

SOME NOTES ON RAYLEIGH DISTRIBUTION

by

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SUMMARY

In this paper, we emphasize some aspects concerning the Rayleigh distribution. After a brief presentation of the literature devoted to this variable, we give a characterization of the Rayleigh distribution by using a Gerbach idea (1967). Then, some properties are put into light and in the latter part of the work sequential tests are constructed in order to test a simple and a compound hypothesis concerning a Rayleigh parameter.

1. *Introduction.* It is well-known that a random variable T obeys a Rayleigh law, if its density function has the form :

$$T : f(t; \lambda) = 2 \lambda^{-2} t \exp \{-\lambda^{-2} t^2\}, \quad t \geq 0, \quad \lambda > 0. \quad (1)$$

It can be easily obtained by taking in the expression of the Weibull law :

$$T_w : f(t; \gamma, p) = \gamma p t^{p-1} \exp \{-\gamma t^p\}, \quad t \geq 0, \quad \gamma, p > 0, \quad (2)$$

respectively, $\gamma = \lambda^{-2}$, and $p = 2$.

Bruscantini (1968) enlists some important occurrences of the Rayleigh law in various practical situations, namely :

- in naval research, to simulate the sea waves ;
- in telecommunications, to describe the signal fluctuations due to multipath effects in the line-of-sight links ;

- in bombing problems, to describe the distribution of distances from target to the actual impact points.

In some cases, Rayleigh law appears naturally : let us consider a system built up by n pairs of elements, connected in series. In every pair, each element is functioning and a failure occurs if at least a pair is failed. Suppose that the reliability of each element obeys an exponential law and let λ_k be the hazard rate of the k^{th} pair. If $\theta = \left(\sum_{k=1}^n \lambda_k^2 \right)^{\frac{1}{2}}$ is fixed and $\lim_{n \rightarrow \infty} \sum_{k=1}^n \lambda_k^3 = 0$, then the asymptotic reliability function of the system is given by :

$$AsP(t) = \lim_n P(t) = \exp \{ -\theta^2 t^2 \} , \quad t \geq 0 , \theta > 0 , \quad (3)$$

that is a Rayleigh law (see Gnedenko-Belyaev-Solov'yev, 1965).

It is to note also that the Rayleigh law can be obtained not only from the Weibull distribution but even as a peculiar case of other important probability laws .

Let us remember the "chi-distribution" and so-called "generalized Rayleigh variable" whose density is :

$$f(t; \theta, k) = \frac{2 \theta^{k+1}}{\Gamma(k+1)} t^{2k+1} \exp \{ -\theta^2 t^2 \} , \quad t \geq 0 , \theta, k > 0 , \quad (4)$$

where, if we take $\theta = \lambda^{-2}$ and $k=0$, simple Rayleigh law is obtained (see for details, Vodň, 1972) .

Finally, let us note that if $M(X, Y)$ is a point in a rectangular coordinate - system, where X and Y are random variables each of class $N(0 ; \sigma^2)$, then the distance $d = (X^2 + Y^2)^{\frac{1}{2}}$ is distributed as a random variable which obeys

Rayleigh law . (See Freeman, 1963 p. 155) .

In various statistical textbooks (see, for instance, Maisel, 1971) the Rayleigh density function appears as below :

$$f(x; \alpha) = \alpha^{-2} x \exp\{-x^2 / 2 \alpha^2\}, x \geq 0, \alpha > 0 \quad (5)$$

The specific literature devoted to the Rayleigh distribution is not so very wide but most of the properties of this distribution can be derived from properties of the Weibull law .

We shall mention the works of Geldston (1962) who has investigated the variable $Y = k \log(1 + bX)$ where X has the density (1), Kryszewski (1963) who has applied method of moments in estimating parameters of a mixture of two Rayleigh distributions (in this connection see also Siddiqui and Weiss, 1963) and the paper of Archer (1967) who has presented some properties of Rayleigh distributed random variables and of their sums and products .

2. *A characterization of Rayleigh distribution.* Let us consider the following function :

$$F(t; \lambda, \xi) = 1 - \exp(-t^2 / \lambda^2) . \quad P(t; \xi) \text{ for } t \geq 0, \lambda, \xi > 0, \quad (6)$$

where

$$P(0; \xi) = 1 \quad \text{and} \quad P(\infty; \xi) = 1 \quad (7)$$

It is clear that in this way , $F(t; \lambda, \xi)$ may be interpreted as a distribution function and more than that, $P(t; \xi)$ may be not necessarily a reliability function (see for instance Gerchbach, 1967) .

