

Some Remarks on Multi-Parameter Processes

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Barndorff-Nielsen, Pedersen, Sato:

"Multivariate Subordination, Self-decomposability and Stability", 2001

Khoshnevisan: "Multiparameter Processes", 2002

Pedersen, Sato:

"Relations between cone-parameter Levy processes and convolution semigroups", 2004

Some Motivation

- From a pure analytic point of view the question arises: How can strongly continuous semigroups T_t be extended to the case of an N-dimensional parameter t ?
- analytic \longleftrightarrow probabilistic: How do N-parameter semigroups correspond to processes with N-dimensional time-parameter.

Definition 1. Let $q : \mathbb{R}^n \times \mathbb{R}^n \longrightarrow \mathbb{C}$ be continuous and negative definite with respect to the second variable, i.e. $q(x, \cdot) : \mathbb{R}^n \longrightarrow \mathbb{C}$ is continuous and negative definite. On $C_0^\infty(\mathbb{R}^n)$ we define the **pseudo-differential operator** $q(x, D)$ by

$$q(x, D)u(x) := (2\pi)^{-\frac{n}{2}} \int_{\mathbb{R}^n} e^{ix\xi} q(x, \xi) \hat{u}\xi \, d\xi.$$

We call q the **symbol** of the pseudo-differential operator $q(x, D)$.

N-Parameter Markov-Semigroups of Operators

Definition 2. An N-parameter convolution semigroup of subprobability measures $(\mu_t)_{t \geq 0, t \in \mathbb{R}_+^N}$, on \mathbb{R}^n is defined by

- (i) $\mu_t(\mathbb{R}^n) \leq 1$ for all $t \geq 0$,
- (ii) $\mu_s * \mu_t = \mu_{s+t}$ for all $s, t \geq 0$,
and $\mu_0 = \varepsilon_0$,
- (iii) $\mu_t \longrightarrow \varepsilon_0$ vaguely as $t \longrightarrow 0$.

Example 3. Let $(\mu_{t_1})_{t_1 \geq 0}$ and $(\nu_{t_2})_{t_2 \geq 0}$ be 1-parameter convolution semigroups of bounded Borel measures on \mathbb{R}^n , then

$$\eta_{t_1, t_2} := \mu_{t_1} \otimes \nu_{t_2} \quad \text{for all } t_1, t_2 \geq 0$$

defines a two-parameter convolution semigroup of Borel measures on \mathbb{R}^{2n} .

Definition 4. A. An N -parameter family $(T_t)_{t \geq 0}$, $t \in \mathbb{R}_+^N$, of bounded linear operators $T_t : X \rightarrow X$ with

- $T_0 = id$
- $T_{s+t} = T_s \circ T_t$ for all $s, t \in \mathbb{R}_+^N$.

is called an **N -parameter semigroup of operators.**

B. We call it **strongly continuous** if for all $x \in X$

$$\lim_{t \rightarrow 0} \|T_t u - u\|_X = 0.$$

C. The semigroup $(T_t)_{t \geq 0}$ is a **contraction semigroup**, if for all $t \geq 0$

$$\|T_t\| \leq 1.$$

Example 5. Let $(\mu_{t_1})_{t_1 \geq 0}$ and $(\nu_{t_2})_{t_2 \geq 0}$ be two-parameter convolution semigroups on \mathbb{R}^n . On the Banach space $(C_\infty(\mathbb{R}^{2n}), \|\cdot\|_\infty)$ we define for all $t \in \mathbb{R}_+^2$ the operator

$$\mathbb{T}_t u(x) := \int_{\mathbb{R}^{2n}} u(x - y) (\mu_{t_1} \otimes \nu_{t_2})(dy). \quad (1)$$

Then $(\mathbb{T}_t)_{t \succeq 0}$ is a strongly continuous contraction semigroup, which is positivity preserving.

