Sankhyā: The Indian Journal of Statistics 2001, Volume 63, Series A, Pt. 1, pp. 107–112

MUTUAL CHARACTERIZATIONS OF THE GAMMA AND THE GENERALIZED INVERSE GAUSSIAN LAWS BY CONSTANCY OF REGRESSION

By VANAMAMALAI SESHADRI

McGill University, Montreal, Canada

and

JACEK WESOŁOWSKI

Politechnika Warszawska, Warszawa, Poland

SUMMARY. Let X and Y be non-negative independent random variables. Characterizations of the generalized inverse Gaussian and gamma distributions through the constancy of regression of $X^{-1} - (X + Y)^{-1}$ on X + Y are considered.

1. Introduction

Tweedie (1957), in a fundamental paper devoted to the inverse Gaussian distribution, proved that if X_1, \ldots, X_n are independent identically distributed (iid) inverse Gaussian random variables (rv's) then $\sum_{i=1}^n X_i$ and $\sum_{i=1}^n X_i^{-1} - n^2 (\sum_{i=1}^n X_i)^{-1}$ are independent. For the converses of this one can consult Khatri (1962) and Letac and Seshadri (1985). In Seshadri (1983) the inverse Gaussian law is characterized by the constancy of regression of $\sum_{i=1}^n X_i^{-1} - n^2 (\sum_{i=1}^n X_i)^{-1}$ on $\sum_{i=1}^n X_i$ under appropriate moment conditions. For a review of characterization problems for the inverse Gaussian distribution, see Seshadri (1993).

For the gamma law an important characterization through an independence property is due to Lukacs (1955), based on the independence of Y_1/Y_2 and $Y_1 + Y_2$, where Y_1 , Y_2 are independent random variables. The regression version of this statement was obtained in Wesołowski (1990) (see also Hall and Simons (1969), Wang (1981) and Li, Huang and Huang (1994) for related results).

These are separate results for the inverse Gaussian and the gamma distributions. A characterization of the generalized inverse Gaussian (GIG) that involves the gamma distribution was given by Letac and Seshadri (1983). It goes as follows: if X has the same law as $(X + Y)^{-1}$, where X and Y are independent and Y is gamma, then X follows the GIG law.

Paper received November 1999; revised September 2000.

AMS (1991) subject classification. 60E05, 62E10.

Key words and phrases. Bessel equation, constancy of regression, characterizations of probability distributions, gamma distribution, generalized inverse Gaussian distribution.

Let us denote by $\mu_{-p,a,b}$ the GIG law defined by the density:

 $f(x) = C_1 x^{-p-1} \exp(-(ax + b/x)/2) I_{(0,\infty)}(x);$

and by $\gamma_{p,a/2}$ the gamma law with the density

$$g(y) = C_2 y^{p-1} \exp(-ay/2) I_{(0,\infty)}(y),$$

where C_1 and C_2 are appropriate norming constants and p, a, b are positive numbers.

Recently Matsumoto and Yor (1998) observed that if $X \sim \mu_{-p,a,b}$ and $Y \sim \gamma_{p,a/2}$ are independent then U = 1/(X+Y) and V = 1/X - 1/(X+Y) are also independent with distributions $\mu_{-p,b,a}$ and $\gamma_{p,b/2}$, respectively. (Actually their result was proved for the case a = b.) The converse of this statement, relying only on the independence property and being a simultaneous characterization of the GIG and gamma laws, has been proved in Letac and Wesołowski (2000). A closely related functional equation has been studied recently in Wesołowski (2000).

Observe that if we consider the scaled rv's $\dot{\tilde{X}} = (a/b)^{1/2}X$, $\tilde{Y} = (a/b)^{1/2}Y$, then $\tilde{X} \sim \mu_{-p,\sqrt{ab},(ab)^{1/2}}$, $\tilde{Y} \sim \gamma_{p,(ab)^{1/2}/2}$ and for $\tilde{U} = 1/(\tilde{X} + \tilde{Y})$ and $\tilde{V} = 1/\tilde{X} - \tilde{U}$ it follows immediately that $\tilde{U} \sim \mu_{-p,(ab)^{1/2},(ab)^{1/2}}$ and $\tilde{V} \sim \gamma_{p,(ab)^{1/2}/2}$.

