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A LIMIT THEOREM FOR FUNCTIONALS OF A POISSON PROCESS
Nguyen van HUU

Abstract: Let μ be a random point measure defined on a locally compact topological space X with countable basis and let μ have the Poisson distribution Q_ν with intensity measure ν . The asymptotic behaviour of the distribution function of the random variable $Z_n(\mu) = \mu(hI_{K_n})$ as the compact subset $K_n \uparrow X$ is considered. This work also deals with the rate of convergence to the limit distribution.

Key words: Stochastic point process, asymptotic normality, intensity measure, exponential trend.

Classification: 60F05

§ 1. Introduction. Poisson processes form an important class of point processes. Many interesting problems of statistical analysis of Poisson processes on the line have been considered in [1] by D.R. Cox and P.A.W. Lewis and (on more general spaces) by M. Brown [3]. This article is concerned with the limit distribution of certain linear functionals of a Poisson process. Limit theorems will be stated in Section 2. The rate of convergence to the limit distribution function will be considered in Section 3. Section 4 contains some applications of the results obtained in Section 2.

§ 2. Limit theorem. Following [4],[5] let us consider a locally compact topological space X with countable basis. Let $\beta(X)$ be the σ -algebra of Borel subsets of X , $\mathcal{M} = \mathcal{M}(X)$ the family of Radon measures on $(X, \beta(X))$ and \mathcal{H}_c - the class of continuous functions with compact supports defined on X .

Let us also consider a Poisson process Q_ν on X with intensity measure ν ($\nu \in \mathcal{M}(X)$), i.e., a probability distribution defined on the σ -algebra $\mathcal{L}(\mathcal{M})$ generated by all open subsets with respect to the topology of vague convergence ^{x)} with the characteristic functional defined by

$$(1) \hat{Q}_\nu(f) = \int_{\mathcal{M}} \exp(i\mu(f)) Q_\nu(d\mu) = \exp(\nu(e^{if} - 1)), f \in \mathcal{H}_c,$$

where $\nu(f) = \int_X f(x) \nu(dx)$.

Suppose that $\mu \in \mathcal{M}$ is a realization of Q_ν . Usually one can only observe the realization μ on some compact set K of X , as X too large.

Let us consider a statistic of the form

$$(2) Z_K(\mu) = \mu(hI_K),$$

where I_K is the indicator of K , h is some measurable function on X .

The statistic $Z_K(\mu)$ plays an important role for many problems of testing hypothesis and estimating the parameters of Poisson processes. The distribution function of $Z_K(\mu)$ depends on h , K and ν , and is rather complicated, the asympto-

^{x)} $\{\mu_n\}$ is called to be vaguely convergent to μ iff $\mu_n(f) \rightarrow \mu(f)$ for all $f \in \mathcal{H}_c$.

tic theory for such statistics is therefore convenient for practical purposes.

Suppose that K_n is a sequence of compact sets such that $K_n \uparrow X$. Let $Z_n(\mu) = Z_{K_n}(\mu)$, and let us consider the asymptotic behaviour of the distribution law under Q_ν of the random variable of the form

$$(3) \quad Y_n(\mu) = (Z_n(\mu) - a_n) / b_n,$$

where a_n, b_n ($b_n > 0$, for all n) are constants.

Note that

$$Z_n(\mu) = \pm \infty \text{ iff } A = \{x: h(x) = \pm \infty\} \subset K_n \cap \text{supp } \mu.$$

Consequently, letting

$$R_n = \{\mu: \mu(hI_{K_n}) = Z_n(\mu) \neq \pm \infty\}$$

we obtain (see [4])

$$Q_\nu(R_n) = \exp(-\nu(K_n A)).$$

Consequently, for the existence of the limit distribution of $Y_n(\mu)$ the necessary condition is

$$Q_\nu\{Z_n(\mu) = \pm \infty\} = 1 - Q_\nu(R_n) = 1 - \exp(-\nu(AK_n)) \rightarrow 1 - \exp(-\nu(A)) = 0$$

or

$$(4) \quad \nu(AK_n) \rightarrow \nu(A) = \nu\{x: h(x) = \pm \infty\} = 0$$

Therefore, in the following theorems we always assume that (4) is fulfilled.

Let $\lambda_n = \nu(K_n)$, $\nu_K(\cdot)$ be the restricted measure of ν on K , i.e. $\nu_K(A) = \nu(AK)$, for all K and $A \in \mathcal{B}(X)$, and $\nu_n(\cdot) = \nu_{K_n}(\cdot)$, $G_n(t)$ be the characteristic function ch.f. of $Z_n(\mu)$ under Q_ν .

