

RECURSIVE KERNEL ESTIMATION OF THE INTENSITY FUNCTION OF AN INHOMOGENEOUS POISSON PROCESS

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In this paper we give an extension of the results considered in [1] to the recursive algorithms. Recursive estimation is particularly useful in large sample size since the result can be easily updated with each additional observation. The structure of the estimate is similar to the recursive kernel estimate of density function introduced in [2, 3], and thoroughly examined in [4]. The estimate is constructed on a single realization of a Poisson process on fixed interval $[0, T]$. Specificity of the statistic (1) is due to a random sample size.

Let $\{t_i, i = \overline{1, N}, 0 \leq t_i \leq T\}$ be a realization of a Poisson point process having unknown intensity function $\lambda(\cdot)$ on $[0, T]$, where N is the number of points falling into $[0, T]$, denote $\Lambda(a, b) = \int_a^b \lambda(t) dt$. Consider the following expression as an estimate of the function $\lambda(\cdot)/\Lambda(0, T)$ at a point t

$$S_N = \frac{1}{N} \sum_{i=1}^N \frac{1}{h_i} K\left(\frac{t - t_i}{h_i}\right) = \left(1 - \frac{1}{N}\right) S_{N-1} + \frac{1}{Nh_N} K\left(\frac{t - t_N}{h_N}\right) \quad (1)$$

where $(h_n) \downarrow 0$ is a sequence of real numbers, $nh_n \rightarrow \infty$, $K(\cdot)$ is a kernel function; let $S_N = 0$ if $N = 0$.

Consider an asymptotic behavior of statistic (1) under following scheme of series: let series of observations are done on $[0, T]$ with the intensity of the process in n -th trial equals to $\lambda_n(\cdot) = n\lambda(\cdot)$. Denote the value of the statistic (1) in n -th trial

$$S_n = \frac{1}{N_n} \sum_{i=1}^{N_n} \frac{1}{h_{in}} K\left(\frac{t - t_{in}}{h_{in}}\right), \quad (2)$$

where N_n and (t_{in}) – respectively the number of observations and the realization of the process in n -th trial.

Theorem 1 (asymptotic unbiasedness). Let the kernel $K(\cdot)$ is a compact bounded real valued Borel function on $[-T, T]$ and $\left| \int_{-T}^T K(u) u^m du \right| \leq M \forall m = 1, 2, \dots$; the intensity function $\lambda(\cdot)$ is continuous at the point $t \in (0, T)$, $\Lambda(t, T) \neq 0$; (h_{in}) is monotonically no increasing with i sequence, $(h_{in}) \rightarrow 0$ as $i \rightarrow \infty$, and $\left| \sum_{i=0}^m (-1)^i C_m^i h_{(i+N)n}^m \right| \leq Ch_{Nn}^m \forall m$ with some constant C . Then the estimate (1) is asymptotically unbiased: $\lim_{n \rightarrow \infty} E(S_n) - \lambda(t)/\Lambda(0, T) = 0$.

Proof. Let $p_i(x/n)$ be the conditional density function t_i given $N = n \geq 1$, $p_n = P(N = n) = \Lambda(0, T)^n e^{-\Lambda(0, T)} / n!$, $n \geq 0$. Then the expected value

$$\begin{aligned} E(S_N) &= \sum_{n=1}^{\infty} p_n \frac{1}{n} \sum_{i=1}^n \frac{1}{h_i} \int_0^T K\left(\frac{t-x}{h_i}\right) p_i(x/n) dx = \\ &= \sum_{n=1}^{\infty} \frac{1}{n} \sum_{i=1}^n \frac{1}{h_i} \int_0^T K\left(\frac{t-x}{h_i}\right) \frac{\Lambda(0,x)^{i-1}}{(i-1)!} \frac{\Lambda(x,T)^{n-i}}{(n-i)!} \frac{\lambda(x)}{\Lambda(0,T)} dx \frac{e^{-\Lambda(0,T)}}{1-e^{-\Lambda(0,T)}}. \end{aligned} \quad (3)$$

