

## DISTRIBUTIONAL PROPERTIES OF SQUARES OF LINEAR STATISTICS

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**ABSTRACT:** Relations between distributions of squares of linear forms in independent random variables and distributions of the parent random variables are studied. Some new characterizations of the normal law involving chi-square distribution are obtained.

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**KEY WORDS AND PHRASES:** squares of linear statistics, symmetry, chi-square distribution, normal distribution, characterization of probability measures.

### 1. INTRODUCTION

In a series of papers: Geisser and Roberts (1966), Roberts (1971), Geisser (1973), relations between the normal distribution of random variables (r.v's) and the chi-square distribution of linear forms in them were investigated – see also 4.2.4 in Patel and Read (1982) and Ahsanullah (1989). The main result of these papers states that if some linear statistics in independent r.v's are  $\chi^2(1)$  (chi-square with one degree of freedom) then the parent distribution is normal.

This note is a complement to the papers mentioned above. In Section 2 we show that some of the results for the  $\chi^2(1)$  law may be generalized not only to the gamma distribution – this case is considered in Geisser and Roberts (1966) – but to an arbitrary absolutely continuous distribution on  $(0, +\infty)$  – Theorem 1. A new characterization of symmetry together with an extension of Theorem 5 from Roberts (1971) follows next. In Section 3 we consider the special case of linear forms distributed according to the  $\chi^2(1)$  law. Some new characterizations of the normal law complementary to those from Roberts (1971), Geisser (1973) and Ahsanullah (1989) are obtained.

A new contribution to this theme, involving the concept of sub-independence, is contained in Ahsanullah et al. (1992).

### 2. GENERAL SCHEME AND SYMMETRY

The scheme used for the gamma law in Geisser and Roberts (1966) may also be applied for an arbitrary positive absolutely continuous distribution.

For a random variable  $W$  define its symmetrization  $W_s$  by the formula

$$W_s = U W,$$

where  $U$  is independent of  $W$  and  $P(U=1)=P(U=-1)=1/2$ . The following distributional relations involving  $W$ ,  $W_S$  and  $W^2$  are obvious:

Let  $F_2$  and  $G$  be distribution functions of  $W_S$  and  $W^2$ , respectively. Then

$$F_S(x) = \begin{cases} (1-G(x^2^-))/2 & \text{for } x < 0 \\ (1+G(x^2))/2 & \text{for } x \geq 0, \end{cases} \quad (1)$$

$$G(x) = 2F_S(\sqrt{x}) - 1 \quad \text{for } x \in R. \quad (2)$$

Denote by  $\phi$  and  $\phi_S$  characteristic functions of  $W$  and  $W_S$ , respectively. Obviously  $\phi_S$  is determined by the distribution of  $W^2$  (by (1)). On the other hand

$$\frac{1}{2}[\phi(t) + \phi(-t)] = \phi_S, \quad t \in R. \quad (3)$$

If  $W^2$  is absolutely continuous with the density  $g$  then  $W$  and  $W_S$  are also absolutely continuous and their densities,  $f$  and  $f_S$ , respectively, fulfill the relations

$$\frac{1}{2}[f(x) + f(-x)] = f_S, \quad x \in R. \quad (4)$$

$$f_S(x) = |x|g(x^2), \quad x \in R \text{ (by (1)).} \quad (5)$$

Consequently, we have the following result:

**Theorem 1.** If  $W^2$  is absolutely continuous with density  $g$  then density  $f$  of  $W$  has the form

$$f(x) = h(x)|x|g(x^2), \quad h(x) + h(-x) = 2, \quad x \in R. \quad (6)$$

and its characteristic function  $\phi$  fulfills the equation

$$\phi(t) + \phi(-t) = 2 \int_R e^{itx} |x|g(x^2) dx, \quad t \in R. \quad (7)$$

Conversely, if the characteristic function  $\phi$  of  $W$  fulfils (7) or its density  $f$  fulfils (6), where  $g$  is a non-negative function, then  $g$  is the density of  $W^2$ .

**Proof:** Assume that  $g$  and  $f$  are the densities of  $W^2$  and  $W$ , respectively. Define

$$h(x) = \begin{cases} f(x)/[|x|g(x^2)] & \text{if only } |x|g(x^2) \neq 0 \\ 1 & \text{if only } |x|g(x^2) = 0. \end{cases}$$

Then by (4) we have

$$f_S(x) = \frac{1}{2}|x|g(x^2)[h(x) + h(-x)], \quad x \in R$$

Now (6) follows from (5) and the above formula. Equation (7) is an immediate consequence of (3) and (5).