We have the following theorems

Theorem 1. Assume that $\lambda = \nu(X) < \infty$, then

$$(5) \quad G_n(t) \rightarrow \exp(\lambda [g(t) - 1]) = G(t), \text{ say,}$$

holds, where

$$(6) \quad g(t) = \nu(\exp(it h)) / \lambda$$

is the ch.f. of random variable $h(T)$ with T being a random element in X possessing the distribution law $\nu(\cdot) / \lambda$.

Further, $G(t)$ is the ch.f. of the random variable $\gamma \xi$, where γ is some constant, ξ has the Poisson distribution with the mean value λ , iff $h = \gamma$, ν -a.e. $G(t)$ is always the ch.f. of a nonnormal random variable.

The case $\lambda = \infty$ is more interesting.

Theorem 2. Suppose that $\lambda = \infty$. Then the following conditions (i), (ii) are sufficient for the existence of number sequences $\{a_n\}$ and $\{b_n\}$ with $b_n \rightarrow \infty$ such that

$$(7) \quad F_n(y) = Q_{\nu}\{Y_n(\mu) < y\} = Q_{\nu}\{(Z_n(\mu) - a_n) / b_n < y\} \rightarrow F(y)$$

where, here and in the sequel, the convergence is meant in the weak sense.

$$(i) \quad a_n / b_n \lambda_n^{1/2} \rightarrow \alpha \quad (\alpha - \text{finite})$$

$$(ii) \quad P\{(S_n - a_n) / b_n < y\} \rightarrow K(y)$$

where

$$S_n = \sum_{k=1}^{[\lambda_n]} h(x_{nk})$$

is the sum of independent random variables $h(x_{nk})$ with x_{nk} , $k=1, 2, \dots, [\lambda_n]$, being identically distributed independent random elements in X possessing the common distribution law $\nu_n(\cdot) / \lambda_n$ for each n and with $[\lambda_n]$ denoting the entire of λ_n .

Further, F has the form

$$(8) \quad F(y) = K * \phi_{\infty}(y)$$

where $\phi_{\infty}(y)$ is the normal distribution function with mean value zero and variance ∞^2 .

Notice. $\phi_0(y)$ is the distribution function with jump one at zero, whereas $\phi_1(y)$ is redented by $\phi(y)$.

Proof of Theorem 1. It is easy to see that the ch.f. $G_n(t)$ of Z_n under Q_n is defined by

$$(9) \quad \begin{aligned} G_n(t) &= E_{Q_n}(\exp(it, \mu(hI_{K_n}))) = \hat{Q}_n(t) \\ &= \exp(\nu_n(\exp(it) - 1)) = \exp(\lambda_n[g_n(t) - 1]) \end{aligned}$$

with

$$(10) \quad g_n(t) = \nu_n(\exp(it)) / \lambda_n$$

Since $\lambda_n \rightarrow \lambda$ as $K_n \uparrow X$, $g_n(t)$ converges to $g(t)$ and (5) follows from (9).

The second statement of Theorem 1 comes true iff $g(t) = \exp(i\gamma t)$, but this occurs iff $h(x) = \gamma$, ν -a.e..

As to the last statement, let us suppose inversely that $G(t) = \exp(iat - b^2 t^2 / 2)$, then $g(t) = 1 + iat / \lambda - b^2 t^2 / 2\lambda$.

However, the right hand side of this equality is not a ch.f.. This proves the last statement.

Proof of Theorem 2. Let V_n be a Poisson distributed random variable with mean λ_n . For the sake of simplicity we suppose that λ_n is an integer.

$$\text{Put } \bar{A}_n(y) = P\{V_n < y\}$$

$$A_n(y) = P\{(V_n - \lambda_n) / \lambda_n^{1/2} < y\} = \bar{A}_n(y \lambda_n^{1/2} + \lambda_n)$$

It is obvious that $A_n(y) \rightarrow \phi(y)$ since $\lambda_n \rightarrow \lambda = \infty$.

