

# Supplementary Material to "Model selection for the robust efficient signal processing observed with small Lévy noise". \*

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This supplementary document contains Appendix in which we prove all technical results.

## 1 Appendix

### A.1 Proof of Proposition 3

We use here the same method as in Konev and Pergamenshchikov (2009). First, note that from the definitions (19) and (35) we obtain

$$\hat{\tau}_{j,\varepsilon} = \tau_j + \varepsilon \eta_j, \quad (\text{A.1})$$

where

$$\tau_j = \int_0^1 S(t) \text{Tr}_j(t) dt \quad \text{and} \quad \eta_j = \int_0^1 \text{Tr}_j(t) d\check{\xi}_t.$$

So, we have

$$\hat{\varkappa}_\varepsilon = \sum_{j=[1/\varepsilon]+1}^n \tau_j^2 + 2\varepsilon \check{M}_\varepsilon + \varepsilon^2 \sum_{j=[1/\varepsilon]+1}^n \eta_j^2, \quad (\text{A.2})$$

where  $\check{M}_\varepsilon = \sum_{j=[1/\varepsilon]+1}^n \tau_j \eta_j$ . Note that for the continuously differentiable functions (see, for example, Lemma A.6 in Konev and Pergamenshchikov (2009a))

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the Fourier coefficients  $(\tau_j)$  for any  $n \geq 1$  satisfy the following inequality

$$\sum_{j=[1/\varepsilon]+1}^{\infty} \tau_j^2 \leq 4\varepsilon \left( \int_0^1 |\dot{S}(t)| dt \right)^2 \leq 4\varepsilon \|\dot{S}\|^2. \quad (\text{A.3})$$

We recall, that  $\dot{S}$  is the derivative of  $S$ . To estimate the term  $\check{M}_\varepsilon$  note, that it can be represented as

$$\check{M}_\varepsilon = \check{I}_1 \left( \sum_{j=[1/\varepsilon]+1}^n \tau_j \text{Tr}_j \right).$$

Therefore, using the last equality in (19) and orthonormality of the functions  $(\text{Tr}_j)_{j \geq 1}$ , we obtain that

$$\mathbf{E}_Q \check{M}_\varepsilon^2 = \check{\varkappa}_Q \sum_{j=[1/\varepsilon]+1}^n \tau_j^2 \leq 4\varepsilon \varkappa_Q \|\dot{S}\|^2.$$

Moreover, taking into account that for  $j \geq 2$  the expectation  $\mathbf{E} \eta_j^2 = \check{\varkappa}_Q$  we can represent the last term in (A.2) as

$$\varepsilon^2 \sum_{j=[1/\varepsilon]+1}^n \eta_j^2 = \check{\varkappa}_Q (\varepsilon^2 n - \varepsilon^2 [1/\varepsilon]) + \varepsilon B_{2,\varepsilon}(x'),$$

where  $x' = (x'_j)_{1 \leq j \leq n}$  with  $x'_j = \varepsilon \mathbf{1}_{\{1/\varepsilon < j \leq n\}}$  and the function  $B_{2,\varepsilon}(x')$  is defined in (20). We remind that  $n = [1/\varepsilon^2]$  and that for the trigonometric basis (14) the upper bound  $\phi^* = \sqrt{2}$ . Therefore, in view of Proposition 2 and the definition of  $\bar{a}$  in (27) we obtain that

$$\mathbf{E}_Q \left| \varepsilon^2 \sum_{j=[\sqrt{1/\varepsilon}]+1}^n \eta_j^2 - \check{\varkappa}_Q \right| \leq \varepsilon \left( 2 \varkappa_Q + \sqrt{U_Q} + 4\sqrt{\varkappa_Q} \right).$$

So, we obtain the bound (36). Hence Proposition 3.  $\square$

## A.2 Property of Lemma 1

In view of the definition of  $\text{Err}_\varepsilon(\lambda)$  and the equation (19) one has

$$\begin{aligned} \text{Err}_\varepsilon(\lambda) &= \sum_{j=1}^n \left( (\lambda(j) - 1)\theta_j + \varepsilon \lambda(j) \bar{\xi}_j \right)^2 + \sum_{j=n+1}^{\infty} \theta_j^2 \\ &\geq 2\varepsilon \sum_{j=1}^n (\lambda(j) - 1)\theta_j \lambda(j) \bar{\xi}_j + \varepsilon^2 \sum_{j=1}^n \lambda^2(j) \bar{\xi}_j^2. \end{aligned}$$

