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## Local Asymptotic Normality of Statistical Experiments in an Inhomogeneous Competing Risks Model

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**Abstract.** In this paper we consider an inhomogeneous competing risks model. For the likelihood ratio statistics (LRS), proved the theorem on the locally asymptotic normality of statistical experiment.

**Keywords:** competing risks model, random censoring, local asymptotic normality, likelihood ratio statistics.

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### 1. Introduction and preliminaries

In a inhomogeneous competing risks model (CRM) it's interesting to investigate the independent random variables (r.v.)  $\{X_m, m \geq 1\}$  with a distribution function (d.f.)  $\{H_m(x; \theta), m \geq 1\}$  with values in a measurable space  $(\mathcal{X}_m, \mathcal{B}_m)$ ,  $m \geq 1$ , where  $\mathcal{X}_m$  is set of possible values of r.v.  $X_m$  and  $\mathcal{B}_m = \sigma(\mathcal{X}_m)$ . D.f. of r.v.  $X_m$  also depends on the scalar parameter  $\theta \in \Theta$ , where  $\Theta$ -parametrical space; is open set in  $R^1$ . In CRM with r.v.  $X_m$  pairwise disjoint and forming a complete group of events  $\{A_m^{(1)}, \dots, A_m^{(k)}\}$  observed. We observe the sample of size  $n$ :  $\left\{ \left( X_m; \delta_m^{(0)}, \delta_m^{(1)}, \dots, \delta_m^{(k)} \right), m = 1, \dots, n \right\}$ , where  $\delta_m^{(i)} = I(A_m^{(i)})$  is indicator of events  $A_m^{(i)}$ . Let  $H_m^{(i)}(x; \theta) = P_\theta(X_m < x, \delta_m^{(i)} = 1)$  are subdistributions such that  $H_m^{(1)}(x; \theta) + \dots + H_m^{(n)}(x; \theta) = H(x; \theta)$  for all  $(x; \theta) \in R^1 \times \Theta$ . By  $h_m^{(i)}$  we define density of subdistributions  $H_m^{(i)}$ . Then note that there is a density  $h_m^{(1)}(x; \theta) + \dots + h_m^{(k)}(x; \theta) = \frac{\partial H_m(x; \theta)}{\partial x} = h_m(x; \theta)$  for all  $(x; \theta) \in R^1 \times \Theta$ . We define  $\Lambda_m^{(i)}(x; \theta) = \int_{-\infty}^x (1 - H_m(u; \theta))^1 dH_m(u; \theta)$  integral and  $\lambda_m^{(i)}(x; \theta)$  density of failure rate of the  $i$ -th risk. Then

$$\lambda_m^{(1)}(x; \theta) + \dots + \lambda_m^{(k)}(x; \theta) = \frac{h_m(x; \theta)}{1 - H_m(x; \theta)},$$

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for all  $(x; \theta) \in R^1 \times \Theta$ .

We introduce the functional of the exponential intensity function [1, 3]:

$$1 - F_m^{(i)}(x; \theta) = \exp(-\Lambda_m^{(i)}(x; \theta)), \quad i = 1, \dots, k; \quad m = 1, \dots, n.$$

Note that this functional has all the properties of a d.f. [1].

By  $f_m^{(i)}(x; \theta)$  we define density of  $F_m^{(i)}(x; \theta)$ . It is easy to see that

$$f_m^{(i)}(x; \theta) = \exp \left\{ \Lambda^{(i)}(x; \theta) \right\} \frac{h_m^{(i)}(x; \theta)}{1 - H_m(x; \theta)}, \quad i = 1, \dots, k.$$

