

Viorel Gh. Vodă

Inferential procedures on a generalized Rayleigh variate. II

Aplikace matematiky, Vol. 21 (1976), No. 6, 413--419

Persistent URL: <http://dml.cz/dmlcz/103664>

Terms of use:

© Institute of Mathematics AS CR, 1976

Institute of Mathematics of the Academy of Sciences of the Czech Republic provides access to digitized documents strictly for personal use. Each copy of any part of this document must contain these *Terms of use*.



This paper has been digitized, optimized for electronic delivery and stamped with digital signature within the project *DML-CZ: The Czech Digital Mathematics Library* <http://project.dml.cz>

INFERENCEAL PROCEDURES ON A GENERALIZED
RAYLEIGH VARIATE (II)

V. GH. VODĀ

(Received March 28, 1974)

1. CONFIDENCE INTERVALS

The GRV introduced in the first part has the property that if X is the underlying variable then X^2 is a Gamma variate with certain parameters.

In this way, if x_1, x_2, \dots, x_n is an independent sample on X then it is easy to prove that the statistic $2\theta\zeta$ where $\zeta = \sum_{i=1}^n x_i^2$ is distributed as a chi-square variable with $2(k + 1)n$ degrees of freedom, k being the shape parameter of the GRV and θ the scale parameter.

Therefore, we can determine two numbers l_i and l_s such that for a given confidence – say $(1 - \gamma)$ – we have

$$(1) \quad \text{Prob} \{l_i < 2\theta\zeta < l_s\} = 1 - \gamma.$$

The length of the interval for θ is

$$(2) \quad Q = \frac{1}{2\zeta}(l_s - l_i)$$

and if we look for Q -minimum, we obtain after some tedious algebra:

$$(3) \quad \int_{l_i}^{l_s} x^{n(k+1)-1} e^{-x/2} dx = (1 - \gamma) 2^{nk+n} \Gamma(nk + n), \quad \left(\frac{l_s}{l_i}\right)^{nk+n-1} = \\ = \exp \left\{ \frac{1}{2}(l_s - l_i) \right\}$$

(see also Vodā [3]) which may provide values for l_i and l_s . In this situation seems to be more convenient to look for confidence intervals for $1/\theta$. We have

$$(4) \quad \text{Prob} \left\{ \frac{2\zeta}{l_s} < \frac{1}{\theta} < \frac{2\zeta}{l_i} \right\} = 1 - \gamma.$$

The length is now $\tilde{Q} = 2\zeta(1/l_i - 1/l_s)$ and the minimum condition yields finally

$$(5) \quad \int_{l_i}^{l_s} x^{nk+n-1} e^{-x/2} dx = (1-\gamma) 2^{nk+n} \Gamma(nk+n), \quad \left(\frac{l_s}{l_i}\right)^{nk+n+1} = \\ = \exp\left\{\frac{1}{2}(l_s - l_i)\right\}$$

which can be used for concrete solutions with the aid of Tate-Klett tables [2] but entering in the cell corresponding to $2(k+1)n$ degrees of freedom.

From (4) we obtain easily

$$(6) \quad \text{Prob} \left\{ \frac{\Gamma(k + \frac{3}{2})}{\Gamma(k+1)} \sqrt{\frac{2\zeta}{l_s}} < E(X) < \frac{\Gamma(k + \frac{3}{2})}{\Gamma(k+1)} \sqrt{\frac{2\zeta}{l_i}} \right\} = 1 - \gamma$$

or

$$(7) \quad \text{Prob} \{ \delta(l_s)^{-\frac{1}{2}} < E(X) < \delta(l_i)^{-\frac{1}{2}} \} = 1 - \gamma$$

where

$$(8) \quad \delta = \frac{\Gamma(k + \frac{3}{2})}{\Gamma(k+1)} \cdot \left(2 \sum_{i=1}^n x_i^2 \right)^{\frac{1}{2}}.$$

The above relation may be interpreted as a confidence interval with minimum length for the expected - life in a GR model. In the table below we give the values of the constant

$$(9) \quad \omega = \frac{2^{\frac{1}{2}} \Gamma(k + \frac{3}{2})}{\Gamma(k+1)} \quad \text{where} \quad \delta = \omega \left(\sum_{i=1}^n x_i^2 \right)^{\frac{1}{2}}.$$

N	Density function	ω	Degrees of freedom
1	Rayleigh ($k = 0$)	1.2533141373	$2n$
2	Maxwell ($k = \frac{1}{2}$)	1.595769121	$3n$

Table 1. Useful constants for computing confidence intervals.

For an application of the method we must take into account that Tate-Klett tables [2] are computed for $n = 2(1)30$. Therefore the sample sizes must be limited to stay within the range of the tables (in Rayleigh case: $n \leq 15$ and in Maxwell case: $n \leq 10$).

