

Some McKean–Vlasov problems on the half-line

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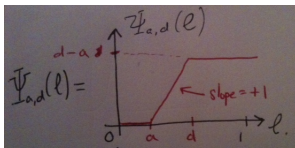
- 1 Loss-dependent correlation model
- 2 Contagion model

Motivation

- Initial motivation: portfolio credit derivatives
- $N \geq 1$ defaultable assets, default times $\{\tau^{i,N}\}_{1 \leq i \leq N}$,
- Care about options on proportional loss process

$$L_t^N = \frac{1}{N} \sum_{i=1}^N \mathbf{1}_{\tau^{i,N} \leq t}, \quad \text{value} = \mathbf{E} \Psi((L^N)_{t \in [0, T]})$$

- Think of Ψ of form



- Correlations matter: high $\rho \implies$ higher probability of defaulting together
- But also higher chance of surviving together

Model

Basic model — Bush, Hambly, Howarth, Jin, Reisinger (2011)

$$dX_t^i = \mu dt + \rho dW_t + \sqrt{1 - \rho^2} dW_t^i$$

$$\tau^i = \inf\{t > 0 : X_t^i \leq 0\}$$

$$X_0^i \sim \nu_0$$

- Take a limit as $N \rightarrow \infty$,
- Study the empirical processes

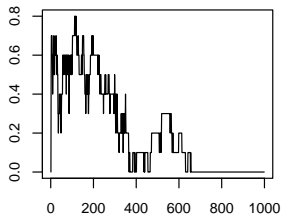
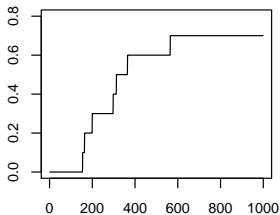
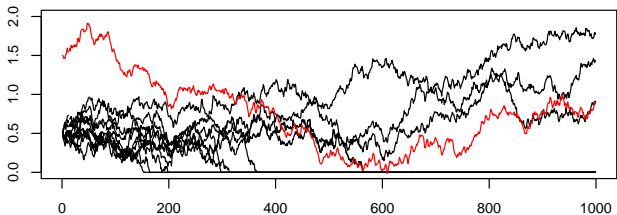
$$\nu_t^N = \frac{1}{N} \sum_{i=1}^N \mathbf{1}_{t < \tau^i} \delta_{X_t^i} \in \mathcal{M}$$

- Everything conditionally iid, so (marginal) convergence is easy

$$\nu_t^N(\phi) \rightarrow \nu_t(\phi) := \mathbf{E}[\phi(X_t^1) \mathbf{1}_{t < \tau^1} | W], \quad L_t^N \rightarrow \nu_t(\mathbf{1}_{(0, \infty)}) = \mathbf{P}(\tau \leq t | W)$$

- Price with $L_t := \mathbf{P}(\tau \leq t | W)$

Basic model

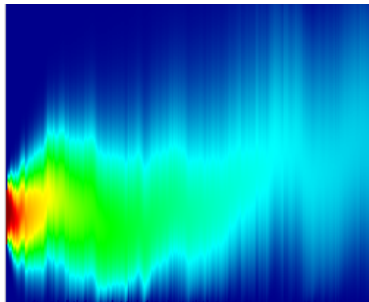
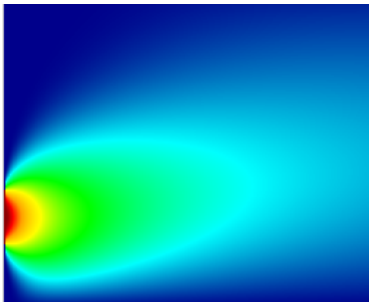


Basic model

SPDE — Bush, Hambly, Howarth, Jin, Reisinger (2011)

$$d\nu_t(\phi) = \mu\nu_t(\partial_x\phi)dt + \frac{1}{2}\nu_t(\partial_{xx}\phi)dt + \rho\nu_t(\partial_x\phi)dW_t$$

$$\phi(0) = 0.$$



Basic model

Proof

Easy! Conditional independence, Linear evolution equation.

Existence: Ito (drop μ)

$$\begin{aligned}\phi(X_t^i) &= \phi(X_0^i) + \frac{1}{2} \int_0^t \phi''(X_s^i) ds + \int_0^t \rho \phi'(X_s^i) dW_s \\ &\quad + \int_0^t \sqrt{1 - \rho^2} \phi'(X_s^i) dW_s^i\end{aligned}$$

If $\phi(0) = 0$ stopping incorporates b.c.: $\phi(X_{t \wedge \tau}^i) = \phi(X_t^i) \mathbf{1}_{t < \tau^i}$.

