

Additional File 2 of "Goodness-of-fit tests and nonparametric adaptive estimation for spike train analysis": proof of Theorem 2 of [1]

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In this additional file, the intensity $\lambda(\cdot)$ should be understood as a function on the whole real line which is null outside $[0, T_{max}]$. In particular, the Poisson process $N^{a,n}$, with intensity $n\lambda(\cdot)$, exists now on \mathbb{R} , but there is no points outside $[0, T_{max}]$ and $N^{a,n}(\mathbb{R}) = N^{a,n}([0, T_{max}])$. The proof is inspired by Proposition 2 of [2]. We set:

$$\chi = (1 + \eta)(1 + \|K\|_1)\|K\|_2.$$

Therefore, we have:

$$A(h) = \sup_{h' \in \mathcal{H}} \left\{ \|\hat{\lambda}_n^{h, h'} - \hat{\lambda}_n^{K_{h'}}\|_2 - \frac{\chi \sqrt{N^{a,n}(\mathbb{R})}}{n\sqrt{h'}} \right\}_+$$

and

$$\hat{h} = \arg \min_{h \in \mathcal{H}} \left\{ A(h) + \frac{\chi \sqrt{N^{a,n}(\mathbb{R})}}{n\sqrt{h}} \right\}.$$

For any $h \in \mathcal{H}$,

$$\|\hat{\lambda}_n^{GL} - \lambda\|_2 \leq A_1 + A_2 + A_3,$$

with

$$A_1 := \|\hat{\lambda}_n^{GL} - \hat{\lambda}_n^{\hat{h}, h}\|_2 \leq A(h) + \frac{\chi \sqrt{N^{a,n}(\mathbb{R})}}{n\sqrt{\hat{h}}},$$

$$A_2 := \|\hat{\lambda}_n^{\hat{h}, h} - \hat{\lambda}_n^{K_{\hat{h}}}\|_2 \leq A(\hat{h}) + \frac{\chi \sqrt{N^{a,n}(\mathbb{R})}}{n\sqrt{\hat{h}}}$$

and

$$A_3 := \|\hat{\lambda}_n^{K_h} - \lambda\|_2.$$

By definition of \hat{h} , we have:

$$A_1 + A_2 \leq 2A(h) + \frac{2\chi\sqrt{N^{a,n}(\mathbb{R})}}{n\sqrt{h}}.$$

Therefore, by setting

$$\zeta_n(h) := \sup_{h' \in \mathcal{H}} \left\{ \|(\hat{\lambda}_n^{h,h'} - \mathbb{E}[\hat{\lambda}_n^{h,h'}]) - (\hat{\lambda}_n^{K_{h'}} - \mathbb{E}[\hat{\lambda}_n^{K_{h'}}])\|_2 - \frac{\chi\sqrt{N^{a,n}(\mathbb{R})}}{n\sqrt{h'}} \right\}_+$$

we have:

$$\begin{aligned} A_1 + A_2 &\leq 2\zeta_n(h) + 2 \sup_{h' \in \mathcal{H}} \|\mathbb{E}[\hat{\lambda}_n^{h,h'}] - \mathbb{E}[\hat{\lambda}_n^{K_{h'}}]\|_2 + \frac{2\chi\sqrt{N^{a,n}(\mathbb{R})}}{n\sqrt{h}} \\ &\leq 2\zeta_n(h) + 2 \sup_{h' \in \mathcal{H}} \|K_h \star K_{h'} \star \lambda - K_{h'} \star \lambda\|_2 + \frac{2\chi\sqrt{N^{a,n}(\mathbb{R})}}{n\sqrt{h}} \\ &\leq 2\zeta_n(h) + 2\|K\|_1 \|K_h \star \lambda - \lambda\|_2 + \frac{2\chi\sqrt{N^{a,n}(\mathbb{R})}}{n\sqrt{h}}. \end{aligned}$$

