get the acceptance region  $(\theta_0-r, \theta_0+r)$ . Namely, our test is to reject  $H_0$  if  $Y \in (-\infty, \theta_0-r] \cup [\theta_0+r, \infty)$  and accept  $H_0$  if  $Y \in (\theta_0-r, \theta_0+r)$ . Let  $\phi(y)$  be a randomized test which chooses between two decisions, rejection or acceptance of  $H_0$  with probabilities  $\phi(y)$  and  $1-\phi(y)$ , respectively. Let  $\phi^*(y)$  be a randomized test defined by

$$(2.6) \quad \phi^*(y) = \begin{cases} 1, & \text{if } y \leq \theta_0 - r \text{ or } \theta_0 + r \leq y \\ 0, & \text{if } \theta_0 - r < y < \theta_0 + r. \end{cases}$$

In the next section we show that the power function of our test is minimized at  $\theta_0$  and exhibit its exact form.

### §3. The power function.

Let  $y_1 = \theta_0 - r$  and  $y_2 = \theta_0 + r$  where  $r$  is given by (2.4). We define the power function  $\pi(\theta)$  of our test as follows:

$$\pi(\theta) = E_{\theta}(\phi^*(Y)) = 1 - \int_{y_1}^{y_2} g_Y(y|\theta) dy$$

where  $g_Y(y|\theta)$  is defined by (2.1). From (2.2) with  $x_2 = -r_1 (=r)$  we have that  $\pi(\theta_0) = \alpha$ .

From (2.3) and  $d\pi(\theta)/d\theta = g_Y(y_2|\theta) - g_Y(y_1|\theta)$  we obtain that  $[d\pi(\theta)/d\theta]_{\theta=\theta_0} = 0$ . Furthermore, from  $d\pi(\theta)/d\theta = g_Y(y_2|\theta) - g_Y(y_1|\theta)$  and the definitions of  $g_Y(y_i|\theta)$  ( $i=1, 2$ ) we can easily check that  $d\pi(\theta)/d\theta < 0$  for  $\theta < \theta_0$  and  $d\pi(\theta)/d\theta > 0$  for  $\theta_0 < \theta$ . Since  $\pi(+\infty) = \pi(-\infty) = 1$  and  $\pi(\theta_0) = \alpha$ ,  $\pi(\theta)$  is minimized at  $\theta = \theta_0$  with the minimum value  $\pi(\theta_0) = \alpha$ .

We can find the power function directly as follows:

$$(3.1) \quad \pi(\theta) = \begin{cases} 1, & \text{for } \theta \leq \theta_0 - r - c/2 \\ 1 - \{1 - 2c^{-1}(\theta_0 - r - \theta)\}^n / 2, & \text{for } \theta_0 - r - c/2 \leq \theta < \theta_0 + r - c/2 \\ 1 - \{1 - 2c^{-1}(\theta_0 - r - \theta)\}^n / 2 + \{1 - 2c^{-1}(\theta_0 + r - \theta)\}^n / 2, & \text{for } \theta_0 + r - c/2 \leq \theta < \theta_0 - r \\ \{1 - 2c^{-1}(\theta - \theta_0 + r)\}^n / 2 + \{1 - 2c^{-1}(\theta_0 + r - \theta)\}^n / 2, & \text{for } \theta_0 - r \leq \theta < \theta_0 + r \\ 1 - \{1 - 2c^{-1}(\theta - \theta_0 - r)\}^n / 2 + \{1 - 2c^{-1}(\theta - \theta_0 + r)\}^n / 2, & \text{for } \theta_0 + r \leq \theta < \theta_0 - r + c/2 \\ 1 - \{1 - 2c^{-1}(\theta - \theta_0 - r)\}^n / 2, & \text{for } \theta_0 - r + c/2 \leq \theta < \theta_0 + r + c/2 \\ 1, & \text{for } \theta_0 + r + c/2 \leq \theta. \end{cases}$$

Table. The Values of  $\alpha^{1/n}$ .