Take average over $1 \leq i \leq N$

$$\text{eqn}(\nu^N)_t = \frac{1}{N} \sum_{i=1}^N \int_0^t \sqrt{1 - \rho^2} \phi'(X_s^i) dW_s^i = O_{\text{m.s.}}(N^{-1})$$

In limit: $\text{eqn}(\nu)_t = 0$.

Basic model

Proof

Uniqueness: Smooth with heat kernel

$$T_\varepsilon \nu_t(x) := \int_0^\infty \frac{1}{\sqrt{2\pi\varepsilon}} \left(e^{-\frac{(x-x_0)^2}{2\varepsilon}} - e^{-\frac{(x+x_0)^2}{2\varepsilon}} \right) \nu_t(dx_0) \in C^\infty$$

eqn($T_\varepsilon \nu$) = rem(ν, ε). Manipulate classically using energy estimation

$$\mathbf{E} \|T_\varepsilon \nu_t\|_2^2 + \mathbf{E} \int_0^t \|\partial_x T_\varepsilon \nu_s\|_2^2 ds \leq \|T_\varepsilon \nu_0\|_2^2 + \mathbf{E} \|\text{rem}(\nu, \varepsilon)\|_2^2$$

Control remainder with the quantity

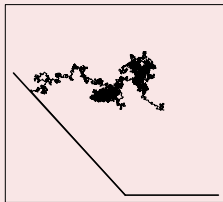
$$\mathbf{E} \nu_t(0, \varepsilon)^2 = \mathbf{P}(X_t^1, X_t^2 \in (0, \varepsilon), t < \tau^1 \wedge \tau^2)$$

need $o(\varepsilon^{3+\delta})$ control

Basic model

Proof

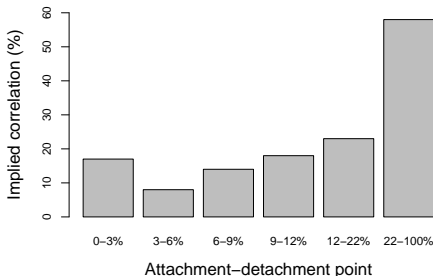
$$\mathbf{E} \nu_t(0, \varepsilon)^2 = \mathbf{P}(X_t^1, X_t^2 \in (0, \varepsilon), t < \tau^1 \wedge \tau^2)$$



Explicit formula, or note harmonic function with zero b.c. in wedge decays like $O(r^{1+\delta})$ □

Extension

- Model is too simple, can't choose one ρ to match all traded tranche spreads, *correlation skew* or *smile*,



- Why not make ρ a function of the loss in the system?

Loss-dependent model

$$dX_t^{i,N} = \mu(L_t^N)dt + \rho(L_t^N)dW_t + \sqrt{1 - \rho^2(L_t^N)}dW_t^i$$

Extension

- Can have diffusions, but drop for talk

Loss-dependent model

$$dX_t^{i,N} = \rho(L_t^N) dW_t + \sqrt{1 - \rho^2(L_t^N)} dW_t^i$$

- Piecewise constant ρ across tranches desirable
- Allow at least finitely many discontinuities, piecewise Lipschitz ρ
- Need $0 \leq \rho(\ell) \leq \rho_{\max} < 1$, stop degeneracy
- Challenges: need to deal with boundary effects but correlation too complicated to do explicit calculations,
- For convergence, discontinuous ρ bad, key to show limit points must have strictly increasing loss
- Associated SPDE is non-linear

Results

Tightness/Weak existence

The sequence of triples $(\nu^N, L^N, W)_{N \geq 1}$ are tight (with suitable topology). If (ν^*, L^*, W) realises a limiting law, then