Assume conversely that (6) holds. Then (4) yields

$$f_1(x) = \frac{1}{2} |x| g(x^2) [h(x) + h(-x)] = |x| g(x^2), \quad x \in \mathbb{R}.$$

Similarly from (7) and (3) we have (5), too. Consequently by (2)

$$\begin{aligned} P(W^2 \leq x) &= G(x) = 2 \int_{-\infty}^{\sqrt{x}} f_1(u) du - 1 = \\ &= 2 \int_{-\infty}^0 (-u) g(u^2) du + 2 \int_0^{\sqrt{x}} u g(u^2) du - 1 = \int_0^x g(u) du, \end{aligned}$$

for any positive  $x$ . Hence  $g$  is the density of  $W^2$ .  $\square$

Immediate consequences of Theorem 1 are Theorem and Corollary 1 from Geisser and Roberts (1966) and Corollaries 1, 2 from Roberts (1971).

Now we work towards an extension of Theorem 5 from Roberts (1971). The assumption of absolute continuity is relaxed.

**Theorem 2.** Let  $X, Y$  be i.i.d. r.v.'s with the characteristic function  $\phi$ . Then

$$[\operatorname{Re} \phi(t)]^2 = [\psi_1(t) + \psi_2(t)] / 2, \quad (8)$$

$$[\operatorname{Im} \phi(t)]^2 = [\psi_1(t) - \psi_2(t)] / 2, \quad t \in \mathbb{R} \quad (9)$$

where  $\psi_1$  and  $\psi_2$  are the characteristic functions of  $(X+Y)_s$  and  $(X-Y)_s$ , respectively.

**Proof:** Observe that  $\phi^2(t)$  and  $\phi(t)\phi(-t)$  are the characteristic functions of  $X+Y$  and  $X-Y$ , respectively. Consequently by (3)

$$\phi^2(t) + \phi^2(-t) = 2\psi_1(t), \quad (10)$$

$$\phi(t)\phi(-t) = \psi_2(t), \quad (11)$$

$t \in \mathbb{R}$ . Hence

$$4[\operatorname{Re} \phi(t)]^2 = [\phi(t) + \phi(-t)]^2 = 2[\psi_1(t) + \psi_2(t)]$$

$$4[\operatorname{Im} \phi(t)]^2 = [\phi(t) - \phi(-t)]^2 = 2[\psi_1(t) - \psi_2(t)]$$

$t \in \mathbb{R}$ . The above equations imply (8) and (9).

As a consequence of Theorem 2 we obtain an extension of Theorem 5 from Roberts (1971).

**Corollary 1.** Let  $X, Y$  be i.i.d. r.v.'s. The r.v.'s  $(X+Y)^2$  and  $(X-Y)^2$  are equidistributed iff  $X$  is symmetric.

**Proof:** Since  $(X+Y)^2 \stackrel{d}{=} (X-Y)^2$  then also  $(X+Y)_{\mathcal{S}} \stackrel{d}{=} (X-Y)_{\mathcal{S}}$ . Consequently in Theorem 2 we have  $\psi_1 = \psi_2$  and by (9) the characteristic function of  $X$  is real. Hence  $X$  has a symmetric distribution.

If  $X$  is symmetric then (10) and (11) imply equidistribution of  $(X+Y)_{\mathcal{S}}$  and  $(X-Y)_{\mathcal{S}}$ . By (2)  $(X+Y)^2$  and  $(X-Y)^2$  are also identically distributed.

**Remark.** Assume that  $(X+Y)^2 \stackrel{d}{=} (X-Y)^2$ . Observe that if the distribution of  $(X+Y)_{\mathcal{S}}$  is uniquely determined by its moments or its characteristic function is non-vanishing then by Theorem 2 the distribution of the  $X$  is uniquely determined by that of  $(X+Y)^2$ . On the other hand the distribution of  $X$  determines that of  $(X+Y)^2$  and  $(X-Y)^2$  even without the symmetry assumption.

In Theorem 5 from Roberts (1971) the special case

$$(X+Y)^2 \stackrel{d}{=} (X-Y)^2 \stackrel{d}{=} X^2(1)$$

was considered. Another straightforward generalization of this result is given in Section 3.

### 3. THE NORMAL AND $x^2$ DISTRIBUTIONS.

In this section we study relations between the  $x^2(1)$  and normal distributions via properties of squares of linear statistics in independent random variables. Only upon introducing  $x^2(1)$  law can we sharpen some of the previously known results.

A useful tool in such investigations is a result of Roberts (1971) (it is also a consequence of our Theorem 1).

