

Fractional Diffusions, Lévy Processes*

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 - ★ **Pure:** Investigations on the “fine” potential and spectral theoretic properties for subclasses of Lévy processes

Some References

- **General Theory**

1. J. Bertoin, *Lévy Processes*, Cambridge University Press, 1996.
2. D. Applebaum. *Lévy Processes and Stochastic Calculus*, Cambridge University Press, 2004

- **Applications**

1. B. Øksendal and A. Sulem. *Applied Stochastic Control of Jump Diffusions*, Springer, 2004
2. R. Cont and P. Tankov, *Financial Modelling with Jump Processes*, Chapman & Hall/CRC, 2004

- **Recent analytical, probabilistic, geometric, developments:** Work of R. Bass, K. Burdzy, K. Bogdan, Z. Chen, T. Kulczycki, R. Song, J-M. Wu, . . .

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Outline

- **Definitions, Structure**
- **Examples**
- **Semigroups, generator, . . .**
- **“Boundary” value problems for rotationally invariant stable processes– the boundary value problems for the “fractional” Laplacian, Some Questions, Some Results**
- **Applications to problems for the Laplacian (Brownian motion) which result from the investigations of the more general operators.**

Definition

A **Lévy Processes** is a stochastic process $\eta = (\eta_t), t \geq 0$ with

- η has independent and stationary increments
- Each $\eta_0 = 0$ (with probability 1)
- η is *stochastically continuous*: For all $\varepsilon > 0$,

$$\lim_{t \rightarrow s} P\{|\eta_t - \eta_s| > \varepsilon\} = 0$$

Note: Not the same as a.s. continuous paths. However, it gives “cadlag” paths:
Right continuous with left limits.

- **Stationary increments:** $0 \leq s \leq t < \infty$, $A \in \mathbb{R}^n$ Borel

$$P\{\eta_t - \eta_s \in A\} = P\{\eta_{t-s} - \eta_0 \in A\}$$

- **Independent increments:** For any given sequence of ordered times

$$0 \leq t_1 \leq t_2 \cdots \leq t_m < \infty,$$

the random variables

$$\eta_{t_1} - \eta_0, \eta_{t_2} - \eta_{t_1}, \dots, \eta_{t_m} - \eta_{t_{m-1}}$$

are independent.

The characteristic function of η_t is

$$\varphi_t(\xi) = E \left(e^{i\xi \cdot X_t} \right) = \int_{\mathbb{R}^n} e^{i\xi \cdot x} p_t(dx)$$

where p_t is the distribution of η_t .

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Lévy-Khintchine Formula: $\varphi_t(\xi) = e^{t\rho(\xi)}$ where

$$\rho(\xi) = ib \cdot \xi - \frac{1}{2}\xi \cdot A\xi + \int_{\mathbb{R}^n \setminus \{0\}} \left(e^{i\xi \cdot x} - 1 - i\xi \cdot x 1_{\{|x| < 1\}}(x) \right) \nu(dx)$$

for some $b \in \mathbb{R}^n$, a non-negative definite symmetric $n \times n$ matrix A and a Borel measure ν on $\mathbb{R}^n \setminus \{0\}$ with

$$\int_{\mathbb{R}^n \setminus \{0\}} \min(|x|^2, 1) \nu(dx) < \infty$$

$\rho(\xi)$ is called the *symbol* of the process or the *characteristic exponent*. The triple (b, A, ν) is called the *characteristics of the process*.

Converse also true. Given such a triple we can construct a Lévy process.

Examples

1. Standard Brownian motion:

With $(0, I, 0)$, I the identity matrix,

$$\eta_t = B_t, \quad \text{Standard Brownian motion}$$

2. Gaussian Processes, “General Brownian motion”:

$(0, A, 0)$, η_t is “generalized” Brownian motion, mean zero, covariance

$$E(\eta_s^j \eta_t^i) = a_{ij} \min(s, t)$$

X_t has the normal distribution

$$\frac{1}{(2\pi t)^{n/2} \sqrt{\det(A)}} \exp\left(-\frac{1}{2t} x \cdot A^{-1} x\right)$$

3. “Brownian motion” plus drift: With $(b, A, 0)$ get Brownian motion with a drift:

$$\eta_t = bt + B_t$$

4. **Poisson Process:** The Poisson Process $\eta_t = N_\lambda(t)$ of intensity $\lambda > 0$ is a Lévy process with $(0, 0, \lambda\delta_1)$ where δ_1 is the Dirac delta at 1.