Definition 6. A strongly continuous N-parameter contraction semigroup of operators on $(C_\infty(\mathbb{R}^{2n}), \|\cdot\|_\infty)$ which is positivity preserving is called in N-parameter **Feller semigroup**.

Definition 7. Let $(T_t)_{t \geq 0}$ be an N -parameter semigroup of operators, for $j = 1, \dots, N$ we define its **j -th marginal (1-parameter) semigroup** by

$$T_{t_j}^{(j)} := T_{t_j \cdot e_j} \quad \text{for all } t_j \geq 0.$$

Remark 8. With this definition every N -parameter semigroup can be represented as a composition of its marginal semigroups:

$$T_t = T_{t_1}^{(1)} \circ T_{t_2}^{(2)} \circ \dots \circ T_{t_N}^{(N)}.$$

The semigroups are commuting, i.e.

$$\left[T_{t_i}^{(i)}, T_{t_j}^{(j)} \right] = 0, \quad \text{for all } i, j = 1, \dots, N$$

and for each marginal semigroup $T_{t_j}^{(j)}$ we denote its generator by $A^{(j)}$.

Example 9. Consider two independent n -dimensional Brownian motions $B^{(1)}$ and $B^{(2)}$ and define for $t \in \mathbb{R}_+^2$

$$X_t = B_{t_1}^{(1)} + B_{t_2}^{(2)},$$

then X_t is a 2-parameter process, called the n -dimensional additive Brownian motion and with

$$T_t u(x) := (2\pi\{t_1 + t_2\})^{-\frac{n}{2}} \int_{\mathbb{R}^n} e^{-\frac{\|y-x\|^2}{2\{t_1+t_2\}}} u(y) dy,$$

we have

$$\mathbb{E}^x [u(X_t)] = T_t u(x)$$

and

$$\mathbb{P}^x (X_t \in A) = T_t \chi_A(x).$$

Lemma 10. For $u \in D(A^{(1)} \circ \dots \circ A^{(N)})$ and $(T_t)_{t \geq 0}$ as above we have

$$\frac{\partial^N}{\partial t_1 \dots \partial t_N} T_t u = A^{(1)} \circ \dots \circ A^{(N)} \circ T_t u.$$

Theorem 11. Let $(T_t)_{t \geq 0, t \in \mathbb{R}_+^N}$, be a strongly continuous N -parameter contraction semigroup on a Banach space $(X, \|\cdot\|_X)$ and $(\eta_s)_{s \geq 0, s \in \mathbb{R}_+^M}$, be an M -parameter convolution semigroup on \mathbb{R}^N with $\text{supp } \eta_s \subset [0, \infty)^N$, for all $s \in \mathbb{R}_+^M$. Now, we define $(T_s^\eta)_{s \geq 0}$ using the M -dimensional Bochner integral:

$$T_s^\eta u := \int_{\mathbb{R}_+^M} T_t u \eta_s(dt). \quad (2)$$

Then the integral is well-defined and $(T_s^\eta)_{s \geq 0}$ is a strongly continuous contraction semigroup on X and is called the subordinate (in the sense of Bochner) to $(T_t)_{t \geq 0}$ with respect to the convolution semigroup $(\eta_s)_{s \geq 0}$.

Corollary 12. *Let $(T_t)_{t \geq 0}$ be either an N -parameter Feller semigroup on $C_\infty(\mathbb{R}^n; \mathbb{R})$ or an N -parameter sub-Markovian semigroup on $L^p(\mathbb{R}^n; \mathbb{R})$, $1 \leq p \leq \infty$, further let $(\eta_s)_{s \geq 0}$ be an N -parameter convolution semigroup on \mathbb{R}_+^M with $\text{supp } \eta_s \subset \mathbb{R}_+^N$, i.e. with positive support. Then $(T_s^\eta)_{s \geq 0}$ is an M -parameter Feller or sub-Markovian semigroup, respectively.*