In the present paper we replace the independence of U and V by the constancy of regression of V on U under suitable moment conditions. Also a dual problem for the constancy of regression of 1/V on U will be considered. The price we pay for relaxing independence is that the characterization is not simultaneous because we have to assume one law to obtain the other.

2. Constancy of Regression of V on U

Let X and Y be two independent positive rv's, such that E(1/X), E(X) and E(Y) are finite. For U = 1/(X+Y) and V = 1/X - 1/(X+Y) we consider the following constancy of regression condition, namely

$$E(V|U) = c,$$

where c = E(V). Since V = Y/(X(X+Y)) equivalently we can write this condition as

$$E(Y|X|X+Y) = c(X+Y).$$
 (1)

Observe that (1) is equivalent to

$$E\left((Y/X)e^{s(X+Y)}\right) = cE\left((X+Y)e^{s(X+Y)}\right),\tag{2}$$

for any s, for which both sides are finite. To see this, use conditioning with respect to (X + Y) on the left side of (2). Then (2) follows from (1) directly and (1) follows from (2) by the uniqueness of the Laplace transform since (2) implies

$$E(E(Y|X|X+Y)e^{s(X+Y)}) = E(c(X+Y)e^{s(X+Y)})$$

108

for any $s \leq 0$ and at both sides we have the Laplace transforms of finite measures on $(0, \infty)$, which are equal.

Define now

$$h_X(s) = E\left(X^{-1}e^{sX}\right), \quad L_Y(s) = E\left(e^{sY}\right), \quad s \in \Theta,$$

where $\Theta = \{s : h_X(s) < \infty, L_Y(s) < \infty\}$ and $(-\infty, 0] \subset \Theta$. Then (2) can be rewritten as

$$h_X(s)L'_Y(s) = c \, h''_X(s)L_Y(s) + c \, h'_X(s)L'_Y(s), \quad s \in \Theta.$$
(3)

We start with a characterization of the GIG law assuming that \boldsymbol{Y} has a gamma distribution.

THEOREM 1. Assume E(1/X), E(X) are finite and that $Y \sim \gamma_{p,a/2}$ for some positive numbers p and a. If the regression of V on U is constant, i.e. (1) holds, then $X \sim \mu_{-p,a,b}$, where b = 2p/c and c = E(V).

PROOF. Since $Y \sim \gamma_{p,a/2}$

$$L_Y(s) = \left(\frac{a}{a-2s}\right)^p, \quad L'_Y(s) = \frac{2pa^p}{(a-2s)^{p+1}}, \quad s \in (-\infty, a/2).$$

Hence (3) takes the form

$$(a-2s) h_X''(s) + 2p h_X'(s) - b h_X(s) = 0$$
(4)

with b = 2p/c, and the above equation holds in some set Θ_1 containing $(-\infty, 0]$. Define now a function w by

$$h_X(s) = (a - 2s)^{(p+1)/2} w(s), s \in \Theta_1.$$

Consequently

$$\begin{aligned} h'_X(s) &= -(p+1)(a-2s)^{(p-1)/2}w(s) + (a-2s)^{(p+1)/2}w'(s), \\ h''_X(s) &= (p^2-1)(a-2s)^{(p-3)/2}w(s) - 2(p+1)(a-2s)^{(p-1)/2}w'(s) \\ &+ (a-2s)^{(p+1)/2}w''(s), \quad s \in \Theta_1. \end{aligned}$$

Plugging these into equation (4) we obtain

$$(a-2s)w''(s) - 2w'(s) - [(p+1)^2/(a-2s) + b]w(s) = 0, \quad s \in \Theta_1.$$
 (5)

Now for $z = ((a-2s)b)^{1/2}$ and for the function v defined by v(z) = w(s), we get

$$\frac{dw}{ds} = -\frac{dv}{dz}\frac{b}{z},$$
$$\frac{d^2w}{ds^2} = \frac{d^2v}{dz^2}\left(\frac{b}{z}\right)^2 - \frac{dv}{dz}\frac{b^2}{z^3},$$

z > 0. Inserting the above in (5) we arrive at the modified Bessel equation

$$z^{2}v''(z) + zv'(z) - (z^{2} + (p+1)^{2})v(z) = 0, \quad z > 0.$$

The space of the solutions of the equation is spanned by two modified Bessel functions $K_{(p+1)}$ and $I_{(p+1)}$ (see, for instance, Abramovitz and Stegun (1964)).

