

# WEAK ASYMPTOTICS OF THE BAYES ESTIMATOR OF THE RELIABILITY FUNCTION IN THE KOZIOL-GREEN MODEL

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**ABSTRACT.** The Bayesian estimator (assuming conjugate prior density) of the reliability function in the Koziol-Green model with exponential distribution is considered, and the weak convergence of the estimator process to the Gaussian process in  $C[0, \infty]$  is proved.

## 1. INTRODUCTION

We are dealing with the Koziol-Green model of random censorship with an exponential distribution, the model, which is described in many reliability and survival data literature. We continue the works [3] and [4], where Bayesian estimators and above all their asymptotic properties and Bayesian risk are studied. [6] deals with the asymptotic properties of the maximum likelihood estimator, in [5] a test of fit with the Koziol-Green model is given. A review of models and methods of estimation is found in [7].

Let us have two independent random samples:  $X_1, X_2, \dots, X_n$  (failure times) distributed  $\text{Exp}(\lambda)$  ( $\lambda > 0$ ), the exponential distribution with the reliability function

$$R(t) = e^{-\lambda t}, \quad t \geq 0,$$

and  $T_1, T_2, \dots, T_n$  (time censors) distributed  $\text{Exp}(\lambda\gamma)$  ( $\gamma > 0$ ). The reliability function of  $T$  equals  $R^\gamma$  and satisfies thus the assumption of the Koziol-Green model — to be a power of that of  $X$ . Instead of  $\gamma$  we can consider the parameter  $p = (1 + \gamma)^{-1}$ .

Due to the censoring by  $T_j$ 's, the information available is fully contained in

$$W_j = \min(X_j, T_j) \quad \text{and} \quad I_j = \chi_{[X_j \leq T_j]}.$$

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If we denote the total time of an experiment and the number of uncensored items by

$$W = \sum_{j=1}^n W_j \quad \text{and} \quad I = \sum_{j=1}^n I_j,$$

the likelihood function for the parameters of the model,  $\lambda$  and  $\gamma$ , can be written as

$$L(\lambda, \gamma; W, I) = \lambda^n e^{-\lambda(1+\gamma)W} \gamma^{n-I}, \quad \lambda > 0, \gamma > 0.$$

We choose a prior density

$$k_{a,r,s}(\lambda, \gamma) = \frac{a^{r+s}}{\Gamma(r)\Gamma(s)} \lambda^{r+s-1} e^{-\lambda(1+\gamma)} \gamma^{s-1}, \quad \lambda, \gamma > 0,$$

from the natural conjugate system for  $L$ ,  $\mathcal{K} = \{k_{a,r,s}, a, r, s > 0\}$ . Then the Bayesian estimators of  $\lambda$  and  $R(t)$  are

$$\hat{\lambda} = \frac{I+r}{W+a} \quad \text{and} \quad \widehat{R(t)} = \left( \frac{W+a}{W+a+t} \right)^{I+r},$$

respectively.

Given  $\lambda, \gamma$ , the convergences

$$\hat{\lambda} \xrightarrow[n \rightarrow \infty]{\text{a. s.}} \lambda, \quad \sqrt{n}(\hat{\lambda} - \lambda) \xrightarrow[n \rightarrow \infty]{\mathcal{D}} N(0, \lambda^2/p)$$

and

$$(1) \quad \sqrt{n} \left( \widehat{R(1)} - R(1) \right) \xrightarrow[n \rightarrow \infty]{\mathcal{D}} N(0, (R(1) \ln R(1))^2/p).$$

are proved in [4]. In this paper we will consider  $\widehat{R(t)}$  and  $R(t)$  as the trajectories of continuous processes  $\widehat{R}$  and  $R$  and extend (1) to the process  $\sqrt{n}(\widehat{R} - R)$ .

## 2. WEAK CONVERGENCE OF $\widehat{R}$

**Lemma.** *Let  $U(t) = (I+r) \ln(1+t/(a+W)) - \lambda t$ . Then*

$$(2) \quad \sqrt{n} e^{-\lambda t} U(t) \xrightarrow[n \rightarrow \infty]{\mathcal{D}} Z(t) \quad \text{in } C([0, \infty]),$$

where  $Z$  is a zero mean Gaussian process with  $\text{cov}(Z(s), Z(t)) = st e^{-\lambda(s+t)} \lambda^2/p$ .