Remark 13. *Especially, subordinating with respect to $(\varepsilon_{\tau e_j})_{\tau \in \mathbb{R}_+}$ we get the j -th marginal semigroup, i.e.*

$$T_\tau^{(j)} u := \int_{\mathbb{R}_+^N} T_t u \varepsilon_{\tau e_j}(dt).$$

Example 14. *Considering two pseudo-differential operators $q_1(x, D)$ and $q_2(x, D)$ on $\mathcal{S}(\mathbb{R}^2)$, then*

$$[q_1(x, D), q_2(x, D)] = 0$$

holds for example, if

$$q_1(x, D) = \tilde{q}_1(x_1, D_{x_1})$$

and

$$q_2(x, D) = \tilde{q}_2(x_2, D_{x_2}),$$

with $\tilde{q}_1(x_1, D_{x_1})$ and $\tilde{q}_2(x_2, D_{x_2})$ being pseudo-differential operators on $\mathcal{S}(\mathbb{R})$.

Beyond N-Parameter Markov-Semigroups - A Case Study

Let $\psi : \mathbb{R}^n \longrightarrow \mathbb{C}$ be a continuous negative definite function. We consider

$$\phi_{s,t}(\xi) := (2\pi)^{-\frac{n}{2}} e^{-st\psi(\xi)},$$

which is for $s, t \geq 0$ a continuous and positive definite function, i.e. by Bochner's Theorem each $\phi_{s,t}$ is the Fourier transform of a measure on \mathbb{R}^n . This leads us to the definition of the following **two-parameter family of measures** $(\mu_{s,t})_{s,t \geq 0}$:

$$\hat{\mu}_{s,t}(\xi) := (2\pi)^{-\frac{n}{2}} e^{-st\psi(\xi)}, \quad (3)$$

and to the definition of a **two-parameter family of operators**

$$T_{s,t}u(x) := \int_{\mathbb{R}^n} u(x-y) \mu_{s,t}(dy).$$

Fix $s = s_0 \neq 0$, then

- $(\mu_{s_0,t})_{t \geq 0}$ is a convolution semigroup in t
- $(T_{s_0,t})_{t \geq 0}$ a semigroup of operators w.r.t. t

and for $s_0 = 0$ the operator $T_{s_0,t}$ is the identity.

The Bernstein function $f_1(x) = x^{1/2}$ is associated with the convolution semigroup

$$\eta_r^{(1)} = h_r(\cdot)\lambda^{(1)}, \quad h_r(x) = \chi_{(x>0)} \frac{1}{\sqrt{2\pi}} \cdot r \cdot x^{-\frac{3}{2}} \cdot e^{-\frac{r^2}{4x}}.$$

The Bernstein function $f_2(x) = \log(1+x)$ is associated with the convolution semigroup

$$\eta_r^{(2)} = g_r(\cdot)\lambda^{(1)}, \quad g_r(x) = \chi_{(x>0)} \frac{1}{\Gamma(r)} x^{r-1} e^{-x}.$$

Subordinating $(\mu_{s,t})_{s,t \geq 0}$ by the Bernstein function $f(x) = x^{1/2}$ with respect to parameter s gives us a new family of measures $(\nu_{r,t})_{r,t \geq 0}$, which can be characterized by its Fourier transform

$$\widehat{\nu}_{r,t}(\xi) = e^{-rt^{1/2}[\psi(\xi)]^{1/2}}.$$

Then subordinating $(\nu_{r,t})_{r,t \geq 0}$ by $g(x) = \log(1+x)$, which is a Bernstein function, with respect to parameter t leads to another family $(\tau_{r,p})_{r,t \geq 0}$ of measures:

$$\begin{aligned} \hat{\tau}_{r,p}(\xi) &= \frac{-\Gamma(2p)}{(\Gamma(p))^2} \cdot 2^{1-2p} \cdot [\psi(\xi)]^{(1/2)} \cdot \sqrt{\pi} \\ &\quad \times r \cdot {}_1F_1\left(\frac{1}{2} + p; \frac{3}{2}; \frac{1}{4}r^2\psi(\xi)\right) \\ &\quad + \frac{\Gamma(2p) \cdot 2^{1-2p}}{\Gamma(p)\Gamma(p + \frac{1}{2})} \cdot \sqrt{\pi} \\ &\quad \times {}_1F_1\left(p; \frac{1}{2}; \frac{1}{4}r^2\psi(\xi)\right), \end{aligned}$$

