Density Bounds for some Degenerate Stable Driven SDEs.

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Degenerate SDE?

We consider the following degenerate stable-driven SDE:

$$\begin{array}{ll} dX_t^1 = & \left(a_t^{1,1}X_t^1 + \ a_t^{1,2}X_t^2 + \ \cdots + \ a_t^{1,n-1}X_t^{n-1} + a_t^{1,n}X_t^n\right)dt + \sigma(t,X_{t^-})dZ_t \\ dX_t^2 = & \left(a_t^{2,1}X_t^1 + \ a_t^{2,2}X_t^2 + \ \cdots + \ a_t^{2,n-1}X_t^{n-1} + a_t^{2,n}X_t^n\right)dt \\ dX_t^3 = & \left(a_t^{3,2}X_t^2 + \ \cdots + \ a_t^{3,n-1}X_t^{n-1} + a_t^{3,n}X_t^n\right)dt \\ & \vdots \\ dX_t^n = & \left(a_t^{n,n-1}X_t^{n-1} + a_t^{n,n}X_t^n\right)dt \end{array}$$

with initial condition $X_0 = x \in \mathbb{R}^{nd}$, and where

- Z is an \mathbb{R}^d valued symmetric α stable process $(\alpha \in (0,2))$,
- $\sigma: \mathbb{R}_+ \times \mathbb{R}^{nd} \to \mathbb{R}^d \otimes \mathbb{R}^d$, Hölder continuous, uniformly elliptic, bounded.
- $a^{i,j}: \mathbb{R}_+ \to \mathbb{R}^d \otimes \mathbb{R}^d$,
- for $x \in \mathbb{R}^{nd}$, we denote $x = (x^1, \dots, x^n)$, with $x^i \in \mathbb{R}^d$.

(1)

Motivation

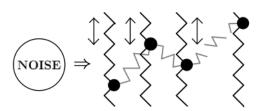
Two examples.

• For n = 2, the pricing of the Asian option in jump diffusion models.

$$X_{t}^{1} = x^{1} + \int_{0}^{t} a_{s}^{1} X_{s}^{1} ds + \int_{0}^{t} \sigma(s, X_{s-}) dZ_{s}$$

$$X_{t}^{2} = x^{2} + \int_{0}^{t} a_{t}^{2} X_{s}^{1} ds$$

In Physics, a system of perturbated oscillators:



The Stable Process

• The Lévy measure factorizes as $\nu(dz) = C \frac{d|z|}{|z|^{1+\alpha}} \mu(d\bar{z})$ and its Fourier transform writes:

$$\mathbb{E}\left(e^{i\langle p,Z_{\mathbf{t}}\rangle}\right) = \exp\left(-t\int_{\varsigma d-1} |\langle p,\varsigma\rangle|^{\alpha} \mu(d\varsigma)\right).$$

• When μ the spectral measure is non degenerate, the density exists and we have the estimates:

$$c\frac{t^{-\frac{d}{\alpha}}}{\left[1+\frac{|z|}{t^{\frac{1}{\alpha}}}\right]^{d+\alpha}} \leq p_{Z_{\mathbf{t}}}(z) \leq C\frac{t^{-\frac{d}{\alpha}}}{\left[1+\frac{|z|}{t^{\frac{1}{\alpha}}}\right]^{d+\alpha}}.$$

• When $|z| \leq t^{\frac{1}{\alpha}}$, we say that the **diagonal** regime holds, and:

$$ct^{-\frac{d}{\alpha}} \leq p_{Z_t}(z) \leq Ct^{-\frac{d}{\alpha}}.$$

ullet When $|z| \geq t^{rac{1}{lpha}}$ we say that the **off-diagonal** regime holds, and

$$c\frac{t}{|z|^{d+\alpha}} \leq p_{Z_{\mathbf{t}}}(z) \leq C\frac{t}{|z|^{d+\alpha}}.$$

Related works

- In the case n=1 and $\alpha=2$, the brownian estimate under uniform elliptic setting is known from Friedman 1964 [3].
- When n = 1, Kolokoltsov in 2000 [4] showed that for a stable diffusion:

$$c\frac{(s-t)^{-\frac{d}{\alpha}}}{\left[1+\frac{|y-x|}{(s-t)^{\frac{1}{\alpha}}}\right]^{d+\alpha}} \leq p(t,s,x,y) \leq C\frac{(s-t)^{-\frac{d}{\alpha}}}{\left[1+\frac{|y-x|}{(s-t)^{\frac{1}{\alpha}}}\right]^{d+\alpha}}.$$