Denote by  $\mathcal{F}_t^{(i)} = \sigma \left\{ X_m I(X_m \leq t), \delta_m^{(i)} I(X_m \leq t), 1 \leq m \leq n \right\}$ , flow of  $\sigma$ -algebra, generated by pairs of observations  $\left\{ (X_m, \delta_m^{(i)}), 1 \leq m \leq n \right\}$ ,  $i = \overline{1, k}$ . Let define a sequence of martingale processes for  $m = \overline{1, n}$  and  $i = \overline{1, k}$ :

$$\mu_m^{(i)}(t) = I(X_m \leq t, \delta_m^{(i)} = 1) - \int_{-\infty}^t I(X_m > s) d\Lambda^{(i)}(s; \theta_0), \quad t \geq 0,$$

where  $\theta_0$  is true value of parameter  $\theta$ ,  $\theta_0 \in \Theta$ . These martingales have zero mea

$$\begin{aligned} E_{\theta_0} \mu_m^{(i)} &= H_m^{(i)}(t; \theta_0) - \int_0^t (1 - H(s; \theta_0)) d\Lambda_m^{(i)}(s; \theta_0) = \\ &= H_m^{(i)}(t; \theta_0) - \int_0^t \frac{1 - H_m(s; \theta_0)}{1 - H_m(s; \theta_0)} dH_m^{(i)}(s; \theta_0) = H_m^{(i)}(t; \theta_0) - H_m^{(i)}(t; \theta_0) = 0, \end{aligned}$$

$i = \overline{1, k}$ ;  $m = \overline{1, n}$ . They are members of class  $\mathcal{M}^2(\mathcal{F}_t^{(i)})$ , i.e. square-integrable martingales with a predictable quadratic characteristics [4]:

$$\langle \mu_m^{(i)}, \mu_m^{(j)} \rangle = \begin{cases} \Lambda^{(i)}(t), & i = j, \\ 0, & i \neq j. \end{cases}$$

Therefore, these martingales are orthogonal. According to this

$$\sum_{m=1}^n \mu_m^{(i)}(t) \in \mathcal{M}^2(\mathcal{F}_t^{(i)}), \quad \sum_{i=1}^k \sum_{m=1}^n \mu_m^{(i)}(t) \in \mathcal{M}^2(\mathcal{F}_t),$$

where  $\mathcal{F}_t = \bigcap_{i=1}^k \mathcal{F}_t^{(i)}$ . To prove local asymptotic normality (LAN) we need some regularity conditions for  $f_m^{(i)}(x; \theta)$ :

(C1) The supports  $N_{f_m^{(i)}} = \{x : f_m^{(i)}(x; \theta) > 0\}$  are independent from  $\theta$ ,  $i = \overline{1, k}$ ;  $m = \overline{1, n}$ ;

(C2) There exist and are continuous for all  $x$  derivatives  $\frac{\partial^l}{\partial \theta^l} f_m^{(i)}(x; \theta)$ ,  $l = 1, 2$ ;  $i = \overline{1, k}$ ;  $m = \overline{1, n}$ ;

(C3) Fisher information  $\sum_{m=1}^n I_m(\theta) = \sum_{m=1}^n \sum_{i=1}^k I_m^{(i)}(\theta)$  is finite and continuous in  $\theta$  and positive at the point  $\theta_0$ , where

$$I_m^{(i)}(\theta) = \int_{-\infty}^{\infty} \left( \frac{\partial}{\partial \theta} \log \lambda^{(i)}(s; \theta) \right)^2 dH_m^{(i)}(s; \theta).$$

Such a representation of Fisher information in terms of intensity density  $\lambda_m^{(i)}(x; \theta)$  was established by the authors of [2]. Let  $\varphi(n) = \left( \sum_{m=1}^n I_m(\theta_0) \right)^{-1/2}$ . Let for every  $u \in R^1$ ,  $\theta_n = \theta_0 + u\varphi(n) \in \Theta$ . We have a likelihood ratio statistics (LRS) of the sample  $Z^{(n)} = (X_1, \dots, X_n)$ :  $dQ_{\theta_n}^{(n)}(Z^{(n)})/dQ_{\theta_0}^{(n)}(Z^{(n)})$ , where

$$Q_{\theta}^{(n)}(Z^{(n)}) = p_n(Z^{(n)}; \theta) = \prod_{m=1}^n \prod_{i=1}^k \left\{ f_m^{(i)}(X_m; \theta) \cdot \prod_{\substack{j=1 \\ j \neq i}}^k [1 - F_m^{(j)}(X_m; \theta)]^{\delta_m^{(j)}} \right\}.$$

## 2. LAN for LRS.

The following theorem asserts LAN for LRS.