$$P\{N_\lambda(t) = m\} = \frac{e^{-\lambda t}(\lambda t)^m}{m!}, \quad m = 1, 2, \dots$$

$N_\lambda(t)$ has continuous paths except for jumps of size 1 at the random times

$$\tau_m = \inf\{t > 0 : N_\lambda(t) = m\}$$

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6. **Relativistic Brownian motion** According to quantum mechanics, a particle of rest mass m moving with momentum p has kinetic energy

$$E(p) = \sqrt{m^2c^4 + c^2|p|^2} - mc^2$$

where c is speed of light. Then $\eta(p) = -E(p)$ is the symbol of a Lévy process, called “*relativistic Brownian motion.*”

7. **The zeta process:** Consider the Riemann zeta function

$$\zeta(z) = \sum_{n=1}^{\infty} \frac{1}{n^z} = \prod_{p \in \mathbb{P}} \frac{1}{1 - p^{-z}}, \quad z = x + iy \in \mathbb{C}$$

Khintchine: For every fix $x > 1$,

$$\rho_x(y) = \log \left(\frac{\zeta(x + iy)}{\zeta(x)} \right)$$

is the symbol of a Lévy process.

Biane-Pitman-Yor: “Probability laws related to the Jacobi theta and Riemann Zeta functions and Brownian excursions, Bull. Amer math. Soc., 2001.

M Yor: A note about Selberg’s integrals with relation with the beta–gamma algebra, 2006.

8. **The rotationally invariant Lévy stable processes:** These are self-similar processes in \mathbb{R}^n with symbol

$$\rho(\xi) = -|\xi|^\alpha, \quad 0 < \alpha \leq 2.$$

That is,

$$\varphi_t(\xi) = E \left(e^{i\xi \cdot X_t} \right) = e^{-t|\xi|^\alpha}$$

$\alpha = 2$ is *Brownian motion*. $\alpha = 1$ is the *Cauchy processes*. $\alpha = 3/2$ is called the *Haltmark distribution* used to model gravitational fields of stars.

Transition probabilities:

$$P_x\{\eta_t \in A\} = \int_A p_t^\alpha(x - y) dy, \quad \text{any Borel } A \subset \mathbb{R}^n$$

$$p_t^\alpha(x) = \int_{\mathbb{R}^n} e^{-i\xi \cdot x} e^{-t|\xi|^\alpha} d\xi$$

$$p_t^2(x) = \frac{1}{(4\pi t)^{n/2}} e^{-\frac{|x|^2}{4t}}, \quad \alpha = 2, \quad \text{Brownian motion}$$

$$p_t^1(x) = \frac{C_n t}{(|x|^2 + t^2)^{\frac{n+1}{2}}}, \quad \alpha = 1, \quad \text{Cauchy Process}$$

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For any $a > 0$, the two processes

$$\{\eta_{(at)}; t \geq 0\} \quad \text{and} \quad \{a^{1/\alpha} \eta_t; t \geq 0\},$$

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In the same way, the transition probabilities scale similarly to those for BM:

$$p_t^\alpha(x) = t^{-n/\alpha} p_1^\alpha(t^{-1/\alpha} x)$$

Lévy–Itô decomposition

Lévy processes can be written as

$$\eta(t) = \eta^c(t) + \sum_{s:0 \leq s \leq t} \Delta\eta(s)$$

where

$$\Delta\eta(s) = \eta(s) - \eta(s^-) = \text{jump at time } s.$$

The *Poisson random measure*, aka “jump measure” is

$$N(t, U) = N(t, U, \omega) = \sum_{s:0 < s \leq t} \chi_U(\Delta\eta(s))$$

Lévy–Itô: The Lévy process $X(t)$ has the representation

$$\eta(t) = bt + B_A(t) + \int_{\{0 < |x| < 1\}} x \left[N(t, dx) - t d\nu(x) \right] + \int_{\{|x| \geq 1\}} x N(t, dx)$$

A Lévy stochastic differential equation (Lévy SDE) is of the form

$$dX(t) = \alpha(t, X(t))dt + \sigma(t, X(t))dB(t) + \int \gamma(t, X(t), z)\tilde{N}(t, dz)$$

(with certain “Lipschitz” assumptions on the coefficients it has a unique solution).