It follows from (9) that

$$(11) \quad G_n(t) = \exp(-\lambda_n) \sum_{k=0}^{\infty} \frac{\lambda_n^k}{k!} [g_n(t)]^k = \\ = \int_0^{\infty} [g_n(t)]^y d\bar{A}_n(y) = \int_{-\infty}^{\infty} [g_n(t)]^y \lambda_n^{1/2 + \lambda_n} dA_n(y)$$

Let $k(t)$, $f(t)$ be the ch.f. corresponding to K , F and $H_n(t)$, $k_n(t)$ be the ch.f. of Y_n , $(S_n - a_n)/b_n$, respectively. It is easy to see from (11) that

$$H_n(t) = \exp(-ita_n/b_n) G_n(t/b_n) = \\ = \int_{-\infty}^{\infty} [k_n(t)]^{1+y/\lambda_n^{1/2}} \exp(iya_n/b_n \lambda_n^{1/2}) dA_n(y) \rightarrow \\ \rightarrow k(t) \int_{-\infty}^{\infty} \exp(i\alpha ty) d\phi(y) = k(t) \exp(-\alpha^2 t^2/2)$$

This proves Theorem 2.

Remark. According to Theorem 2 the problem of investigating the convergence of $F_n(y)$ reduces to the classical limit problem for the sum S_n of independent random variables, and with the aid of this theorem we can obtain a large class of limit distributions of Z_n .

The following theorem states conditions for asymptotic normality of Z_n .

We say that Z_n is asymptotically normal $N(a_n, b_n^2)$ if

$$\sup_{-\infty < y < \infty} |F((Z_n - a_n)/b_n < y) - \phi(y)| \rightarrow 0$$

Theorem 3. Assume that $\lambda_n \rightarrow \infty$. Then necessary and sufficient conditions for the existence of number sequences $\{a_n\}$ and $\{b_n\}$ with $b_n > 0$ and $b_n \rightarrow \infty$ such that

(i) $g_n(\varepsilon) = \nu(K_n \cap \{x: |h(x)| > \varepsilon b_n\}) \rightarrow 0$ as $n \rightarrow \infty$ for all $\varepsilon > 0$

(ii) Z_n is asymptotically normal $N(a_n, b_n^2)$

are that there exists a number sequence $\{d_n\}$ with $d_n \rightarrow \infty$ such that

$$(a) \quad C_n^2 = \nu(h^2 I_{K_n S_n}) \rightarrow \infty \quad \text{where } S_n = \{x: |h(x)| \leq d_n\}$$

$$(b) \quad d_n = o(C_n), \quad \nu(K_n S_n^c) \rightarrow 0$$

Further, in this case the constants a_n, b_n can be defined by

$$(12) \quad a_n = \nu(h I_{K_n S_n}), \quad b_n^2 = C_n^2 = \nu(h^2 I_{K_n S_n}).$$

Proof of necessity. Suppose that (i), (ii) are fulfilled. Since $g_n(\varepsilon) \rightarrow 0$, there exists a sequence $\{\varepsilon_n\}$ such that $\varepsilon_n \downarrow 0$ and $g_n(\varepsilon_n) \rightarrow 0$.

Putting $d_n = \varepsilon_n b_n = o(b_n)$, we obtain $\nu(K_n S_n^c) \rightarrow 0$

Further, the logarithm of the ch.f. $H_n(t)$ of $Y_n = (Z_n - a_n)/b_n$ can be extended in the following form (see (9))

$$(13) \quad \begin{aligned} \ln H_n(t) = & -i t a_n / b_n + \nu([\exp(i t h / b_n) - 1] I_{K_n}) = \\ & -i t a_n / b_n + \nu([\exp(i t h / b_n) - 1] I_{K_n S_n}) + o(1) \end{aligned}$$

since

$$|\nu([\exp(i t h / b_n) - 1] I_{K_n S_n^c})| \leq 2 \nu(K_n S_n^c) \rightarrow 0$$

Furthermore,

$$(14) \quad \begin{aligned} \nu([\exp(i t h / b_n) - 1] I_{K_n S_n}) = & i t \nu(h I_{K_n S_n}) / b_n - \\ & - t^2 \nu(h^2 I_{K_n S_n}) / 2 b_n^2 + o(|t|^3 (d_n / b_n)) \nu(h^2 I_{K_n S_n}) / 6 b_n^2 \end{aligned}$$

with $|\theta| \leq 1$.

It follows from (13), (14) and from the assumption of asymptotic normality of Z_n that

$$(15) \quad \ln H_n(t) = -it a_n / b_n + it \nu(h I_{K_n S_n}) / b_n - t^2 \nu(h^2 I_{K_n S_n}) / 2b_n^2 + \\ \Theta |t|^3 (d_n / 6b_n^3) \nu(h^2 I_{K_n S_n}) + o(1) \rightarrow -t^2 / 2.$$

(15) holds iff

$$\nu(h^2 I_{K_n S_n}) / b_n^2 \rightarrow 1, \text{ or } C_n^2 / b_n^2 \rightarrow 1$$

and it follows from $d_n / b_n \rightarrow 0$ that $d_n / C_n \rightarrow 0$.

