

# 2 Results on singular SDEs

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# Singular Diffusions

$$dX_t = b(t, X_t) dt + d\text{"Noise"} \quad \text{distribution}$$

here:  $\alpha$ -stable process,  $\alpha \in (1, 2]$   
 $(\alpha = 2: \text{Brownian motion})$

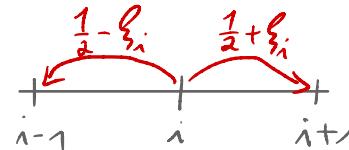
## Questions:

- Why are we interested?
- How to solve them?
- Qualitative aspects?

# Singular Diffusions as Processes in Random Media

Sinai's random walk in random environment:

$(\xi_j)_{j \in \mathbb{Z}}$  iid, centred, values in  $(-\frac{1}{2}, \frac{1}{2})$



$$\mathbb{P}(S_{n+1} = i+1 | S_n = i, (\xi_j)) = \frac{1}{2} + \xi_i \quad \mathbb{P}(S_{n+1} = i-1 | S_n = i, (\xi_j)) = \frac{1}{2} - \xi_i$$

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Continuous analogue: Brox diffusion

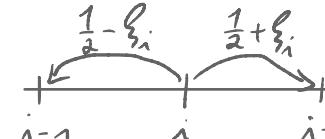
$$dX_t = \xi(X_t)dt + dW_t$$

white noise =  $B^1$        $\curvearrowleft$  1d B.m.

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Continuous analogue: Brox diffusion

$$dX_t = \xi(X_t)dt + dW_t$$

↑                      ↗ 1d B.m.

white noise =  $B'$

- scaling limit if  $\xi_i$  "small"
- easier to analyze because of scaling invariance
- $\alpha^{-2} X_{\alpha x}$  converges,  $\alpha \rightarrow \infty$

# Singular Diffusions as Processes in Random Media

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Continuous analogue: Brox diffusion → faster for  $W \rightarrow$  Lévy noise?  
(compare Imkeller-Pavlyukevich '06)

$$dX_t = \xi(X_t)dt + dW_t$$

white noise =  $B'$

↑  
1d B.m.

- scaling limit if  $\xi_i$  "small"
- easier to analyze because of scaling invariance
- $\alpha^{-2} X_{e^\alpha}$  converges,  $\alpha \rightarrow \infty$

very slow!

# Singular diffusions as random polymer measures

Random directed polymer measure

Alberts - Khanin - Quastel '14, De la Rue - Dierl '16

$$dP = \frac{1}{Z} \exp \left( \int_0^T \xi(s, B_s) ds \right) dW$$

space-time white noise

Wiener measure

$\Rightarrow$  under  $P$  the coordinate process solves

$$dX_t = J_x h(t, X_t) dt + dW_t$$

solution to KPZ, non-differentiable

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$$dX_t = J_x h(t, X_t) dt + dW_t$$

solution to KPZ, non-differentiable

1d

Comizzaro - Chouk '18: Similar story for space white noise,  $d = 1, 2, 3$ .

# Singular diffusions as stochastic characteristics I

KPZ equ.:  $\partial_t h = -\left(\frac{1}{2} \partial_{xx} h + \frac{1}{2} (\partial_x h)^2 + \xi\right), \quad h(T) = \bar{h}$

Gubinelli-P'17  
 $\Rightarrow h(t, x) = \sup_v \mathbb{E}_{t,x} \left[ \int_t^T \xi(s, X_s^v) ds + \bar{h}(X_T^v) \right]$

$$dX_t^v = v_t dt + dW_t$$

Optimal  $v_t = \partial_x h(t, X_t^v)$

↑ distributional drift

→ leads to good  $L^\infty$  bounds for  $h$

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Expect:

- Feynman-Kac type representations for general linear singular SPDEs (parabolic)
- Nonlinear equ. via BSDEs, McKean-Vlasov, stoch. control

## What's new here?

- Extension of Cammazza - Chouk'18 to Lévy noise
- "Quantitative" heat kernel bounds for singular diffusions in the "  
"Young regime"