**Proposition 1.** The r.v.  $W^2 \stackrel{d}{=} x^2(1)$  iff the characteristic function  $\phi$  of  $W$  satisfies

$$\phi^2(t) + \phi^2(-t) = 2\exp(-t^2/2), \quad t \in \mathbb{R}. \quad \square$$

The following extension of the Theorem from Geisser (1973) and Theorem 2 from Ahsanullah (1989) relaxes the assumption of equidistribution of the r.v.'s involved.

**Theorem 3.** Let  $X$  and  $Y$  be independent r.v.'s. Then  $W_1^2 = (aX+bY)^2 / (a^2+b^2) \stackrel{d}{=} x^2(1) \stackrel{d}{=} W_1^2 = (aX-bY)^2 / (a^2+b^2)$  for some  $a, b \neq 0$  iff at least one of the pair  $X, Y$  is  $N(0,1)$  and  $X^2 \stackrel{d}{=} Y^2 \stackrel{d}{=} x^2(1)$ .

**Proof:** As in Geisser (1973) we have

$$\exp[-(a^2 + b^2)t^2 / 2] = \phi_X(at)\phi_Y(bt), \quad t \in \mathbb{R} \quad (12)$$

Hence by the Cramer theorem  $X_{\mathcal{S}} \stackrel{d}{=} Y_{\mathcal{S}} \stackrel{d}{=} N(0,1)$ . This fact implies  $X^2 \stackrel{d}{=} Y^2 \stackrel{d}{=} x^2(1)$ . Since  $W_1^2 \stackrel{d}{=} x^2(1)$  then by Proposition 1

$$2 \exp[-(a^2 + b^2)t^2 / 2] = \phi_X(at)\phi_Y(bt) + \phi_X(-at)\phi_Y(-bt)$$

$t \in \mathbb{R}$ . Now by (3) applied to  $X_S$  and  $Y_S$  from (12)

$$[\phi_X(at) - \exp(-a^2t^2)][\phi_Y(bt) - \exp(-b^2t^2)] = 0, \quad t \in \mathbb{R}$$

Consequently, there is a sequence  $(t_n), t_n \neq 0$ , such that

$$\phi_X(t_n) = \exp(-t_n^2/2) \quad \text{or} \quad \phi_Y(t_n) = \exp(-t_n^2/2), \quad n \geq 1$$

By Corollary 1 to Lemma 1.2.1 in Kagan et al. (1973) at least one of the pair  $X, Y$  is  $N(0,1)$  r.v.

The sufficiency part: Assume that  $X \stackrel{d}{=} N(0,1)$  and  $Y \stackrel{d}{=} x^2(1)$ . Then by Proposition 1

$$\begin{aligned} \phi_X(at)\phi_Y(bt) + \phi_X(-at)\phi_Y(-bt) &= \\ = \exp(-a^2t^2/2)[\phi_Y(bt) + \phi_Y(-bt)] &= \exp(-a^2t^2/2) \end{aligned}$$

for  $t \in \mathbb{R}$ . Consequently once again applying Proposition 1 we find that  $W_1^2$  is  $x^2(1)$  r.v. Similar computation proves the result for  $W_2^2$ .  $\square$

Corollary 3 from Geisser (1973) is a natural complement of Theorem 3. However the proof of this Corollary seems to be unsatisfactory since the equation  $f(t)g(t)=0, t \in \mathbb{R}$ , does not yield  $f \equiv 0$  or  $g \equiv 0$ , in general. Fortunately the argument based on a sequence  $(t_n), t_n \neq 0$ , from the proof of Theorem 3 works.

Also in Geisser (1973) the following interesting problem was stated: Assume that  $X$  and  $Y$  are i.i.d. r.v.'s and  $(X+Y)^2/2$  has the  $x^2(1)$  distribution. Is the distribution of  $X$  normal  $N(0,1)$ ? Seemingly slight modification: "-" instead of "+" makes the problem very easy (see also Ahsanullah (1989)).

**Proposition 2.** Assume that  $X, Y$  are i.i.d. r.v.'s. The r.v.  $(X-Y)^2/2$  is  $x^2(1)$  iff  $X$  is  $N(0,1)$ .

Observe that it is a straightforward generalization of Theorem 5 from Roberts (1971), where it is assumed that  $(X+Y)^2/2$  is  $x^2(1)$  additionally.

**Proof:** By Proposition 1

$$\phi(t)\phi(-t) = \exp(-t^2), \quad t \in \mathbb{R}$$

and the result follows from the Cramer theorem. The sufficiency is obvious.

The original Geisser conjecture is answered positively in Ramachandran and Lau (1991).

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