Finally, since $(a + b + c)^2 \leq 3a^2 + 3b^2 + 3c^2$,

$$\begin{aligned} \mathbb{E}[(A_1 + A_2)^2] &\leq 12 \mathbb{E}[\zeta_n^2(h)] + 12\|K\|_1^2 \|K_h \star \lambda - \lambda\|_2^2 + \frac{12\chi^2 \mathbb{E}[N^{a,n}(\mathbb{R})]}{n^2 h} \\ &\leq 12 \mathbb{E}[\zeta_n^2(h)] + 12\|K\|_1^2 \|K_h \star \lambda - \lambda\|_2^2 + \frac{12\chi^2 \|\lambda\|_1}{nh}. \end{aligned}$$

For the last term, we obtain:

$$\begin{aligned} \mathbb{E}[A_3^2] &= \int \text{Var}(\hat{\lambda}_n^{K_h}(x)) dx + \|K_h \star \lambda - \lambda\|_2^2 \\ &= \frac{1}{n^2} \iint K_h^2(x-u) n\lambda(u) du dx + \|K_h \star \lambda - \lambda\|_2^2 \\ &= \frac{\|\lambda\|_1}{nh} \|K\|_2^2 + \|K_h \star \lambda - \lambda\|_2^2. \end{aligned}$$

Finally, replacing χ with its definition, we obtain: for any $h \in \mathcal{H}$,

$$\begin{aligned} \mathbb{E}\|\hat{\lambda}_n^{GL} - \lambda\|_2^2 &\leq 2 \mathbb{E}[(A_1 + A_2)^2] + 2 \mathbb{E}[A_3^2] \\ &\leq 2(1 + 12\|K\|_1^2) \|K_h \star \lambda - \lambda\|_2^2 + 2(1 + 12(1 + \eta)^2 (1 + \|K\|_1)^2) \frac{\|\lambda\|_1}{nh} \|K\|_2^2 + 24 \mathbb{E}[\zeta_n^2(h)]. \end{aligned}$$

The result follows by using the next lemma.

Lemma 1. *If $\mathcal{H} \subset \{D^{-1} : D = 1, \dots, D_{\max}\}$ with $D_{\max} = \delta n$ for some $\delta > 0$, and if $\|\lambda\|_\infty < \infty$, then there exists a constant C depending on $\delta, \eta, \|K\|_2, \|K\|_1, \|\lambda\|_1$ and $\|\lambda\|_\infty$ such that for any $h \in \mathcal{H}$*

$$\mathbb{E}[\zeta_n^2(h)] \leq Cn^{-1}.$$

Proof. First, we notice that for any $h \in \mathcal{H}$

$$\begin{aligned}\zeta_n(h) &\leq \sup_{h' \in \mathcal{H}} \left\{ \|\hat{\lambda}_n^{h,h'} - \mathbb{E}[\hat{\lambda}_n^{h,h'}]\|_2 + \|\hat{\lambda}_n^{K_{h'}} - \mathbb{E}[\hat{\lambda}_n^{K_{h'}}]\|_2 - \frac{\chi \sqrt{N^{a,n}(\mathbb{R})}}{n\sqrt{h'}} \right\}_+ \\ &\leq \sup_{h' \in \mathcal{H}} \left\{ (\|K\|_1 + 1) \|\hat{\lambda}_n^{K_{h'}} - \mathbb{E}[\hat{\lambda}_n^{K_{h'}}]\|_2 - \frac{\chi \sqrt{N^{a,n}(\mathbb{R})}}{n\sqrt{h'}} \right\}_+ \\ &\leq (\|K\|_1 + 1) S_n,\end{aligned}$$

with

$$S_n := \sup_{h \in \mathcal{H}} \left\{ \|\hat{\lambda}_n^{K_h} - \mathbb{E}[\hat{\lambda}_n^{K_h}]\|_2 - \frac{\|K\|_2(1+\eta)\sqrt{N^{a,n}(\mathbb{R})}}{n\sqrt{h}} \right\}_+.$$

