

## Reliability and Ratio of Two Independent Exponential Distributions

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### Abstract

We shall consider an estimation of the reliability  $P(Y < X)$ , and derive moments of the ratio  $X/(X+Y)$  in two independent exponential random variables.

**Keywords** : Ratio, Reliability

### 1. Introduction

Let  $X_1, X_2, \dots, X_n$  be independently identical random variables with the pdf

$$f(x; \theta) = \frac{1}{\theta} e^{-\frac{x-\theta}{\theta}}, \quad x > \theta > 0. \quad (1.1)$$

In reliability applications the exponential distribution has been considered by many authors in Johnson et al(1994)). If the reliability  $R = P(Y < X)$  depends on a parameter  $\rho$  only and  $R$  is a monotone function of  $\rho$ , then inference on  $\rho$  is equivalent to inference on  $R$  in McCool(1991).

The reliability has often applied to engineering, biological phenomenon and physics. In a recent, Kim, et al(2003) studied a UMVUE of the reliability in an exponential distribution. The distribution of the ratio of two independent random variables arises in a model of ionic current fluctuation in biological membranes. The distribution of the ratio of independent gamma variate was studied by Bowman and Shenton(1998).

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Here we shall consider an estimation of the reliability  $P(Y < X)$ , and derive a distribution of the ratio  $X/(X+Y)$  when  $X$  and  $Y$  are two independent exponential random variables each having the exponential distribution (1.1) with the different parameters.

## 2. Reliability

We shall consider an estimation of the reliability in the exponential distribution (1.1). Assume  $X$  and  $Y$  be two independent random variables each having the following exponential densities:

$$\begin{aligned} f_X(x) &= \frac{1}{\theta_1} \exp\left(-\frac{x-\theta_1}{\theta_1}\right), \quad x > \theta_1 > 0 \\ f_Y(y) &= \frac{1}{\theta_2} \exp\left(-\frac{y-\theta_2}{\theta_2}\right), \quad y > \theta_2 > 0. \end{aligned} \quad (2.1)$$

Then the reliability can be obtained:

$$R \equiv P(X < Y) = \begin{cases} 1 - \frac{1}{1+\rho} e^{1-\rho}, & \text{if } \rho \geq 1 \\ \frac{\rho}{1+\rho} e^{1-\frac{1}{\rho}}, & \text{if } 0 < \rho < 1 \end{cases} \quad (2.2)$$

where  $\rho \equiv \frac{\theta_2}{\theta_1}$ . (see Kim et al(2003))

From the result (2.2), the reliability  $R = P(Y < X)$  depends on  $\rho$  only and is a monotone function of  $\rho$ . Because  $R$  is a monotone function of  $\rho$ , inference on  $\rho$  is equivalent to inference on  $R$  in McCool(1991). And hence it's sufficient for us to consider an estimation of  $\rho$  in stead of estimating the reliability  $R$  itself.

Assume  $X_1, X_2, \dots, X_m$  and  $Y_1, Y_2, \dots, Y_n$  be two independent samples from the preceding random variables  $X$  and  $Y$  each having the density (2.1).

Based on the MLE of parameters  $\theta_1$  and  $\theta_2$ , two estimators of  $\rho$  can be defined as:

$$\hat{\rho} = \frac{Y_{(1)}}{X_{(1)}},$$

and

$$\tilde{\rho} = \frac{m \sum_{i=1}^n Y_i}{n \sum_{i=1}^m X_i} .$$

From the densities of the first order statistics  $X_{(1)}$  and  $Y_{(1)}$  in Johnson et al(2004), and the formulas 2.9 and 2.14 in Oberhettinger & Badii(1973), we can obtain the expectation and variance of  $\hat{\rho}$

$$\begin{aligned} E(\hat{\rho}) &= (1 + \frac{1}{n})m e^m [-E_i(-m)]\rho , \\ Var(\hat{\rho}) &= [\frac{1}{n^2} + (1 + \frac{1}{n})^2][m + m^2 e^m E_i(-m)]\rho^2 - E^2(\hat{\rho}) \end{aligned} \quad (2.3)$$

where  $-E_i(-x) = \int_x^\infty \frac{e^{-t}}{t} dt$ ,  $x > 0$  is an exponential integral.

From the densities of two sum statistics  $\sum_{i=1}^m X_i$  and  $\sum_{i=1}^n Y_i$  in Kim et al(2003), and the formula 3.8 in Oberhettinger and Badii(1973) and the formula 3.5 in Oberhettinger(1974), we can obtain the expectation and variance of  $\tilde{\rho}$ :

$$\begin{aligned} E(\tilde{\rho}) &= 2m^m e^m \Gamma(-(m-1), m)\rho , \\ Var(\tilde{\rho}) &= \frac{4m+1}{n} m^{\frac{m+1}{2}} e^{m/2} W_{-\frac{m}{2}-\frac{1}{2}, \frac{m}{2}-1}(m) \rho^2 - E^2(\tilde{\rho}) \end{aligned} \quad (2.4)$$

where  $\Gamma(-n, x) = \frac{(-1)^n}{n!} [-E_i(-x) - e^{-x} \sum_{i=0}^{n-1} (-1)^i \frac{i!}{x^{i+1}}]$

and  $W_{a,b}(x)$  is the Whittaker function .