If, in particular, there exists a value $\xi = \xi_0$ such that for every $t \geq 0$, $P(t; \xi_0) = 1$, then the Rayleigh distribution function is rediscovered .

Let $P(t; \xi)$ be of the following type :

$$P(t; \xi) = \exp \{ \alpha(t) r(\xi) + \beta(t) + s(\xi) \}, t \geq 0, \xi > 0 \quad (8)$$

and preserving the properties (7).

In this situation one can give the following characterization of the Rayleigh law:

LEMMA 1: Let T_1, T_2, \dots, T_n be independent and identically distributed random variables. Then the distribution of $\min(T_1, T_2, \dots, T_n)$ is of type (6) if and only if the T_i 's distributions are of the same type.

Proof: Let $T_{(1)} < T_{(2)} < \dots < T_{(n)}$ be the ordered variables T_1, \dots, T_n , each with cumulative distribution function given by (6). Then, we can write:

$$\begin{aligned} F_0(t) &= \text{Prob} \{ T_{(1)} < t \} = 1 - \text{Prob} \{ T_{(1)} > t \} = 1 - \text{Prob} \{ \text{all } T_i > t \} = 1 - [1 - F(t; \lambda, \xi)]^n \\ &= 1 - \exp(-n t^2 / \lambda^2) [P(t; \xi)]^n = \exp(-n t^2 / \lambda^2) \cdot \exp \{ n \alpha(t) r(\xi) + n \beta(t) + n s(\xi) \} \quad (9) \end{aligned}$$

If we denote $n \lambda^{-2} = \theta^{-2}$, $\theta > 0$, $\alpha(t) = \tilde{\alpha}(t)$, $n r(\xi) = \tilde{r}(\xi)$, $n \beta(t) = \tilde{\beta}(t)$ and $n s(\xi) = \tilde{s}(\xi)$, we obtain:

$$F_0(t) = 1 - \exp(-t^2 / \theta^2) \tilde{P}(t; \xi), \quad (10)$$

that is a distribution of the type (6).

Lemma is proved.

3. Estimation problems. The expected-value and the variance of a Rayleigh variate are:

$$E(T) = \frac{\sqrt{\pi}}{2} \lambda \quad \text{and} \quad \text{Var}(T) = \left(1 - \frac{\pi}{4}\right) \lambda^2 \quad (11)$$

The square of a Rayleigh variate T^2 obeys an exponential law:

$$T^2 : f_{T^2}(t; \lambda) = \frac{1}{\lambda^2} \exp(-t/\lambda^2), t \geq 0, \lambda > 0 \quad (12)$$

As a consequence, the maximum likelihood estimate of λ^2 , that is:

$$(\hat{\lambda}^2)_{ML} = \frac{1}{n} \sum_{i=1}^n t_i^2, \quad (13)$$

t_1, t_2, \dots, t_n being independent random sample drawn from T , is unbiased and with minimum variance .

On the other hand, the statistic :

$$b(t) = \frac{\sqrt{\pi}}{2} \cdot t \quad (14)$$

is an unbiased estimate for λ with respect to the loss-function :

$$W[\lambda; b(t)] = \frac{1}{\lambda^2} [b(t) - \lambda]^2 \quad (15)$$

The associated risk is :

$$R_{b(t)}(\lambda) = E\{W[\lambda; b(t)]\} = 1 - \frac{\pi}{4} \quad (16)$$

and the variance :

$$\text{Var}(b(t)) = E[b(t) - E(b(t))]^2 = \frac{\pi}{4} \left(1 - \frac{\pi}{4}\right) \lambda^2 \quad (17)$$

In the following we shall derive the maximum likelihood estimate for λ^2 in the case of a special incomplete sample.

Therefore, suppose that from n observations we have the possibility to record only :

$$t_{(m)} < t_{(m+1)} < \dots < t_{(m+k)} \quad (18)$$

where $1 < m < m+k < n$.

The lack of the first $m-1$ observations may be interpreted as "infant mortality" of certain cutting-tools and the absence of the last observations may be due to the technological restrictions .