$\alpha$	.10	.05	.01	.001
n=4	.56	.47	.32	.18
5	.63	.55	.40	.25
6	.69	.61	.46	.32
7	.72	.65	.52	.37
10	.79	.74	.63	.50
50	.95	.94	.91	.87
100	.98	.97	.95	.93

Here, we remark that for ( $\alpha \geq 0.001; n \geq 10$ ), ( $\alpha \geq 0.01; n \geq 7$ ), ( $\alpha \geq 0.05; n \geq 5$ ) and ( $\alpha \geq 0.10; n \geq 4$ ), we have that  $\alpha^{1/n} \geq 1/2$ . (See Table.) So, calculations led to (3.1) depend on  $\alpha$  and  $n$  such that  $\alpha^{1/n} \geq 1/2$ .

In the last section we show that our test has the greatest power among size- $\alpha$  tests symmetric about  $\theta_0$ . To do so we use the generalized Neyman-Pearson lemma.

#### §4. Best symmetric two-sided test.

In this section we shall prove the following theorem:

Theorem. Let  $X_1, \dots, X_n$  be a random sample from the p.d.f.  $f(x|\theta)$  given by (1.1). The test  $\phi^*$  given by (2.6) for testing the hypothesis  $H_0: \theta = \theta_0$  versus the alternative hypothesis  $H_1: \theta \neq \theta_0$  has the greatest power among size- $\alpha$  tests symmetric about  $\theta_0$ .

Proof.) We first introduce a size- $\alpha$  test  $\phi$  symmetric about  $\theta_0$ . Using the generalized Neyman-Pearson Lemma we obtain the test  $\phi$  given by (4.8) which has the greatest power. Hence, the test  $\phi^*$  given by (2.6) will be the best symmetric two-sided test of size- $\alpha$ .

Let  $X_{(i)}$  be the  $i$ -th smallest observation such that  $X_{(1)} \leq X_{(2)} \leq \dots \leq X_{(n)}$ . Let  $Y$  be as defined by (1.2). The p.d.f.  $g_Y(y|\theta)$  of  $Y$  is given by (2.1). Let  $\phi$  be a size- $\alpha$  test symmetric about  $\theta_0$  (namely,  $\phi(y) = \phi(2\theta_0 - y)$ ,  $\forall y$ ). Then, it follows that

$$(4.1) \quad E_{\theta_0}(\phi(Y)) = \alpha$$

and

$$(4.2) \quad E_{\theta_0}(Y\phi(Y)) = \theta_0 E_{\theta_0}(\phi(Y)) = \theta_0 \alpha.$$

Above (4.2) holds because  $E_{\theta_0}((Y - \theta_0)\phi(Y)) = 0$  and (4.1) holds.

Hence, by generalized Neyman-Pearson Lemma  $\phi$  defined by (4.8) maximizes the

integral

$$\int \phi(y) g_Y(y|\theta') dy \text{ for } \theta' \neq \theta_0$$

out of all functions  $\phi$ ,  $0 \leq \phi \leq 1$ , satisfying (4.1) and (4.2), when there exist real constants  $k_1$  and  $k_2$  such that in almost everywhere sense of either of two probability measures

$$(4.3) \quad \phi(y) = \begin{cases} 1, & \text{if } (c-2|y-\theta'|)^{n-1} I_{(-c/2, c/2)}(y-\theta') \\ & \geq (k_2+k_1 y)(c-2|y-\theta_0|)^{n-1} I_{(-c/2, c/2)}(y-\theta_0) \\ 0, & \text{if } (c-2|y-\theta'|)^{n-1} I_{(-c/2, c/2)}(y-\theta') \\ & < (k_2+k_1 y)(c-2|y-\theta_0|)^{n-1} I_{(-c/2, c/2)}(y-\theta_0). \end{cases}$$

$$(4.4) \quad < (k_2+k_1 y)(c-2|y-\theta_0|)^{n-1} I_{(-c/2, c/2)}(y-\theta_0).$$

We check existence of such  $k_1$  and  $k_2$  and show that the test  $\phi$  is of form (4.8).