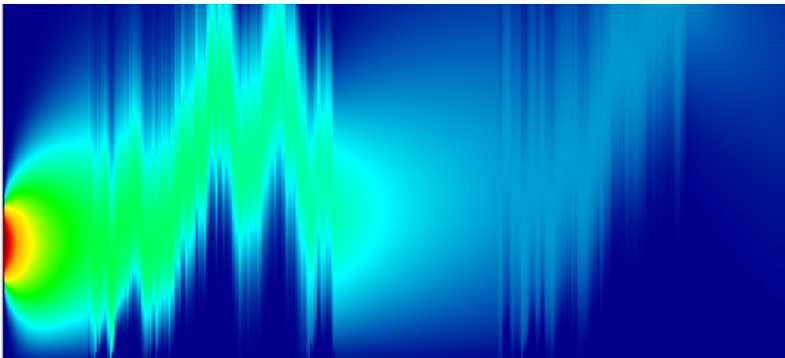
$$d\nu_t^*(\phi) = \frac{1}{2} \nu_t^*(\partial_{xx}\phi) dt + \rho(L_t) \nu_t^*(\partial_x\phi) dW_t$$
$$L_t^* = 1 - \nu_t^*(0, \infty).$$

[+ some regularity conditions.]

Pathwise uniqueness/LLN

For a given W , the SPDE has a most one solution ν . Hence there is a unique law of a solution (ν, L, W) and so the sequence converges to it.

Results



Results

- Weak existence + pathwise uniqueness gives strong solution (on rich enough space), so

M–V problem

With (ν, L, W) be the unique soln. For any independent B.M. W^\perp there exists a process X satisfying

$$dX_t = \rho(L_t)dW_t + \sqrt{1 - \rho(L_t)^2} dW_t^\perp$$

$$\tau := \inf\{t > 0 : X_t \leq 0\}$$

$$\nu_t(\phi) = \mathbf{E}[\phi(X_t)\mathbf{1}_{t < \tau} | W],$$

$$L_t = \mathbf{P}(\tau \leq t | W).$$

The law of (X, W) is unique.

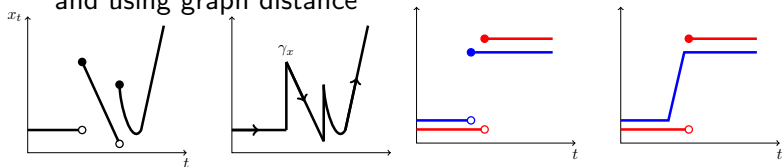
- Conditional/stochastic M–V problem
- Sznitman, *Topics in propagation of chaos* (1991)

Methods

Proof outline

Pick suitable topology \rightarrow prove tightness \rightarrow show limiting laws solve SPDE \rightarrow uniqueness for Lip. $\rho(\cdot)$ \rightarrow extend to all $\rho(\cdot)$.

- Skorokhod M1 topology is useful
- $\text{dist}(x, y)$ defined on $D_{\mathbf{R}}$ by joining up discontinuity points and using graph distance



- Tightness controlled by modulus of continuity:

$$w(x; \delta) = \sup_t \sup_{(t_1, t_2, t_3) \in \text{Trip}_{t, \delta}} |x_{t_2} - [x_{t_1}, x_{t_3}]_{\mathbf{R}}|_{\mathbf{R}}$$

- Vanishes if x is monotone

Topology

- Useful for L.L.N. in queuing theory, Whitt, *Stochastic process limits* (2002)
- We extend to infinite-dim. range space,
- Work on $D_{S'}$, space of distribution-valued càdlàg paths
- $\mathcal{M} \subseteq S'$, recovering $f \in \mathcal{M}$ generally easy, allows for CLT (perhaps)

Tightness characterisation

$(\nu^N)_{N \geq 1}$ tight on $(D_{S'}, M1)$ iff $(\nu^N(\phi))_{N \geq 1}$ tight on $(D_{\mathbf{R}}, M1)$ for all $\phi \in C_0^\infty(\mathbf{R})$.

- Extends Mitoma (1983)
- Useful as

$$\nu_t^N(\phi) = \frac{1}{N} \sum_{i=1}^N \phi(X_t^{i,N}) \mathbf{1}_{t < \tau^{i,N}} = \frac{1}{N} \sum_{i=1}^N \phi(X_{t \wedge \tau^{i,N}}^{i,N}) - \phi(0) L_t^N$$

Loss-dependent SPDE

- Existence easy once tightness established, Care that topology strong enough to recover SPDE
- Also need to know $t \mapsto L_t^*$ strictly increasing, long estimates,
- For uniqueness estimates, work in space H^{-1}

$$\mathbf{E} \partial_x^{-1} \|T_\varepsilon \nu_t\|_2^2 + \mathbf{E} \int_0^t \|T_\varepsilon \nu_s\|_2^2 ds \leq \|\partial_x^{-1} T_\varepsilon \nu_0\|_2^2 + \mathbf{E} \|\partial_x^{-1} \text{rem}(\nu, \varepsilon)\|_2^2$$