• When $\alpha = 2$, Delarue and Menozzi in 2010 [2]. For the brownian chain, the estimate holds:

$$C^{-1}t^{-n^{2}\frac{d}{2}}\exp\left(-C\left|(\mathbb{T}_{s-t}^{2})^{-1}(y-R_{s,t}(x))\right|^{2}\right)$$

$$\leq p(t,s,x,y) \leq$$

$$Ct^{-n^{2}\frac{d}{2}}\exp\left(-C^{-1}\left|(\mathbb{T}_{s-t}^{2})^{-1}(y-R_{s,t}(x))\right|^{2}\right).$$

• Consider the simple case:
$$dX_t = \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} X_t dt + \begin{pmatrix} dZ_t \\ 0 \end{pmatrix}$$
.

This writes:

$$X_s^1 = x_1 + Z_s,$$

 $X_s^2 = x_2 + sx_1 + \int_0^s Z_t dt.$

- The first component is at scale $s^{1/\alpha}$, the second is at scale $s^{1+1/\alpha}$.
- We can put all the component at the same scale by normalizing by

$$\mathbb{T}_{s}^{\alpha} = \begin{pmatrix} s^{\frac{1}{\alpha}}I_{d} & 0\\ 0 & s^{1+\frac{1}{\alpha}}I_{d} \end{pmatrix}.$$

→ In the degenerate framework, the typical behavior is given by

$$\left| \left(\mathbb{T}_s^{\alpha} \right)^{-1} \left(y - R_s x \right) \right| \approx \frac{\left| y^1 - R_s^1 x \right|}{s^{\frac{1}{\alpha}}} + \frac{\left| y^2 - R_s^2 x \right|}{s^{1 + \frac{1}{\alpha}}}.$$

Hypotheses [H]

[H-1]: (Hölder regularity) $\exists H > 0, \ \eta \in (0,1], \ \forall x,y \in \mathbb{R}^{nd} \text{ and } \forall t \geq 0,$

$$||\sigma(t,x)-\sigma(t,y)|| \leq H|x-y|^{\eta}.$$

[H-2]: (Non degeneracy of the spectral measure)

$$\exists \Lambda_1, \Lambda_2 \in \mathbb{R}_+^*, \ \forall u \in \mathbb{R}^d,$$
$$\Lambda_1 |u|^{\alpha} \leq \int_{S^{d-1}} |\langle u, \varsigma \rangle|^{\alpha} \mu(d\varsigma) \leq \Lambda_2 |u|^{\alpha}.$$

[H-3]: (Ellipticity) $\exists \ \overline{c}, \ \underline{c} > 0$, $\forall \xi \in \mathbb{R}^d$, $\forall z \in \mathbb{R}^{nd}$ and $\forall t \geq 0$,

$$\underline{c}|\xi|^2 \leq \langle \xi, \sigma\sigma^*(t,z)\xi \rangle \leq \overline{c}|\xi|^2.$$

[H-4]: (Hörmander-like condition) $\exists \overline{\alpha}, \ \underline{\alpha} \in \mathbb{R}_+^*, \ \forall \xi \in \mathbb{R}^{nd} \ \text{and} \ \forall t \geq 0,$ $\forall i \in [\![2,n-1]\!],$ $\underline{\alpha}|\xi|^2 \leq \langle a_t^{i,i-1}\xi,\xi \rangle \leq \overline{\alpha}|\xi|^2.$

Also, for all $(i,j) \in [1,n]^2$, $||a_t^{i,j}|| \leq \overline{\alpha}$.

Weak Uniqueness

Theorem

Under **[H]**, the martingale problem associated with the generator of the degenerate equation (1) admits a unique solution, i.e.: $\forall x \in \mathbb{R}^{nd}$, $\exists ! \mathbb{P}$ probability measure on $\Omega = \mathcal{D}(\mathbb{R}_+, \mathbb{R}^{nd})$ such that $\forall f \in \mathcal{C}_0^{1,1}(\mathbb{R}_+ \times \mathbb{R}^{nd}, \mathbb{R})$,

$$\mathbb{P}(X_0 = x) = 1$$
 and $f(t, X_t) - \int_0^t (\partial_u + L_u) f(u, X_u) du$ is a \mathbb{P} - martingale.

Hence, weak uniqueness holds.

