

No. 713

A New View on Statistical Inference.

(Part I)

--- Case of the Uniform Distribution $U[\theta, \theta+1)$ ---

by

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February 1997

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Abstract.

In this paper we again use the uniform distribution with density $f(x|\theta)=1$ for $\theta \leq x < \theta+1$; $=0$, otherwise for real θ . The purpose of this paper is to consider the problem of testing the null hypothesis $H_0: \theta = \theta_0$ versus the alternative hypothesis $H_1: \theta \neq \theta_0$, and construct the two-sided unbiased test ϕ^* of size α which is uniformly most powerful (UMP). From direct computation we also see that the power function of ϕ^* is convex (from below).

§ 1. Introduction.

Let $I_A(x)$ be an indicator function such that $I_A(x)=1$ for $x \in A$; $=0$ for $x \notin A$. Throughout this paper the underlined distribution is the uniform with density $f(x|\theta)=I_{(\theta, \theta+1)}(x)$ for real θ . We also let $X_{(i)}$ ($i=1,2, \dots, n$) denote the i -th smallest observation of X_1, \dots, X_n , taken randomly from the population with density $f(x|\theta)$. If we need to test the null hypothesis $H_0: \theta=\theta_0$ against an alternative $H_1: \theta \neq \theta_0$, then the likelihood-ratio tests based on the minimal sufficient statistics are useless.

So, in this paper we use an unbiased estimator $Y \triangleq (X_{(1)} + X_{(n)} - 1)/2$ to construct the two-sided size- α test ϕ^* which is unbiased and see this ϕ^* is uniformly most powerful (UMP) (in Section 2). In Section 3 the power function based on ϕ^* is computed, exactly. Eventually, we can see that for finite n , this power function is convex (from below).

For notational conveniences we denote the defining property by \triangleq and also let $h'(x) \triangleq dh(x)/dx$.

§ 2. Optimal Two-Sided Tests.

Let α be a real number such that $0 < \alpha < 1$. In this section we consider the problem of testing $H_0: \theta=\theta_0$ versus $H_1: \theta \neq \theta_0$ for real θ_0 and construct the optimal two-sided size- α test ϕ^* . Let $Y \triangleq (X_{(1)} + X_{(n)} - 1)/2$.

Let ϕ be a two-sided test of the following form:

$$(1) \quad \phi(Y) = \begin{cases} 1, & \text{if } Y < Y_1 \text{ or } Y > Y_2, \\ \gamma_1, & \text{if } Y = Y_1, \quad i=1,2, \\ 0, & \text{if } Y_1 < Y < Y_2, \end{cases}$$

where Y_1, Y_2, γ_1 and γ_2 are chosen so that $E_{\theta_0}(\phi(Y)) = \alpha$.

We let C be the critical region given by

$$C = \{y: y \leq y_1 \text{ or } y \geq y_2\}.$$

To find optimal y_1 and y_2 we find the p.d.f. of Y .

Applying a variable-transformation to the joint density of $(X_{(1)}, X_{(n)})$ and taking marginal probability density function (p.d.f.) we obtain the p. d. f. of Y as follows:

$$(2) \quad g(y|\theta) = \begin{cases} n(1-2|y-\theta|)^{n-1} & \text{for } -1/2 < y-\theta < 1/2, \\ 0, & \text{elsewhere.} \end{cases}$$

Let $\pi_C(\theta) \doteq E_\theta(\phi(Y))$. To get size- α unbiased test, we determine y_1, y_2 which satisfy $\pi_C(\theta_0) = \alpha$ and minimise $\pi_C(\theta)$ at $\theta = \theta_0$. Then, we get optimal region

$$(3) \quad C^* = \{y: y \leq \theta_0 - r, \quad y \geq \theta_0 + r\}$$

where

$$(4) \quad r = (1 - \alpha^{1/n})/2.$$

Let ϕ^* be the two-sided test based on above C^* ;

$$(5) \quad \phi^*(y) = \begin{cases} 1, & \text{if } y \leq \theta_0 - r, \quad y \geq \theta_0 + r \\ 0, & \text{if } \theta_0 - r < y < \theta_0 + r. \end{cases}$$

From the generalized Fundamental Lemma, we can show that UMP unbiased size- α test of testing $H_0: \theta = \theta_0$ versus $H_1: \theta \neq \theta_0$ is of form (1).

Hence, ϕ^* is certainly UMP-unbiased.

§ 3. The convex power function.

In this section we shall exhibit the power function $\pi_C^*(\theta)$ and show that it is convex.

$$\begin{aligned}
 (5) \quad \pi_C^*(\theta) = & \begin{cases} 1, & (\theta < \theta_0 - r - 1/2), \\ \\ 1 - \{1 - 2(\theta_0 - r - \theta)\}^n / 2, & (\theta_0 - r - 1/2 \leq \theta < \theta_0 + r - 1/2) \\ \\ 1 - [\{1 - 2(\theta_0 - r - \theta)\}^n - \{1 - 2(\theta_0 + r - \theta)\}^n] / 2, & (\theta_0 + r - 1/2 \leq \theta < \theta_0 - r) \\ \\ [\{1 - 2(\theta - \theta_0 + r)\}^n + \{1 - 2(\theta_0 + r - \theta)\}^n] / 2, & (\theta_0 - r \leq \theta < \theta_0 + r), \\ \\ 1 - [\{1 + 2(\theta_0 + r - \theta)\}^n - \{1 - 2(\theta - \theta_0 + r)\}^n] / 2, & (\theta_0 + r \leq \theta < \theta_0 - r + 1/2), \\ \\ 1 - \{1 + 2(\theta_0 + r - \theta)\}^n / 2, & (\theta_0 - r + 1/2 \leq \theta < \theta_0 + r + 1/2), \\ \\ 1, & (\theta_0 + r + 1/2 \leq \theta). \end{cases}
 \end{aligned}$$

(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, below.) So, calculations led to (5) depend on α and n

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

α	.10	.05	.01	.001
$n=4$.56	.47	.32	.18
5	.63	.55	.40	.25
6	.68	.61	.46	.32
7	.72	.65	.52	.37
10	.79	.74	.63	.50
50	.95	.94	.91	.87
100	.98	.97	.96	.93

such that $\alpha^{1/n} \geq 2^{-1}$.)

Since, $\pi_C^*(\theta) < 0$ for $\theta < \theta_0$,
 $\pi_C^*(\theta) > 0$ for $\theta > \theta_0$ and $\pi_C^*(\theta_0) = 0$,
 and since $\pi_C^*(\theta_0) = \alpha$ and $\pi_C^*(+\infty) =$
 $\pi_C^*(-\infty) = 1$, $\pi_C^*(\theta)$ is a convex (from

below) function of θ and the test based on C^* is an unbiased size- α test.

We also see that for all θ with $\theta \neq \theta_0$, $\pi_C^*(\theta) \uparrow 1$ as $n \uparrow \infty$.