We first check existence of such  $k_1$  and  $k_2$  until the fifth line below (4.7).

When  $\theta' \leq \theta_0 - c/2$  or  $\theta_0 + c/2 \leq \theta'$ , we take  $k_1=0$  and  $k_2=1$ . When  $\theta' \leq \theta_0 - c$  or  $\theta_0 + c \leq \theta'$ , the inequality (4.3) or (4.4) trivially holds. When  $\theta_0 - c < \theta' \leq \theta_0 - c/2$ , the inequality (4.4) holds for  $(\theta_0 + \theta')/2 < y < \theta_0 + c/2$  and the inequality (4.3) holds for  $\theta' - c/2 < y \leq (\theta_0 + \theta')/2$ . When  $\theta_0 + c/2 \leq \theta' < \theta_0 + c$ , the inequality (4.4) holds for  $\theta_0 - c/2 < y \leq (\theta_0 + \theta')/2$  and the inequality (4.3) holds for  $(\theta_0 + \theta')/2 < y < \theta' + c/2$ .

We now consider the case of  $\theta_0 - c/2 < \theta' < \theta_0 + c/2$ . Let  $\theta_0 < \theta' < \theta_0 + c/2$ . Then, for  $\theta_0 - c/2 < y < \theta' - c/2$ , the inequality (4.4) holds when  $k_1=0$  and  $k_2=1$ . On the other hand, for  $\theta_0 + c/2 < y < \theta' + c/2$ , the inequality (4.3) always holds for any  $k_1$  and  $k_2$ . Let  $\theta_0 - c/2 < \theta' < \theta_0$ . Then, for  $\theta' - c/2 < y < \theta_0 - c/2$ , the inequality (4.3) holds for any real  $k_1$  and  $k_2$ . On the other hand, for  $\theta' + c/2 < y < \theta_0 + c/2$ , the inequality (4.4) holds when  $k_1=0$  and  $k_2=1$ . Henceforth, it is enough to consider the  $y$ 's in  $(\theta' - c/2, \theta_0 + c/2)$  for  $\theta_0 < \theta' < \theta_0 + c/2$  or in  $(\theta_0 - c/2, \theta' + c/2)$  for  $\theta_0 - c/2 < \theta' < \theta_0$ . We let

$$h(y) = (c-2|y-\theta'|) / (c-2|y-\theta_0|) \quad (\geq 0)$$

and

$$z(y) = (k_2 + k_1 y)^{1/(n-1)} \quad (\geq 0).$$

We also let

$$(4.5) \quad y_0 = -k_2/k_1.$$

For  $\theta_0 < \theta' < \theta_0 + c/2$ , take  $y_0$  such that  $\theta' - c/2 < y_0 < \theta_0$ . For  $\theta_0 - c/2 < \theta' < \theta_0$ , take  $y_0$  such that  $\theta_0 < y_0 < \theta' + c/2$ . Let  $p$  be a given number such that  $0 < p < 1$ . Take  $y_0 = (1-p)\theta_0 + p(\theta' - c/2)$  for  $\theta_0 < \theta' < \theta_0 + c/2$  and take  $y_0 = (1-p)\theta_0 + p(\theta' + c/2)$  for  $\theta_0 - c/2 < \theta' < \theta_0$ . Then, from (4.5),  $k_2$  is taken as follows:

$$(4.6) \quad k_2 = \begin{cases} -k_1 \{(1-p)\theta_0 + p(\theta' - c/2)\}, & \text{for } \theta_0 < \theta' < \theta_0 + c/2 \\ -k_1 \{(1-p)\theta_0 + p(\theta' + c/2)\}, & \text{for } \theta_0 - c/2 < \theta' < \theta_0. \end{cases}$$