The associated semigroup is strong Feller.

Theorem (**Density Estimates**)

Under **[H]**, the the unique weak solution of (1) has for every s > 0 a density. Let $R_{s,t}$ be the resolvent of the deterministic ODE, and let d = 1 and n = 2.

For fixed T, K > 0, $\exists C \ge 1$, s.t. $\forall (x, y) \in (\mathbb{R}^2)^2$, $\forall 0 \le t < s \le T$,

$$p(t,s,x,y) \leq C\bar{p}_{\alpha}(t,s,x,y) \left(1 + \log\left(K \vee \left| (\mathbb{T}_{s-t}^{\alpha})^{-1} (y - R_{s,t}x) \right| \right) \right),$$

where

$$\bar{p}_{\alpha}(t,s,x,y) = C_{\alpha} \frac{(s-t)^{-\frac{1}{\alpha}-\left(1+\frac{1}{\alpha}\right)}}{\left[K + \frac{|y^{1}-R^{1}_{s,t}x|}{(s-t)^{\frac{1}{\alpha}}} + \frac{|y^{2}-R^{2}_{s,t}x|}{(s-t)^{1+\frac{1}{\alpha}}}\right]^{2+\alpha}}.$$

Also, when $|(\mathbb{T}_{s-t}^{\alpha})^{-1}(y-R_{s,t}x)| \leq K$,

$$p(t,s,x,y) \geq C^{-1}(s-t)^{-\frac{1}{\alpha}-\left(1+\frac{1}{\alpha}\right)}.$$

The Frozen process

The proxy is constructed as follows:

- let T > 0 arbitrary deterministic time,
- let $y \in \mathbb{R}^{nd}$ final freezing point,
- recall $R_{s,T}(y)$ satisfies $\frac{d}{ds}R_{s,T}=A_sR_{s,T}$, with $R_{T,T}=I_{nd}$ in $\mathbb{R}^{nd}\otimes\mathbb{R}^{nd}$.
- We define the frozen process as follows:

$$d\tilde{X}_{s}^{T,y} = A_{s}\tilde{X}_{s}^{T,y}ds + B\sigma(s, R_{s,T}(y))dZ_{s},$$

$$\tilde{X}_{0}^{T,y} = x,$$

- We follow the deterministic system backwards.
- We set $\sigma(u, R_{u,T}(y)) = \sigma_u$.

Proposition

Fix $(t,x) \in [0,T] \times \mathbb{R}^{nd}$. The Frozen Process $\tilde{X}^{T,y}$ starting from x at time t writes:

$$\tilde{X}_{s}^{t,x,T,y} = R_{s,t}x + \int_{t}^{s} R_{s,u}B\sigma_{u}dZ_{u},$$

and has a density $p_{\alpha}^{T,y}(t,s,x,z)$, that satisfies the bounds:

$$C^{-1}\bar{p}_{\alpha}(t,s,x,z) \leq \tilde{p}_{\alpha}^{T,y}(t,s,x,z) \leq C\bar{p}_{\alpha}(t,s,x,z),$$

where

$$\bar{p}_{\alpha}(t,s,x,y) = C_{\alpha} \frac{\det(\mathbb{T}_{s-t}^{\alpha})^{-1}}{\left[K \vee \left| (\mathbb{T}_{s-t}^{\alpha})^{-1} (y - R_{s,t}x) \right| \right]^{nd+\alpha}}.$$

The Parametrix Series

Assume first that $p(t, s, x, z) = \mathbb{P}(X_s \in dz | X_t = x)$ exists and is smooth. We set $\tilde{p}_{\alpha}(t, s, x, z) = \tilde{p}_{\alpha}^{s,z}(t, s, x, z)$.

$$p(t,T,x,y) - \tilde{p}(t,T,x,y) = \int_{t}^{T} d\tau \partial_{\tau} \int_{\mathbb{R}^{nd}} p(t,\tau,x,z) \tilde{p}(\tau,T,z,y) dz$$

Differentiating formally under the integral:

$$p(t, T, x, y) - \tilde{p}_{\alpha}(t, T, x, y) = \int_{t}^{T} d\tau \int_{\mathbb{R}^{nd}} \partial_{\tau} p(t, \tau, x, z) \tilde{p}_{\alpha}(\tau, T, z, y) dz$$
$$+ \int_{t}^{T} d\tau \int_{\mathbb{R}^{nd}} p(t, \tau, x, z) \partial_{\tau} \tilde{p}_{\alpha}(\tau, T, z, y) dz$$