- Only required $o(\varepsilon^{1+\delta})$ control on remainder,
- Enough to use trivial bound

$$\mathbf{E} \nu_t(0, \varepsilon)^2 \leq \mathbf{E} \nu_t(0, \varepsilon) = \lim_{N \rightarrow \infty} \mathbf{P}(X_t^{1,N} \in (0, \varepsilon), t < \tau^{1,N})$$

- Eliminates correlation, much easier to work with
- Use a stopping argument for Lip. and uniqueness

A neuroscience model

- Many neurons, look at typical behaviour
- Voltage level $t \mapsto X_t$
- Start at $X_0 > 0$, distributed as ν_0
- Experiences own noise, $(B_t)_{t \geq 0}$, and spikes when hits level 0
- But receives a kick towards the origin of size α times the proportion of spiking particles:

M–V problem (mean field)

$$\begin{aligned}X_t &= X_0 + B_t - \alpha L_t \\ \tau &= \inf\{t > 0 : X_t \leq 0\} \\ L_t &= \mathbf{P}(\tau \leq t)\end{aligned}$$

- (Then restarts back on $[0, \infty)$ according to ν_0 — but problem is hard enough)

PDE and integral equation

M–V problem (mean field)

$$\begin{aligned}X_t &= X_0 + B_t - \alpha L_t \\ \tau &= \inf\{t > 0 : X_t \leq 0\} \\ L_t &= \mathbf{P}(\tau \leq t)\end{aligned}$$

- What does it mean to solve? Find L
- Can rewrite as a PDE and IE
- Let $\nu_t(\phi) := \mathbf{E}[\phi(X_t)\mathbf{1}_{t < \tau}]$, so $L_t = 1 - \nu_t(\mathbf{1}_{(0, \infty)})$,

PDE problem (large population distn.)

$$\begin{aligned}d\nu_t(\phi) &= \frac{1}{2}\nu_t(\partial_{xx}\phi)dt - \alpha\nu_t(\partial_x\phi)dL_t \\ L_t &= 1 - \nu_t(\mathbf{1}_{(0, \infty)})\end{aligned}$$

PDE and integral equation

PDE problem (large population distn.)

$$d\nu_t(\phi) = \frac{1}{2}\nu_t(\partial_{xx}\phi)dt - \alpha\nu_t(\partial_x\phi)dL_t$$

$$L_t = 1 - \nu_t(\mathbf{1}_{(0,\infty)})$$

- If solution is nice, V_t density of ν_t

PDE problem (large population distn.)

$$\partial_t V_t(x) = \frac{1}{2}\partial_{xx} V_t(x)dt + \alpha\partial_t L_t\partial_x V_t(x)$$

$$L_t = 1 - \int_0^\infty V_t(x)dx$$

$$V_t(0) = 0, \quad x \mapsto V_0(x) \text{ given}$$

$$[\quad \partial_t L_t = \partial_x V_t(0) \quad]$$

PDE and integral equation

- Or, given Brownian motion with drift $t \mapsto \ell_t$

$$X_t^\ell = X_0 + B_t - \alpha \ell_t,$$

the function $\Gamma[\ell]_t := \mathbf{P}(\tau^\ell \leq t)$ satisfies

$$\int_0^\infty \Phi\left(-\frac{x - \alpha \ell_t}{t^{1/2}}\right) V_0(x) dx = \int_0^t \Phi\left(\alpha \frac{\ell_t - \ell_s}{(t-s)^{1/2}}\right) d\Gamma[\ell]_t$$

- Also: hitting law of Brownian motion on boundary $t \mapsto \alpha \ell_t$
- (Laplace transform on this law to get IE — Volterra IE of first kind)

IE (fixed point)

Solve $\Gamma[L] = L$

Solutions

Delarue, Inglis, Rubenthaler, Tanré (2015)

With $\nu_0 = \delta_{x_0}$, there exists $\alpha_0 > 0$ such that for all $\alpha < \alpha_0$ there exists a unique solution to (M-V) with $t \mapsto L_t \in C^1[0, T]$

Cáceres, Carrillo, Perthame (2011)

For any initial condition, there exists $\alpha_1 > 0$ such that for all $\alpha > \alpha_1$ no continuous function $t \mapsto L_t$ solves (M-V)

