

Heat kernel estimates for subordinate Brownian motions

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Probability and Analysis

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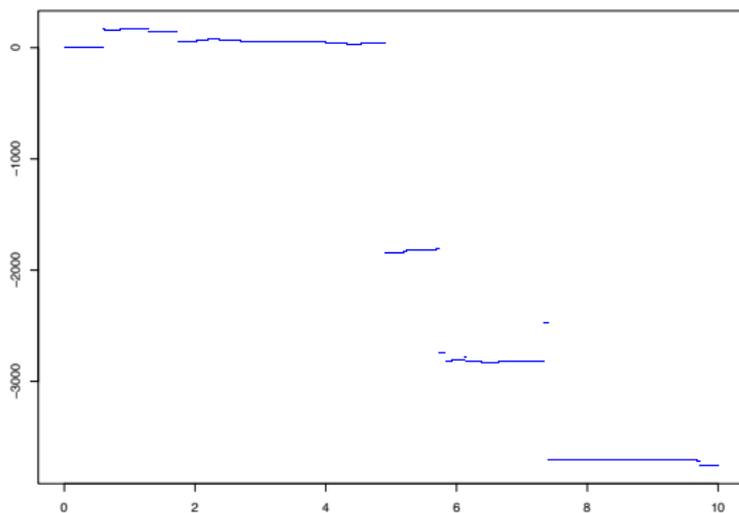
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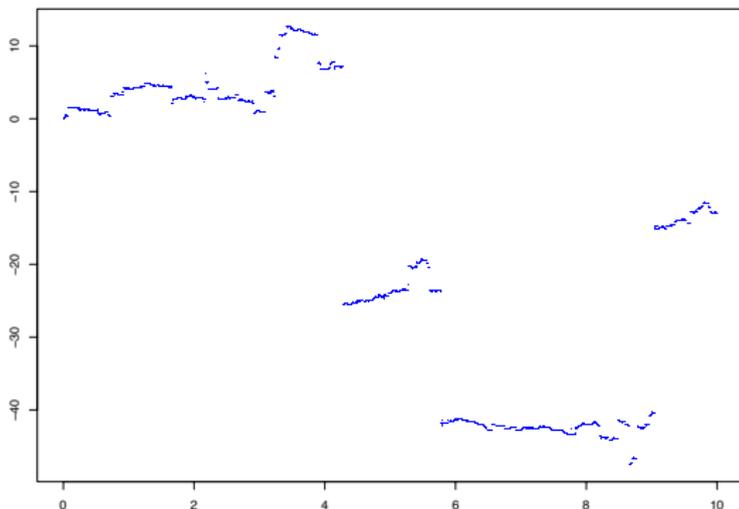