Now, from Kolmogorov's equation:

$$\partial_{\tau}p(t,\tau,x,z)=L_{\tau}^{*}p(t,\tau,x,z),\ \partial_{\tau}\tilde{p}_{\alpha}(\tau,T,z,y)=-\tilde{L}_{\tau}\tilde{p}_{\alpha}(\tau,T,z,y).$$

Thus,

$$p(t, T, x, y) - \tilde{p}_{\alpha}(t, T, x, y) = \int_{t}^{T} d\tau \int_{\mathbb{R}^{nd}} L_{\tau}^{*} p(t, \tau, x, z) \tilde{p}_{\alpha}(\tau, T, z, y) dz$$
$$- \int_{t}^{T} d\tau \int_{\mathbb{R}^{nd}} p(t, \tau, x, z) \tilde{L}_{\tau} \tilde{p}_{\alpha}(\tau, T, z, y) dz$$

Taking the adjoint in the first integral yields:

$$p(t, T, x, y) - \tilde{p}_{\alpha}(t, T, x, y)$$

$$= \int_{t}^{T} d\tau \int_{\mathbb{R}^{nd}} dz p(t, \tau, x, z) \underbrace{\left(L_{\tau} - \tilde{L}_{\tau}\right) \tilde{p}_{\alpha}(\tau, T, z, y)}_{H(\tau, T, z, y)}$$

That is:

$$p(t, T, x, y) = \tilde{p}_{\alpha}(t, T, x, y) + p \otimes H(t, T, x, y),$$

where we defined the time space convolution:

$$f \otimes g(t, T, x, y) = \int_t^T du \int_{\mathbb{R}^{nd}} dz f(t, u, x, z) g(u, T, z, y).$$

The Parametrix Series

Proposition

Let $P_{t,T}f(x) = \mathbb{E}[f(X_T)|X_t = x]$, the transition of the solution of (1), we have:

$$P_{t,T}f(x) = \int_{\mathbb{R}^{nd}} dy \left(\sum_{r=0}^{+\infty} \tilde{p}_{\alpha} \otimes H^{(r)}(t,T,x,y) \right) f(y),$$

where H is the parametrix kernel:

$$\forall 0 \leq t < T, \ (x,y) \in (\mathbb{R}^{nd})^2, \ H(t,T,x,y) := (L_t - \tilde{L}_t) \tilde{p}_{\alpha}^{T,y}(t,T,x,y),$$

$$L_t \varphi(x) = \langle A_t, \nabla \varphi(x) \rangle - \int_{S^{d-1}} |\langle \frac{\partial}{\partial x}, B\sigma(t,x)\varsigma \rangle|^{\alpha} \cdot \varphi(x) \mu(d\varsigma).$$

$$\tilde{L}_t^{T,y} \varphi(x) = \langle A_t, \nabla \varphi(x) \rangle - \int_{S^{d-1}} |\langle \frac{\partial}{\partial x}, B\sigma(t,R_{t,T}(y))\varsigma \rangle|^{\alpha} \cdot \varphi(x) \mu(d\varsigma).$$

Lemma (Control of the kernel)

There exists constants C > 0, $\delta > 0$ s.t. for all $T \in (0, T_0]$ and $(t, x, y) \in [0, T) \times (\mathbb{R}^{nd})^2$:

$$|H(t,T,x,y)| \leq C \frac{\delta \wedge |x-R_{t,T}(y)|^{\eta(\alpha \wedge 1)}}{T-t} \tilde{p}_{\alpha}^{T,y}(t,T,x,y).$$

and the following bound holds:

$$\int_{\mathbb{R}^{nd}} |H(\tau,T,z,y)| dz \leq C(T-\tau)^{\eta(\frac{1}{\alpha}\wedge 1)-1}.$$

- The last bound reflects the smoothing property (in time) of H.
- Singularity in $\frac{1}{T-t}$, same as in the Brownian setting.
- This is an important ingredient in the proof of the weak uniqueness.
- This holds for any dimension d and any number of oscillators n.

Uniqueness to the Martingale Problem

Suppose we are given two solutions \mathbb{P}^1 and \mathbb{P}^2 of the martingale problem associated to $(L_s)_{s\in[t,T]}$, starting in x at time t.