**Theorem.** *Under regularity conditions (C1)–(C3) for LRS we have representation*

$$\frac{dQ_{\theta_n}^{(n)}(Z^{(n)})}{dQ_{\theta_0}^{(n)}(Z^{(n)})} = \exp \left\{ u\Delta_n(\theta_0) + \frac{u^2}{2} + R_n(u) \right\}, \tag{1}$$

where  $\Delta_n = \varphi(n) \sum_{m=1}^n \sum_{i=1}^k \int_{-\infty}^{\infty} \left( \frac{\partial}{\partial \theta} \log \lambda_m(s; \theta_0 + u\varphi(n)\gamma) \right) d\mu_m^{(i)}(s)$ ,  $0 < \gamma < 1$ ,  $L(\Delta_n/Q_{\theta_0}^{(n)}) \rightarrow N(0; 1)$  and  $R_n(u) \xrightarrow{Q_{\theta_0}^{(n)}} 0$  at  $n \rightarrow \infty$ .

*Proof of Theorem.* Let represent logarithm of LRS in terms of martingale-processes as follows:

$$\begin{aligned} L_n(u) &= \log \left\{ \frac{dQ_{\theta_n}^{(n)}(Z^{(n)})}{dQ_{\theta_0}^{(n)}(Z^{(n)})} \right\} = \log \left[ \frac{p_n(Z^{(n)}; \theta_n)}{p_n(Z^{(n)}; \theta_0)} \right] = \\ &= \sum_{m=1}^n \sum_{i=1}^k \delta_m^{(i)} \log \left\{ \frac{f^{(i)}(X_m; \theta_n) \prod_{j=1}^k (1 - F^{(j)}(X_m; \theta_n))}{f^{(i)}(X_m; \theta_0) \prod_{j=1}^k (1 - F^{(j)}(X_m; \theta_0))} \right\} = \\ &= \sum_{m=1}^n \sum_{i=1}^k \delta_m^{(i)} \log \left[ \frac{\lambda_m^{(i)}(X_m; \theta_n)}{\lambda_m^{(i)}(X_m; \theta_0)} \right] + \sum_{m=1}^n \log \left[ \frac{1 - H_m(X_m; \theta_n)}{1 - H_m(X_m; \theta_0)} \right] = \\ &= \sum_{m=1}^n \sum_{i=1}^k \delta_m^{(i)} \log \left[ \frac{\lambda_m^{(i)}(X_m; \theta_n)}{\lambda_m^{(i)}(X_m; \theta_0)} \right] + \sum_{m=1}^n (\Lambda_m(X_m; \theta_n) - \Lambda_m(X_m; \theta_0)) = \tag{2} \\ &= \sum_{m=1}^n \sum_{i=1}^k \int_{-\infty}^{\infty} \log \left[ \frac{\lambda_m^{(i)}(s; \theta_n)}{\lambda_m^{(i)}(s; \theta_0)} \right] d\mu_m^{(i)}(s) + \\ &\quad + \left\{ \sum_{m=1}^n \sum_{i=1}^k \int_{-\infty}^{\infty} \log \left[ \frac{\lambda_m^{(i)}(s; \theta_n)}{\lambda_m^{(i)}(s; \theta_0)} \right] I(X_m > s) d\Lambda_m^{(i)}(s; \theta_0) - \right. \\ &\quad \left. - \sum_{m=1}^n (\Lambda_m(X_m; \theta_n) - \Lambda_m(X_m; \theta_0)) \right\} = A_n(u) + R_n(u). \end{aligned}$$