Taking into account $\sum_{n=1}^{\infty} \frac{1}{n} \sum_{i=1}^n \frac{\Lambda(0,x)^{i-1}}{(i-1)!} \frac{\Lambda(x,T)^{n-i}}{(n-i)!} = \sum_{n=1}^{\infty} \frac{\Lambda(0,T)^{n-1}}{n!} = \frac{e^{\Lambda(0,T)} - 1}{\Lambda(0,T)}$ and

boundedness of the kernel we know that series (3) converges absolutely. Let $h_i < \min(t/T, 1-t/T) \forall i$, and since $K(\cdot)$ is a compact function on $[-T, T]$ we have

$$\frac{1}{h_i} \int_0^T K\left(\frac{t-x}{h_i}\right) \frac{\Lambda(0,x)^{i-1}}{(i-1)!} \frac{\Lambda(x,T)^{n-i}}{(n-i)!} \lambda(x) dx = \int_{-T}^T K(x) \frac{\Lambda(0,t-h_i x)^{i-1}}{(i-1)!} \frac{\Lambda(t-h_i x, T)^{n-i}}{(n-i)!} \lambda(t-h_i x) dx.$$

Note that $\lambda(t-h_i x) = \lambda(t) + \tilde{C} h_i x$ for sufficiently small h_i , then

$$\begin{aligned} E(S_n) &= \frac{e^{-n\Lambda(0,T)}}{1-e^{-n\Lambda(0,T)}} \sum_{k=1}^{\infty} \frac{1}{k} \sum_{i=1}^k n^k \int_{-T}^T K(x) \frac{\Lambda(0,t-h_i x)^{i-1}}{(i-1)!} \frac{\Lambda(t-h_i x, T)^{k-i}}{(k-i)!} \lambda(t-h_i x) dx = \\ &= \frac{e^{-n\Lambda(0,T)}}{1-e^{-n\Lambda(0,T)}} \sum_{k=1}^{\infty} \frac{1}{k} \sum_{i=1}^k n^k \int_{-T}^T K(x) \frac{\Lambda(0,t-h_i x)^{i-1}}{(i-1)!} \frac{\Lambda(t-h_i x, T)^{k-i}}{(k-i)!} (\lambda(t) + \tilde{C} h_i x) dx = \lambda(t) L_n + L_n^o. \end{aligned}$$

For the proof of the theorem we establish the convergence $\lim_{n \rightarrow \infty} L_n = \Lambda(0,T)^{-1}$.

Using binomial expansion and the linearization $\Lambda(t-h_i x, t) = \int_{t-h_i x}^t \lambda(t) dt = r h_i x$, where

$r > 0$ – some constant, we obtain

$$\begin{aligned} &\sum_{k=1}^{\infty} \frac{1}{k} \sum_{i=1}^k n^k \int_{-T}^T K(x) \frac{\Lambda(0,t-h_i x)^{i-1}}{(i-1)!} \frac{\Lambda(t-h_i x, T)^{k-i}}{(k-i)!} dx = \\ &= \sum_{k=1}^{\infty} \frac{1}{k} \sum_{i=1}^k n^k \int_{-T}^T K(x) \sum_{l=0}^{i-1} (-1)^{i-1-l} \frac{\Lambda(0,t)^l}{l!} \frac{(r h_i x)^{i-1-l}}{(i-1-l)!} \sum_{m=0}^{k-i} \frac{\Lambda(t,T)^m}{m!} \frac{(r h_i x)^{k-i-m}}{(k-i-m)!} dx = \\ &= \sum_{k=0}^{\infty} \frac{n}{k+1} \sum_{l=0}^k \frac{\Lambda(0,t)^l}{l!} n^k \sum_{i=0}^{k-l} (-1)^i \int_{-T}^T K(x) \frac{(r h_i x)^i}{i!} \sum_{m=0}^{k-i-l} \frac{\Lambda(t,T)^m}{m!} \frac{(r h_i x)^{k-i-l-m}}{(k-i-l-m)!} dx = \\ &= \sum_{k=0}^{\infty} \frac{n}{k+1} \sum_{l=0}^k \frac{\Lambda(0,t)^l n^l}{l!} \sum_{m=0}^{k-l} \frac{\Lambda(t,T)^m n^m}{m!} n^{k-l-m} \sum_{i=0}^{k-l-m} (-1)^i \int_{-T}^T K(x) \frac{(r h_i x)^{k-l-m}}{i!(k-i-l-m)!} dx = \\ &= \sum_{k=0}^{\infty} \frac{n}{k+1} \sum_{l=0}^k \frac{\Lambda(0,t)^l n^l}{l!} \frac{\Lambda(t,T)^{k-l} n^{k-l}}{(k-l)!} + \Delta_n, \end{aligned} \quad (4)$$