Moreover, using here the definition (20) we obtain

$$\mathbf{E}_Q \text{Err}_\varepsilon(\lambda) \geq \varepsilon^2 \sum_{j=1}^n \lambda^2(j) \mathbf{E}_Q \bar{\xi}_j^2 = P_\varepsilon(\lambda) - \varepsilon^2 B_{1,\varepsilon}(\lambda^2),$$

where  $\lambda^2 = (\lambda^2(j))_{1 \leq j \leq n}$ . Now, Proposition 1 implies Lemma 1.  $\square$

### A.3 Proof of Proposition 4

First of all note that, the density (59) on the process  $\xi$  is bounded with respect to  $\theta_j \in \mathbb{R}$  and for any  $1 \leq j \leq d$

$$\limsup_{|\theta_j| \rightarrow \infty} f(\xi, \theta) = 0. \quad \text{a.s.}$$

Now, we set

$$\tilde{\Phi}_j = \tilde{\Phi}_j(x, \theta) = \frac{\partial (f(x, \theta) \Phi(\theta)) / \partial \theta_j}{f(x, \theta) \Phi(\theta)}.$$

Taking into account the condition (60) and integrating by parts yield

$$\begin{aligned} \tilde{\mathbf{E}} \left( ((\hat{g} - g(\theta)) \tilde{\Phi}_j) \right) &= \int_{\mathcal{X} \times \mathbb{R}^d} ((\hat{g}(x) - g(\theta)) \frac{\partial}{\partial \theta_j} (f(x, \theta) \Phi(\theta))) d\theta \mathbf{P}_\xi(dx) \\ &= \int_{\mathcal{X} \times \mathbb{R}^{d-1}} \left( \int_{\mathbb{R}} g'_j(\theta) f(x, \theta) \Phi(\theta) d\theta_j \right) \left( \prod_{i \neq j} d\theta_i \right) \mathbf{P}_\xi(dx) = \lambda_j. \end{aligned}$$

Now by the Bouniakovskii-Cauchy-Schwarz inequality we obtain the following lower bound for the quadratic risk

$$\tilde{\mathbf{E}}((\hat{g} - g(\theta))^2) \geq \frac{\Lambda_j^2}{\tilde{\mathbf{E}} \tilde{\Phi}_j^2}.$$

To study the denominator in the left hand of this inequality note that in view of the representation (59)

$$\frac{1}{f(y, \theta)} \frac{\partial f(y, \theta)}{\partial \theta_j} = \frac{1}{\varrho_1} \int_0^1 \phi_j(t) dw_t.$$

Therefore, for each  $\theta \in \mathbb{R}^d$ ,

$$\mathbf{E}_\theta \frac{1}{f(y, \theta)} \frac{\partial f(y, \theta)}{\partial \theta_j} = 0$$

and

$$\mathbf{E}_\theta \left( \frac{1}{f(y, \theta)} \frac{\partial f(y, \theta)}{\partial \theta_j} \right)^2 = \frac{1}{\varrho_1^2} \int_0^1 \phi_j^2(t) dt = \frac{1}{\varrho_1^2} \|\phi_j\|^2.$$

Taking into account that

$$\tilde{\Phi}_j = \frac{1}{f(x, \theta)} \frac{\partial f(x, \theta)}{\partial \theta_j} + \frac{1}{\Phi(\theta)} \frac{\partial \Phi(\theta)}{\partial \theta_j},$$

we get

$$\tilde{\mathbf{E}} \tilde{\Phi}_j^2 = \frac{1}{\varrho_1^2} \|\phi_j\|^2 + I_j.$$