Proof of sufficiency. Suppose that (a), (b) are satisfied. Then putting in (15) $a_n = \nu(h I_{K_n S_n})$, $b_n = C_n$, we obtain

$$\ln H_n(t) \rightarrow -t^2 / 2$$

i.e. (ii) is fulfilled, whereas (i) follows immediately from (b) with $b_n = C_n$.

Remark. The statement on the sufficiency of conditions (a), (b) of Theorem 3 may be considered as a corollary of Theorem 2.

Indeed, according to Theorem 2.3 in [2], $F_n(y) = P\{(Z_n - a_n) / b_n < y\} \rightarrow \phi(y)$ iff for any subsequence $\{n'\}$ of $\{n\}$ there exists a subsequence $\{k\}$ of $\{n'\}$ such that $F_k(y) \rightarrow \phi(y)$. We shall show that the statement holds, provided (a), (b) are satisfied.

Note that if a_n, b_n are given by (12) we have

$$|a_n / b_n \lambda_n^{1/2}| \leq 1$$

The logarithm of the ch.f. $k_n(t)$ of $(S_n - a_n) / b_n$ defined in Theorem 2 is given by

$$(16) \quad \ln k_n(t) = -it a_n / b_n + \lambda_n \ln g_n(t),$$

where

$$(17) \quad g_n(t) = \nu(\exp(i t h / b_n) I_{K_n}) / \lambda_n.$$

On the other hand,

$$|\nu(\exp(iht/b_n)I_{K_n S_n^c})| \leq \nu(K_n S_n^c) \rightarrow 0$$

hence

$$\begin{aligned} g_n(t) &= \nu(\exp(iht/b_n)I_{K_n S_n}) / \lambda_n + o(\lambda_n^{-1}) = \\ (18) \quad &= 1 + it a_n / b_n \lambda_n - t^2 / 2 \lambda_n + o(d_n / b_n \lambda_n) + o(\lambda_n^{-1}) = \\ &= 1 + it a_n / b_n \lambda_n - t^2 / 2 \lambda_n + o(\lambda_n^{-1}). \end{aligned}$$

From (16) - (18) we obtain easily

$$(19) \quad \ln k_n(t) = -t^2/2 + t^2 a_n^2 / 2 \lambda_n b_n^2 + o(1).$$

On the other hand, for any subsequence $\{n'\}$ of $\{n\}$ there exists a subsequence $\{k\}$ of $\{n'\}$ such that $a_k^2 / \lambda_k b_k^2 \rightarrow \alpha^2$, hence

$$(20) \quad \lim_{k \rightarrow \infty} \ln k_k(t) = (\alpha^2 - 1)t^2/2, \quad \alpha^2 \leq 1.$$

Consequently, by Theorem 2, $F_k(y) \rightarrow K * \phi_\alpha(y)$, where K is the distribution function corresponding to the ch.f., the logarithm of which is equal to the right hand part of (20).

The logarithm of the ch.f. of $K * \phi_\alpha(y)$ is therefore equal to

$$(\alpha^2 - 1)t^2/2 - \alpha^2 t^2/2 = -t^2/2.$$

Consequently, $K * \phi_\alpha(y) = \phi(y)$. This proves the "sufficiency" part of Theorem 3.

Corollary 1. Assume that

$$b_n^2 = \nu(h^2 I_{K_n}) < \infty, \quad b_n \rightarrow \infty \quad \text{and} \quad \sup_{x \in K_n} |h(x)| = o(b_n).$$

Then Z_n is asymptotically normal $N(a_n, b_n^2)$ with $a_n = \nu(h I_{K_n})$.

Proof. Corollary 1 follows immediately from Theorem 3

by putting

$$d_n = \begin{cases} \sup_{K_n} |h(x)| & \text{if } \sup_{K_n} |h(x)| \rightarrow \infty \\ b_n^{1/2} & \text{if } \sup_{K_n} |h(x)| \not\rightarrow \infty \end{cases}$$

Corollary 2. (Theorem of Brown (1972).) Let ν_1, ν_2, φ be Radon measures on $(X, \mathcal{B}(X))$ and $\nu_1, \nu_2 \ll \varphi$. Further, suppose that the following conditions (i), (ii), (iii), or (i), (ii), (iv) are satisfied:

(i) $\nu_2 \ll \nu_1, f_1 = d\nu_1/d\varphi, f_2 = d\nu_2/d\varphi$.