where

$$\begin{aligned} \Delta_n &= \sum_{k=1}^{\infty} \frac{n}{k+1} \sum_{l=0}^{k-1} \frac{\Lambda(0,t)^l n^l}{l!} \sum_{m=0}^{k-l-1} \frac{\Lambda(t,T)^m n^m}{m!} n^{k-l-m} \sum_{i=0}^{k-l-m} (-1)^i \int_{-T}^T K(x) \frac{(r h_i x)^{k-l-m}}{i!(k-i-l-m)!} dx + \\ &+ n + \sum_{k=0}^{\infty} \frac{\Lambda(0,t)^k n^{k+1}}{(k+1)!}. \end{aligned}$$

In the next step we prove the convergence $\lim_{n \rightarrow \infty} e^{-n\Lambda(0,T)} \Delta_n = 0$:

$$\begin{aligned}
 & \lim_{n \rightarrow \infty} e^{-n\Lambda(0,T)} \left| \sum_{k=1}^{\infty} \frac{n}{k+1} \sum_{l=0}^{k-1} \frac{\Lambda(0,t)^l n^l}{l!} \frac{\Lambda(t,T)^{k-l} n^{k-l}}{(k-l)!} \sum_{m=1}^{k-l} \frac{C_{k-l}^m r^m}{\Lambda(t,T)^m} \sum_{i=0}^m (-1)^i C_m^i h_{(i+l+1)n}^m \int_{-T}^T K(x) x^m dx \right| \leq \\
 & \leq MC \lim_{n \rightarrow \infty} e^{-n\Lambda(0,T)} \sum_{k=1}^{\infty} \frac{n}{k+1} \sum_{l=0}^{k-1} \frac{\Lambda(0,t)^l n^l}{l!} \frac{\Lambda(t,T)^{k-l} n^{k-l}}{(k-l)!} \sum_{m=1}^{k-l} \frac{C_{k-l}^m r^m h_{(l+1)n}^m}{\Lambda(t,T)^m} = \\
 & = MC \lim_{n \rightarrow \infty} e^{-n\Lambda(0,T)} \sum_{k=1}^{\infty} \frac{n}{k+1} \sum_{l=0}^{k-1} \frac{\Lambda(0,t)^l n^l}{l!} \frac{\Lambda(t,T)^{k-l} n^{k-l}}{(k-l)!} \left(\left(1 + \frac{rh_{(l+1)n}}{\Lambda(t,T)} \right)^{k-l} - 1 \right) = 0.
 \end{aligned} \tag{5}$$

It follows from

$$\begin{aligned}
 & \lim_{n \rightarrow \infty} e^{-n\Lambda(0,T)} \sum_{k=1}^{\infty} \frac{n}{k+1} \sum_{l=0}^{k-1} \frac{\Lambda(0,t)^l n^l}{l!} \frac{\Lambda(t,T)^{k-l} n^{k-l}}{(k-l)!} = \lim_{n \rightarrow \infty} e^{-n\Lambda(0,T)} \sum_{k=1}^{\infty} \frac{n}{k+1} \sum_{l=0}^k \frac{\Lambda(0,t)^l n^l}{l!} \frac{\Lambda(t,T)^{k-l} n^{k-l}}{(k-l)!} = \\
 & = \lim_{n \rightarrow \infty} e^{-n\Lambda(0,T)} \sum_{k=1}^{\infty} \frac{n^{k+1}}{k+1} \frac{\Lambda(0,T)^k}{k!} = \lim_{n \rightarrow \infty} e^{-n\Lambda(0,T)} \frac{(e^{n\Lambda(0,T)} - 1 - n\Lambda(0,T))}{\Lambda(0,T)} = \frac{1}{\Lambda(0,T)}, \\
 & \frac{1}{\Lambda(0,T)} = \lim_{n \rightarrow \infty} e^{-n\Lambda(0,T)} \sum_{k=1}^{\infty} \frac{n}{k+1} \sum_{l=0}^{k-1} \frac{\Lambda(0,t)^l n^l}{l!} \frac{\Lambda(t,T)^{k-l} n^{k-l}}{(k-l)!} \leq \\
 & \leq \lim_{n \rightarrow \infty} e^{-n\Lambda(0,T)} \sum_{k=1}^{\infty} \frac{n}{k+1} \sum_{l=0}^{k-1} \frac{\Lambda(0,t)^l n^l}{l!} \frac{\Lambda(t,T)^{k-l} n^{k-l}}{(k-l)!} \left(1 + \frac{rh_{(l+1)n}}{\Lambda(t,T)} \right)^{k-l} \leq \\
 & \leq \lim_{n \rightarrow \infty} e^{-n\Lambda(0,T)} \sum_{k=1}^{\infty} \frac{n}{k+1} \sum_{l=0}^{k-1} \frac{\Lambda(0,t)^l n^l}{l!} \frac{\Lambda(t,T)^{k-l} n^{k-l}}{(k-l)!} \left(1 + \frac{rh_{2n}}{\Lambda(t,T)} \right)^{k-l} = \\
 & = \lim_{n \rightarrow \infty} \frac{\exp \{ n[\Lambda(0,t) + \Lambda(t,T)(1 + rh_{2n}/\Lambda(t,T))] - n\Lambda(0,T) \}}{\Lambda(0,t) + \Lambda(t,T)(1 + rh_{2n}/\Lambda(t,T))} = \frac{1}{\Lambda(0,T)}. \tag{6}
 \end{aligned}$$