Define for a bounded Borel function $f:[0,T]\times\mathbb{R}^{nd}\to\mathbb{R}$,

$$S^i f = \mathbb{E}^i \left(\int_t^T f(s, X_s) ds \right), \ i \in \{1, 2\},$$

where $(X_s)_{s\in[t,\mathcal{T}]}$ stands for the canonical process associated with $(\mathbb{P}^i)_{i\in\{1,2\}}$.

We denote:

$$S^{\Delta}f=S^1f-S^2f.$$

We show that $S^{\Delta}f=0$ for all $f:[0,T]\times\mathbb{R}^{nd}\to\mathbb{R}$ measurable and bounded, by proving:

$$\|S^{\Delta}\| := \sup_{|f|_{\infty} < 1} |S^{\Delta}f| = 0.$$

Exploiting the fact that $(\mathbb{P}^i)_{i \in \{1,2\}}$ are solutions to the martingale problem, we have $\forall f \in \mathcal{C}_0^{1,1}([0,T) \times \mathbb{R}^{nd},\mathbb{R})$:

$$f(t,x) + \mathbb{E}^i \left(\int_t^T (\partial_s + L_s) f(s,X_s) ds \right) = 0, \ i \in \{1,2\}.$$

Thus, for all $f \in \mathcal{C}^{1,1}_0([0,T) \times \mathbb{R}^{nd},\mathbb{R})$,

$$S^{\Delta}\Big((\partial_{\cdot}+L_{\cdot})f\Big)=0.$$

Inserting the frozen generator:

$$\underbrace{S^{\Delta}\Big((\partial_{\cdot} + \tilde{L}_{\cdot})f\Big)}_{\text{Estimates on }\tilde{p}_{\alpha}} + \underbrace{S^{\Delta}\Big((L_{\cdot} - \tilde{L}_{\cdot})f\Big)}_{\text{Estimates on }H} = 0$$

We take as f the following function:

$$\Psi_{\varepsilon}(t,x) = \int_{\mathbb{R}^{nd}} dy \int_{t}^{T} ds \tilde{p}_{\alpha}^{s+\varepsilon,y}(t,s+\varepsilon,x,y) h(s,y),$$

where h is a, arbitrary test function.

Applying $\partial_t + L_t$ and introducing the frozen generator \tilde{L}_t :

$$(\partial_t + L_t)\Psi_{\varepsilon}(t,x) = (\partial_t + \tilde{L}_t)\Psi_{\varepsilon}(t,x) + (L_t - \tilde{L}_t)\Psi_{\varepsilon}(t,x)$$

= $I_1^{\varepsilon} + I_2^{\varepsilon}$.

We can show that:

$$I_1^{\varepsilon} \underset{\varepsilon \to 0}{\longrightarrow} -h(t,x), \text{ and } |I_2^{\varepsilon}| \leq 1/2|h|_{\infty}.$$

Also, recall that $S^{\Delta}((\partial_{\cdot} + L_{\cdot})\Psi_{\varepsilon}) = 0$ so that $|S^{\Delta}(I_{1}^{\varepsilon})| = |S^{\Delta}(I_{2}^{\varepsilon})|$. We deduce:

$$|\mathcal{S}^{\Delta}h| = \lim_{\varepsilon \to 0} |\mathcal{S}^{\Delta}I_1^{\varepsilon}| = \lim_{\varepsilon \to 0} |\mathcal{S}^{\Delta}I_2^{\varepsilon}| \leq \|\mathcal{S}^{\Delta}\| \limsup_{\varepsilon \to 0} |I_2^{\varepsilon}| \leq 1/2 \|\mathcal{S}^{\Delta}\| |h|_{\infty}.$$

Hence, $\|S^{\Delta}\| \le 1/2\|S^{\Delta}\|$. But since $\|S^{\Delta}\| \le T - t$, we deduce that $\|S^{\Delta}\| = 0$.



R.F. Bass, and E. Perkins. A new technique for proving uniqueness for martingale problems. Asterisque 327 (2009): 47.



F. Delarue, S. Menozzi, Density estimates for a random noise propagating through a chain of differential equations, J. Funct. Anal. 259, (2010).



A. Friedman. Partial differential equation of parabolic type. Prentice-Hall, (1964).



V. Kolokoltsov, Symmetric stable laws and stable like diffusions, London mathematical society (2000).



S. Menozzi, Parametrix techniques and martingale problems for some degenerate Kolmogorov equations. Electronic Communications in Probability 16 (2011): 234-250.