here ${}_1F_1$ is the confluent hypergeometric function. The following table gives $\hat{\tau}_{r,p}$, if $(\mu_{s,t})_{s,t \geq 0}$ is subordinated in the first step by $f(x) = x^{1/n}$.

$$\frac{1}{2} \left| \frac{e^{r^2\psi/4}}{\Gamma(p)} \left(\Gamma(p) {}_1F_1\left(\frac{1}{2} - p; \frac{1}{2}; -\frac{r^2\psi}{4}\right) - r\sqrt{\psi}\Gamma\left(\frac{1}{2} + p\right) {}_1F_1\left(1 - p; \frac{3}{2}; -\frac{r^2\psi}{4}\right) \right) \right|$$

$$\frac{1}{3} \left| \frac{1}{2\Gamma(p)} \left(2\Gamma(p) {}_1F_2\left(p; \frac{1}{3}, \frac{2}{3}; -\frac{r^3\psi}{27}\right) - 2r\psi^{1/3}\Gamma\left(\frac{1}{3} + p\right) {}_1F_2\left(\frac{1}{3} + p; \frac{2}{3}, \frac{4}{3}; -\frac{r^3\psi}{27}\right) \right. \right. \\ \left. \left. + r^2\psi^{2/3}\Gamma\left(\frac{2}{3} + p\right) {}_1F_2\left(\frac{2}{3} + p; \frac{4}{3}, \frac{5}{3}; -\frac{r^3\psi}{27}\right) \right) \right|$$

$$\frac{1}{4} \left| {}_1F_3\left(p; \frac{1}{4}, \frac{1}{2}, \frac{3}{4}; \frac{r^4\psi}{256}\right) - \frac{\Gamma\left(\frac{1}{4} + p\right)}{\Gamma(p)} r\psi^{1/4} {}_1F_3\left(\frac{1}{4} + p; \frac{1}{2}, \frac{3}{4}, \frac{5}{4}; \frac{r^4\psi}{256}\right) \right. \\ \left. + \frac{\Gamma\left(\frac{1}{2} + p\right)}{2\Gamma(p)} r^2\psi^{1/2} {}_1F_3\left(\frac{1}{2} + p; \frac{3}{4}, \frac{5}{4}, \frac{3}{2}; \frac{r^4\psi}{256}\right) \right. \\ \left. + \frac{\Gamma\left(\frac{3}{4} + p\right)}{6\Gamma(p)} r^3\psi^{3/4} {}_1F_3\left(\frac{3}{4} + p; \frac{5}{4}, \frac{3}{2}, \frac{7}{4}; \frac{r^4\psi}{256}\right) \right|$$

$$\frac{1}{5} \left| {}_1F_4\left(p; \frac{1}{5}, \frac{2}{5}, \frac{3}{5}, \frac{4}{5}; -\frac{r^5\psi}{3125}\right) - \frac{\Gamma\left(\frac{1}{5} + p\right)}{\Gamma(p)} r\psi^{1/5} {}_1F_4\left(\frac{1}{5} + p; \frac{2}{5}, \frac{3}{5}, \frac{4}{5}, \frac{6}{5}; -\frac{r^5\psi}{3125}\right) \right. \\ \left. + \frac{\Gamma\left(\frac{2}{5} + p\right)}{2\Gamma(p)} r^2\psi^{2/5} {}_1F_4\left(\frac{2}{5} + p; \frac{3}{5}, \frac{4}{5}, \frac{6}{5}, \frac{7}{5}; -\frac{r^5\psi}{3125}\right) \right. \\ \left. + \frac{\Gamma\left(\frac{3}{5} + p\right)}{6\Gamma(p)} r^3\psi^{3/5} {}_1F_4\left(\frac{3}{5} + p; \frac{4}{5}, \frac{6}{5}, \frac{7}{5}, \frac{8}{5}; -\frac{r^5\psi}{3125}\right) \right. \\ \left. + \frac{\Gamma\left(\frac{4}{5} + p\right)}{24\Gamma(p)} r^4\psi^{4/5} {}_1F_4\left(\frac{4}{5} + p; \frac{6}{5}, \frac{7}{5}, \frac{8}{5}, \frac{9}{5}; -\frac{r^5\psi}{3125}\right) \right|$$