2. PARAMETER ESTIMATION IN THE CASE OF A MIXTURE OF TWO GR VARIABLES

Consider now a random variable X_{mix} characterized by the following density:

$$(10) \quad X_{\text{mix}} : f_{\text{mix}}(x; \theta_1, \theta_2, p, k) = pf(x; \theta_1, k) + (1-p)f(x; \theta_2, k)$$

where $x > 0$, $\theta_1, \theta_2 > 0$, $0 < p < 1$ and $k \geq 0$ are assumed to be known and $f(x; \theta, k)$ is the density of a GRV.

Let our task be to estimate the parameters θ_1, θ_2 and p .

In this way, we shall generalize a former work of Kryszicki [13] which concerns the mixture of two simple Rayleigh laws.

We shall apply the same method — namely the method of moments.

It is interesting also to investigate the behaviour of the density (10) with respect to the modal value.

We have

$$(11) \quad f'_{\text{mix}}(x) = \frac{2(1-p)\theta_2^{k+2}x^{2k}(2x^2 - \theta_1^{-1}(2k+1))}{\Gamma(k+1)\exp(\theta_1x^2)} \cdot \left[\frac{2x^2 - \theta_2^{-1}(2k+1)}{\theta_1^{-1}(2k+1) - 2x^2} \exp\{(\theta_1 - \theta_2)x^2\} - \frac{p}{1-p} \left(\frac{\theta_1}{\theta_2}\right)^{k+2} \right].$$

To find the modal value, we must impose

$$(12) \quad f'_{\text{mix}}(x; \theta_1, \theta_2, p, k) = 0.$$

It is clear that the product (11) vanishes if

$$(13) \quad x = \left(\frac{2k+1}{2\theta_1}\right)^{\frac{1}{2}}.$$

But this value is not a solution of (12), therefore we have in fact to solve the equations:

$$(14) \quad \frac{2x^2 - \theta_2^{-1}(2k+1)}{\theta_1^{-1}(2k+1) - 2x^2} \exp\{(\theta_1 - \theta_2)x^2\} = \frac{p}{1-p} \left(\frac{\theta_1}{\theta_2}\right)^{k+2}$$

which is a transcendental equation.

Since $(\theta_1/\theta_2)^{k+2} > 0$ for every $\theta_1, \theta_2 > 0$ and $k \geq 0$ and as $0 < p < 1$, the right-hand side is an increasing function of p , due to the factor $p/(1-p)$.

The left-hand side becomes infinite for x given by (13) and vanishes for

$$(15) \quad x = \left(\frac{2k+1}{2\theta_2}\right)^{\frac{1}{2}}.$$

Therefore, for x lying in the interval

$$(16) \quad \mathcal{L} \equiv \left(\left(\frac{2k+1}{2\theta_1}\right)^{\frac{1}{2}}, \left(\frac{2k+1}{2\theta_2}\right)^{\frac{1}{2}} \right)$$

the left-hand side is positive if $\theta_1/\theta_2 > 1$.

Let us denote for brevity the left-hand side by $g(x)$.

It follows that to study the behaviour of $g(x) = (p/(1-p)) (\theta_1/\theta_2)^{k+2}$ in the interval \mathcal{L} , we have to take the derivative of $g(x)$. We have

$$(17) \quad g'(x) = \frac{8(\theta_1 - \theta_2)x}{[(2k+1)\theta_1^{-1} - 2x^2]^2} \exp [(\theta_1 - \theta_2)x^2] \cdot \left\{ -x^4 + \frac{1}{2}(2k+1) \left(\frac{1}{\theta_1} + \frac{1}{\theta_2} \right) x^2 - \frac{(2k+1)(2k+3)}{4\theta_1\theta_2} \right\}.$$

The sign of $g'(x)$ is determined by the expression in parentheses.

The discriminant of the equation in parentheses is

$$(18) \quad \delta = \frac{1}{4}(2k+1)^2 \left(\frac{1}{\theta_2^2} - 2 \frac{2k+5}{2k+1} \cdot \frac{1}{\theta_1\theta_2} + \frac{1}{\theta_1^2} \right).$$

Therefore

$$(19) \quad \delta \geq 0 \quad \text{if} \quad \frac{1}{\theta_2} \geq \frac{2k+5+2(4k+6)^{\frac{1}{2}}}{(2k+1)\theta_1}.$$

Under this condition we obtain for $g'(x)$ two points $x^{(1)}$ and $x^{(2)}$ for which $g'(x^{(i)}) = 0$, $i = 1, 2, \dots$

They are given by

$$(20) \quad x^{(i)} = \left[\frac{(2k+1)(\theta_1 + \theta_2)}{4\theta_1\theta_2} \mp \frac{1}{2}\delta^{\frac{1}{2}} \right]^{\frac{1}{2}}.$$

The behaviour of g' is indicated in Table 2.

x	$\frac{2k+1}{2\theta_1}$	$x^{(1)}$	$x^{(2)}$	$\left(\frac{2k+1}{2\theta_2} \right)^{\frac{1}{2}}$
$g'(x)$	$+\infty$	$\searrow 0 \nearrow$ min	$+$	$\nearrow 0 \searrow$ Max
	$-$		$-$	0

Table 2. Behaviour of $g'(x)$ for $x \in \mathcal{L}$.