(ii) There exists a finite positive number M such that

$$\nu_1\{x: |\ln(f_2/f_1(x))| > M\} < \infty.$$

(iii) $\nu_1([\ln(f_2/f_1)^2 - 1]^2 I_{D_c}) = \infty$ for all $c > 0$,

where

$$D_c = \{x: |[\ln(f_2(x)/f_1(x))]^2 - 1| < c\}.$$

(iv) There exists a finite number M_0 such that

$$\nu_1\{x: |\ln(f_2/f_1)| \geq M_0\} = \infty.$$

Then, as $K_n \uparrow X$, $(\mu(I_{K_n} \ln(f_2/f_1)))$ is asymptotically normal $N(a_n, b_n^2)$ under Q_{ν_1} , where

$$a_n = \nu_1(\ln(f_2/f_1) I_{K_n S_M}), \quad b_n^2 = \nu_1(\ln^2(f_2/f_1) I_{K_n S_M})$$

with $S_M = \{x: |\ln(f_2/f_1)| \leq M\}$.

Proof. Corollary 2 can be obtained immediately from Theorem 3 by putting $h = \ln(f_2/f_1)$.

Indeed, for $\nu_2 \ll \nu_1$, $\ln(f_2/f_1)$ is well defined ν_1 -a.e.. Let us now suppose that (ii), (iii) hold, then

$$(f_2/f_1)^2 - 1 \sim 2\ln(f_2/f_1) \text{ as } |(f_2/f_1)^2 - 1| < c,$$

hence it follows from (iii) that $\nu_1(h^2 I_{S_c/2}) = \infty$ and

$$(21) \quad \nu_1(h^2 I_{K_n S_M}) \geq \nu_1(h^2 I_{K_n S_c}) \rightarrow \infty \quad \text{if } c \leq M$$

If (ii), (iv) hold, then $M_0 < M$ and

$$(22) \quad \begin{aligned} b_n^2 &= \nu_1(h^2 I_{K_n S_M}) \geq \int_{K_n \cap \{|h| \leq M\}} h^2 \nu_1(dx) \geq \\ &\geq M_0^2 \nu_1(\{|h| \leq M\} \cap K_n) = M_0^2 [\nu_1(\{|h| \geq M_0\} \cap K_n) - \\ &\quad - \nu_1(\{|h| \geq M\} \cap K_n)] \rightarrow \infty \end{aligned}$$

Consequently, choosing $d_n = o(b_n)$, $d_n \rightarrow \infty$, then it follows from (21), (22) that

$$\nu_1(h^2 I_{K_n S_{d_n}}) \rightarrow \infty.$$

Further, $\nu_1(K_n S_{d_n}^c) \leq \nu_1(S_{d_n}^c) \rightarrow 0$, since

$$\begin{aligned} \nu_1(S_M^c) &= \sum_{i=1}^{\infty} \nu_1(d_i < |h| \leq d_{i+1}) < \infty \text{ implies } \nu_1(S_{d_n}^c) = \\ &= \sum_{j=n}^{\infty} \nu_1(d_j < |h| \leq d_{j+1}) \rightarrow 0, \text{ letting } M = d_1 < d_2 < \dots. \text{ Thus} \\ &\text{the conditions (a), (b) of Theorem 3 are satisfied. The condition } \nu_1(h = \pm \infty) = 0 \text{ is also fulfilled since} \end{aligned}$$

$$\nu_1(h = \pm \infty) \leq \nu_1(S_{d_n}^c) \rightarrow 0.$$

Consequently, the statements of Corollary follows from Theorem 3.

Remark 1. We observe that the assumptions of Theorem 3 are strictly weaker than those of the cited theorem of Brown.

In fact, let ν_1 be Lebesgue measure on the half line $X = [0, \infty)$, $\nu_2 \ll \nu_1$ with $d\nu_2/d\nu_1 = \exp(t) = f_2(t)$, $f_1(t) \equiv 1$.

Then $h(t) = \ln f_2(t) = t$. It is obvious that condition (ii) of the theorem of Brown is not fulfilled, since

$$\nu_1 \{ |h| > M \} = \nu_1 \{ t : t > M \} = \infty \text{ for all } M > 0.$$

Theorem 3 is, however, utilizable. Indeed, if $K_n = [0, T_n]$ with $T_n \uparrow \infty$, letting $T_n = d_n$ we have $K_n \cap S_n^c = \emptyset$,

$$b_n^2 = \int_0^{d_n} t^2 dt = d_n^3/3, \text{ so that } d_n = o(b_n).$$

Consequently, by Theorem 3, Z_n is asymptotically normal $N(a_n, b_n^2)$ with

$$a_n = \int_0^{T_n} t dt = T_n^2/2, \quad b_n^2 = T_n^3/3.$$

§ 3. The rate of convergence to limit distribution.