Substituting these values into  $z(y)$  we have that

$$(4.7) \quad z(y) = \begin{cases} [k_1 \{y - ((1-p)\theta_0 + p(\theta' - 2^{-1}c))\}]^{1/(n-1)}, & \text{for } \theta_0 < \theta' < \theta_0 + 2^{-1}c, \\ [k_1 \{y - ((1-p)\theta_0 + p(\theta' + 2^{-1}c))\}]^{1/(n-1)}, & \text{for } \theta_0 - 2^{-1}c < \theta' < \theta_0 \end{cases}$$

which are drawn by stripe lines in the figure below. Since we must accept  $H_0$  for  $y = \theta_0$ , we must have  $h(\theta_0) < z(\theta_0)$ . We take  $k_1 = 2(pc)^{-1}$  for  $\theta_0 < \theta' < \theta_0 + 2^{-1}c$  and  $k_1 = -2(pc)^{-1}$  for  $\theta_0 - 2^{-1}c < \theta' < \theta_0$ . Then,  $(z(\theta_0))^{n-1} = h(\theta_0)$ . Hence, we have  $h(\theta_0) < z(\theta_0)$  because  $0 < 1 - 2|\theta_0 - \theta'|/c < 1$ .  $k_2$  is obtained by substituting these values of  $k_1$  into (4.6).

To show that the test  $\psi$  is of form (4.8) we check the existence of two intersection points of  $h(y)$  and  $z(y)$ . Since for  $\theta_0 < \theta' < \theta_0 + c/2$

$$h(y) = \begin{cases} 1 - \{2(\theta' - \theta_0)/(2y + c - 2\theta_0)\}, & \text{for } \theta' - c/2 < y \leq \theta_0, \\ -1 - \{2(c - (\theta' - \theta_0))/(2y - c - 2\theta_0)\}, & \text{for } \theta_0 < y \leq \theta', \\ 1 - \{2(\theta' - \theta_0)/(2y - c - 2\theta_0)\}, & \text{for } \theta' < y < \theta_0 + c/2, \end{cases}$$

$h(y)$  is an increasing function for  $\theta' - c/2 < y < \theta_0 + c/2$ . Since for  $\theta_0 - c/2 < \theta' < \theta_0$

$$h(y) = \begin{cases} 1 + \{2(\theta_0 - \theta')/(2y + c - 2\theta_0)\}, & \text{for } \theta_0 - c/2 < y \leq \theta', \\ -1 + \{2c - 2(\theta_0 - \theta')/(2y + c - 2\theta_0)\}, & \text{for } \theta' < y \leq \theta_0, \\ 1 + \{2(\theta_0 - \theta')/(2y - c - 2\theta_0)\}, & \text{for } \theta_0 < y < \theta' + c/2, \end{cases}$$

$h(y)$  is a decreasing function for  $\theta_0 - c/2 < y < \theta' + c/2$ . On the other hand, when  $\theta_0 < \theta' < \theta_0 + c/2$   $dz(y)/dy > 0$  for all  $y > y_0$  and when  $\theta_0 - c/2 < \theta' < \theta_0$   $dz(y)/dy < 0$  for all  $y < y_0$ . Since  $0 = z(y_0) < h(y_0) (< 1)$  and since for  $\theta_0 < \theta' < \theta_0 + c/2$   $z((\theta_0 + c/2)^-) < \lim_{y \rightarrow (\theta_0 + c/2)^-} h(y) = +\infty$  and for  $\theta_0 - c/2 < \theta' < \theta_0$   $z((\theta_0 - c/2)^+) < \lim_{y \rightarrow (\theta_0 - c/2)^+} h(y) = +\infty$ ,