We then have:

$$\mathbb{E}[\zeta_n^2(h)] \leq (\|K\|_1 + 1)^2(A + B)$$

with

$$A := \mathbb{E}[S_n^2 1_{\{N^{a,n}(\mathbb{R}) \leq (1-\alpha)^2 n \|\lambda\|_1\}}],$$

$$B := \mathbb{E}[S_n^2 1_{\{N^{a,n}(\mathbb{R}) > (1-\alpha)^2 n \|\lambda\|_1\}}]$$

for all $\alpha \in (0, 1)$. We have:

$$\begin{aligned}S_n^2 &\leq 2 \sup_{h \in \mathcal{H}} \|\hat{\lambda}_n^{K_h}\|_2^2 + 2 \sup_{h \in \mathcal{H}} \|\mathbb{E}[\hat{\lambda}_n^{K_h}]\|_2^2 \\ &\leq 2n^{-2} \sup_{h \in \mathcal{H}} \int dx \left(\int K_h(x-u) dN^{a,n}(u) \right)^2 + 2 \sup_{h \in \mathcal{H}} \|K_h \star \lambda\|_2^2 \\ &\leq 2n^{-2} \sup_{h \in \mathcal{H}} \iint K_h^2(x-u) dN^{a,n}(u) dx \times N^{a,n}(\mathbb{R}) + 2 \sup_{h \in \mathcal{H}} \|K_h \star \lambda\|_2^2 \\ &\leq 2n^{-2} \sup_{h \in \mathcal{H}} \frac{\|K\|_2^2}{h} \times N^{a,n}(\mathbb{R})^2 + 2 \sup_{h \in \mathcal{H}} \frac{\|K\|_2^2}{h} \|\lambda\|_1^2 \\ &\leq 2\delta n \|K\|_2^2 \left(\frac{N^{a,n}(\mathbb{R})^2}{n^2} + \|\lambda\|_1^2 \right).\end{aligned}$$

Therefore,

$$A \leq 4\delta n \|K\|_2^2 \|\lambda\|_1^2 \times \mathbb{P}(N^{a,n}(\mathbb{R}) \leq (1-\alpha)^2 n \|\lambda\|_1).$$

To bound the last term, we use, for instance, Inequality (5.2) of [3] (with $\xi = (2\alpha - \alpha^2)n\|\lambda\|_1$ and with the function $f \equiv -1$), which shows that there exists $\alpha' > 0$ only depending on α such that

$$\mathbb{P}(N^{a,n}(\mathbb{R}) \leq (1-\alpha)^2 n \|\lambda\|_1) \leq \exp(-\alpha' \|\lambda\|_1 \times n).$$

This shows that

$$A \leq C_A n^{-1},$$

where C_A depends on α , δ , $\|K\|_2$ and $\|\lambda\|_1$. Now, we deal with the term B by fixing the previous value α : we set $\alpha = \min(\eta/2, 1/4)$. This implies

$$(1 + \eta)(1 - \alpha) \geq 1 + \frac{\eta}{4}$$

and

$$\begin{aligned} B &= \mathbb{E}[S_n^2 1_{\{N^{a,n}(\mathbb{R}) > (1-\alpha)^2 n \|\lambda\|_1\}}] \\ &\leq \mathbb{E} \left[\sup_{h \in \mathcal{H}} \left\{ \|\hat{\lambda}_n^{K_h} - \mathbb{E}[\hat{\lambda}_n^{K_h}]\|_2 - \frac{(1 + \frac{\eta}{4})\|K\|_2 \sqrt{\|\lambda\|_1}}{\sqrt{nh}} \right\}_+^2 \right] \\ &= \int_0^{+\infty} \mathbb{P} \left(\sup_{h \in \mathcal{H}} \left\{ \|\hat{\lambda}_n^{K_h} - \mathbb{E}[\hat{\lambda}_n^{K_h}]\|_2 - \frac{(1 + \frac{\eta}{4})\|K\|_2 \sqrt{\|\lambda\|_1}}{\sqrt{nh}} \right\}_+ \geq x \right) dx \\ &\leq \sum_{h \in \mathcal{H}} \int_0^{+\infty} \mathbb{P} \left(\left\{ \|\hat{\lambda}_n^{K_h} - \mathbb{E}[\hat{\lambda}_n^{K_h}]\|_2 - \frac{(1 + \frac{\eta}{4})\|K\|_2 \sqrt{\|\lambda\|_1}}{\sqrt{nh}} \right\}_+ \geq x \right) dx. \end{aligned}$$

