

*Mathematics*

## On the Estimation of a Maximum Likelihood of Truncated Exponential Distributions

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**ABSTRACT.** The problem of estimation of parameters of truncated exponential distribution by the maximum likelihood method is studied. It is shown that maximum likelihood equation for truncated exponential distribution has a unique solution which gives an asymptotic effective estimator of the parameter. © 2013 Bull. Georg. Natl. Acad. Sci.

**Key words:** truncation, exponential distribution, maximum likelihood estimate.

Let  $X$  be a truncated exponentially distributed random value with density

$$f(x; \theta, \alpha, \beta) = \begin{cases} \frac{\theta e^{-\theta x}}{e^{-\alpha\theta} - e^{-\beta\theta}}, & \alpha < x \leq \beta \\ 0, & x \leq \alpha, x > \beta \end{cases}, \quad (1)$$

where  $\alpha < \beta$  and  $\alpha$ ,  $\beta$ , and  $\theta$  are the unknown parameters. Let  $X_1, X_2, \dots, X_n$  be a random sampling of size  $n$  taken from the truncated exponential distributions given by (1). It is required to estimate  $\alpha$ ,  $\beta$ , and  $\theta$  by these observations. We do this by applying the maximum likelihood estimator.

A likelihood function has the form

$$L(x; \theta, \alpha, \beta) = \theta^n (e^{-\alpha\theta} - e^{-\beta\theta})^{-n} \cdot \exp\left(-\theta \sum_{i=1}^n x_i\right) = \theta^n (e^{-\alpha\theta} - e^{-\beta\theta})^{-n} \cdot \exp(-n\theta \bar{X}). \quad (2)$$

The logarithm of likelihood functions is defined and differentiated on the open interval  $(0, \infty)$ , whereas a maximum value, if it exists, takes place at a stationary point at which

$$\frac{\partial \ln L(x; \theta, \alpha, \beta)}{\partial \theta} = 0,$$

but does not take place at a boundary point lying on the interval  $(0, \infty)$ .

**Lemma [1].** Let the following conditions be fulfilled:

(1) the distribution function  $F(x, \theta)$  is continuous with respect to both variables and has the continuous derivative  $f(x, \theta) = \frac{\partial F(x, \theta)}{\partial x}$ ;

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(2) the likelihood function  $L$  has the absolute maximum  $\theta = \bar{\theta}_n$ .

Then  $\bar{\theta}_n$  is an asymptotically consistent and asymptotically effective estimate of the true value of the parameter  $\theta$ .

The logarithm of likelihood functions is written as

$$\ln L(x; \theta, \alpha, \beta) = n \ln \theta - n \ln(e^{-\alpha\theta} - e^{-\beta\theta}) - n\theta \bar{X}.$$

Consider the equation

$$\begin{aligned} \frac{\partial \ln L(x; \theta, \alpha, \beta)}{\partial \theta} &= 0, \\ \frac{\partial \ln L(x; \theta, \alpha, \beta)}{\partial \theta} &= \frac{n}{\theta} - \frac{n(\beta \cdot e^{-\theta\beta} - \alpha \cdot e^{-\theta\alpha})}{e^{-\alpha\theta} - e^{-\beta\theta}} - n\bar{X} = 0, \\ \frac{1}{\theta} - (\beta \cdot e^{-\beta\theta} - \alpha \cdot e^{-\alpha\theta}) \cdot (e^{-\alpha\theta} - e^{-\beta\theta})^{-1} - \bar{X} &= 0. \end{aligned} \quad (3)$$

Let us investigate this equation for  $\theta \rightarrow 0$ . Note that for  $\theta \rightarrow 0$  we have  $e^{-\alpha\theta} - e^{-\beta\theta} \rightarrow 0$ .

According to l'Hospital's rule, we obtain

$$\begin{aligned} \lim_{\theta \rightarrow 0} \left( \frac{1}{\theta} - (\beta e^{-\beta\theta} - \alpha e^{-\alpha\theta})(e^{-\alpha\theta} - e^{-\beta\theta})^{-1} \right) &= \lim_{\theta \rightarrow 0} \frac{\frac{e^{-\alpha\theta} - e^{-\beta\theta}}{\beta e^{-\beta\theta} - \alpha e^{-\alpha\theta}} - \theta}{(e^{-\alpha\theta} - e^{-\beta\theta}) \cdot \theta} = \\ &= \lim_{\theta \rightarrow 0} \frac{-\theta(-\beta^2 e^{-\beta\theta} + \alpha^2 e^{-\alpha\theta})}{\theta(e^{-\alpha\theta} - e^{-\beta\theta}) + (e^{-\alpha\theta} - e^{-\beta\theta})} = \\ &= \lim_{\theta \rightarrow 0} \frac{-\theta(\beta^3 e^{-\beta\theta} - \alpha^3 e^{-\alpha\theta}) - (-\beta^2 e^{-\beta\theta} + \alpha^2 e^{-\alpha\theta})}{(\beta e^{-\beta\theta} - \alpha e^{-\alpha\theta}) + \theta(\beta e^{-\beta\theta} - \alpha e^{-\alpha\theta}) - \alpha e^{-\alpha\theta} + \beta e^{-\beta\theta}} = \\ &= \frac{\beta^2 - \alpha^2}{2(\beta - \alpha)} = \frac{\beta + \alpha}{2}, \end{aligned} \quad (4)$$

i.e.