Therefore

$$(21) \quad g_{\min} = g(x^{(1)}) \quad \text{and} \quad g_{\max} = g(x^{(2)}).$$

It follows also that the values of $(p/(1-p))(\theta_1/\theta_2)^{k+2}$ vary between g_{\min} and g_{\max} . Hence we can determine an interval for p as

$$(22) \quad \frac{g_{\min}}{g_{\min} + (\theta_1/\theta_2)^{k+2}} < p < \frac{g_{\max}}{g_{\max} + (\theta_1/\theta_2)^{k+2}}.$$

We shall show now that

$$(23) \quad g_{\min} = g(x^{(1)}) = g^*(\theta_1/\theta_2),$$

$$(24) \quad g_{\max} = g(x^{(2)}) = g^{**}(\theta_1/\theta_2).$$

This facts can be easily seen if we write (18) in the form

$$(25) \quad \delta = \frac{(2k+1)^2}{4\theta_1^2} \left(\frac{\theta_1^2}{\theta_2^2} - 2 \frac{2k+5}{2k+1} \frac{\theta_1}{\theta_2} + 1 \right),$$

$$(26) \quad g_{\min} = \frac{1 - r - \left(r^2 - 2 \frac{2k+5}{2k+1} r + 1 \right)^{\frac{1}{2}}}{1 - r + \left(r^2 - 2 \frac{2k+5}{2k+1} r + 1 \right)^{\frac{1}{2}}}$$

$$\cdot \exp \left\{ \frac{2k+1}{4} \left(r - \frac{1}{r} \right) - \frac{2k+1}{4} \left(1 - \frac{1}{r} \right) \left(r^2 - 2 \frac{2k+5}{2k+1} r + 1 \right)^{\frac{1}{2}} \right\}$$

where we have denoted $\theta_1/\theta_2 = r$.

We have by similar calculation

$$(27) \quad g_{\max} = \frac{1 - r + \left(r^2 - 2 \frac{2k+5}{2k+1} r + 1 \right)^{\frac{1}{2}}}{1 - r - \left(r^2 - 2 \frac{2k+5}{2k+1} r + 1 \right)^{\frac{1}{2}}}$$

$$\cdot \exp \left\{ \frac{2k+1}{4} \left(r - \frac{1}{r} \right) + \frac{2k+1}{4} \left(1 - \frac{1}{r} \right) \left(r^2 - 2 \frac{2k+5}{2k+1} r + 1 \right)^{\frac{1}{2}} \right\}.$$

Now it is clear that we must require

$$(28) \quad r \geq \frac{2k+5 + 2(4k+6)^{\frac{1}{2}}}{2k+1}$$

taking into account (19).

For instance, if we wish to tabulate limits for the values of p for different particular densities we must begin from a value of r given by (28) where we have insert the specific value of k .

Example. In the case of Rayleigh distribution we have $k = 0$; therefore

$$(29) \quad r \geq 5 + 2\sqrt{6} \cong 9.899$$

and tabulation may begin from $r = 10$.

The values g_{\min} and g_{\max} are respectively

$$(30) \quad g_{\min}^{(k=0)} = \frac{1 - r - (r^2 - 10r + 1)^{\frac{1}{2}}}{1 - r + (r^2 - 10r + 1)^{\frac{1}{2}}} \exp \left\{ \frac{1}{4} (r - \frac{1}{4}) (r^2 - 10r + 1)^{\frac{1}{2}} \right\},$$

$$(31) \quad g_{\max}^{(k=0)} = \frac{1 - r + (r^2 - 10r + 1)^{\frac{1}{2}}}{1 - r - (r^2 - 10r + 1)^{\frac{1}{2}}} \cdot \exp \left\{ \frac{1}{4} \left(r - \frac{1}{r} \right) + \frac{1}{4} \left(1 - \frac{1}{r} \right) (r^2 - 10r + 1)^{\frac{1}{2}} \right\}.$$

As concerns the estimation, let x_1, x_2, \dots, x_n be an independent sample form the underlying population. Therefore

$$(32) \quad E(X_{\min}^j) = \frac{\Gamma(k + \frac{1}{2}j + 1)}{\Gamma(k + 1)} [p\theta_1^{-\frac{1}{2}j} + (1 - p)\theta_2^{-\frac{1}{2}j}].$$

Since three unknown parameters are involved we take for j successively the values 1, 2, 3.