Hence, the likelihood function in this case has the form :

$$L = [1 - F(t_{(m+k)}; \lambda)]^{n-m-k} \cdot [F(t_{(m)}; \lambda)]^{m-1} \prod_{i=m}^{m+k} f(t_{(i)}; \lambda) \quad (19)$$

Hence :

$$\ln L = (n-m-k) \ln [1 - F(t_{(m+k)}; \lambda)] + (m-1) \ln F(t_{(m)}; \lambda) + \sum_{i=m}^{m+k} \ln f(t_{(i)}; \lambda) \quad (20)$$

or in our particular case :

$$\ln L = (n-m-k) \cdot \left(-\frac{t_{(m+k)}^2}{\lambda^2} \right) + (m-1) \ln [1 - \exp \{ -t_{(m)}^2 / \lambda^2 \}] + \sum_{i=m}^{m+k} \left(\ln \frac{2}{\lambda^2} + \ln t_{(i)} - \frac{t_{(i)}^2}{\lambda^2} \right) \quad (21)$$

$$\begin{aligned} \frac{\partial \ln L}{\partial (\lambda^2)} &= (n-m-k) \cdot \left(\frac{t_{(m+k)}^2}{\lambda^4} \right) + (m-1) \frac{-\frac{1}{\lambda^4} t_{(m)}^2 \exp \{ -t_{(m)}^2 / \lambda^2 \}}{1 - \exp \{ -t_{(m)}^2 / \lambda^2 \}} - (k+1) \frac{1}{\lambda^2} + \\ &+ \frac{1}{\lambda^4} \cdot \sum_{i=1}^{m+k} t_{(i)}^2 \end{aligned} \quad (22)$$

which may be rearranged as :

$$(n-m-k) \frac{t_{(m+k)}^2}{\hat{\lambda}^4} - \frac{m-1}{2\hat{\lambda}^2} \cdot t_{(m)} \Lambda(t_{(m)}) - \frac{k+1}{\hat{\lambda}^2} + \frac{1}{\hat{\lambda}^4} \cdot \sum_{i=m}^{m+k} t_{(i)}^2 = 0 \quad (23)$$

where by $\Lambda(t_{(m)})$ we understand the hazard rate :

$$\Lambda(t) = \frac{f(t; \lambda)}{1 - F(t; \lambda)} \quad (24)$$

calculated in the point $t_{(m)}$.

We obtain finally :

$$\hat{\lambda}^2 = \frac{(n-m-k) t_{(m+k)}^2 + \sum_{i=m}^{m+k} t_{(i)}^2}{k+1 + \frac{1}{2} (m-1) t_{(m)} \cdot \Lambda(t_{(m)})} \quad (25)$$

It is easy to see that if $m=1$ and $m+k=n$, the classical m.l.e. for λ^2 is obtained. (For linear estimations of λ , see Iliescu-Vodă, 1973).

4. *Testing statistical hypotheses.* We shall construct now a SPRT for testing the simple hypothesis :

$$H_0 : \lambda = \lambda_0 \quad \text{versus} \quad H_1 : \lambda = \lambda_1 \quad (\lambda_0 < \lambda_1) \quad (26)$$

Up to a certain constant, this one can be interpreted as a hypothesis on the mean-life of the considered Rayleigh process.

Let $t_1, t_2, \dots, t_n, \dots$ be a sequential sample drawn from T and let $\text{Prob} \{ H_1 | H_0 \} = \alpha$ and $\text{Prob} \{ H_0 | H_1 \} = \beta$ be the usual risks in sequential analysis.

Straightforward computations provide the log-likelihood ratio as :

$$\ln \frac{\prod_{i=1}^n f(t_i; \lambda_1)}{\prod_{i=1}^n f(t_i; \lambda_0)} = \ln r_n = 2n \ln \frac{\lambda_0}{\lambda_1} + \left(\frac{1}{\lambda_0^2} - \frac{1}{\lambda_1^2} \right) \sum_{i=1}^n t_i^2 \quad (27)$$

Therefore, in accordance with Wald's rules (1960) we can present the decisions as below :

$$(1) \text{ if } \sum_{i=1}^n t_i^2 \leq \frac{\ln \frac{\beta}{1-\alpha}}{\frac{1}{\lambda_0^2} - \frac{1}{\lambda_1^2}} + n \frac{2 \ln \frac{\lambda_1}{\lambda_0}}{\frac{1}{\lambda_0^2} - \frac{1}{\lambda_1^2}}, \text{ then accept } H_0;$$

$$(2) \text{ if } \sum_{i=1}^n t_i^2 \geq \frac{\ln \frac{1-\beta}{\alpha}}{\frac{1}{\lambda_o^2} - \frac{1}{\lambda_1^2}} + n \cdot \frac{2 \ln \frac{\lambda_1}{\lambda_o}}{\frac{1}{\lambda_o^2} - \frac{1}{\lambda_1^2}}, \text{ then accept } H_1 ;$$

$$(3) \text{ if } \frac{\ln \frac{\beta}{1-\alpha}}{\frac{1}{\lambda_o^2} - \frac{1}{\lambda_1^2}} + n \frac{2 \ln \frac{\lambda_1}{\lambda_o}}{\frac{1}{\lambda_o^2} - \frac{1}{\lambda_1^2}} < \sum_{i=1}^n t_i^2 < \frac{\ln \frac{1-\beta}{\alpha}}{\frac{1}{\lambda_o^2} - \frac{1}{\lambda_1^2}} + n \frac{2 \ln \frac{\lambda_1}{\lambda_o}}{\frac{1}{\lambda_o^2} - \frac{1}{\lambda_1^2}},$$

then continue the experiment.