Solutions to Lévy SDE’s in the *time homogeneous case*, i.e.,

$$\alpha(t, x) = \alpha(x)$$

$$\sigma(t, x) = \sigma(x)$$

$$\gamma(t, x, z) = \gamma(x, z)$$

are called a **“jump diffusion”** or **“Lévy diffusion”**

The diffusion generator is:

$$\begin{aligned} Af(x) &= \lim_{t \rightarrow 0} \frac{1}{t} \left(E_x[f(X(t))] - f(x) \right) \\ &= \sum_{i=1}^n \alpha_j(x) \frac{\partial f}{\partial x_i} + \frac{1}{2} \sum_{i,j=1}^n (\sigma \sigma^T)(x) \frac{\partial^2 f}{\partial x_i \partial x_j}(x) \\ &+ \int \left[f(x + \gamma(x, z)) - f(x) - \nabla f(x) \cdot \gamma(x, z) \right] \nu(dz) \end{aligned}$$

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For the Lévy process $\{\eta(t); t \geq 0\}$,

$$T_t f(x) = E[f(\eta(t)) | \eta_0 = x] = E_x[f(\eta(t))], \quad f \in \mathcal{S}(\mathbb{R}^n)$$

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$$Af(x) = \int_{\mathbb{R}^n} e^{-ix \cdot \xi} \rho(\xi) \hat{f}(\xi) d\xi$$

Using the Lévy–Khintchine formula one can show

$$Af(x) = \sum_{i=1}^n b_i \partial_i f(x) + \frac{1}{2} a_{i,j} \partial_i \partial_j f(x) + \int \left[f(x+y) - f(x) - y \cdot \nabla f(x) \chi_{\{|y|<1\}} \right] \nu(dy)$$

Examples:

- Standard Brownian motion:

$$Af(x) = \frac{1}{2} \Delta f(x)$$

- Poisson Process of intensity λ :

$$Af(x) = \lambda \left[f(x+1) - f(x) \right]$$

- Rotationally Invariant Stable Processes of order $0 < \alpha < 2$, **Fractional Diffusions:**

$$\begin{aligned} Af(x) &= -(-\Delta)^{\alpha/2} f(x) \\ &= C_{\alpha,n} \lim_{\varepsilon \rightarrow 0} \int_{\{|x-y|>\varepsilon\}} \frac{f(y) - f(x)}{|x-y|^{n+\alpha}} dy \end{aligned}$$

Fractional Diffusions in regions of \mathbb{R}^n

From now on $\eta(t)$ is rotationally invariant stable with symbol

$$\rho(\xi) = -|\xi|^\alpha, \quad 0 < \alpha < 2.$$

Let D be a bounded connected subset of \mathbb{R}^n . The first exit time of $\eta(t)$ from D is

$$\tau_D = \inf\{t > 0 : \eta(t) \notin D\}$$

Heat Semigroup in D

$$T_t^D f(x) = E_x \left[f(\eta(t)); \tau_D > t \right], \quad f \in L^2(D)$$

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$$p_t^D(x, y) = p_t^\alpha(x, y) - E^x(\tau_D < t; p_{t-\tau_D}^\alpha(\eta(\tau_D), y)).$$

$p_t^D(x, y)$ is called the **Heat Kernel for the stable process in D** .

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The general theory of heat semigroups gives an orthonormal basis of eigenfunctions

$$\{\varphi_m\}_{m=1}^\infty \text{ on } L^2(D)$$

with eigenvalues $\{\lambda_m\}$ satisfying

$$0 < \lambda_1 < \lambda_2 \leq \lambda_3 \leq \dots \rightarrow \infty$$

That is,

$$T_t^D \varphi_m(x) = e^{-\lambda_m t} \varphi_m(x), \quad x \in D.$$

From general heat semigroup theory: λ_1 is simple, φ is strictly positive in D , the spectral gap $\lambda_2 - \lambda_1 > 0$.

Some “basic” problems

- Regularity of heat kernels, eigenfunctions, general solutions of “heat equation”, Sobolev, log-Sobolev inequalities, “intrinsic ultracontractivity,” . . .
- “Boundary” regularity of solutions
- Estimates on eigenvalues, including the ground state λ_1 and the spectral gap $\lambda_2 - \lambda_1$.
- Number of “nodal” domains (Courant–Hilbert Nodal domain Theorem).
- Geometric properties of eigenfunctions, including a “Brascamp–Lieb” log-concavity type theorem for φ_1 .

Many open even when the domain $D = (-1, 1)$.