$$\lim_{\theta \rightarrow 0} \frac{\partial \ln L(x; \theta, \alpha, \beta)}{\partial \theta} = \frac{\beta + \alpha}{2} - \bar{X}.$$

Analogously, for  $\theta \rightarrow \infty$  we have  $\lim_{\theta \rightarrow \infty} \frac{\partial \ln L(x; \theta, \alpha, \beta)}{\partial \theta} = -\bar{X}$ . If  $0 < \bar{X} < \frac{\beta + \alpha}{2}$ , then there exists a stationary point, say  $\theta^*$ , such that

$$\left. \frac{\partial \ln L(x; \theta, \alpha, \beta)}{\partial \theta} \right|_{\theta=\theta^*} = 0.$$

Let us check that at this point the second derivative is negative. We write the second derivative in the form

$$\begin{aligned} & \frac{\partial^2 \ln L(x; \theta, \alpha, \beta)}{\partial \theta^2} = \\ &= -\frac{n}{\theta^2} - \frac{n(\alpha^2 e^{-\alpha\theta} - \beta^2 e^{-\beta\theta})(e^{-\alpha\theta} - e^{-\beta\theta})}{(e^{-\alpha\theta} - e^{-\beta\theta})^2} - \frac{(\beta e^{-\beta\theta} - \alpha e^{-\alpha\theta})(\beta e^{-\theta\beta} - \alpha e^{-\theta\alpha})}{(e^{-\alpha\theta} - e^{-\beta\theta})^2} = \\ &= \frac{n \cdot e^{-\theta(\beta+\alpha)} \cdot (\beta - \alpha)^2}{(e^{-\alpha\theta} - e^{-\beta\theta})^2} - \frac{n}{\theta^2} = n \left( \frac{e^{-\theta(\beta+\alpha)} \cdot (\beta - \alpha)^2}{(e^{-\alpha\theta})^2 (1 - e^{-\theta(\beta-\alpha)})^2} - \frac{1}{\theta^2} \right). \end{aligned}$$

If we use the elementary inequality  $1 - e^x < -x$ , then for  $x \neq 0$  and  $\frac{e^{-\theta\beta}}{e^{-\theta\alpha}} < 1$  we obtain

$$\frac{\partial^2 \ln L(x; \theta, \alpha, \beta)}{\partial \theta^2} < n \left( \frac{e^{-\theta(\beta+\alpha)} \cdot (\beta - \alpha)^2}{(e^{-\alpha\theta})^2 \cdot \theta^2 (\beta - \alpha)^2} - \frac{1}{\theta^2} \right) = n \left( \frac{e^{-\theta\beta}}{e^{-\theta\alpha} \cdot \theta^2} - \frac{1}{\theta^2} \right) < 0.$$

Therefore equation (3) has a unique solution  $\theta^*$ ,  $0 < \theta^* < \infty$  and it is unique at a maximum point. A maximum likelihood estimate  $\hat{\theta}$  for the parameter  $\theta$  is written in the following form

$$\hat{\theta} = \begin{cases} \theta^* & \text{if } 0 < \bar{X} < \frac{\beta + \alpha}{2} \\ \text{not exist} & \text{if } \bar{X} \geq \frac{\beta + \alpha}{2}. \end{cases}$$

In view of the assertion of Lemma, the following theorem is valid.

**Theorem.** Assume that we have the sample  $X_1, X_2, \dots, X_n$  of random values which are distributed according to law (1) where  $\alpha$ ,  $\beta$  and  $\theta$  are the unknown parameters.

If  $0 < \bar{X} < \frac{\beta + \alpha}{2}$ , then the maximum likelihood estimate for  $\theta$  exists and is the unique root of the equation

$$\frac{1}{\theta} - (\beta e^{-\beta\theta} - \alpha e^{-\alpha\theta})(e^{-\alpha\theta} - e^{-\beta\theta})^{-1} - \bar{X} = 0. \tag{5}$$

This estimate is consistent and asymptotically effective.

Also,  $\alpha = X_{(1)} = \min(X_1, \dots, X_n)$ ,  $\beta = X_{(n)} = \max(X_1, \dots, X_n)$ .

**Remark.** If  $\alpha = 0$  and  $\beta = \infty$ , then from (5) we obtain the classical case

$$\theta = \frac{1}{\frac{1}{n} \sum_{i=1}^n x_i}.$$

Computerized simulation of an exponential distribution with parameters  $\alpha = 1$ ,  $\beta = 2$  and  $\theta = 2$  was carried out. For the sample of size  $n = 10000$ , we obtained the estimate  $\bar{X} = 1,344$ ,  $\hat{\theta} = 2,005$ .

*მათემატიკა*

## მოკვეთილი ექსპონენციალური განაწილების მაქსიმალური დასაჯერობის შეფასების შესახებ

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