Going back to the formula connecting h_X and w we see that, since h_X is a Laplace transform of a bounded measure on $(0, \infty)$, it goes to zero as s goes to $-\infty$; consequently the same is true for w. Now by the definition of v it follows that $\lim_{z\to\infty} v(z) = 0$. But $I_{(p+1)}(z)$ explodes as $z \to \infty$, and hence the only possible solution is the modified Bessel function of the third kind $K_{(p+1)} = K_{-(p+1)}$.

We conclude that $h_X(s) = c_3(a-2s)^{(p+1)/2}K_{-(p+1)}((b(a-2s))^{1/2}), s \in \Theta_1$, where c_3 is a suitable constant. Hence $\Theta_1 = (-\infty, a/2)$ and h_X is the Laplace transform of the function $x^{-1}f(x)$, where f is the density of $\mu_{-p,a,b}$. \Box

Now we consider a parallel situation assuming that the distribution of X is GIG and seek the distribution of Y.

THEOREM 2. Assume that $X \sim \mu_{-p,a,b}$ for some positive numbers p, a and b and that E(Y) is finite. If the regression of V on U is constant, and equal to 2p/b, *i.e.* (1) holds with c = 2p/b, then $Y \sim \gamma_{p,a/2}$.

PROOF. Since $X \sim \mu_{-p,a,b}$

$$h_X(s) = c_4(b(a-2s))^{(p+1)/2} K_{-(p+1)}((b(a-2s))^{1/2}), \quad s < a/2$$

where c_4 is an appropriate constant.

Let $z = (b(a - 2s))^{1/2}$. Then z' = -b/z and $z'' = -b^2/z^3$. Now, to avoid cumbersome notation we will supress the indices of the K function. Then

$$h_X(s) = c_4 z^{p+1} K(z),$$

$$h'_X(s) = -c_4 b[(p+1) z^{p-1} K(z) + z^p K'(z)],$$

$$h''_X(s) = c_4 b^2 z^{p-3} [z^2 K''(z) + (2p+1) z K'(z) + (p^2 - 1) K(z)],$$

(the primes on the left sides denote derivatives taken with respect to s, while those on the right sides are derivatives taken with respect to z). But from the modified Bessel equation we have $z^2K''(z) + zK' - (z^2 + (p+1)^2)K(z) = 0$ and this then gives

$$h_X''(s) = c_4 b^2 z^{p-3} [(z^2 + 2p(p+1))K(z) + 2pzK'(z)], \quad s \in \Theta_1.$$

where Θ_1 is some set containing $(-\infty, 0]$. We plug the above formulas into (3) to obtain

$$L'_{Y}(s) \left(z^{p+1}K(z) + 2p[(p+1)z^{p-1}K(z) + z^{p}K'(z)] \right)$$

= 2pbL_Y(s)z^{p-3}[(z² + 2p(p+1))K(z) + 2pzK'(z)].

After obvious simplifications we arrive at

$$z^2 L'_Y(s) = 2pbL_Y(s) , \quad s \in \Theta_1.$$

110

Now since $z^2 = b(a - 2s)$ we immediately obtain $L_Y(s) = (a/(a - 2s))^p$, $s \in \Theta_1$. Hence $\Theta_1 = (-\infty, a/2)$ and consequently L_Y is the Laplace transform of the $\gamma_{p,a/2}$ distribution.

3. Constancy of Regression of 1/V on U

Let X and Y be two independent positive rv's such that $E(X^2)$, E(1/Y) and E(Y) are finite. With U and V defined in Section 2 we consider the constancy of regression of 1/V on U, i.e. the condition

$$E(V^{-1}|U) = d,$$

where d is a constant equal to E(1/V). An equivalent form reads

$$E(X/Y|X+Y) = d/(X+Y).$$
 (6)

Just as in Section 2, we get

$$L_X''(s)h_Y(s) + L_X'(s)h_Y'(s) = dL_X(s)h_Y'(s),$$
(7)

where

$$L_X(s) = E\left(e^{sX}\right), \quad h_Y(s) = E\left(Y^{-1}e^{sY}\right), \quad s \in \Theta,$$

and $\Theta = \{s : L_X(s) < \infty, h_Y(s) < \infty\}.$

Now assuming that the distribution of Y is gamma we derive a characterization of the GIG law for X.