Hence we obtain the following equations:

$$(33) \quad \hat{p}\hat{\theta}_1^{-\frac{1}{2}} + (1 - \hat{p})\hat{\theta}_2^{-\frac{1}{2}} = \frac{\Gamma(k + 1)}{n \Gamma(k + \frac{3}{2})} \sum_{i=1}^n x_i,$$

$$(34) \quad \hat{p}\hat{\theta}_1^{-1} + (1 - \hat{p})\hat{\theta}_2^{-1} = \frac{1}{n(k + 1)} \sum_{i=1}^n x_i^2,$$

$$(35) \quad \hat{p}\hat{\theta}_1^{-\frac{3}{2}} + (1 - \hat{p})\hat{\theta}_2^{-\frac{3}{2}} = \frac{\Gamma(k + 1)}{n \Gamma(k + \frac{5}{2})} \sum_{i=1}^n x_i^3$$

Let us denote by u and v the following expressions:

$$(36) \quad u = \hat{\theta}_1^{-\frac{1}{2}}, \quad v = \hat{\theta}_2^{-\frac{1}{2}}.$$

We have after some calculations

$$(37) \quad \hat{p}(u - v) = \frac{\Gamma(k + 1)}{n \Gamma(k + \frac{3}{2})} \sum_{i=1}^n x_i - v,$$

$$(38) \quad \hat{p}(u^2 - v^2) = \frac{1}{n(k + 1)} \sum_{i=1}^n x_i^2 - v^2,$$

$$(39) \quad \hat{p}(u^3 - v^3) = \frac{\Gamma(k + 1)}{n \Gamma(k + \frac{5}{2})} \sum_{i=1}^n x_i^3 - v^3.$$

Still other calculations yield

$$(40) \quad u + v = \frac{\frac{\Gamma(k+1)}{n\Gamma(k+\frac{5}{2})} \sum_{i=1}^n x_i^3 - \frac{\Gamma(k+1)}{n^2(k+1)\Gamma(k+\frac{3}{2})} \sum_{i=1}^n x_i \sum_{i=1}^n x_i^2}{\frac{1}{n(k+1)} \sum_{i=1}^n x_i^2 - \left[\frac{\Gamma(k+1)}{n\Gamma(k+\frac{3}{2})} \sum_{i=1}^n x_i \right]^2}$$

$$(41) \quad uv = \frac{\frac{\Gamma(k+1)}{n^2\Gamma(k+\frac{3}{2})\Gamma(k+\frac{5}{2})} \sum_{i=1}^n x_i \sum_{i=1}^n x_i^3 - \frac{1}{n^2(k+1)^2} \left(\sum_{i=1}^n x_i^2 \right)^2}{\frac{1}{n(k+1)} \sum_{i=1}^n x_i^2 - \left[\frac{\Gamma(k+1)}{n\Gamma(k+\frac{3}{2})} \sum_{i=1}^n x_i \right]^2}$$

Supposing that the common denominator of the two ratios is not zero we have a second degree equation.

This equation will provide the moment estimators for $\hat{\theta}_1^{-\frac{1}{2}}$ and $\hat{\theta}_2^{-\frac{1}{2}}$. Then an estimate for p is easily established from (37).

References

- [1] *W. Kryszicki*: Application de la méthode des moments a l'estimation des paramètres d'un melange de deux distributions de Rayleigh. Rev. Stat. Appl. vol. XI (1963), 25–45.
- [2] *R. F. Tate, G. W. Klett*: Optimum confidence intervals for the variance of a normal population. J. Amer. Stat. Assoc. 54 (1959), 674–682.
- [3] *V. Gh. Vodă*: On the "inverse Rayleigh" distributed random variable. Rep. Stat. Appl. Res. JUSE (Japan), vol. 19 (1972), 13–21.
- [4] *V. Gh. Vodă*: Inferential procedures on a generalized Rayleigh variate (I). Apl. mat. 21 (1976), 395–412.

Souhrn

PROCEDURY STATISTICKÉ INDUKCE PRO ZOBECNĚNOU RAYLEIGHOVU PROMĚNNOU (II)

V. GH. VODĂ

V této části se konstruují intervaly spolehlivosti minimální délky pro střední hodnotu zobecněné Rayleighovy proměnné. Dále se studují některé problémy týkající se odhadování ve směsi dvou zobecněných Rayleighových proměnných.

Author's address: Viorel Gh. Vodă, Center of Mathematical Statistics, Știrbey Voda Street No 174, Bucharest 7000 Romania.