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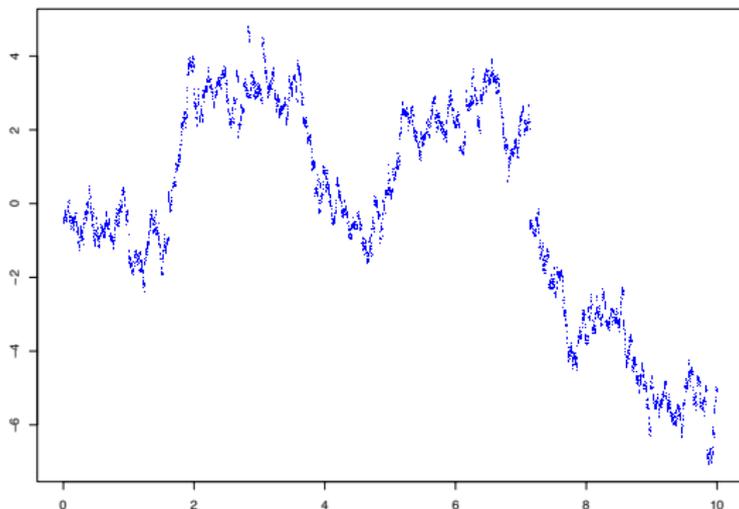
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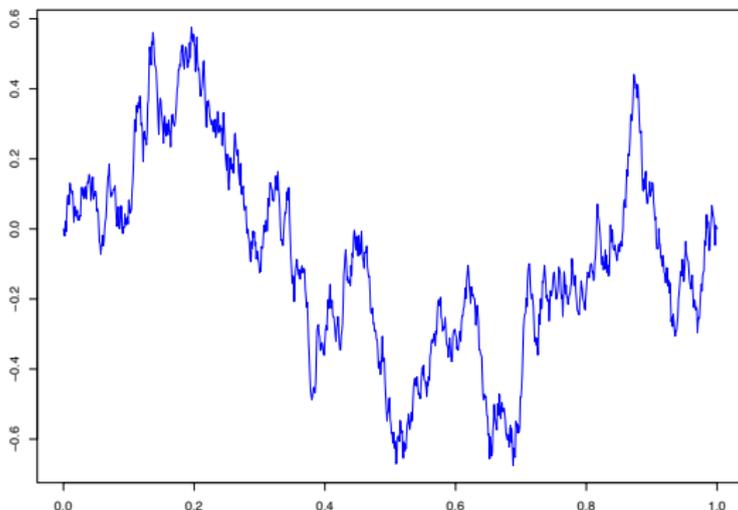
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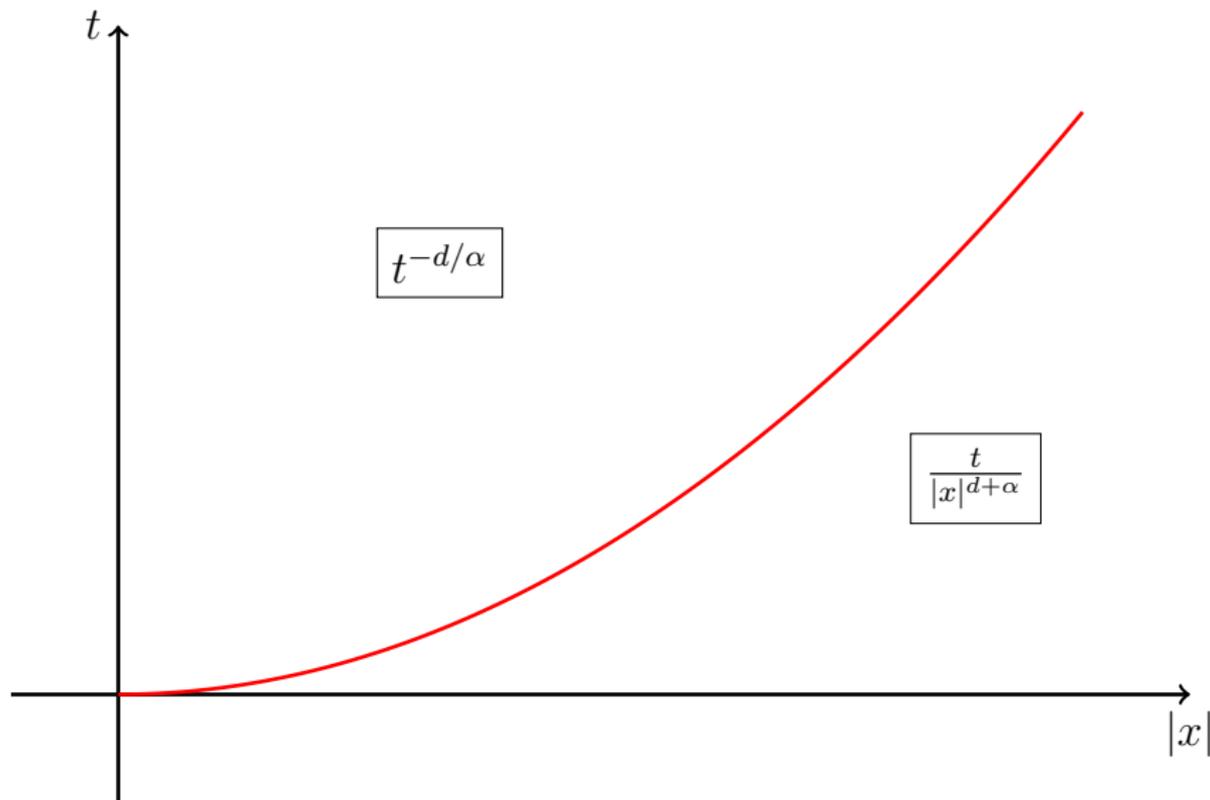
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$$t|x|^{-\alpha} = 1$$