# Solution theories for singular diffusions

- 1d / gradient drift : Itô - McKean construction / Dirichlet forms

Brox '86 / Mathieu '94

- 1d via martingale problem / Zvonkin transform

Flandoli - Russo - Wolf '03 / Bass - Chen '01

$$dX_t = b(X_t)dt + dW_t$$

$$\Rightarrow \text{formally: } f(X_t) - f(X_0) - \int_0^t \mathcal{L}f(X_s) ds = \text{Mart.}$$

$$\mathcal{L}f = b\partial_x f + \frac{1}{2}\partial_{xx} f$$

$f$  smooth  $\Rightarrow \mathcal{L}f$  distribution

# Solution theories for singular diffusions

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$$dX_t = b(X_t)dt + dW_t$$

$$\Rightarrow \text{formally: } f(X_t) - f(X_0) - \int_0^t Lf(X_s)ds = \text{Mart.}$$

$$Lf = b\partial_x f + \frac{1}{2}\partial_{xx} f$$

Idea: find non-smooth  $f$  s.t.  $Lf \in C_b$ .  $f$  smooth  $\Rightarrow Lf$  distribution

E.g. solve  $Lf = g$  (linear ODE in 1d)

- $d \geq 1$  and  $b \in C(\mathbb{R}_+, \mathbb{C}^{-\alpha})$ ,  $\alpha < \frac{1}{2}$  (Young regime):

Flandoli-Issoglio-Russo '17 via regularity analysis for

$$\partial_t u = Lu + f, \quad u(0) = \varphi$$

- $d=1$  and  $b \in C(\mathbb{R}_+, \mathbb{C}^{-\alpha})$ ,  $\alpha < \frac{2}{3}$  (rough path regime):

Delarue-Dier '16 via rough path analysis of  $\partial_t u = Lu + f$

(similar to Hairer's rough stochastic PDEs)

- $d > 1$  and  $b \in C(\mathbb{R}_+, \mathbb{C}^{-\alpha})$ ,  $\alpha < \frac{2}{3}$  (rough regime)  $\rightarrow$

$\alpha < 1$  would be  
subcritical  
(RS?)

Caruza-Chouk '18 via paraccontrolled distrib.

## Extension of Cannizzaro - Chouk to Lévy noise

$$dX_t = b(t, X_t)dt + dL_t$$

↗  $\alpha$ -stable Lévy process, generator  $\simeq (-\Delta)^{\frac{\alpha}{2}}$

Formal martingale problem:

$$f(t, X_t) - f(0, X_0) - \int_0^t (\partial_s + \mathcal{L}) f(s, X_s) ds = \text{Mart.}$$

$$= (\partial_s + b(s, x) \cdot \nabla - (-\Delta)^{\frac{\alpha}{2}}) f$$

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Formal martingale problem:

$$f(t, X_t) - f(0, X_0) - \int_0^t (\partial_s + b) f(s, X_s) ds = \text{Mart.}$$

$$= (\partial_s + b(s, x) \cdot \nabla - (-\Delta)^{\frac{\alpha}{2}}) f$$

→ Kolmogorov backward equation:  $e^{-\gamma t} \Rightarrow$  guess  $u \in C^{\alpha-\gamma} \Rightarrow \nabla u \in C^{\alpha-\gamma-1}$

gains  $C^\alpha$

$$(\partial_t - (-\Delta)^{\frac{\alpha}{2}}) u = b \cdot \nabla u + f$$

$\Rightarrow b \cdot \nabla u$  ok if  
 $-\gamma + (\alpha - \gamma - 1) > 0 \Leftrightarrow \gamma < \frac{\alpha-1}{2}$

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$$= (\partial_s + b(s, x) \cdot \nabla - (-\Delta)^{\frac{\alpha}{2}}) f$$

Young regime,

Athreya - Butkovsky - Mytnik <sup>20</sup>

Ling - Zhao '19

de Raynal - Manozzi '19

Lévy Brox would be

$$\beta \in C^{-\frac{1}{2}}, \frac{1}{2} > \frac{\alpha-1}{2}$$

→ Kolmogorov backward equation:  $\epsilon \in C^{-\gamma} \Rightarrow$  guess  $u \in C^{\alpha-\gamma} \Rightarrow \nabla u \in C^{\alpha-\gamma-1}$

gains  $C^\alpha$

$$(\partial_t - (-\Delta)^{\frac{\alpha}{2}}) u = -b \cdot \nabla u + f$$

$$\Rightarrow b \cdot \nabla u \text{ ok if } -\gamma + (\alpha - \gamma - 1) > 0 \Leftrightarrow \gamma < \frac{\alpha-1}{2}$$