We start again with the two-parameter family defined earlier, now we subordinate by $f(x) = x^{1/2}$ with respect to s and t . Using the Meijer-G-function we can represent $\hat{\tau}_{r,p}$ as follows

$$\hat{\tau}_{r,p}(\xi) = \frac{r \cdot [\psi(\xi)]^{1/2}}{4\pi} G_{0,3}^{3,0} \left(\begin{matrix} - & - & - \\ -\frac{1}{2} & 0 & 0 \end{matrix} \middle| \frac{1}{16} p^2 r^2 \psi \right)$$

$$\frac{1}{2} \frac{pr}{4\pi} \times \sqrt{\psi} \mathbf{G}_{0,3}^{3,0} \left(-\frac{1}{2} \quad 0 \quad 0 \quad \frac{1}{16} p^2 r^2 \psi \right)$$

$$\frac{1}{3} \frac{1}{4\pi} {}_0F_3 \left(\frac{1}{3}, \frac{1}{2}, \frac{2}{3}; \frac{1}{108} p^2 r^3 \psi \right) - 2\sqrt{3}r \sqrt[3]{p^2 \psi} \Gamma \left(\frac{1}{3} \right)^2 {}_0F_3 \left(\frac{2}{3}, \frac{5}{6}, \frac{4}{3}; \frac{1}{108} p^2 r^3 \psi \right) \\ + 8\pi \sqrt{p^2 r^3 \psi} {}_0F_3 \left(\frac{5}{6}, \frac{7}{6}, \frac{3}{2}; \frac{1}{108} p^2 r^3 \psi \right) + 3\sqrt{\pi} (2p^2 r^3 \psi)^{(2/3)} \Gamma \left(\frac{5}{6} \right) {}_0F_3 \left(\frac{7}{6}, \frac{4}{3}, \right)$$

$$\frac{1}{4} \frac{pr^2 \sqrt{\psi}}{32\sqrt{2}\pi^2} \times \mathbf{G}_{0,5}^{5,0} \left(-\frac{1}{2} \quad -\frac{1}{4} \quad 0 \quad 0 \quad \frac{1}{4} ; \frac{1}{1024} p^2 r^4 \psi \right)$$