Theorem 4. Suppose that $\nu(|h|^3 I_{K_n}) < \infty$ and let $a_n = \nu(h I_{K_n})$, $b_n^2 = \nu(h^2 I_{K_n})$, $\gamma_n = \nu(|h|^3 I_{K_n})$,

$$b_n'^2 = b_n^2/\lambda_n; \quad \gamma_n' = \gamma_n/\lambda_n, \quad F_n(y) = Q_{\nu} \{ (Z_n - a_n)/b_n < y \}.$$

Then

$$(23) \quad \sup_{-\infty < y < \infty} |F_n(y) - \Phi(y)| \leq \Lambda \gamma_n / b_n^3 = \Lambda \gamma_n' / b_n'^3 \lambda_n^{1/2}$$

where Λ may be taken the value

$$\Lambda = (3/2\pi)^{1/2} + C^2(1/\pi) / (2\pi)^{3/2}$$

with $C(t)$ being the solution of the equation

$$\int_0^{C(t)} (\sin^2 u / u^2) du = \pi/4 + 1/8t$$

Proof. Let $H_n(t)$ be the ch.f. corresponding to F_n .

Then (see (9))

$$\begin{aligned}
H_n(t) &= \exp\{-it a_n/b_n + \nu([\exp(i t h/b_n) - 1] I_{K_n})\} = \\
&= \exp\{-it a_n/b_n + it \nu(h I_{K_n}) - t^2 \nu(h^2 I_{K_n})/2b_n^2 + \Theta |t|^3 \gamma_n/6b_n^3\} = \\
&= \exp(-t^2/2 + \Theta |t|^3 \gamma_n/6b_n^3) \text{ with } |\Theta| \leq 1,
\end{aligned}$$

hence

$$\begin{aligned}
|H_n(t) - \exp(-t^2/2)| &= \exp(-t^2/2) |\exp(\Theta |t|^3 \gamma_n/6b_n^3) - 1| \leq \\
&\leq (|t|^3 \gamma_n/6b_n^3) \exp(-t^2/2 + |t|^3 \gamma_n/6b_n^3) \leq |t|^3 \exp(-t^2/6) \gamma_n/6b_n^3
\end{aligned}$$

provided $|t| \leq 2b_n^3/\gamma_n = T$, say.

In accordance with Theorem 2, p. 137, [6], we have

$$\begin{aligned}
&\sup_y |F_n(y) - \phi(y)| \leq \\
&\leq \frac{1}{\pi} \int_{-T}^T \left| \frac{H_n(t) - \exp(-t^2/2)}{t} \right| dt + C^2(1/\pi) / T \pi \sqrt{2\pi}.
\end{aligned}$$

We therefore receive from (25)

$$\begin{aligned}
\sup_y |F_n(y) - \phi(y)| &\leq (\gamma_n/6\pi b_n^3) \int_{-\infty}^{\infty} t^2 \exp(-t^2/6) dt + \\
&+ C^2(1/\pi) \gamma_n/b_n^3 (2\pi)^{3/2} = A \gamma_n/b_n^3.
\end{aligned}$$

Remark 2. If $\gamma_n/b_n^3 \leq M$ (in general, it is usually fulfilled), then

$$\sup_y |F_n(y) - \phi(y)| \leq AM \lambda_n^{-1/2}$$

and this is the best estimation of the deviation between

$F_n(y)$ and $\phi(y)$. Indeed, if $h \equiv 1$ then $Z_n(\mu) = \mu(K_n)$ has Poisson distribution with the mean value $\lambda_n = \nu(K_n)$ and it is to see that

$$\sup_y |F_n(y) - \phi(y)| = O(\lambda_n^{-1/2}) \text{ where } F_n(y) = Q_{\nu} \{ [\mu(K_n) - \lambda_n] \lambda_n^{-1/2} < < y \}$$

Example. Let us consider the example described in Remark 1. We have

$\lambda_n = T_n$, $\gamma_n' = T_n^3/4$, $b_n'^2 = T_n^2/3$, hence $\gamma_n'/b_n'^3 = 3\sqrt{3}/4$, thus, by (23), $\sup|\mathbb{F}_n(\gamma) - \phi(\gamma)| \leq 3A\sqrt{3}/4T_n^{1/2}$.