# Paracontrolled ansatz

$$(\partial_t - (-\Delta)^{\frac{\alpha}{2}}) u = -\nabla u \cdot b + f = -\underbrace{\nabla u \otimes b}_{\in C^{\alpha-2\gamma-1}} + f + \underbrace{-\nabla u \otimes b}_{\in C^{-\gamma}}$$

(better)

"paraproducts"

$\nabla u \otimes b$

isolates  
singular part  
of the product

Thm. (Krengel-P. '20)

For  $\gamma < \frac{2\alpha-2}{3}$ ,  $b \in C_T C^{-\gamma}$ ,  $b \circ \nabla (\partial_t - (-\Delta)^{\frac{\alpha}{2}})^{-1} b \in C_T C^{\alpha-2\gamma-1}$ :

exists! paracontrolled solution  $u$ . Similar to Cannizzaro-Chouk with  $\Delta \rightarrow -(-\Delta)^{\frac{\alpha}{2}}$

# Paraccontrolled ansatz

$$(\partial_t - (-\Delta)^{\frac{\alpha}{2}}) u = -\nabla u \cdot b + f = -\underbrace{\nabla u \otimes b}_{C^\alpha C^{\alpha-2\gamma-1}} + \underbrace{-\nabla u \otimes b}_{C^{-\gamma}}$$

"paraproducts"

(better)

isolates  
singular part  
of the product

→ white noise:  $\gamma = \frac{1}{2} +$ , ok if  $\alpha > \frac{7}{4}$ . Critical regularity:

$$\alpha = \frac{3}{2} \quad (\text{via RS?})$$

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For  $\gamma < \frac{2\alpha-2}{3}$ ,  $b \in C_T C^{-\gamma}$ ,  $b \circ \nabla (\partial_t - (-\Delta)^{\frac{\alpha}{2}})^{-1} b \in C_T C^{\alpha-2\gamma-1}$ :

exists! paraccontrolled solution  $u$ . Similar to Cannizzaro-Chouk with  $\Delta \rightarrow -(-\Delta)^{\frac{\alpha}{2}}$

## Construction of the process

$$dX_t^n = b_n(t, X_t^n) dt + dL_t$$

$b_n \rightarrow b$  in "rough path norm"

Need: • Tightness of  $(X^n)$ .

• Any limit point  $X$  is s.t.  $\forall u$  solving  $(\partial_t - (-\delta)^{\alpha/2})u = -f \cdot \nabla u + f$ :

$$u(t, X_t) - u(0, X_0) - \int_0^t f(s, X_s) ds = \text{Mart.}$$

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Kremp-P. '20: • Tightness via stoch. calculus for Poisson random measures

(Campbell's formula), get  $E\left[\left|\int_s^t b_n(r, X_r^n) dr\right|^p\right] \leq |t-s|^{p(1-\frac{1}{\alpha})}$

• Martingale property easy once we know  $P(\Delta X_t \neq 0) = 0 \forall t \geq 0$ .

# Brox diffusion with Lévy noise

$$dX_t = \xi(X_t) dt + dL_t, \quad d=1$$

↑ take periodic white noise (avoid weighted spaces)

- Clear:  $\xi \in C^{-\frac{1}{2}}$
- Kremp - P. '20: can construct  $\xi \circ \nabla (\partial_t - (-\Delta)^{\alpha/2})^{-1} \xi \in C^{\alpha-2}$    
 if  $\alpha > \frac{3}{2}$  : critical regularity
- get periodic Brox diffusion for  $\alpha > \frac{7}{4}$ .