Subordinator

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$$\phi(\lambda) = d\lambda + \int_{(0, \infty)} (1 - e^{-\lambda t}) \mu(dt)$$

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- ϕ is a Bernstein function: $\phi \geq 0$ and

$$(-1)^{n+1} \phi^{(n)}(\lambda) \geq 0 \quad \text{for all } n \in \mathbb{N} \text{ and } \lambda > 0$$

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- transition density $\mathbb{P}_x(X_t \in A) = \int_A p(t, x, y) dy$

$$p(t, x, y) = p(t, y-x) = \int_{(0, \infty)} (4\pi t)^{-d/2} e^{-\frac{|x-y|^2}{4s}} \mathbb{P}(S_t \in ds)$$

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$$-(-\Delta)^{\alpha/2} u(x) = \int_{\mathbb{R}^d} (u(x+y) - u(x) - \langle \nabla u(x), y \rangle 1_{\{|y| \leq 1\}}) \underbrace{\frac{C_{d,\alpha}}{|y|^{d+\alpha}} dy}_{\text{Lévy measure}}$$

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- asymptotic expression for J **not** comparable to $\frac{\phi(|y|^{-2})}{|y|^d}$

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There exist $\gamma > 0$, $\lambda_L \geq 0$ and $C_L > 0$ such that

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Upper scaling condition (U)

There exist $\delta > 0$, $\lambda_U \geq 0$ and $C_U > 0$ such that

$$\frac{f(\lambda x)}{f(\lambda)} \leq C_U x^\delta \quad \text{for all } \lambda > \lambda_U \text{ and } x \geq 1.$$

- Model examples are functions f that vary regularly at infinity with index $\rho > 0$, i.e.

$$\lim_{\lambda \rightarrow \infty} \frac{f(\lambda x)}{f(\lambda)} = x^\rho \quad \text{for all } \lambda > 0 \text{ and } x > 0.$$

since for any $\varepsilon > 0$ there exists $C > 0$ and $\lambda_0 > 0$ such that

$$C^{-1} \left(\frac{t}{s}\right)^{\rho-\varepsilon} \leq \frac{f(t)}{f(s)} \leq C \left(\frac{t}{s}\right)^{\rho+\varepsilon}, \quad \lambda_0 \leq s \leq t$$

(Theorem of Potter).

- If f varies at 0 with index $\tilde{\rho} > 0$, then $\lambda_L = \lambda_U = 0$.
- $f(\lambda) = \log(1 + \lambda)$ does not satisfy **(L)**
- Bernstein function ϕ always satisfies **(U)** with $\delta = 1$:

$$\phi(\lambda x) \leq \phi(\lambda)x \quad \text{for all } \lambda > 0 \text{ and } x \geq 1.$$

We define a function $H : (0, \infty) \rightarrow (0, \infty)$ by

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- $\frac{H(\lambda x)}{H(\lambda)} \leq x^2$ for all $\lambda > 0$ and $x \geq 1$

Theorem (Heat kernel estimates, M. '15)

Assume that ϕ satisfies **(L)** and H satisfies **(L)** and **(U)** with $\delta < 2$. Then there exist constants $\kappa_1, \kappa_2, a_L, a_U > 0$ such that for $0 < t < \kappa_1 \phi(\lambda_{L,\phi})^{-1}$ and

$$|x| < \kappa_2 \sqrt{\lambda_{L,H}^{-1} \wedge \lambda_{U,H}^{-1}}$$

$$p(t, x) \asymp \begin{cases} \phi^{-1}(t^{-1})^{d/2} & t\phi(|x|^{-2}) > 1 \\ \frac{tH(|x|^{-2}) \vee e^{-a|x|^2\phi^{-1}(t^{-1})}}{|x|^d} & t\phi(|x|^{-2}) \leq 1, \end{cases}$$

where $a = a_L$ for the lower and $a = a_U$ for the upper bound.