To conclude, it remains to control for any $x \geq 0$, the probability inside the integral. For this purpose, we apply Corollary 2 of [3] and we set

$$U(x) = \hat{\lambda}_n^{K_h}(x) - \mathbb{E}[\hat{\lambda}_n^{K_h}(x)] = \frac{1}{n} \int K_h(x-u) dN^{a,n}(u) - (K_h \star \lambda)(x).$$

If \mathcal{A} is a countable dense subset of the unit ball of $\mathbb{L}_2(\mathbb{R})$, we have:

$$\begin{aligned} \|U\|_2 &= \sup_{a \in \mathcal{A}} \int a(t) U(t) dt \\ &= \sup_{a \in \mathcal{A}} \int a(t) \left(\frac{1}{n} \int K_h(t-u) dN^{a,n}(u) - \int K_h(t-u) \lambda(u) du \right) dt \\ &= \sup_{a \in \mathcal{A}} \int \psi_a(u) (dN^{a,n}(u) - n\lambda(u)) du, \end{aligned}$$

with for any $a \in \mathcal{A}$ and any $u \in \mathbb{R}$,

$$\psi_a(u) = \frac{1}{n} \int a(t) K_h(t-u) dt.$$

We have for any u ,

$$\psi_a^2(u) \leq \frac{1}{n^2} \int a^2(t) dt \int K_h^2(t-u) dt = \frac{\|K\|_2^2}{n^2 h}.$$

So, if $b = \frac{\|K\|_2}{n\sqrt{h}}$, $\|\psi_a\|_\infty \leq b$. We have

$$\mathbb{E}[\|U\|_2^2] = \int \mathbb{E}[U^2(t)] dt = \int \text{var}(\hat{\lambda}_n^{K_h}(t)) dt = \frac{\|\lambda\|_1}{nh} \|K\|_2^2,$$

which implies

$$\mathbb{E}[\|U\|_2] \leq \frac{\sqrt{\|\lambda\|_1}}{\sqrt{nh}} \|K\|_2.$$

Finally,

$$\begin{aligned} v &:= \sup_{a \in \mathcal{A}} \int \psi_a^2(x) n \lambda(x) dx \\ &= \sup_{a \in \mathcal{A}} \int \left(\int \frac{a(t)}{n} K_h(t-x) dt \right)^2 n \lambda(x) dx \\ &\leq \frac{1}{n} \int \lambda(x) dx \int a^2(t) |K_h(t-x)| dt \int |K_h(t-x)| dt \\ &\leq \frac{\|K\|_1^2 \|\lambda\|_\infty}{n}. \end{aligned}$$

Corollary 2 of [3] gives: for any $\epsilon > 0$, for any $u > 0$,

$$\mathbb{P} \left(\|\hat{\lambda}_n^{K_h} - \mathbb{E}[\hat{\lambda}_n^{K_h}]\|_2 \geq (1 + \epsilon) \frac{\sqrt{\|\lambda\|_1}}{\sqrt{nh}} \|K\|_2 + \sqrt{\frac{12}{n} \|K\|_1^2 \|\lambda\|_\infty} u + \left(\frac{5}{4} + \frac{32}{\epsilon} \right) \frac{\|K\|_2}{n\sqrt{h}} u \right) \leq \exp(-u).$$

Then, we conclude by using exactly the same computation as in the proof of Lemma 1 of [4, p 32-34] that gives:

$$B \leq C_B n^{-1},$$

where C_B is a constant depending on δ , η , $\|K\|_2$, $\|K\|_1$ and $\|\lambda\|_\infty$. □

The lemma leads to the result.

References

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