Next: heat kernel estimates in the Brownian case

Motivation: Mass asymptotics for 2d PAM

$$\partial_t u = \Delta u + u \xi, \quad \xi \text{ 2d white noise}$$

$$U(t) := \int_{\mathbb{R}^d} u(t, x) dx$$

Goal: study  $U(t)$  for  $t \rightarrow \infty$ .  
(a.s. limit)

## Known results

- $\xi$  smooth Gaussian potential  $\Rightarrow \frac{1}{t(\log t)^{\frac{1}{2}}} \log U(t) \rightarrow \sqrt{2d \mathbb{E}[\xi(0)^2]}$
- Carmona - Molchanov '95, Gärtner - König - Molchanov '00: 2nd order asymptotics  
 $\mathbb{E}[\xi(x)\xi(y)] = |x-y|^{-\alpha}, 0 < \alpha < \min\{2, d\}$   
 $\Rightarrow \frac{1}{t(\log t)^{\frac{2}{4-\alpha}}} \log U(t) \rightarrow c_\alpha \quad \text{Chen '14}$
- $\xi$  1d white noise  $\Rightarrow \frac{1}{t(\log t)^{\frac{2}{3}}} \log U(t) \rightarrow c \quad \text{Chen '14}$
- $\xi = \partial_t B^H$   $\Rightarrow \frac{1}{t(\log t)^{\frac{1}{1+H}}} \log U(t) \rightarrow c$   
fractional B.m., Hurst index  $H \in (0,1)$   
Chakraborty - Chen - Gao - Tindel '18

$$\frac{1}{t(\log t)^{\frac{1}{2}}}$$

$$\frac{1}{t(\log t)^{\frac{2}{4-\alpha}}}$$

$$\alpha \in (0, 2 \wedge d)$$

$$\frac{1}{t(\log t)^{\frac{2}{3}}}$$

$$\frac{1}{t(\log t)^{\frac{1}{1+H}}}$$

$$H \in (0, 1)$$

Note: Always  $\frac{1}{t(\log t)^\gamma} \log U(t) \rightarrow c$ ,  $\gamma \in [\frac{1}{2}, 1)$

$\gamma = 1$  is natural barrier,  $\gamma \geq 1$  needs renormalization.

Expect:  $\zeta \in C^{-k^-} \Rightarrow \gamma = \frac{1}{2-k} \vee \frac{1}{2}$

$$\rightsquigarrow \gamma < 1 \Leftrightarrow k < 1$$

$$\frac{1}{t(\log t)^{\frac{1}{2}}}$$

$$\frac{1}{t(\log t)^{\frac{2}{4-\alpha}}}$$

$$\alpha \in (0, 2)$$

$$\frac{1}{t(\log t)^{\frac{2}{3}}}$$

$$\frac{1}{t(\log t)^{\frac{1}{1+H}}}$$

$$H \in (0, 1)$$

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$\gamma = 1$  is natural barrier,  $\gamma \geq 1$  needs renormalization.

Expect:  $\xi \in C^{-k}$   $\Rightarrow \gamma = \frac{1}{2-k} + \frac{1}{2}$

2d white noise:

$$\rightarrow \gamma < 1 \Leftrightarrow k < 1$$

$$k=1$$

What breaks at  $K=1$ ?

$$\partial_t u = \Delta u + u \xi, \quad u(0) = \delta_0$$

All proofs based on Feynman-Kac:  $U(t) = \mathbb{E}[\exp(\int_0^t \xi(W_s) ds)]$

But:  $\xi \in C^{-1}$   $\Rightarrow$  Cannot construct

$$\int_0^t \xi(W_s) ds$$


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Krieg-van Tuylgen-P.'20: Still can show

$$\lim_{t \rightarrow \infty} \frac{1}{t \log t} \log U(t) \rightarrow \chi \quad \text{for 2d white noise}$$

via "transformed Feynman-Kac formula"

# Making the Feynman-Kac formula work

Let  $-\Delta Z = \xi$

$$\Rightarrow Z(B_t) = Z(B_0) + \int_0^t \nabla Z(B_s) \cdot dB_s + \underbrace{\int_0^t \xi Z(B_s) ds}_{\stackrel{\text{"1/2=1"}}{\downarrow}} = - \int_0^t \xi(B_s) ds$$

$$\Rightarrow \int_0^t \xi(B_s) ds = Z(B_0) - Z(B_t) + \int_0^t \nabla Z(B_s) \cdot dB_s$$

ok if  $\xi \in C^{-1+\varepsilon}$  ( $\Rightarrow Z \in C^{1+\varepsilon}$ )