By the mean value theorem, from (2) we have

$$\begin{aligned} A_n(u) &= \sum_{m=1}^n \sum_{i=1}^k \int_{-\infty}^{\infty} \left[ \log \lambda_m^{(i)}(s; \theta_0 + u\varphi(n)) - \log \lambda_m^{(i)}(s; \theta_0) \right] d\mu_m^{(i)}(s) = \\ &= u\varphi(n) \sum_{m=1}^n \sum_{i=1}^k \int_{-\infty}^{\infty} \frac{\partial}{\partial \theta} \log \lambda_m^{(i)}(s; \theta_0 + u\varphi(n)\gamma) d\mu_m^{(i)}(s) = u\Delta_n(\theta_0), \end{aligned}$$

where

$$\Delta_n(\theta_0) = \varphi(n) \sum_{m=1}^n \sum_{i=1}^k \int_{-\infty}^{\infty} \frac{\partial}{\partial \theta} \log \lambda_m^{(i)}(s; \theta_0 + u\varphi(n)\gamma) d\mu_m^{(i)}(s). \quad (3)$$

Integrating function in (3) is a continuous real-valued function of  $s$ , hence it is a predictable and is a square-integrable martingale:  $\Delta_n \in \mathcal{M}^2(\mathcal{F}_t)$ . We need to establish that

$$\mathcal{L} \left( \Delta_n(\theta_0) / Q_{\theta_0}^{(n)} \right) \rightarrow N(0; 1). \quad (4)$$

It is easy to see that  $E_{\theta_0} \Delta_n(\theta_0) = 0$  (since  $E_{\theta_0} \mu_m^{(i)}(t) = 0$ ). Moreover, the quadratic variation of the martingale  $\Delta_n(\theta_0)$  is

$$\begin{aligned} &< \Delta_n(\theta_0), \Delta_n(\theta_0) > = \\ &>= \varphi^2(n) \sum_{m=1}^n \sum_{i=1}^k \int_{-\infty}^{\infty} \left( \frac{\partial \log \lambda_m^{(i)}(s; \theta_0 + u\varphi(n)\gamma)}{\partial \theta} \right)^2 (1 - H_m(s; \theta_0)) d\Lambda_m^{(i)}(s; \theta_0) = \\ &= \left( \sum_{m=1}^n I_m(\theta_0) \right)^{-1} \sum_{m=1}^n I_m(\theta_0 + u\varphi(n)\gamma) \rightarrow 1, \quad n \rightarrow \infty, \end{aligned}$$

in view of condition (C3). Consider the Lindberg conditions

$$\begin{aligned} &\varphi^2(n) \sum_{m=1}^n \sum_{i=1}^k \int_{-\infty}^{\infty} \left( \frac{\partial \log \lambda_m^{(i)}(s; \theta_0 + u\varphi(n)\gamma)}{\partial \theta} \right)^2 \times \\ &\quad \times I \left( \left| \frac{\partial \log \lambda_m^{(i)}(s; \theta_0 + u\varphi(n)\gamma)}{\partial \theta} \right| > \varepsilon\varphi(n) \right) d < \mu_m^{(i)}(t), \mu_m^{(j)}(t) > = \\ &= \varphi^2(n) \sum_{m=1}^n \sum_{i=1}^k \int_{-\infty}^{\infty} \left( \frac{\partial \log \lambda_m^{(i)}(s; \theta_0 + u\varphi(n)\gamma)}{\partial \theta} \right)^2 \times \\ &\quad \times I \left( \left| \frac{\partial \log \lambda_m^{(i)}(s; \theta_0 + u\varphi(n)\gamma)}{\partial \theta} \right| > \varepsilon\varphi(n) \right) (1 - H_m(s; \theta_0)) d\Lambda_m^{(i)}(s; \theta_0) = \\ &= \varphi^2(n) \sum_{m=1}^n \sum_{i=1}^k \int_{-\infty}^{\infty} \left( \frac{\partial \log \lambda_m^{(i)}(s; \theta_0 + u\varphi(n)\gamma)}{\partial \theta} \right)^2 \times \\ &\quad \times I \left( \left| \frac{\partial \log \lambda_m^{(i)}(s; \theta_0 + u\varphi(n)\gamma)}{\partial \theta} \right| > \varepsilon\varphi(n) \right) dH_m^{(i)}(s; \theta_0) \rightarrow 0, \quad n \rightarrow \infty, \end{aligned}$$

where the convergence to zero of the integral follows from the requirement that the Fisher information  $\sum_{m=1}^n I_m(\theta)$  is finite in view of condition (C3). Consequently, weak convergence (4)

follows from the central limit theorem for martingales [5]. Consider second addition in (2). Second addition in (2) converges in probability to  $-\frac{u^2}{2}$ :  $R_n(u) \xrightarrow{Q_{\theta_0}^{(n)}} -\frac{u^2}{2}$ ,  $n \rightarrow \infty$ . Therefore, it remains to show that

$$R_n(u) + \frac{u^2}{2} \xrightarrow{Q_{\theta_0}^{(n)}} 0, \quad n \rightarrow \infty. \tag{5}$$