A simpler proof

Suppose a continuous solution exists. Set $\phi(x) = x$ into PDE.

$$0 < \nu_t(\phi) = \nu_0(\phi) - \alpha \int_0^t \nu_s(1) dL_s = \nu_0(\phi) - \alpha \int_0^t (1 - L_s) dL_s$$



Solutions

...A simpler proof

Can do this integral since continuous and increasing

$$\frac{\alpha}{2}(1 - (1 - L_t)^2) \leq \nu_0(\phi).$$

By ignoring drift, easy to see $L_t \rightarrow 1$ as $t \rightarrow \infty$. Therefore

$$\frac{\alpha}{2} \leq \nu_0(\phi).$$

So $\alpha_1 = 2\nu_0(\phi)$ will do. □

- Obviously not an optimal choice of α_1 , so ...

Solutions

Critical value?

Fix an initial condition (come back to this!) does there exist a critical value $\alpha_c > 0$ such that

$\alpha < \alpha_c \implies$ a continuous solution exists

$\alpha > \alpha_c \implies$ a continuous solution cannot exist (blow-up)?

- Obvious monotonicity arguments delicate: depends on rate, not absolute value, of L
- Location of blow-up tricky:

Monotone?

Let $\bar{\alpha} > \alpha > \alpha_c$, does $L^{\bar{\alpha}}$ blow-up before L^α ?

How to build solutions

- Three ways
- First is fixed point problem:
- Set $\ell = 0$, put in $\Gamma[\ell]$, iterate $\Gamma[\Gamma[\ell]]$, ..., $\Gamma[\dots\Gamma[\Gamma[\ell]]\dots]$
- Done in [DIRT15] on subspace of C^1 with sup-norm ($W^{1,\infty}$)
- ...but requires assumptions from theorem
- Second approach: delay the equation:

$$X_t^\delta = X_0 + B_t - \alpha L_t^\delta, \quad L_t^\delta = \mathbf{P}(\tau^\delta \leq t - \delta)$$

- An easy problem: Initially no contribution (when $t < \delta$), so solve problem on $[0, \delta)$, then use that information to solve on $[\delta, 2\delta)$, etc...

How to build solutions

- Delay prevents blow-up, so $L^\delta \in C^1[0, T]$
- Limit as $\delta \rightarrow 0$ shown to converge in [DIRT15a]
- Also in that paper: Third approach: The microscopic model
- This is important as it contains the physical meaning
- The problem is that we want blow-up to occur, models synchronisation of neurons
- Should have existence and uniqueness theory that incorporates blow-up — i.e. a way to restart solutions
- To solve the problem, let solutions L have jumps (restrict to càdlàg)
- How?

Global solutions

- Underspecified:

M–V problem (mean field)

$$\begin{aligned}
 X_t &= X_0 + B_t - \alpha L_t \\
 \tau &= \inf\{t > 0 : X_t \leq 0\} \\
 L_t &= \mathbf{P}(\tau \leq t)
 \end{aligned}$$

what if L jumps? $\Delta L_t := L_t - L_{t-}$

- $L_t - L_s = \mathbf{P}(X_s + \inf_{s < u < t} \{B_{s,u} - \alpha L_{s,u}\} \leq 0, \tau > s)$, send $s \uparrow t$
- $\Delta L_t = \mathbf{P}(X_{t-} \in (0, \alpha \Delta L_t), \tau > t) = \nu_{t-}(0, \alpha \Delta L_t)$
- So jump at t solves $J = \nu_{t-}(0, \alpha J)$, Still underspecified!

Global solutions

- Obvious solution: look for càdlàg solutions with jumps as small as they need to be at each time

Defn: Minimal-jump solution

If L solves (M-V) and is càdlàg, then L is a *minimal-jump solution* if whenever \bar{L} is another such solution agreeing with L on $[0, t)$ then

$$\Delta L_t \leq \Delta \bar{L}_t.$$

Characterisation

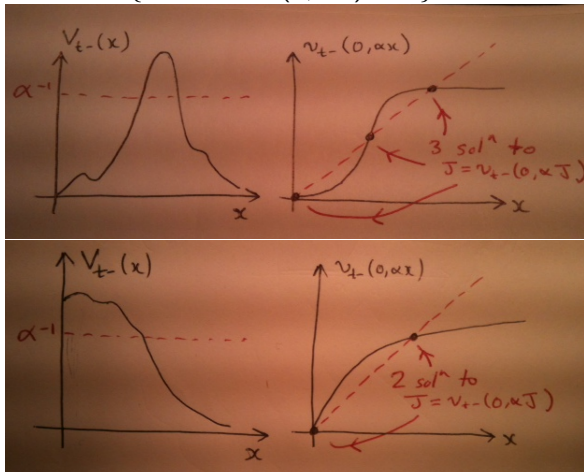
L is a minimal-jump solution iff L solves (M-V) and

$$\Delta L_t = \inf\{x > 0 : \nu_{t-}(0, \alpha x) < x\}, \quad \text{for all } t$$