here: just fails

## Making Feynman-Kac work

$$\int_0^t \zeta(B_s) ds = Z(B_0) - Z(B_t) + \int_0^t \nabla Z(B_s) \cdot dB_s$$

$$\Rightarrow U(t) = \mathbb{E} \left[ \exp \left( Z(B_0) - Z(B_t) + \int_0^t \nabla Z(B_s) \cdot dB_s \right) \right]$$

$$\begin{aligned} &= \mathbb{E} \left[ \exp \left( Z(0) - Z(B_t) + \left( \int_0^t \nabla Z(B_s) \cdot dB_s - \frac{1}{2} \int_0^t |\nabla Z(B_s)|^2 ds \right) \right. \right. \\ &\quad \left. \left. + \frac{1}{2} \int_0^t |\nabla Z(B_s)|^2 ds \right) \right] \end{aligned}$$

# Making Feynman-Kac work

$$\int_0^t \zeta(B_s) ds = Z(B_0) - Z(B_t) + \int_0^t \nabla Z(B_s) \cdot dB_s$$

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$$= \mathbb{E} \left[ \exp \left( Z(0) - Z(X_t) + \frac{1}{2} \int_0^t |\nabla Z(X_s)|^2 ds \right) \right] + \frac{1}{2} \int_0^t |\nabla Z(B_s)|^2 ds$$

$$X_t = \int_0^t \nabla Z(X_s) ds + B_t \quad (\text{weakly})$$

# Making Feynman-Kac work

$$\int_0^t \zeta(B_s) ds = Z(B_0) - Z(B_t) + \int_0^t \nabla Z(B_s) \cdot dB_s$$

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$|\nabla Z|^2$  ill-defined ( $Z \in C^{1-\epsilon}$ )

$$X_t = \int_0^t \nabla Z(X_s) ds + B_t \quad (\text{weakly})$$

Probabilistic argument: construct  
 $|\nabla Z|^2$  as (random) distribution  
 in  $C^{-\epsilon}$

# Rigorous Feynman-Kac for 2d PAM

It's trick once more:

$$-\Delta Z = \xi, \quad (-\Delta - \nabla Z \cdot \nabla) Y = -|\nabla Z|^2, \quad W = Z + Y$$

$\hookrightarrow C^{-\varepsilon} \Rightarrow Y \in C^{2-\varepsilon}$   
 $\Rightarrow |\nabla Y|^2 \text{ ok}$

$$dX_t = \nabla W(X_t) dt + dB_t$$

$$\Rightarrow U(t) = \mathbb{E} \left[ \exp \left( W(0) - W(X_t) - \frac{1}{2} \int_0^t |\nabla Y|^2(X_s) ds \right) \right]$$

→ Get asymptotics of  $U(t)$  if we understand  $X$  well.  
(upper + lower tail bounds)

# Heat kernel estimates for singular diffusions

$$dX_t = b(t, X_t)dt + dB_t \quad b \in C(\mathbb{R}_+, \mathbb{C}^{-\alpha}), \alpha < \frac{1}{2}$$

→ not hard to show: ∃ transition density

$$\Gamma_{s,t}(x,y)dy = \mathbb{P}(X_t \in dy | X_s = x)$$

Zhang-Zhao '17: ∃ M > 1, ∀ T > 0 ∃ C > 0 depending only on T and

$$\|b\|_{C(\mathbb{R}, T, \mathbb{C}^{-\alpha})} :$$

$$\frac{1}{C} \gamma\left(\frac{t}{M}, x-y\right) \leq \Gamma_t(x,y) \leq C \gamma(Mt, x-y)$$

(if  $\alpha < \frac{1}{2}$ )

↑ standard Gaussian kernel

based on  
"Zwanzig  
transform"

# Heat kernel estimates for singular diffusions

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→ not hard to show:  $\exists$  transition density

$$\Gamma_t(x, y) dy = P_x(X_t \in dy)$$

We need explicit dependence on  $T$ ,  
||b||.