§ 4. Some applications

1. Estimating the parameter of exponential trend. Let us consider a family of Poisson processes $\{Q_\theta = Q_{\varphi_\theta}, \theta \in \Theta\}$ on $(X, \mathcal{B}(X))$, where the intensity measure φ_θ possesses the density with respect to some Radon measure λ

$$d\varphi_\theta/d\lambda = \exp(\theta T(x)), \theta \in \Theta - \text{an open interval of } R^1.$$

Usually we can only observe a realization μ of the process Q_θ on a compact set K_n of X . In this case let us consider the σ -algebra \mathcal{A}_{K_n} generated by $\{\mu(A) : A \subseteq K_n\}$. Then, according to [4] the restrictions $Q_\theta^{(n)}, Q_\lambda^{(n)}$ of Q_θ, Q_λ on \mathcal{A}_{K_n} have the property that $Q_\theta^{(n)} \ll Q_\lambda^{(n)}$, and the logarithm of the likelihood function of the process is given by

$$(26) \quad L_n(\theta) = \ln(dQ_\theta^{(n)}/dQ_\lambda^{(n)}) = \lambda(K_n) - \varphi_\theta(K_n) + \theta \mu(TI_{K_n}).$$

Let

$$h_n(\theta) = \varphi_\theta(K_n) = \lambda(I_{K_n} \exp(\theta T))$$

Suppose that $h_n(\theta)$ satisfied the following conditions:

- (i) $dh_n(\theta)/d\theta = \lambda(I_{K_n} T \exp(\theta T)) = a_n(\theta)$, say, and $a_n(\theta)$ is finite,
- (ii) $d^2h_n(\theta)/d\theta^2 = \lambda(I_{K_n} T^2 \exp(\theta T)) = b_n^2(\theta) < \infty$, and $b_n(\theta) \rightarrow \infty$ as $n \rightarrow \infty$,

(iii) $C_n(\theta) = \lambda(|T|^3 \exp(\theta T) I_{K_n})$ is finite and there exists a number $\sigma'(\theta) > 0$ such that

$$\sup \{ |C_n(\theta')|, |\theta' - \theta| < \sigma' \} / b_n^3(\theta) \rightarrow 0 \text{ as } n \rightarrow \infty$$

It is obvious that $\{dQ_\theta^{(n)}/dQ_\lambda^{(n)}, \theta \in \Theta\}$ is an exponential family of one parameter and $Z_n(\mu) = \mu(TI_{K_n})$ is a complete sufficient statistic for θ and is an unbiased estimate of $a_n(\theta)$. In particular, $Z_n(\mu)$ takes in the form of the statistic considered in Theorem 2 and 3. We have the following statement:

Proposition. Assume that the above conditions (i), (ii), (iii) are satisfied. Then the likelihood equation $dL_n(\theta)/d\theta = 0$ or $a_n(\theta) - Z_n(\mu) = 0$ has under Q_{θ_0} unique solution $\hat{\theta}(\mu)$ as $n \rightarrow \infty$ and with probability approaching to 1, and $\hat{\theta}(\mu)$ is asymptotically normal $N(\theta_0, b_n^{-2}(\theta_0))$.

Proof. At first let us remark that according to (23) of Theorem 4

$$(27) \sup_y |Q_{\theta_0} \{ (Z_n - a_n)/b_n(\theta_0) < y \} - \phi(y)| \leq AC_n(\theta_0)/b_n^3(\theta_0) \rightarrow 0$$

Further,

$$(28) \ a_n(\theta_0 \pm \sigma) = a_n(\theta_0) \pm \sigma b_n^2(\theta_0) + \beta \sigma^2 C_n(\theta_0 + \alpha \sigma) / 2,$$

$$|\beta|, |\alpha| \leq 1.$$

Choosing $\sigma = u_n/b_n(\theta_0)$ so that $u_n/b_n \rightarrow 0$ and $u_n(\theta_0) \rightarrow \infty$, $u_n^2(\theta_0) = O(b_n^3/C_n)$ (this is always fulfilled) we obtain from (28)

$$\frac{a_n(\theta_0 \pm \sigma) - Z_n(\mu)}{b_n(\theta_0)} = \frac{a_n(\theta_0) - Z_n(\mu)}{b_n(\theta_0)} \pm u_n(\theta_0) + O(1)$$