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Theorem 3: The Faber–Krahn inequality: For any region the eigenvalue is minimized for a ball of same volume.

$$\lambda_1(D^*) \leq \lambda_1(D), \quad D^* = \text{ball of same volume as } D$$

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Theorem 4: Brascamp–Lieb true for the Cauchy processes ($\alpha = 1$ case) in an interval in one dimension.

Only known case!

The Faber–Krahn inequality (finite dimensional distributions)

Apply the semigroup to the function $f(x) = 1, x \in D$

$$T_t^D f(x) = E_x[\chi_D(\eta_t); \tau_D > t] = \int_D p_t(x, y) dy$$

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$$= \sum_{m=1}^{\infty} e^{-t\lambda_m} \varphi_m(x) \int_D \varphi_m(y) dy$$

$$\Rightarrow \lim_{t \rightarrow \infty} \frac{1}{t} \log P_x\{\tau_D > t\} = -\lambda_1$$

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More:

$$\lim_{t \rightarrow \infty} e^{t\lambda_1} P_x\{\tau_D > t\} = \varphi_1(x) \int_D \varphi_1(y)$$

Finite Dimensional Distributions

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More general: Study the finite dimensional distributions:

$$P_x\{\eta(t_1) \in D_1, \eta(t_2) \in D_2, \dots, \eta(t_m) \in D_m\}$$

for any **sequence of times and any sequence of domains**:

$$0 < t_1 < t_2 < \dots < t_m; \quad D_1 \subset \mathbb{R}^n, \dots, D_m \subset \mathbb{R}^n,$$

as a function of $x \in D_1$ and the domains.

Sample Theorem 1: Take $D_1 = D_2 = \dots = D_m = D$, $x \in D$.

$$\begin{aligned} & P_x\{\eta(t_1) \in D, \eta(t_2) \in D, \dots, \eta(t_m) \in D\} \\ & \leq P_0\{\eta(t_1) \in D^*, \eta(t_2) \in D^*, \dots, \eta(t_m) \in D^*\} \end{aligned}$$

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Proof: From Brascamp–Lieb–Luttinger re-arrangments of multiple integrals.

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$$\begin{aligned} & P_x\{\eta(t_1) \in D, \eta(t_2) \in D, \dots, \eta(t_m) \in D\} \\ & \leq P_0\{\eta(t_1) \in D^*, \eta(t_2) \in D^*, \dots, \eta(t_m) \in D^*\} \end{aligned}$$

Proof: From Brascamp–Lieb–Luttinger re-arrangments of multiple integrals.

Sample Theorem 2: Let $I = (-1, 1)$. For any order sequence of times t_1, t_2, \dots, t_m ,

$$F(x) = P_x\{\eta(t_1) \in I, \eta(t_2) \in I, \dots, \eta(t_m) \in I\}, \quad x \in I$$

is unimodal in I with mode at 0 and “mid-concave”

Conjecture on Dirichlet Spectral gaps of Schrödinger operators

$H = -\Delta + V$ with Dirichlet conditions in the bounded convex domain $D \subset \mathbb{R}^n$ of finite diameter d_D , $V \geq 0$ is bounded and convex in D . We have eigenvalues

$$0 < \lambda_1(D, V) < \lambda_2(D, V) \leq \lambda_3(D, V) \dots$$

Conjecture (M. van den Berg 1983, Ashbaugh–Benguria 1987, and Problem #44 in Yau's 1990 "open problems in geometry")

$$gap(D, V) = \lambda_2(D, V) - \lambda_1(D, V) > \frac{3\pi^2}{d_D^2}$$

with the lower bound approached when $V = 0$ and the domain becomes a thin rectangular box.

False for non-convex domains even with $V = 0$.

I.Singer–B.Wang–S.T.Yau–S.S.T.Yau (1985) : Using the “p-function” (Max Principle) techniques of Payne, Philippin, . . .

$$\frac{\pi^2}{4d_D^2} \leq \lambda_2(D, V) - \lambda_1(D, V)$$

Many extensions–best lower bound until recently: $\frac{\pi^2}{d_D^2}$

Using inequalities for finite dimensional distributions (which come out of the study of stable processes) leads to the conjecture under some symmetry for D and some potentials:

$$\lambda_2(D, V) - \lambda_1(D, V) \geq \frac{3\pi^2}{d_D^2}$$

Work with P. Médez–Hernández.