The theorem is proved.

Theorem 2 (the variance convergence). Let the assumptions of Theorem 1 hold,

$$\int_{-T}^T K^2(x) x^m dx < \tilde{M} \quad \forall m = 1, 2, \dots; \quad ih_i \rightarrow \infty \text{ as } i \rightarrow \infty. \text{ Then } \lim_{n \rightarrow \infty} \text{Var}(S_n) = 0.$$

$$\begin{aligned}
 & \text{Proof. By } P(N_n \geq 2) \rightarrow 1 \text{ as } n \rightarrow \infty \text{ the variance } \text{Var}(S_n) = \\
 & = E \left\{ \frac{1}{(N_n)^2} \left[\sum_{i=1}^{N_n} \frac{1}{h_{in}^2} K^2 \left(\frac{t-t_{in}}{h_{in}} \right) + 2 \sum_{i>j} \frac{1}{h_{in} h_{jn}} K \left(\frac{t-t_{in}}{h_{in}} \right) K \left(\frac{t-t_{jn}}{h_{jn}} \right) \right] \right\} - \left\{ E \left[\frac{1}{N_n} \sum_{i=1}^{N_n} \frac{1}{h_{in}} K \left(\frac{t-t_{in}}{h_{in}} \right) \right] \right\}^2 = \\
 & = C_1(n) \sum_{k=1}^{\infty} \frac{n}{k^2} \sum_{i=1}^k \frac{1}{h_{in}} \int_{-T}^T K^2(x) \lambda(t-h_{in}x) \frac{\Lambda(0,t-h_{in}x)^i}{(i-1)!} \frac{\Lambda(t-h_{in}x, T)^{k-i}}{(k-i)!} n^k dx e^{-n\Lambda(0,T)} + \\
 & + 2C_2(n) \sum_{k=2}^{\infty} \frac{1}{k^2} \sum_{i=1}^{k-1} \sum_{j=i+1}^k \frac{1}{h_{in} h_{jn}} \int_0^T \int_T^T K \left(\frac{t-x}{h_{in}} \right) \lambda(x) K \left(\frac{t-y}{h_{jn}} \right) \lambda(y) \frac{\Lambda(0,x)^{i-1}}{(i-1)!} \frac{\Lambda(x,y)^{j-i-1}}{(j-i-1)!} \frac{\Lambda(y,T)^{k-j}}{(k-j)!} \times \\
 & \times n^k dy dx e^{-n\Lambda(0,T)} - C_3(n) \left(\sum_{n=1}^{\infty} \frac{1}{n} \sum_{i=1}^n \frac{1}{h_{in}} \int_0^T K \left(\frac{t-x}{h_{in}} \right) \frac{\Lambda(0,x)^{i-1}}{(i-1)!} \frac{\Lambda(x,T)^{n-i}}{(n-i)!} \lambda(x) dx e^{-n\Lambda(0,T)} \right), \tag{7}
 \end{aligned}$$

where $C_i(n) \rightarrow 1$ as $n \rightarrow \infty$, $i = 1, 2, 3$. The second term in (7) converges to $\left(\frac{\lambda(t)}{\Lambda(0,T)} \right)^2$

$$\text{since according to Newton's polynom } \sum_{i=1}^{k-1} \sum_{j=i+1}^k \frac{\Lambda(0,x)^{i-1}}{(i-1)!} \frac{\Lambda(x,y)^{j-i-1}}{(j-i-1)!} \frac{\Lambda(y,T)^{k-j}}{(k-j)!} =$$

$$= \frac{\Lambda(0, T)^{k-2}}{(k-2)!} \text{ (the main part is finding similarly (4)).}$$