Hence Proposition 4.  $\square$

#### A.4 The absolute continuity of distributions for the Lévy processes.

In this section we study the absolute continuity for the the Lévy processes defined as

$$dy_t = S(t)dt + d\xi_t, \quad 0 \leq t \leq T, \quad (\text{A.4})$$

where  $S(\cdot)$  is any arbitrary nonrandom square integrated function, i.e. from  $\mathbf{L}_2[0, T]$  and  $(\xi_t)_{0 \leq t \leq T}$  is the Lévy process of the form (7) with nonzero constants  $\varrho_1$  and  $\varrho_2$ . We denote by  $\mathbf{P}_y$  and  $\mathbf{P}_\xi$  the distributions of the processes  $(y_t)_{0 \leq t \leq 1}$  and  $(\xi_t)_{0 \leq t \leq 1}$  on the Skorokhod space  $\mathbf{D}[0, T]$ . Now for any  $0 \leq t \leq T$  and  $(x_t)_{0 \leq t \leq T}$  from  $\mathbf{D}[0, T]$  we set

$$\Upsilon_t(x) = \exp \left\{ \int_0^t \frac{S(u)}{\varrho_1^2} dx_u^c - \int_0^t \frac{S^2(u)}{2\varrho_1^2} du \right\}, \quad (\text{A.5})$$

where  $x^c$  is the continuous part of the process  $x$  defined in (59). Now we study the measures  $\mathbf{P}_y$  and  $\mathbf{P}_\xi$  in  $\mathbf{D}[0, T]$ .

**Proposition A.1.** *For any  $T > 0$  the measure  $\mathbf{P}_y \ll \mathbf{P}_\xi$  in  $\mathbf{D}[0, T]$  and the Radon-Nikodym derivative is*

$$\frac{d\mathbf{P}_y}{d\mathbf{P}_\xi}(\xi) = \Upsilon_T(\xi).$$

**Proof.** Note that to show this proposition it suffices to check that for any  $0 = t_0 < \dots < t_n = T$  any  $b_j \in \mathbb{R}$  for  $1 \leq j \leq n$

$$\mathbf{E} \exp \left\{ i \sum_{l=1}^n b_l (y_{t_l} - y_{t_{l-1}}) \right\} = \mathbf{E} \exp \left\{ i \sum_{l=1}^n b_l (\xi_{t_l} - \xi_{t_{l-1}}) \right\} \Upsilon_T(\xi).$$

taking into account that the processes  $(y_t)_{0 \leq t \leq T}$  and  $(\xi_t)_{0 \leq t \leq T}$  have the independent homogeneous increments, to this end one needs to check only that for any  $b \in \mathbb{R}$  and  $0 \leq s < t \leq T$

$$\mathbf{E} \exp \{ i b (y_t - y_s) \} = \mathbf{E} \exp \{ i b (\xi_t - \xi_s) \} \frac{\Upsilon_t(\xi)}{\Upsilon_s(\xi)}. \quad (\text{A.6})$$

To check this equality note that the process

$$\Upsilon_t(\xi) = \exp \left\{ \int_0^t \frac{S(u)}{\varrho_1} dw_u - \int_0^t \frac{S^2(u)}{2\varrho_1^2} du \right\}$$

is the gaussian martingale. From here we directly obtain the squation (A.6). Hence Proposition A.1.  $\square$

## A.5 Proof of the limit equality (94)

First, setting  $\zeta_\varepsilon = \sum_{j=1}^d \kappa_j^2 a_j$ , we obtain that

$$\{S_\kappa \notin W_{k,\mathbf{r}}\} = \{\zeta_\varepsilon > \mathbf{r}\}.$$

Moreover, note that one can check directly that

$$\lim_{\varepsilon \rightarrow 0} \mathbf{E} \zeta_\varepsilon = \lim_{\varepsilon \rightarrow 0} \frac{1}{v_\varepsilon} \sum_{j=1}^d s_j^* a_j = \check{\mathbf{r}} = (1 - \check{\gamma})\mathbf{r}.$$

So, for sufficiently small  $\varepsilon$  we obtain that

$$\{S_\kappa \notin W_{k,r}\} \subset \{\tilde{\zeta}_\varepsilon > \mathbf{r}_1\},$$

where  $\mathbf{r}_1 = \mathbf{r}\check{\gamma}/2$ ,  $\tilde{\zeta}_\varepsilon = \zeta_\varepsilon - \mathbf{E} \zeta_\varepsilon = v_\varepsilon^{-1} \sum_{j=1}^d s_j^* a_j \tilde{\eta}_j$  and  $\tilde{\eta}_j = \eta_j^2 - 1$ . Through the correlation inequality (see, Proposition A.1 in Galthouk and Pergamenschikov (2013)) we can get that for any  $p \geq 2$

$$\mathbf{E} \tilde{\zeta}_\varepsilon^p \leq (2p)^{p/2} \mathbf{E} |\tilde{\eta}_1|^p v_\varepsilon^{-p} \left( \sum_{j=1}^d (s_j^*)^2 a_j^2 \right)^{p/2} = O(v_\varepsilon^{-\frac{p}{4k+2}}),$$

as  $\varepsilon \rightarrow 0$ . Therefore, for any  $\iota > 0$  using the Chebychev inequality for  $p > (4k+2)\iota$  we obtain that

$$v_\varepsilon^\iota \mathbf{P}(\tilde{\zeta}_\varepsilon > \mathbf{r}_1) \rightarrow 0 \quad \text{as } \varepsilon \rightarrow 0.$$