THEOREM 3. Assume that $E(X^2)$ is finite and $Y \sim \gamma_{p,a/2}$ for some numbers p > 1 and a > 0. If the regression of 1/V on U is constant, i.e. (6) holds, then $X \sim \mu_{-p,a,b}$, where b = 2(p-1)d and d = E(1/V).

Analogous to Theorem 2 we have

THEOREM 4. Assume that $X \sim \mu_{-p,a,b}$ for some positive numbers p, a and b, and that E(1/Y), E(Y) are finite. If the regression of 1/V on U is constant and equal to b/(2(p-1)), i.e. (6) holds with d = b/(2(p-1)), then $Y \sim \gamma_{p,a/2}$.

Since both the theorems have proofs following the lines of the proofs from the previous section, they are skipped.

In each of the cases of constancy of regression considered in this paper we were not able to obtain simulatneous characterization of both the distributions: GIG and gamma. Then an interesting open question arises, if such a characterization holds true if both the conditions E(V|U) = c and $E(V^{-1}|U) = d$ are satisfied. We expect that the answer is in affirmative.

Acknowledgements. This research has been partially funded by an NSERC grant to the first author. The authors thank Gérard Letac for helpful discussions in the preparation of this paper. The first author also thanks the Department of Mathematics, University of Malaya, for providing excellent research facilities while the research was in progress.

References

- ABRAMOVITZ, M., STEGUN, I.A. (1964). Handbook of Mathematical Functions. National Bureau of Standards, Washington.
- HALL, W.J., SIMONS, G. (1969). On characterizations of the gamma distribution, Sankhyā, Series A, 31, 385–390.
- KHATRI, C.G. (1962). A characterization of the inverse Gaussian distribution, Annals of Mathematical Statistics 33, 800–803.
- LETAC, G., SESHADRI, V. (1983). A characterization of the generalized inverse Gaussian distribution by continued fractions, Zeitschrift für Wahrscheinlichkeitstheorie und Verwandte Gebiete 62, 485–489.
- LETAC, G., SESHADRI, V. (1985). On Khatri's characterization of the inverse-Gaussian distribution, Canadian Journal of Statistics, 13, 249–252.
- LETAC, G., WESOLOWSKI, J. (2000). An independence property for the product of GIG and gamma laws, *Annals of Probability*, **28**, 1371–1383.
- LI, S.-H., HUANG, W.-J., HUANG, M.-N.L. (1994). Characterizations of the Poisson process as a renewal process via two conditional moments, Annals of the Institute of Statistical Mathematics, 46, 351–360.
- LUKACS, E. (1955). A characterization of the gamma distribution, Annals of Mathematical Statistics, 26, 319–324.
- MATSUMOTO, H., YOR, M. (1998). Some extensions of Pitman's theorem involving exponential Brownian functional via generalized inverse Gaussian distributions, *Preprint* Nov.98.
- SESHADRI, V. (1983). The inverse Gaussian distribution: some properties and characterizations, Canadian Journal of Statistics, 11, 131–136.
- SESHADRI, V. (1993). The Inverse Gaussian Distribution A Case Study in Exponential Families. Clarendon Press, Oxford.
- TWEEDIE, M.C.K. (1957). Statistical properties of the inverse Gaussian distribution I, Annals of Mathematical Statistics, 28, 362–377.
- WANG, Y. (1981). Extensions of Lukacs' characterization of the gamma distribution. In Analytic Methods in Probability Theory, Lecture Notes in Mathematics 861, Springer, New York, 166–177.
- WESOLOWSKI, J. (1990). A constant regression characterization of the gamma law, Advances in Applied Probability, 22, 488–489.
- WESOLOWSKI, J. (2000) On a functional equation related to the Matsumoto-Yor property, Aequationes Mathematicae, to appear.

VANAMAMALAI SESHADRI DEPARTMENT OF MATHEMATICS AND STATISTICS MCGILL UNIVERSITY MONTREAL, CANADA E-mail: vansesh@hotmail.com Jacek Wesołowski Wydział Matematyki i Nauk Informacyjnych Politechnika Warszawska Warszawa, Poland E-mail:wesolo@alpha.mini.pw.edu.pl