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ON NEYMAN'S FACTORIZATION THEOREM CONCERNING A SUFFICIENT STATISTICS FOR A PARAMETER

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ABSTRACT

In this paper, it is shown that if X_1, X_2, \dots, X_n denote a random sample of size n from a distribution that has pdf $f(x; \theta)$, $\theta \in \Omega$, then the statistics $Y_1 = u_1(X_1, X_2, \dots, X_n)$ is a sufficient statistics for θ if

$$f(x_1; \theta) f(x_2; \theta) \dots f(x_n; \theta) = \Psi(y_1; \theta)$$

where $\Psi(y_1; \theta)$ is a function of y_1 and θ alone.

KEYWORDS: Best statistics, unbiased statistic, sufficient statistic and Fisher-Neyman Criterion.

1. INTRODUCTION AND STATEMENT OF RESULTS

A famous result known as Factorization Theorem of Neyman [1, 2, 3] concerning a sufficient statistic for a parameter θ can be stated as follows:

THEOREM A. Let X_1, X_2, \dots, X_n denote a random sample of size n from a distribution that has pdf $f(x; \theta)$, $\theta \in \Omega$. The statistic $Y_1 = u_1(X_1, X_2, \dots, X_n)$ is a sufficient statistics for parameter θ if and only if we can find two nonnegative functions, Φ_1 and Φ_2 , such that

$$f(x_1; \theta) f(x_2; \theta) \dots f(x_n; \theta) = \Phi_1(u_1(X_1, X_2, \dots, X_n); \theta) \Phi_2(X_1, X_2, \dots, X_n),$$

where for every fixed value of $y_1 = u_1(x_1, x_2, \dots, x_n)$, $\Phi_2(X_1, X_2, \dots, X_n)$ does not depend upon θ .

In order to verify that a certain statistic $Y_1 = u_1(X_1, X_2, \dots, X_n)$ is a sufficient statistic for a parameter θ with the help of factorization theorem of Neyman, we have to express the joint pdf of X_1, X_2, \dots, X_n as product of two nonnegative functions Φ_1 and Φ_2 where $\Phi_2(X_1, X_2, \dots, X_n)$ does not depend upon θ for every fixed value of $y_1 = u_1(x_1, x_2, \dots, x_n)$. Here we establish an elegant criteria pertaining to a sufficient statistic of a parameter θ . More precisely, we prove:

THEOREM 1. Let X_1, X_2, \dots, X_n denote a random sample of size n and let $Y_1 = u_1(X_1, X_2, \dots, X_n)$. If

$$f(x_1; \theta) f(x_2; \theta) \dots f(x_n; \theta) = \Psi(y_1; \theta)$$

where $\Psi(y_1; \theta)$ is a function of y_1 and θ alone, then $Y_1 = u_1(X_1, X_2, \dots, X_n)$ is a sufficient statistic for θ .

We illustrate Theorem 1 with the help of the following examples:

EXAMPLE 1. Let X_1, X_2, \dots, X_n denote a random sample of size n from a distribution having the pdf

$$f(x; \theta) = \theta \exp(-\theta x), 0 < x < \infty, \theta > 0;$$

$$= 0 \text{ elsewhere.}$$

Consider the statistic $Y_1 = X_1 + X_2 + \dots + X_n$. Since the joint pdf of X_1, X_2, \dots, X_n is

$$f(x_1; \theta) f(x_2; \theta) \dots f(x_n; \theta) = \theta^{-\theta x_1} \theta^{-\theta x_2} \dots \theta^{-\theta x_n}$$

$$= \theta^n e^{-(x_1 + x_2 + \dots + x_n)}$$

$$= \theta^n e^{-\theta y_1}$$

$$= \Psi(y_1; \theta) \quad (\text{say})$$

where $\Psi(y_1; \theta)$ is a function of y_1 and θ only. Hence by Theorem 1, $y_1 = \sum_{i=1}^n x_i$ is a sufficient statistic for θ .

EXAMPLE 2. Let X_1, X_2, \dots, X_n denote a random sample of size n from a distribution having the pdf

$$f(x; \theta) = \theta x^{\theta-1}, 0 < x < 1, \theta > 0;$$

$$= 0 \text{ elsewhere.}$$

By theorem 1, it can be easily seen that the statistic $y_1 = x_1, x_2, \dots, x_n$ is a sufficient statistics of θ because

$$f(x_1; \theta) f(x_2; \theta) \dots f(x_n; \theta) = \Psi(y_1; \theta)$$

$$= \theta^n y_1^{\theta-1}.$$

REMARK 1. The advantage of Theorem 1 is that it tells us at a first glance that a statistic $y_1 = u_1(x_1, x_2, \dots, x_n)$ is a sufficient statistic for a parameter θ if the product

$$f(x_1; \theta) f(x_2; \theta) \dots f(x_n; \theta)$$

is a function of y_1 and θ only.