**Remark 1.** The Whittaker function can be represented by an integral form in Gradshteyn et al(1965) to evaluate variance of  $\tilde{\rho}$  numerically as the following:

$$W_{-\frac{1}{2}-\frac{m}{2}, \frac{m}{2}-1}(m) = \frac{m^{\frac{m}{2}-\frac{1}{2}}}{\Gamma(m)} e^{-m/2} \int_0^\infty e^{-mt} t^{m-1} (1+t)^{-2} dt .$$

From the results (2.3) and (2.4) of expectations and variances of two estimators  $\hat{\rho}$  and  $\tilde{\rho}$ , and an integral form of the Whittaker function in Remark 1, Table 1 shows the numerical values of mean squared errors of two estimators  $\hat{\rho}$  and  $\tilde{\rho}$ .

<Table 1> shows that the MLE  $\hat{\rho}$  is more efficient in a sense of MSE than the estimator  $\tilde{\rho}$  when n=10, 20 and m=5(5)20.

<Table 1> Mean squared errors of two estimators  $\hat{\rho}$  and  $\tilde{\rho}$  (unit:  $\rho^2$ )

n	m	$\hat{\rho}$	$\tilde{\rho}$
10	5	0.05383	0.10871
10	10	0.05307	0.09160
10	15	0.04620	0.09087
10	20	0.04272	0.08924
20	5	0.04833	0.09701
20	10	0.04401	0.08715
20	15	0.04068	0.08051
20	20	0.03721	0.07357

### 3. Distribution of the ratio $\frac{X}{X+Y}$

Let  $X$  and  $Y$  be independent random variables each having two parameter exponential densities (2.1). To find a distribution of the ratio  $V = X/(X+Y)$  when  $X$  and  $Y$  are two independent exponential random variables each having density (2.1), we consider the joint density of the following random variables  $W$  and  $V$

Let  $W \equiv X+Y$  and  $V \equiv \frac{X}{X+Y}$ . Then, from the joint pdf of  $W$  and  $V$  and the formulas 3.381 (2) & (3) in Gradshteyn & Ryzhik(1965), we can obtain the density (3.1) of the ratio  $V$ :

$$f_V(x) = \begin{cases} \frac{re^2}{x^2} e^{-(1+\frac{1-x}{x}r)} \left(1 + \frac{1-x}{x}r\right)^{-2} + \frac{re^2}{x^2} e^{-(1+\frac{1-x}{x}r)} \left(1 + \frac{1-x}{x}r\right)^{-1}, & \text{if } 0 < x < \frac{\theta_1}{\theta_1 + \theta_2}, \\ \frac{e^2}{r(1-x)^2} e^{-(1+\frac{x}{r(1-x)})} \left(1 + \frac{x}{r(1-x)}\right)^{-2} + \frac{e^2}{r(1-x)^2} e^{-(1+\frac{x}{r(1-x)})} \cdot \left(1 + \frac{x}{r(1-x)}\right)^{-1}, & \text{if } \frac{\theta_1}{\theta_1 + \theta_2} \leq x < 1 \end{cases}$$

$$\text{where } r \equiv \frac{\theta_1}{\theta_2}. \quad (3.1)$$

We shall introduce the following integrals to find expectation and variance of the ratio  $V$ . From the formulas 2.7 and 2.14 in Oberhettinger and Badii(1973), we can obtain the following Fact 1:

$$\text{Fact 1. } (1) \int_2^\infty \frac{e^{-x}}{x(x+a)} dx = \frac{1}{a} e^a \cdot E_i(-a-2) - \frac{1}{a} E_i(-2).$$

$$(2) \int_2^{\infty} \frac{e^{-x}}{x^2(x+a)} dx = \frac{a+1}{a^2} E_i(-2) - \frac{1}{a^2} e^a E_i(-a-2) + \frac{1}{2a} e^{-2} .$$

$$(3) \int_2^{\infty} \frac{e^{-x}}{x(x+a)^2} dx = \frac{a+1}{a^2} e^a E_i(-a-2) - \frac{1}{a^2} E_i(-2) - \frac{1}{a(a+2)} e^{-2} .$$

$$(4) \int_2^{\infty} \frac{e^{-x}}{x^2(x+a)^2} dx = \frac{a+4}{2a^2(a+2)} e^{-2} + \frac{a+2}{a^3} E_i(-2) + \frac{a-2}{a^3} e^a E_i(-a-2)$$

where  $-E_i(-x) = \int_x^{\infty} \frac{e^{-t}}{t} dt, \quad x > 0 .$

From the density of the ratio (3.1) and Fact 1 (1) & (2), we can obtain the expectation of  $V$  .

$$E(V) = \begin{cases} r \left[ \frac{r-2}{(r-1)^2} e^{r+1} E_i(-r-1) - \frac{\frac{1}{r}-2}{(r-1)^2} e^{\frac{1}{r}+1} E_i\left(-\frac{1}{r}-1\right) \right], & \text{if } r \neq 1, \\ \frac{1}{2}, & \text{if } r = 1 \end{cases}$$

Next, to find the variance of the ratio, we can obtain 2nd moment of the ratio from the density (3.1) and Fact 1 (1), (2), (3) and (4),

$$\begin{aligned} E(V^2) &= \frac{3r-1}{(1-r)^2} + \frac{1-2r+3r^2-4r^3}{r(1-r)^3} e^{\frac{1}{r}+1} E_i\left(-\frac{1}{r}-1\right) + \frac{r^2(r^2-3)}{(r-1)^3} \\ &\quad \cdot e^{r+1} E_i(-r-1), \quad \text{if } r \neq 1 \\ &= -\frac{1}{3} - \frac{5}{3} e^2 E_i(-2) \doteq 0.26887, \quad \text{if } r = 1 . \end{aligned}$$

Therefore, especially if  $r=1$  ( i.e.  $\theta_1=\theta_2$  ), then  $Var(V) \doteq 0.01887$  .

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