*Proof.* We use Taylor expansion of  $\ln(1+x)$  at  $x=0$  to express

$$\ln \left( 1 + \frac{t}{W+a} \right) = \frac{t}{W+a} + Q_1(t),$$

where

$$(3) \quad Q_1(t) = -\frac{1}{2(1+\xi)^2} \left( \frac{t}{W+a} \right)^2 \quad \text{for some } \xi(t) \in [0, t/(W+a)].$$

Since

$$(4) \quad |\sqrt{n} e^{-\lambda t} (I+r) Q_1(t)| \leq \frac{1}{2} t^2 e^{-\lambda t} \frac{(I+r)\sqrt{n}}{(W+a)^2} \leq K \frac{(I+r)\sqrt{n}}{(W+a)^2} \xrightarrow{n \rightarrow \infty} 0 \quad \text{a. s.,}$$

where  $K > 0$  is a constant depending on  $\lambda$  only, both

$$U'(t) = \sqrt{n} e^{-\lambda t} \left( \frac{I+r}{W+a} t - \lambda t \right) = t e^{-\lambda t} \sqrt{n} \left( \frac{I+r}{W+a} - \lambda \right)$$

and  $\sqrt{n} e^{-\lambda t} U(t) = U'(t) + \sqrt{n} e^{-\lambda t} (I+r) Q_1(t)$  converge to equally distributed processes.

We know that  $\sqrt{n}[(I+r)/(W+a) - \lambda] \rightarrow N(0, \lambda^2/p)$  in distribution as  $n \rightarrow \infty$  and hence for every  $t_1, t_2, \dots, t_k > 0$  finite-dimensional distributions

$$(U'(t_1), \dots, U'(t_k)) \xrightarrow[n \rightarrow \infty]{\mathcal{D}} N_k(\mathbf{0}, \text{var}(Z(t_1), \dots, Z(t_k))).$$

Let  $\varepsilon > 0$  be given.

$$\begin{aligned} \mathbb{P}\left[\sup_{|s-t| \leq \delta} |U'(t) - U'(s)| > \varepsilon\right] &= \mathbb{P}\left[\sup_{|s-t| \leq \delta} |t e^{-\lambda t} - s e^{-\lambda s}| \sqrt{n} \left| \frac{I+r}{W+a} - \lambda \right| > \varepsilon\right] \leq \\ &\leq \mathbb{P}\left[\delta \sqrt{n} \left| \frac{I+r}{W+a} - \lambda \right| > \varepsilon\right] \xrightarrow{n \rightarrow \infty} 2\Phi\left(\frac{-\varepsilon}{\delta \lambda / \sqrt{p}}\right), \end{aligned}$$

where  $\Phi$  is the distribution function of  $N(0, 1)$ . We can conclude that

$$\lim_{\delta \rightarrow 0+} \limsup_{n \rightarrow \infty} \mathbb{P}\left[\sup_{|s-t| \leq \delta} |U'(t) - U'(s)| > \varepsilon\right] = 0$$

(Stone's tightness condition). Thus we have the convergence (2) in  $C([0, T])$  for every  $T > 0$  (e. g. [8], Theorem III.5.7, or [1], Chapter 8). The extension to  $C([0, \infty])$  can be made through the condition on tightness at  $\infty$  (based on [1], Theorem 4.2, an example of application [2], Theorem 6.2.1) which reads

$$\lim_{s \rightarrow \infty} \limsup_{n \rightarrow \infty} \mathbb{P}\left[\sup_{t \geq s} |U'(t) - U'(s)| > \varepsilon\right] = 0 \quad \text{for every } \varepsilon > 0,$$

and this is satisfied here since

$$\begin{aligned} \mathbb{P}\left[\sup_{t \geq s} |t e^{-\lambda t} - s e^{-\lambda s}| \sqrt{n} \left| \frac{I+r}{W+a} - \lambda \right| > \varepsilon\right] &\leq \\ &\leq \mathbb{P}\left[s e^{-\lambda s} \sqrt{n} \left| \frac{I+r}{W+a} - \lambda \right| > \varepsilon\right] \xrightarrow{n \rightarrow \infty} 2\Phi\left(\frac{-\varepsilon}{s e^{-\lambda s} \lambda / \sqrt{p}}\right), \end{aligned}$$

assuming  $s > 1/\lambda$ . □

**Proposition.** *For the Bayesian estimator  $\hat{R}(t) = (\frac{W+a}{W+a+t})^{I+r}$  of the reliability function  $R(t) = e^{-\lambda t}$  we have*

$$\sqrt{n}(\hat{R} - R) \xrightarrow[n \rightarrow \infty]{\mathcal{D}} Z \quad \text{in } C([0, \infty]),$$

where  $Z$  is a zero mean Gaussian process with  $\text{cov}(Z(s), Z(t)) = st e^{-\lambda(s+t)} \lambda^2/p$ .