$$\frac{1}{5} \frac{1}{2\sqrt{\pi}} \left(5p \left(\frac{2}{5} \sqrt{\frac{\pi}{p^2}} {}_0F_5 \left(\frac{1}{5}, \frac{2}{5}, \frac{1}{2}, \frac{3}{5}, \frac{4}{5}, \frac{p^2 r^5 \psi}{12500} \right) \right. \right. \\ - p^{-3/5} \sqrt{\frac{10}{5\pi + \sqrt{5}\pi}} r \psi^{1/5} \Gamma \left(\frac{3}{5} \right) \Gamma \left(\frac{6}{5} \right) {}_0F_5 \left(\frac{2}{5}, \frac{3}{5}, \frac{7}{10}, \frac{4}{5}, \frac{6}{5}, \frac{p^2 r^5 \psi}{12500} \right) \\ + \frac{1}{\sqrt{10(5-\sqrt{5})\pi}} \frac{r^2 \psi^{2/5}}{p^{1/5}} \Gamma \left(\frac{1}{5} \right) \Gamma \left(\frac{2}{5} \right) {}_0F_5 \left(\frac{3}{5}, \frac{4}{5}, \frac{9}{10}, \frac{6}{5}, \frac{7}{5}, \frac{p^2 r^5 \psi}{12500} \right) \\ - \frac{8\sqrt{\pi r^5 \psi}}{15} {}_0F_5 \left(\frac{7}{10}, \frac{9}{10}, \frac{11}{10}, \frac{13}{10}, \frac{3}{2}, \frac{p^2 r^5 \psi}{12500} \right) \\ \left. \left. \left. \frac{r^3 \psi^{3/5}}{30} \left(\frac{p}{2} \right)^{1/5} \Gamma \left(-\frac{1}{10} \right) {}_0F_5 \left(\frac{4}{5}, \frac{11}{10}, \frac{6}{5}, \frac{7}{5}, \frac{8}{5}, \frac{p^2 r^5 \psi}{12500} \right) \right. \right. \right. \\ \left. \left. \left. \frac{r^4 \psi^{4/5}}{120} \left(\frac{p}{2} \right)^{3/5} \Gamma \left(-\frac{3}{10} \right) {}_0F_5 \left(\frac{6}{5}, \frac{13}{10}, \frac{7}{5}, \frac{8}{5}, \frac{9}{5}, \frac{p^2 r^5 \psi}{12500} \right) \right) \right) \right)$$

Constructing a Process

Starting with

$$\hat{\mu}_t(\xi) = (2\pi)^{-n/2} e^{-t_1 t_2 \psi(\xi)}, \text{ for all } t = (t_1, t_2) \in \mathbb{R}_+^2,$$

with ψ continuous, negative definite and being zero at the origin, we construct a projective family of probability measures over \mathbb{R}^n . Let

$$u_1 \preceq u_2 \preceq \dots \preceq u_k$$

and

$$\{u_1, u_2, \dots, u_k\} = K \in \mathcal{H},$$

for the set \mathcal{H} of all finite subset of \mathbb{R}_+^2 .

For K we define the measure P_K on $(\mathbb{R}^n)^k$ by

$$\begin{aligned} P_K(A_1 \times A_2 \times \dots \times A_k) &= \\ &= \int_{\mathbb{R}^n} \int_{A_1} \dots \int_{A_k} p(u_{k-1}, u_k, x_{k-1}, dx_k) \dots \\ &\quad \times p(u_1, u_2, x_1, dx_2) \cdot p(0, u_1, x, dx_1) \nu(dx), \end{aligned}$$

for all $A_1, \dots, A_k \in \mathcal{B}(\mathbb{R}^n)$.

Here

$$p(u_j, u_{j+1}, x_j, A) = \mu_{u_j, u_{j+1}}(A - x_j),$$

for all $A \in \mathcal{B}^{(n)}$. Where $\mu_{u_j, u_{j+1}}(A - x_j)$ is defined by:

$$\hat{\mu}_{u_j, u_{j+1}}(\xi) = (2\pi)^{-\frac{n}{2}} e^{-\left(u_{j+1}^{(1)} u_{j+1}^{(2)} - u_j^{(1)} u_j^{(2)}\right) \psi(\xi)}.$$

This family is projective, and applying Kolomogorov's Theorem it follows the existence of a probability measure P on $\mathcal{B}^{(n)}$ satisfying

$$\pi_K(P) = P_K, \text{ for all } K \in \mathcal{H},$$

and the existence of a stochastic process with state space \mathbb{R}^n , whose finite-dimensional distributions are given by $(P_K)_{K \in \mathcal{H}}$.

More general, we can consider $(2\pi)^{-\frac{n}{2}}e^{-k(t,\xi)}$, where we have to assume, that for fixed $t \in \mathbb{R}_+^N$ $k(t, \cdot)$ is a continuous negative definite function and for all $t \succeq s$ and $\xi \in \mathbb{R}^n$

$$k(t, \xi) \geq k(s, \xi).$$