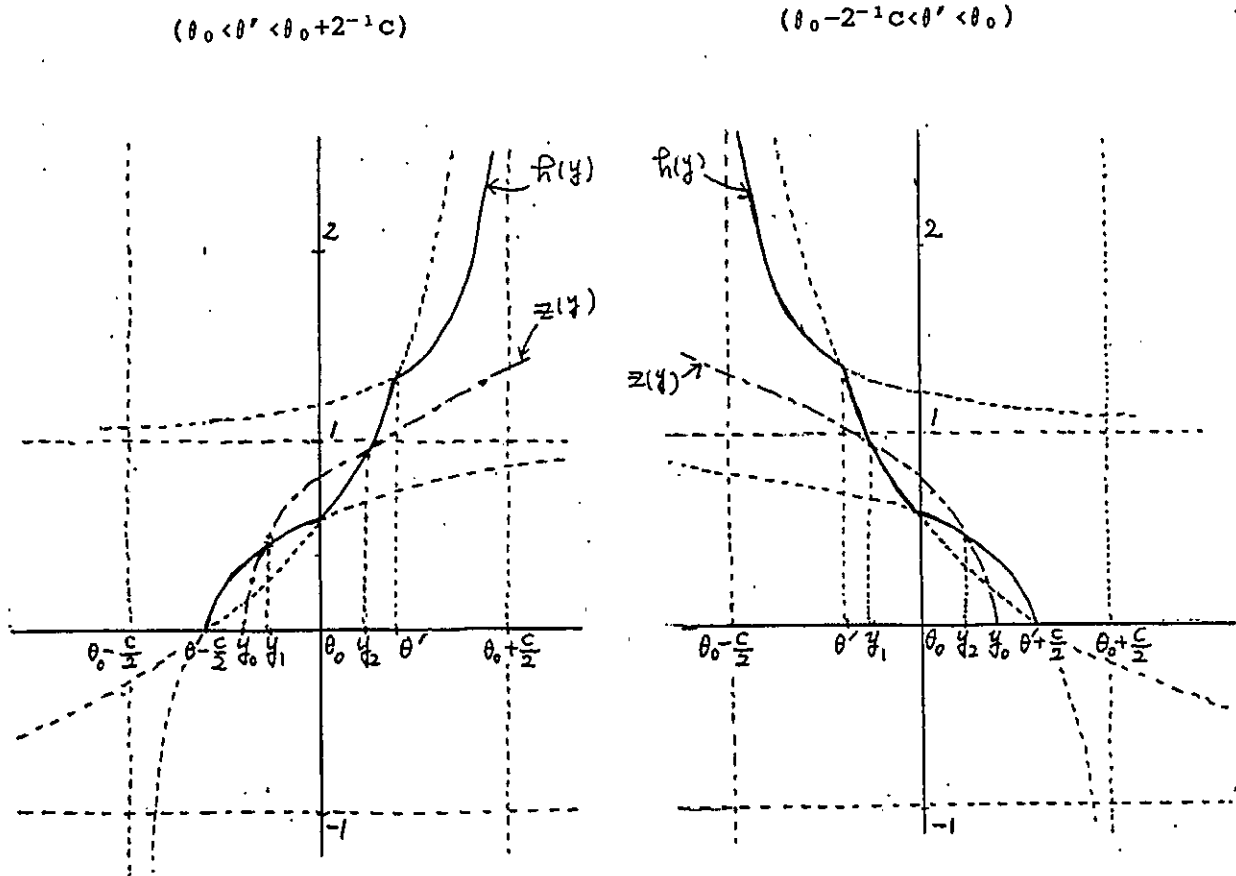
in view of the fact that  $h(\theta_0) < z(\theta_0)$  there must exist two intersection points of  $h(y)$  and  $z(y)$  for  $\theta_0 < \theta' < \theta_0 + c/2$  and for  $\theta_0 - c/2 < \theta' < \theta_0$ , respectively. (See Figure.)

Let  $y_1$  and  $y_2$  be such  $y$ -coordinates of these intersection points with  $y_1 < y_2$ . Then, we finally have the optimal test of form

$$(4.8) \quad \phi(y) = \begin{cases} 1, & \text{if } y \leq y_1 \text{ or } y \geq y_2 \\ 0, & \text{if } y_1 < y < y_2. \end{cases}$$

Hence, the test provided by (2.6) is size- $\alpha$  test with the greatest power among size- $\alpha$  tests symmetric about  $\theta_0$ .

Figure. The graphs of  $h(y)$  and  $z(y)$



(q. e. d.)

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