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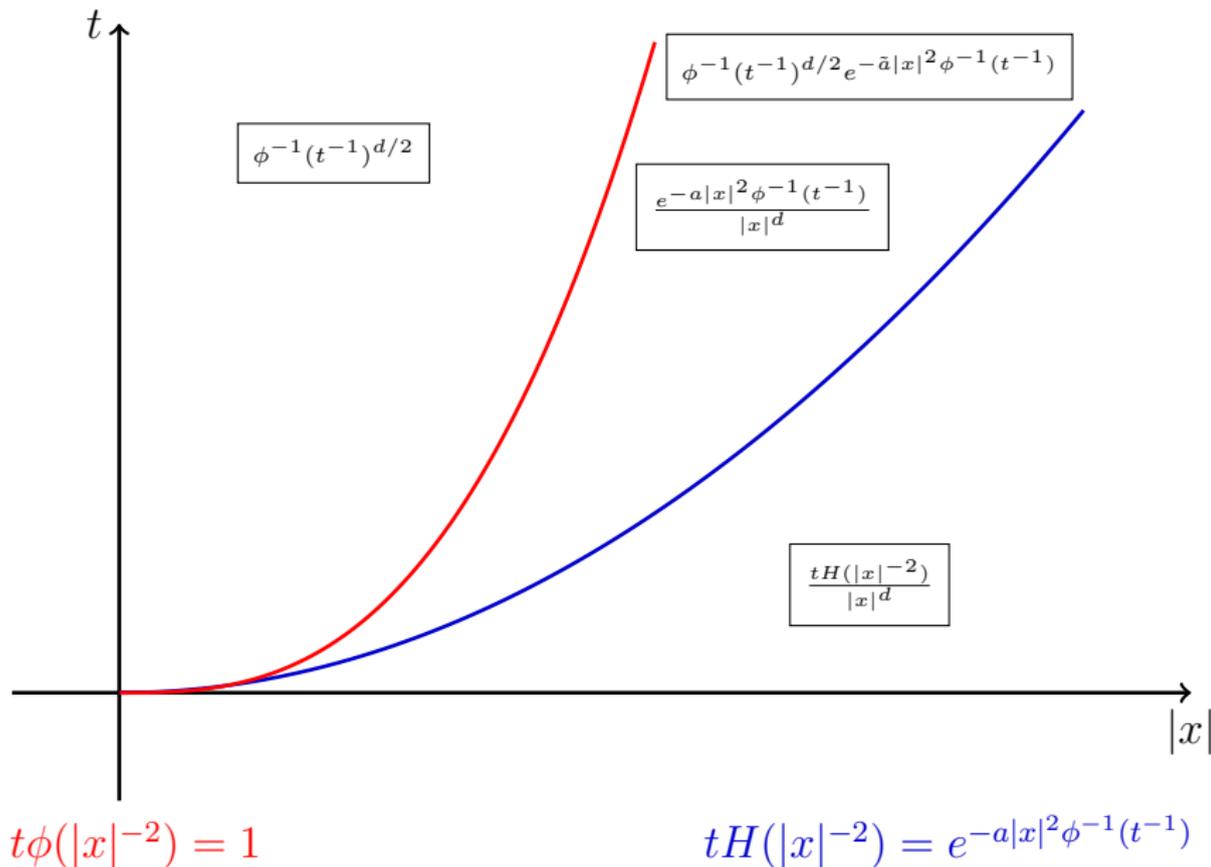
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- $\lambda_{L,\phi} = \lambda_{L,H} = \lambda_{U,H} = 0 \implies$ estimates are global in time and space



'Classical' case

If the scaling is strictly between 0 and 2 we get known estimate:

Corollary

If the Laplace exponent ϕ satisfies **(L)** and **(U)** with $\delta < 1$, then there exists $\kappa > 0$ such that for $0 < t < \phi(\lambda_L)^{-1}$ and $|x| < \kappa\lambda_U^{-1/2}$

$$\begin{aligned} p(t, x) &\asymp \begin{cases} \phi^{-1}(t^{-1})^{d/2} & t\phi(|x|^{-2}) > 1 \\ \frac{t\phi(|x|^{-2})}{|x|^d} & t\phi(|x|^{-2}) \leq 1 \end{cases} \\ &\asymp \phi^{-1}(t^{-1})^{d/2} \wedge \frac{t\phi(|x|^{-2})}{|x|^d}. \end{aligned}$$

Second estimate follows from the following observation

$$t\phi(|x|^{-2}) \leq 1 \iff \phi^{-1}(t^{-1})^{d/2} \geq |x|^{-d}.$$

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- off-diagonal estimate for 'small time and space':