The power of the test is given by $\pi(\lambda) = 1 - L(\lambda)$ where $L(\lambda)$ is the OC-function, defined as below :

$$L(\lambda) \doteq \frac{A^b - 1}{A^b - B^b}, \quad A \doteq \frac{1-B}{\alpha}, \quad B \doteq \frac{B}{1-\alpha}. \quad (28)$$

Here, b is given by the Wald's fundamental lemma, namely :

$$E(e^{bz}) = 1 \quad (b \neq 0), \quad (29)$$

where

$$z = \ln f(t; \lambda_1) - \ln f(t; \lambda_o). \quad (30)$$

In our specific application we have :

$$z = \ln \frac{f(t; \lambda_1)}{f(t; \lambda_o)} = 2 \ln \frac{\lambda_o}{\lambda_1} + \left(\frac{1}{\lambda_o^2} - \frac{1}{\lambda_1^2} \right) t^2. \quad (31)$$

Hence :

$$E(e^{bz}) = C \int_0^\infty t \exp \left[- \left(\frac{b}{\lambda_1^2} - \frac{b}{\lambda_o^2} + \frac{1}{\lambda^2} \right) t^2 \right] dt, \quad (32)$$

where $C = \frac{2}{\lambda^2} \left(\frac{\lambda_o}{\lambda_1} \right)^2$

Imposing $b \lambda_1^{-2} - b \lambda_o^{-2} + \lambda^{-2} > 0$ and denoting $(b \lambda_1^{-2} - b \lambda_o^{-2} + \lambda^{-2}) t^2 = u$

we obtain immediately :

$$E(e^{zb}) = \frac{1}{\lambda^2} \left(\frac{\lambda_o}{\lambda_1} \right)^{2k} \left(\frac{b}{\lambda_1^2} - \frac{b}{\lambda_o^2} \right)^{-1} = 1. \quad (33)$$

From (33) it follows that :

$$\lambda^2 = \frac{\left(\frac{\lambda_o}{\lambda_1} \right)^{2b} - 1}{b \left(\frac{1}{\lambda_1^2} - \frac{1}{\lambda_o^2} \right)}, \quad (34)$$

which together with (28) represents a parametric form for the OC-function .

Now, we compute the ASN for the given test. We have :

$$E_{\lambda}(n) \doteq \frac{L(\lambda) \ln B + (1-L(\lambda)) \ln A}{E_{\lambda}(z)} \quad (35)$$

We have therefore :

$$E_{\lambda}(z) = 2 \ln \frac{\lambda_o}{\lambda_1} + \left(\frac{1}{\lambda_o^2} - \frac{1}{\lambda_1^2} \right) \cdot \frac{2}{\lambda^2} \int_0^{\infty} t^3 \exp \{ -t^2 / \lambda^2 \} dt \quad (36)$$

Taking into account the formula :

$$\int y^n \exp \{ -y^2 \} dy = -\frac{1}{2} y^{n-1} e^{-y^2} + \frac{1}{2} (n-1) \int y^{n-2} \exp \{ -y^2 \} dy \quad (37)$$

we obtain for $n=3$ - needed for (36) - the following expression for $E_{\lambda}(z)$:

$$E_{\lambda}(z) = 2 \ln \frac{\lambda_0}{\lambda_1} + (\lambda_0^{-2} - \lambda_1^{-2}) \lambda^2 \quad (38)$$

Therefore :

$$E_{\lambda}(n) \doteq \frac{b_2 + (b_1 - b_2) L(\lambda)}{q + \lambda^2} \quad (39)$$

where we have introduced following notations :

$$b_1 = \frac{\ln \frac{\beta}{1-\alpha}}{\lambda_0^{-2} - \lambda_1^{-2}} \quad , \quad b_1 = \frac{\ln \frac{1-\beta}{\alpha}}{\lambda_0^{-2} - \lambda_1^{-2}} \quad \text{and} \quad q = \frac{2 \ln \frac{\lambda_1}{\lambda_0}}{\lambda_0^{-2} - \lambda_1^{-2}} \quad (40)$$

We observe that the uncertainty domain of the test can be written as :

$$b_1 + nq < \sum_{i=1}^n t_i^2 < b_2 + nq, \quad (41)$$

that is, bounds for $\sum_{i=1}^n t_i^2$ are respectively acceptance and rejection straight-lines .