2. PROOF OF THE THEOREM

PROOF OF THE THEOREM 1. We shall prove the theorem when the random variables are of the continuous type. In case of random variables of the discrete type, we have to replace integrals by sums only. Suppose that

$$(1) \quad f(x_1; \theta) f(x_2; \theta) \dots f(x_n; \theta) = \Psi[u_1(X_1, X_2, \dots, X_n) ; \theta].$$

Consider one-to-one transformation $y_1 = u_1(x_1, x_2, \dots, x_n), y_2 = u_2(x_1, x_2, \dots, x_n), \dots, y_n = u_n(x_1, x_2, \dots, x_n)$. Let the inverse functions of this transformation be $x_1 = w_1(y_1, y_2, \dots, y_n), x_2 = w_2(y_1, y_2, \dots, y_n), \dots, x_n = w_n(y_1, y_2, \dots, y_n)$. If

$$J = \frac{\partial(x_1, x_2, \dots, x_n)}{\partial(y_1, y_2, \dots, y_n)}$$

denotes the Jacobian of Transformation, then the joint pdf of the statistics Y_1, Y_2, \dots, Y_n is

$$g(y_1, y_2, \dots, y_n; \theta) = f(w_1(y_1, y_2, \dots, y_n; \theta)) f(w_2(y_1, y_2, \dots, y_n; \theta)) \dots f(w_n(y_1, y_2, \dots, y_n; \theta))|J|$$

$$(2) \quad = \Psi(y_1; \theta)|J|.$$

The marginal pdf of $Y_1 = u_1(X_1, X_2, \dots, X_n)$ is given by

$$\begin{aligned} g(y_1; \theta) &= \int_{-\infty}^{\infty} \dots \int_{-\infty}^{\infty} g(y_1, y_2, \dots, y_n; \theta) dy_2 dy_3 \dots dy_n \\ &= \int_{-\infty}^{\infty} \dots \int_{-\infty}^{\infty} \Psi(y_1; \theta) |J| dy_2 dy_3 \dots dy_n \end{aligned}$$

$$(3) \quad = \Psi(y_1; \theta) \int_{-\infty}^{\infty} \dots \int_{-\infty}^{\infty} |J| dy_2 dy_3 \dots dy_n .$$

Since the Jacobian J is a function of Y_1, Y_2, \dots, Y_n and θ is neither involved in the Jacobian nor in the limits of the integration, therefore, the (n-1) fold integral

$$\int_{-\infty}^{\infty} \dots \int_{-\infty}^{\infty} |J| dy_2 dy_3 \dots dy_n$$

is a function of y_1 alone say $\Phi(y_1)$. Thus from (3), we obtain

$$g(y_1; \theta) = \Psi(y_1; \theta) | \phi(y_1) .$$

If $\phi(y_1) = 0$, then $g(y_1; \theta) = 0$. If $\phi(y_1) > 0$, we can write

$$\psi(u_1(x_1, x_2, \dots, x_n); \theta) = \frac{g(u_1(x_1, x_2, \dots, x_n); \theta)}{\phi(u_1(x_1, x_2, \dots, x_n))}$$

$$(4) \quad = g(u_1(x_1, x_2, \dots, x_n); \theta) \psi(u_1(x_1, x_2, \dots, x_n))$$

where $\psi(u_1(x_1, x_2, \dots, x_n)) = \frac{1}{\phi(u_1(x_1, x_2, \dots, x_n))}$

does not depend upon $\theta \in \Omega$ for every fixed value of $y_1 = u_1(x_1, x_2, \dots, x_n)$. Using (1) in (4) we get

$$f(x_1; \theta) f(x_2; \theta) \dots f(x_n; \theta) = g(u_1(x_1, x_2, \dots, x_n); \theta) \psi(u_1(x_1, x_2, \dots, x_n))$$

or
$$\frac{f(x_1; \theta) f(x_2; \theta) \dots f(x_n; \theta)}{g(u_1(x_1, x_2, \dots, x_n); \theta)} = \psi(u_1(x_1, x_2, \dots, x_n))$$

where $\psi(u_1(x_1, x_2, \dots, x_n))$ does not depend upon $\theta \in \Omega$ for every fixed value of $y_1 = u_1(x_1, x_2, \dots, x_n)$. Hence by definition of sufficient statistic, it follows that

$$y_1 = u_1(x_1, x_2, \dots, x_n)$$

is a sufficient statistic for θ . This completes the proof of Theorem 1.

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Professor A. Azizul-Auzeem former Dean Academics and Head, P.G. Department Mathematics and Statistics, University of Kashmir, the first PhD of Kashmir University in Mathematics always guided/supported author as his own younger brother/child. He was among the top mathematicians of Kashmir whose research work was recognized at world level. We lost this great personality due to COVID-19 infection recently. I thank sir from the core of my heart for his guidance and pray Almighty Allah to grant him highest place in Jannah along with my dear ones. Sir, you will remain in my heart always



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