Hence the equality (94).  $\square$

## A.6 Proof of Proposition 5

Substituting (19) and taking into account the definition (95) one gets

$$\|\check{S} - S\|^2 = \sum_{j=1}^{\infty} (1 - \check{\lambda}(j))^2 \theta_j^2 - 2\check{M}_\varepsilon + \varepsilon^2 \sum_{j=1}^{\infty} \check{\lambda}^2(j) \check{\xi}_j^2,$$

where  $\check{M}_\varepsilon = \varepsilon \sum_{j=1}^{\infty} (1 - \check{\lambda}(j)) \check{\lambda}(j) \theta_j \bar{\xi}_j$ . Note now that for any  $Q \in \mathcal{Q}_\varepsilon^*$  the expectation  $\mathbf{E}_{Q,S} \check{M}_\varepsilon = 0$  and, in view of the upper bound (9),

$$\sup_{Q \in \mathcal{Q}_\varepsilon^*} \mathbf{E}_{Q,S} \sum_{j=1}^{\infty} \check{\lambda}^2(j) \bar{\xi}_j^2 \leq \varsigma^* \sum_{j=1}^{\infty} \check{\lambda}^2(j).$$

Therefore,

$$\mathcal{R}_\varepsilon^*(\check{S}, S) \leq \sum_{j=\check{j}_*}^{\infty} (1 - \check{\lambda}(j))^2 \theta_j^2 + \frac{1}{v_\varepsilon} \sum_{j=1}^{\infty} \check{\lambda}^2(j), \quad (\text{A.7})$$

where  $\check{j}_* = j_*(\check{\alpha})$ . Setting

$$\mathbf{u}_\varepsilon = v_\varepsilon^{2k/(2k+1)} \sup_{j \geq \check{j}_*} (1 - \check{\lambda}(j))^2 / a_j,$$

we obtain that for each  $S \in W_r^k$

$$\Upsilon_{1,\varepsilon}(S) = v_\varepsilon^{2k/(2k+1)} \sum_{j=\check{i}}^{\infty} (1 - \check{\lambda}(j))^2 \theta_j^2 \leq \mathbf{u}_\varepsilon \sum_{j=\check{i}}^{\infty} a_j \theta_j^2 \leq \mathbf{u}_\varepsilon r.$$

Taking into account that  $\check{r} \rightarrow r$ , we obtain that

$$\limsup_{\varepsilon \rightarrow 0} \sup_{S \in W_r^k} \Upsilon_{1,\varepsilon}(S) \leq \frac{r^{1/(2k+1)}}{\pi^{2k} (\mathbf{d}_k)^{2k/(2k+1)}} := \Upsilon_1^*.$$

To estimate the last term in the right hand of (A.7), we set

$$\Upsilon_{2,\varepsilon} = \frac{1}{v_\varepsilon^{1/(2k+1)}} \sum_{j=1}^n \check{\lambda}^2(j).$$

It is easy to check that

$$\limsup_{\varepsilon \rightarrow 0} \Upsilon_{2,\varepsilon} \leq \frac{2(r \mathbf{d}_k)^{1/(2k+1)} k^2}{(k+1)(2k+1)} := \Upsilon_2^*.$$

Therefore, taking into account that by the definition of the Pinsker constant in (53) the sum  $\Upsilon_1^* + \Upsilon_2^* = l_*(\mathbf{r})$ , we arrive at the inequality

$$\lim_{\varepsilon \rightarrow 0} v_\varepsilon^{2k/(2k+1)} \sup_{S \in W_r^k} \mathcal{R}_\varepsilon^*(\check{S}, S) \leq l_*(\mathbf{r}).$$

Hence Proposition 5.  $\square$

## References

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