*Proof.* Using Taylor expansion of  $e^x$  at  $x = 0$  we obtain

$$(5) \quad \sqrt{n}(\widehat{R}(t) - R(t)) = \sqrt{n}e^{-\lambda t}(e^{-U(t)} - 1) = \sqrt{n}e^{-\lambda t}(1 - U(t) + Q_2(t) - 1),$$

where

$$Q_2(t) = e^{\xi(t)} U^2(t)/2 \quad \text{for some } \xi(t) \text{ between } 0 \text{ and } -U(t).$$

Let  $\delta(t) = 1$  for  $0 \leq t \leq 1$  and  $\delta(t) = t$  for  $t \geq 1$ . Denoting

$$A(t) = \delta^4(t) e^{-\lambda t + \xi(t)}, \quad B(t) = n^{1/4} \delta^{-2}(t) |U(t)| / \sqrt{2}$$

we can write for  $\varepsilon > 0$

$$(6) \quad \mathbb{P}\left[\sup_{t \in [0, \infty]} \sqrt{n} e^{-\lambda t} |Q_2(t)| > \varepsilon\right] \leq \mathbb{P}\left[\sup_{t \in [0, \infty]} A(t) \cdot \sup_{t \in [0, \infty]} B^2(t) > \varepsilon\right].$$

If  $-U(t) < 0$  then  $A(t) \leq \delta^4(t) e^{-\lambda t} \leq K_1$ , for some  $K_1 > 0$  independent of  $t$  and  $n$ . If  $-U(t) > 0$  then  $A(t) \leq \delta^4(t) e^{-\lambda t - U(t)} = \delta^4(t) (1 + t/(W + a))^{-(I+r)}$ , which is bounded by 1 on  $[0, 1]$ ; function  $t^4 (1 + t/(W + a))^{-(I+r)}$ ,  $t > 0$ , achieves (assuming  $I + r > 4$ ) its maximum at  $t = 4(W + a)/(I + r - 4)$  acquiring the value

$$\left(4 \frac{W + a}{I + r - 4}\right)^4 \left(1 + \frac{4}{I + r - 4}\right)^{-(I+r)} \xrightarrow[n \rightarrow \infty]{\text{a. s.}} \frac{4^4}{\lambda^4} e^{-4}.$$

Thus we have proved existence of constant  $K > \max(K_1, 1, 4^4(\lambda e)^{-4})$  which satisfies  $\mathbb{P}[\sup_{t \in [0, \infty]} A(t) > K] \rightarrow 0$  as  $n \rightarrow \infty$ .

The process  $B(t)$  can be treated similarly as in (3), (4). We have

$$\begin{aligned} |B(t)| &= \frac{1}{\sqrt{2}} n^{1/4} \left| t \delta^{-2}(t) \left( \frac{I + r}{W + a} - \lambda \right) + \delta^{-2}(t) (I + r) Q_1(t) \right| \leq \\ &\leq \frac{1}{\sqrt{2}} n^{-1/4} \left( \sqrt{n} \left| \frac{I + r}{W + a} - \lambda \right| + \frac{(I + r) \sqrt{n}}{2(W + a)^2} \right) \xrightarrow[n \rightarrow \infty]{\mathbb{P}} 0, \end{aligned}$$

hence  $\mathbb{P}[\sup_{t \in [0, \infty]} B^2(t) > \varepsilon] \rightarrow 0$  as  $n \rightarrow \infty$ .

Returning to (6) we have  $\sup_{t \in [0, \infty]} \sqrt{n} e^{-\lambda t} Q_2(t) \rightarrow 0$  in probability. Therefore both (5) and  $-\sqrt{n} e^{-\lambda t} U(t)$  converge to the process  $-Z$  (according to Lemma), with the same distribution as process  $Z$ .  $\square$

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