Zhang-Zhao '17:  $\exists M > 1, \forall T > 0 \quad \exists C > 0$  depending only on  $T$  and

$$\|b\|_{C([0, T], C^{-\alpha})}$$

(if  $\alpha < \frac{1}{2}$ )

$$\frac{1}{C} \gamma\left(\frac{t}{M}, x-y\right) \leq \Gamma_t(x, y) \leq C \gamma(Mt, x-y)$$

standard Gaussian kernel

based on  
"Zwanzig transform"

# "Quantitative" heat kernel bounds for singular diffusions

$$dX_t = b(t, X_t) dt + dB_t \quad , \quad \Gamma_{s,t}(x,y) dy = \mathbb{P}(X_t \in dy \mid X_s = x)$$

Thm. (van Zuijlen - P. '20)

Young regime  
 $\forall \alpha \in (0, \frac{1}{2}), c > 1 \quad \exists C > 0, M > 1:$

$$\frac{1}{C} \exp\left(-C(t-s)\|b\|_{C_t C^{-\alpha}}^{\frac{2}{1-\alpha}}\right) \gamma\left(\frac{t-s}{M}, x-y\right) \leq \Gamma_{s,t}(x,y)$$

$$\leq C \exp\left(C(t-s)\|b\|_{C_t C^{-\alpha}}^{\frac{2}{1-\alpha}}\right) \gamma(c(t-s), x-y)$$

$\forall b \in C(\mathbb{R}_+, \mathbb{C}^\alpha), \forall 0 \leq s < t$

Idea of the proof Fix  $s=0$ ,  $\Gamma_t := \Gamma_{0,t}$

$$\partial_t \Gamma_t = \underbrace{\frac{1}{2} \Delta \Gamma_t + b \cdot \nabla \Gamma_t}_{\mathcal{L} \Gamma_t}, \quad \Gamma_t = \delta_y$$

Parametrix method:  $\partial_t \eta_t = \frac{1}{2} \Delta \eta_t, \eta_0 = \delta_y$

compare Friedman's book

$$\Rightarrow \tau_t := \Gamma_t - \eta_t \text{ solves } (\partial_t - \mathcal{L}) \tau_t = - \underbrace{b \cdot \nabla \Gamma_t}_{=: \Phi_t}, \quad \tau_0 = 0$$

space-time  
"convolution"

$$\leadsto \tau = \Gamma * \Phi$$

Idea of the proof Fix  $\gamma = 0$ ,  $\Gamma_t := \Gamma_{0,t}$

$$\partial_t \Gamma_t = \underbrace{\frac{1}{2} \Delta \Gamma_t + b \cdot \nabla \Gamma_t}_{\mathcal{L} \Gamma_t}, \quad \Gamma_t = \delta_y$$

Parametrix method:  $\partial_t \eta_t = \frac{1}{2} \Delta \eta_t, \eta_0 = \delta_y$

$$\Rightarrow \tau_t := \Gamma_t - \eta_t \text{ solves } (\partial_t - \mathcal{L}) \tau_t = -\underbrace{b \cdot \nabla \Gamma_t}_{=: \underline{\Phi}_t}, \quad \tau_0 = 0$$

space-time  
"convolution"

$$\rightarrow \tau = \Gamma * \underline{\Phi}$$

$$\Gamma = \eta + \tau = \eta * \underline{\Phi} + \Gamma * \underline{\Phi}^{*2}$$

$$\dots \Rightarrow \Gamma = \sum_{n=0}^{\infty} \eta * \underline{\Phi}^{*n}$$

control each term inductively

$$\Gamma = \sum_{n=0}^{\infty} n * \underline{\oplus}^{*n}$$

control each term inductively

$\approx 8$  pages of computations:  
get upper bound for  $|\Gamma - n|$

$$\Gamma = \sum_{n=0}^{\infty} \gamma * \bar{\Phi}^{\ast n}$$

control each term inductively

$\approx 8$  pages of computations:

get upper bound for  $|\Gamma - \gamma|$

Chapman-Kolmogorov

+ "chaining"

$\Rightarrow$  lower bound (cf. Stroock's

PDE book)

# Summary

- Singular diffusions arise as processes in random media  
+ as stochastic characteristics for singular SPDEs.
- Weak solution theory via martingale problem.
- Here 2 new results:
  - Brox diffusion with Lévy noise
  - quant. heat kernel estimates in Brownian Young regime

Thank You