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- Kaleta, Sztonyk '14

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The on-diagonal upper bound

$$p(t, x) \leq C\phi^{-1}(t^{-1})^{d/2}, \quad 0 < t < \phi(\lambda_L)^{-1}, \quad x \in \mathbb{R}^d$$

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$$\implies \frac{\phi^{-1}(at^{-1})}{\phi^{-1}(t^{-1})} \leq b \text{ for all } 0 < t < \phi(\lambda_L)^{-1} \implies \mathbf{(L)} \quad \square$$

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$$\begin{aligned} \mathbb{P}(|X_t| \geq r/2) &\leq \int_{t\phi(2y(r/2)^{-2}) \leq \frac{1}{2}} 2etH(2y(r/2)^{-2}) \mathbb{P}_0(Y \in dy) \\ &\quad + \int_{t\phi(2y(r/2)^{-2}) \geq \frac{1}{2}} \mathbb{P}_0(Y \in dy) \\ &\leq c_1 tH(r^{-2}) \int_{(0, \infty)} (1 + c_2 y^2) \mathbb{P}_0(Y \in dy) \\ &\quad + \mathbb{P}_0(Y \geq c_3 r^2 \phi^{-1}(\frac{t^{-1}}{2})) \\ &\leq c_4 tH(r^{-2}) + c_5 e^{-a_U r^2 \phi^{-1}(t^{-1})}, \end{aligned}$$

since $\mathbb{P}(Y \geq y) = 2^{-d/2} \Gamma(d/2)^{-1} \int_y^\infty s^{d/2-1} e^{-s/2} ds$.

It is left to notice that

$$\begin{aligned}c_6|x|^d p(t, x) &\leq \mathbb{P}_0(|x|/2 \leq X_t \leq |x|) \\ &\leq c_4 t H(r^{-2}) + c_5 e^{-a_U r^2 \phi^{-1}(t^{-1})}.\end{aligned}$$

- For upper bounds, only **(L)** for ϕ is needed, since

$$\frac{H(\lambda x)}{H(\lambda)} \leq x^2 \quad \text{for all } \lambda > 0, x \geq 1$$

always holds.

Application to Green function (potential)

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Proposition

*Assume that $\lambda/\phi(\lambda)$ is also a Bernstein function and that ϕ satisfies **(L)** with $\lambda_L = 0$. Then*

$$G(x) \asymp \frac{1}{|x|^d \phi(|x|^{-2})}, \quad x \in \mathbb{R}^d, x \neq 0.$$

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$$\begin{aligned} G(x) &\geq I_2 \geq c_1 \int_{\phi(|x|^{-2})^{-1}}^{2\phi(|x|^{-2})^{-1}} \phi^{-1}(t^{-1})^{d/2} dt \\ &\geq c_1 \phi^{-1} \left(\frac{\phi(|x|^{-2})}{2} \right)^{d/2} \frac{1}{\phi(|x|^{-2})} \geq \frac{c_1}{|x|^d \phi(|x|^{-2})} \end{aligned}$$

We only prove bound for I_2 :

$$\begin{aligned}
 I_2 &\leq c_2|x|^{-d} \int_0^{\phi(|x|^{-2})^{-1}} \left[tH(|x|^{-2}) + e^{-aV|x|^2\phi^{-1}(t^{-1})} \right] dt \\
 &\leq c_2|x|^{-d} \frac{H(|x|^{-2})}{2\phi(|x|^{-2})^2} + c_3|x|^{-d} \int_0^{\phi(|x|^{-2})^{-1}} \left[|x|^2\phi^{-1}(t^{-1}) \right]^{-1} dt \\
 &\leq \frac{c_4}{|x|^d\phi(|x|^{-2})} + c_5 \int_0^{\phi(|x|^{-2})^{-1}} \frac{\phi^{-1}(\phi(|x|^{-2}))}{\phi^{-1}\left(\frac{\phi(|x|^{-2})}{t\phi(|x|^{-2})}\right)} dt \\
 &\leq \frac{c_4}{|x|^d\phi(|x|^{-2})} + c_6 \int_0^{\phi(|x|^{-2})^{-1}} t\phi(|x|^{-2}) dt \leq \frac{c_7}{|x|^d\phi(|x|^{-2})}.
 \end{aligned}$$

□

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and this is very close to the Green function of Brownian motion (Newtonian potential)

$$G^{(2)}(x) = |x|^{2-d}.$$

Thank

you

for

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attention!