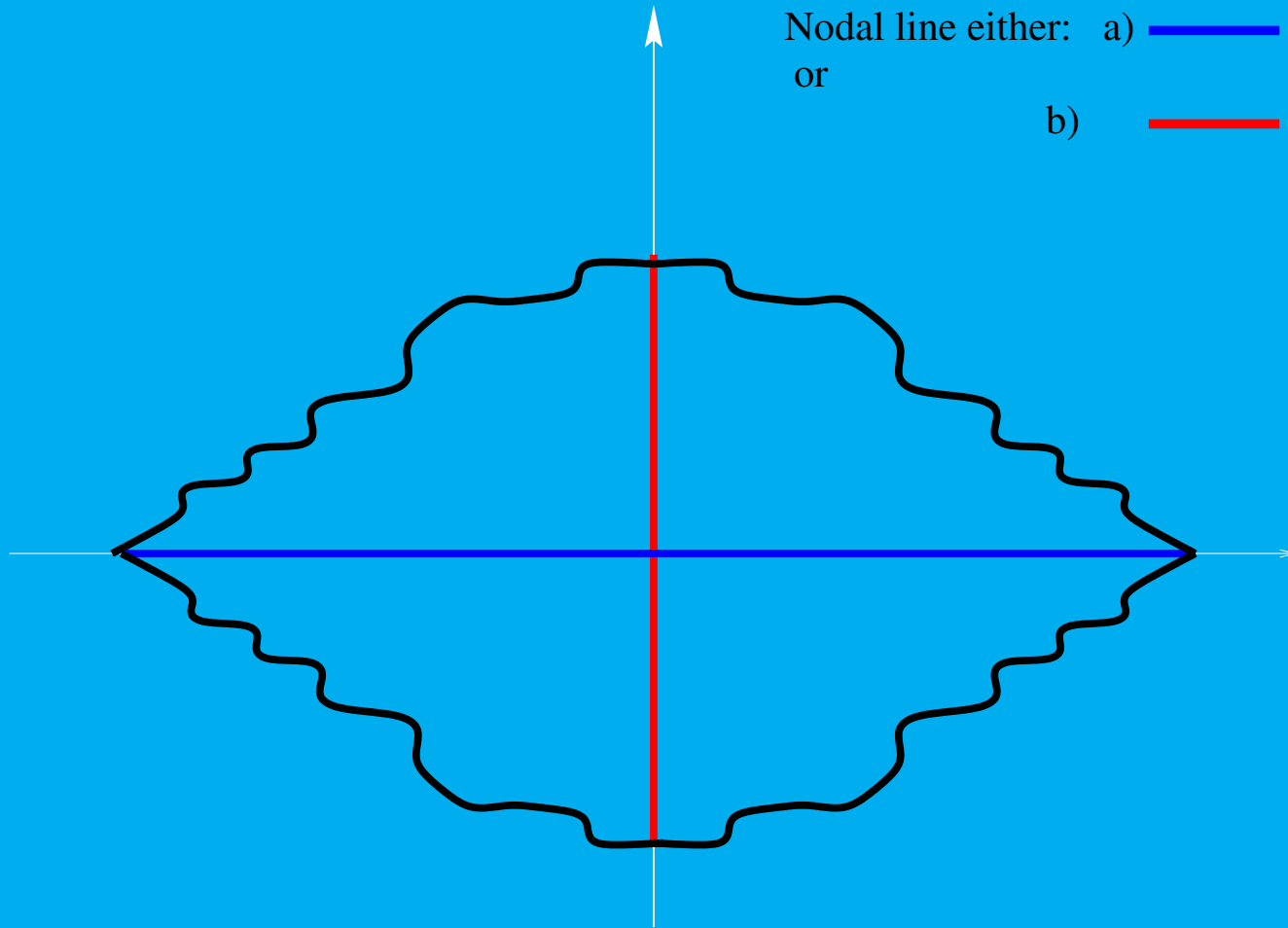
Motivated by the hot–spots conjecture of J. Rauch for the Neumann problem

Conjecture: For any convex domain in $D \subset \mathbb{R}^n$, the function

$$\Psi(x) = \frac{\varphi_2(x)}{\varphi_1(x)}$$

attains its maximum and minimum on the boundary, and only on the boundary, of D .

HERE we are dealing with the eigenfunctions of the “ordinary” Laplacian



$$D^+ = D \cap \{(x, y) \in \mathbb{R}^2 : y > 0\},$$

$$D^- = D \cap \{(x, y) \in \mathbb{R}^2 : y < 0\}.$$