4.1. *Sequential comparison.* Consider now two independent Rayleigh variables X and Y , namely :

$$X : f(x ; \theta) = \frac{2}{\theta^2} x \exp(-x^2/\theta^2) , \quad x \geq 0 , \quad \theta > 0 , \quad (42)$$

$$Y : f(y ; \omega) = \frac{2}{\omega^2} y \exp(-y^2/\omega^2) , \quad y \geq 0 , \quad \omega > 0 , \quad (43)$$

and suppose that we have to test :

$$H : \theta \leq \omega \quad \text{against} \quad H' : \theta > \omega .$$

Up to a certain constant it can be interpreted as a comparison between the mean-life times of two Rayleigh processes .

To solve this problem we shall apply Girshick's procedure ; the likelihood ratio is in this case (see for details Wald, 1960) :

$$R_n = \frac{\prod_{i=1}^n f_1(x_i; y_i)}{\prod_{i=1}^n f_0(x_i; y_i)} = \exp \left[\left(\frac{1}{\theta_o^2} - \frac{1}{\omega_o^2} \right) \sum_{i=1}^n (x_i^2 - y_i^2) \right] \quad (44)$$

with $\theta_o < \omega_o$, or

$$\ln R_n = \delta \sum_{i=1}^n (x_i^2 - y_i^2) \quad (45)$$

where :

$$\delta = -v(\theta_o; \omega_o) = \theta_o^{-2} - \omega_o^{-2} ; \quad v(\theta; \omega) = \omega^{-2} - \theta^{-2}$$

being the so called Girschick's function.

The test itself is as follows :

- (a) if $\sum_{i=1}^n (x_i^2 - y_i^2) \leq \frac{1}{\delta} \ln \frac{\beta}{1-\alpha}$ then accept H ;
- (b) if $\sum_{i=1}^n (x_i^2 - y_i^2) \geq \frac{1}{\delta} \ln \frac{1-\beta}{\alpha}$ then accept H' ;
- (c) if $\frac{1}{\delta} \ln \frac{\beta}{1-\alpha} < \sum_{i=1}^n (x_i^2 - y_i^2) < \frac{1}{\delta} \ln \frac{1-\beta}{\alpha}$ then continue the experiment .

To compute OC-function and the ASN we need first the following .

LEMMA 2 : If X and Y are two independent Rayleigh variables, then

Wald's constant $b(\theta; \omega)$ is given by :

$$b(\theta; \omega) = \frac{(\theta^2 - \omega^2)(\omega_o \theta_o)^2}{(\theta_o^2 - \omega_o^2)(\omega \theta)^2} \quad (46)$$

Proof : Using Obreja's reduction (1969) of Wald's equation

$$E [\exp \{ z b(\theta ; \omega) \}] = 1 \quad (47)$$

We shall obtain :

$$E [\exp \{ \delta b(\theta ; \omega) x^2 \}] E [\exp \{ - \delta b(\theta ; \omega) y^2 \}] = 1 \quad (48)$$

which provides easily :

$$[1 - \theta^2 b(\theta ; \omega) \delta]^2 \cdot [1 + \omega^2 b(\theta ; \omega) \delta]^2 = 1 \quad (49)$$

from it follows immediately :

$$\delta b(\theta ; \omega) = \frac{1}{\omega^2} - \frac{1}{\theta^2} \quad (50)$$

which gives the required relation. Lemma is proved .

Hence, the OC-function has the representation :

$$L(\theta ; \omega) \doteq \frac{A}{A} \frac{\frac{(\theta_o^2 - \omega^2)(\theta_o \omega_o)^2}{(\theta_o^2 - \omega_o^2)(\theta \omega)^2} - 1}{\frac{(\theta^2 - \omega^2)(\theta_o \omega_o)^2}{(\theta_o^2 - \omega_o^2)(\theta \omega)^2} - \frac{(\theta^2 - \omega^2)(\theta_o \omega_o)^2}{(\theta_o^2 - \omega_o^2)(\theta \omega)^2}} \quad (51)$$

To compute more rapidly the ASN we use Obreja-Vodă's form (1973) for the denominator of $E_{\theta ; \omega}(n)$. Direct computations provide :

$$E_{\theta ; \omega}(n) \doteq \frac{(\theta_o \omega_o)^2 [L(\theta ; \omega) \ln B + (1 - L(\theta ; \omega)) \ln A]}{(\omega_o^2 - \theta_o^2)(\omega^2 - \theta^2)} \quad (52)$$

The test now is completely constructed .

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