Consequently, the function $a_n(\theta) - Z_n$ will change its sign on

the interval $(\theta_0 - \sigma, \theta_0 + \sigma)$. Furthermore, by (ii), for n sufficiently large $a_n(\theta)$ is strictly increasing, hence the likelihood equation has only solution $\hat{\theta}$. Further,

$$(29) \quad b_n(\theta_0)[\hat{\theta} - \theta_0] < t \iff a_n(\hat{\theta}) < a_n(\theta_0 + tb_n^{-1}) \iff \\ \iff Z_n(\mu) < a_n(\theta_0 + tb_n^{-1}),$$

whereas $a_n(\theta_0 + tb_n^{-1})$ can be extended in the form (see (28))

$$(30) \quad a_n(\theta_0 + tb_n^{-1}) = a_n(\theta_0) + tb_n(\theta_0) + \beta t^2 C_n(\theta_0 + \alpha tb_n^{-1}) / 2b_n^2$$

It follows from (29), (30), (27) and (iii) that

$$Q_{\theta_0} \{ b_n(\theta_0)[\hat{\theta} - \theta_0] < t \} = Q_{\theta_0} \{ [Z_n(\mu) - a_n] / b_n < t \pm \beta t^2 C_n(\theta_0 + \\ + \alpha tb_n^{-1}) / 2b_n^3 \} \rightarrow \phi(t)$$

as $n \rightarrow \infty$ for any t fixed. This proves the asymptotic normality of $Z_n(\mu)$.

Example. Let $X = [0, \infty)$, $K = [0, T_n]$ with $T_n \uparrow \infty$, $T(x) = x$, λ be Lebesgue measure, $\Theta = (0, \infty)$. Then $\hat{\theta}$ is the unique solution of the equation

$$\int_0^{T_n} x \exp(\theta x) dx = \int_0^{T_n} x \mu(dx) = Z_n(\mu), \text{ say, or equivalently}$$

$$T_n \exp(\theta T_n) / \theta - [\exp(\theta T_n) - 1] / \theta^2 = Z_n(\mu)$$

and it is easy to verify that

$$C_n(\theta') / b_n^3(\theta) \sim \theta^{3/2} \exp([\theta' - \theta - \theta/2] T_n) / \theta' \rightarrow 0 \text{ for all}$$

$\theta': |\theta' - \theta| < \theta/2 = \sigma(\theta)$. Consequently, by the above proposition $\hat{\theta}$ is asymptotically normal $N(\theta, b_n^{-2}(\theta))$ under Q_{θ} with $b_n^2(\theta) \approx \approx T_n^2 \exp(\theta T_n) / \theta$.

Remark. By the theorem of Rao - Blackwell and by the

above proposition estimate $\hat{\theta}$ of θ is asymptotically efficient.

2. Distinguishing two Poisson processes. Let us consider two Poisson processes Q_{ν_1}, Q_{ν_2} and assume that $\nu_1, \nu_2 \ll \ll \lambda$. Further, suppose that we have a realization of μ only on compact subset K at our disposal. Let \mathcal{A}_K be σ -algebra generated by $\{\mu(A): A \subset K\}$. Then (see [4]) the restrictions $Q_{\nu_{iK}}, Q_{\lambda_K}$ of Q_{ν_i}, Q_{λ} on $\mathcal{A}_K, i=1,2$, respectively, have the property that $Q_{\nu_{iK}} \ll Q_{\lambda_K}$ and

$$dQ_{\nu_{iK}}/dQ_{\lambda_K} = \exp \{ \lambda(K) - \nu_i(K) + \mu(I_K \ln(d\nu_i/d\lambda)) \}, i=1,2.$$

Consequently, for testing Q_{ν_1} against Q_{ν_2} we can employ the likelihood ratio test, under which Q_{ν_1} will be rejected if

$$\frac{\exp [\lambda(K) - \nu_2(K) + \mu(\ln(d\nu_2/d\lambda) I_K)]}{\exp [\lambda(K) - \nu_1(K) + \mu(\ln(d\nu_1/d\lambda) I_K)]} > C$$

or equivalently

$$\mu(hI_K) > C_\alpha$$

where $h = \ln \left(\frac{d\nu_2/d\lambda}{d\nu_1/d\lambda} \right)$ and the constant C_α is defined so that the test has significance level $\alpha (0 < \alpha < 1)$. If K is rather large in the sense $\nu_i(K) \rightarrow \infty, i=1,2$ as $K \uparrow X$ we can employ the asymptotical normality of $\mu(hI_K)$ in order to define approximately C_α and the power of the test.

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