We investigate second addition inside the large curly bracket in (2). After elementary transformations, we have

$$\begin{aligned} & \sum_{m=1}^n \sum_{i=1}^k \int_{-\infty}^{\infty} \log \left[ \frac{\lambda_m^{(i)}(s; \theta_n)}{\lambda_m^{(i)}(s; \theta_0)} \right] I(X_m > s) d\Lambda_m^{(i)}(s; \theta_0) - \sum_{m=1}^n (\Lambda_m(X_m; \theta_n) - \Lambda_m(X_m; \theta_0)) = \\ & = \sum_{m=1}^n \sum_{i=1}^k \int_{-\infty}^{\infty} \log \left[ 1 + \frac{\lambda_m^{(i)}(s; \theta_n) - \lambda_m^{(i)}(s; \theta_0)}{\lambda_m^{(i)}(s; \theta_0)} \right] I(X_m > s) d\Lambda_m(s; \theta_0) + \\ & \quad + \sum_{m=1}^n \int_{-\infty}^{+\infty} (\Lambda_m(X_m; \theta_n) - \Lambda_m(X_m; \theta_0)) dI(X_m > s) = \\ & = \sum_{m=1}^n \sum_{i=1}^k \int_{-\infty}^{\infty} \log \left( 1 + \frac{\lambda_m^{(i)}(s; \theta_n) - \lambda_m^{(i)}(s; \theta_0)}{\lambda_m^{(i)}(s; \theta_0)} \right) - \\ & \quad - \int_{-\infty}^{\infty} I(X_m > s) \frac{\lambda_m^{(i)}(s; \theta_n) - \lambda_m^{(i)}(s; \theta_0)}{\lambda_m^{(i)}(s; \theta_0)} d\Lambda_m(s; \theta_0) = \\ & = -\frac{1}{2} \sum_{m=1}^n \sum_{i=1}^k \int_{-\infty}^{\infty} \left[ \frac{\lambda_m^{(i)}(s; \theta_0 + u\varphi(n)\gamma^*) - \lambda_m^{(i)}(s; \theta_0)}{\lambda_m^{(i)}(s; \theta_0 + u\varphi(n)\gamma^*)} \right] I(X_m > s) d\Lambda_m^{(i)}(s; \theta_0) = \\ & = -\frac{u^2 \varphi^2(n)}{2} \int_{-\infty}^{\infty} \left( \frac{\partial}{\partial \theta} \log \lambda(s; \theta_0 + u\varphi(n)\gamma^*) \right)^2 I(X_m > s) d\Lambda_m^{(i)}(s; \theta_0) = \\ & = -\frac{u^2}{2} \left( \sum_{m=1}^n I_m(\theta_0) \right)^{-1} \sum_{m=1}^n I_m(\theta_0 + u\varphi(n)\gamma^*) (1 + o_p(1)) \rightarrow -\frac{u^2}{2}, \quad n \rightarrow \infty, \end{aligned} \tag{6}$$

where  $0 < \gamma^* < 1$ . Now (6) implies (5). The theorem is proved. □

Mutual contiguity of probability measures  $Q_{\theta_n}^{(n)}$  and  $Q_{\theta_0}^{(n)}$  follows from the theorem.

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## **Локальная асимптотическая нормальность статистических экспериментов в неоднородной модели конкурирующих рисков**

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**Аннотация.** В статье рассматривается неоднородная модель конкурирующих рисков. Для статистики отношения правдоподобия доказана теорема о локально асимптотической нормальности статистического эксперимента.

**Ключевые слова:** модель конкурирующих рисков, случайное цензурирование, локальная асимптотическая нормальность, статистика отношения правдоподобия.