To establish the statement of the theorem we must show

$$\begin{aligned} Q_n &= \sum_{k=1}^{\infty} \frac{n}{k^2} \sum_{i=1}^k \frac{1}{h_{in}} \int_{-T}^T K^2(x) \lambda(t - h_{in}x) \frac{\Lambda(0, t - h_{in}x)^i}{(i-1)!} \frac{\Lambda(t - h_{in}x, T)^{k-i}}{(k-i)!} n^k dx = \\ &= o(e^{-n\Lambda(0,T)}) \text{ as } n \rightarrow \infty. \text{ Let us consider } Q_n = \sum_{k=0}^{\infty} \frac{n}{(k+1)^2} \sum_{l=0}^k \frac{\Lambda(0,t)^l n^l}{l!} \frac{\Lambda(t,T)^{k-l} n^{k-l}}{(k-l)!} + \\ &\quad + \sum_{k=1}^{\infty} \frac{n}{(k+1)^2} \sum_{l=0}^{k-1} \frac{\Lambda(0,t)^l n^l}{l!} \sum_{m=0}^{k-l-1} \frac{\Lambda(t,T)^m n^m}{m!} n^{k-l-m} \sum_{i=0}^{k-l-m} \frac{(-1)^i}{h_{in}} \int_{-T}^T K^2(x) \frac{(rh_{in}x)^{k-l-m}}{i!(k-i-l-m)!} dx = \\ &= Q_n^{(1)} + Q_n^{(2)}, \quad Q_n^{(1)} = \frac{1}{\Lambda(0,T)} \sum_{k=1}^{\infty} \frac{\Lambda(0,T)^k n^k}{kk!} = \frac{1}{\Lambda(0,T)} [\text{Ei}(n\Lambda(0,T)) - \ln(n\Lambda(0,T)) - \gamma], \end{aligned}$$

where $\text{Ei}(n)$ – integral exponent, γ – Euler constant, $\lim_{n \rightarrow \infty} \text{Ei}(n)e^{-n} = 0$ [5]. Thus we obtain

$$\lim_{n \rightarrow \infty} Q_n^{(1)} e^{-n} = 0. \text{ Similarly (5) we have}$$

$$\begin{aligned} |Q_n^{(2)}| &\leq \tilde{M} C \sum_{k=1}^{\infty} \frac{n}{(k+1)^2} \sum_{l=0}^{k-1} \frac{\Lambda(0,t)^l n^l}{l!} \frac{\Lambda(t,T)^{k-l} n^{k-l}}{(k-l)! h_{(l+1)n}} \left(\left(1 + \frac{rh_{(l+1)n}}{\Lambda(t,T)} \right)^{k-l} - 1 \right) + o(n) \leq \\ &\leq \tilde{M} C \sum_{k=1}^{\infty} \frac{n}{(k+1)} \sum_{l=0}^{k-1} \frac{\Lambda(0,t)^l n^l}{l!} \frac{\Lambda(t,T)^{k-l} n^{k-l}}{(k-l)!(l+1)h_{(l+1)n}} \left(\left(1 + \frac{rh_{(l+1)n}}{\Lambda(t,T)} \right)^{k-l} - 1 \right) + o(n) \leq \\ &\leq \tilde{M} C \sum_{k=1}^{\infty} \frac{n}{(k+1)} \sum_{l=0}^{k-1} \frac{\Lambda(0,t)^l n^l}{l!} \frac{\Lambda(t,T)^{k-l} n^{k-l}}{(k-l)!} \left(\left(1 + \frac{rh_{(l+1)n}}{\Lambda(t,T)} \right)^{k-l} - 1 \right) + o(n). \end{aligned}$$

Taking (6) into account we have that $\lim_{n \rightarrow \infty} |Q_n^{(2)}| e^{-n} \leq 0$. The theorem is proved.

References

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This work was supported by FP "Scientific and scientific-pedagogical cadres of innovative Russia", project no. 26202.740.11.5190