- Existence of minimal-jump solutions [DIRT2015a]

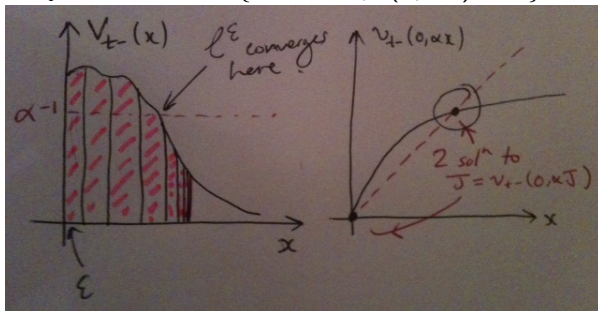
Global solutions

- $\Delta L_t = \inf\{x > 0 : \nu_{t-}(0, \alpha x) < x\}$,

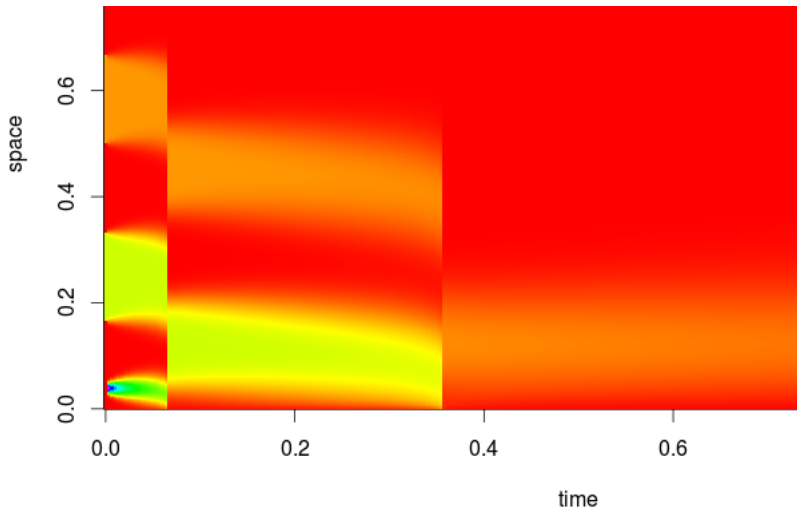


Global solutions

- Eat $l_0^\varepsilon = \varepsilon$ mass,
- Solution takes $l_1^\varepsilon = \varepsilon + \nu_{t-}(0, \alpha l_0^\varepsilon)$
- Then takes $l_2^\varepsilon = \varepsilon + \nu_{t-}(0, \alpha l_1^\varepsilon), \dots$
- $l^\varepsilon = \varepsilon + \nu_{t-}(0, \alpha l^\varepsilon)$, decreasing in ε
- $\rightarrow l = \nu_{t-}(0, \alpha l)$
- Easy to see $l = \inf\{x > 0 : \nu_{t-}(0, \alpha x) < x\}$



Global solutions



Global solutions

Uniqueness?

Is there a unique minimal-jump solution for a given initial condition?

- All current uniqueness arguments exploit smoothness and initial conditions that are too well-behaved— after blow-up $\nu_t(0, x) = O(x)$, \sqrt{t} singularity

P/w C^1 ?

Does the solution behave as C^1 function between jumps?
Weighted space?

Global solutions

- Solution density decays in max value quicker than the heat equation, $t^{-1/2}$,
- So there is a last possible time for a jump to occur,

Frequency?

Is there an upper bound on the number of jumps that can occur for a given α over all possible initial conditions?

Common noise

What about adding a rough (deterministic) noise $t \mapsto z_t$

$$X_t = X_0 + B_t + z_t - \alpha L_t, \quad L_t = \mathbf{P}_B(\tau \leq t)?$$

What if not Holder-1/2?