

Revisiting Clark's robustness problem

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6th ICSAA in Poland, September 2012

- I. CLARK'S ROBUSTNESS PROBLEM
- II. ROBUSTNESS VIA ZAKAI SPDE (joint with Friz)
- III. ROBUSTNESS VIA KALLIANPUR-STRIEBEL FUNCTIONAL
(joint with Crisan, Diehl, Friz)

I. CLARK'S ROBUSTNESS PROBLEM

Nonlinear filtering

Fix $(\Omega, \mathcal{F}, (\mathcal{F}_t), \mathbb{P})$. Consider the pair (X, Z) where

$$\begin{aligned}dX_t &= \mu(X_t) dt + V(X_t) \circ dB_t \text{ (signal in } \mathbb{R}^{d_x}\text{)} \\dZ_t &= h(X_t) dt + d\tilde{B}_t \text{ (observation in } \mathbb{R}^{d_z}\text{)}\end{aligned}$$

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Goal: Given real-valued function f , compute

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There exists a measurable map $\theta_t^f : C([0, t], \mathbb{R}^{d_z}) \rightarrow \mathbb{R}$

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Clark's robustness problem

- ▶ only discrete observations $0 \leq t_1 \leq \dots \leq t_n \leq t$ of Z available
 - ▶ BV path Z^n which approximates Z
 - ▶ BUT $\theta_t^f : C([0, t], \mathbb{R}^{d_V}) \rightarrow \mathbb{R}$ not unique, every $\tilde{\theta}_t^f$
s.t. $\theta_t^f(\cdot) = \tilde{\theta}_t^f(\cdot) \mathbb{P} \circ (Z|_{[0, t]})^{-1}$ - a.s fulfills

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Solution. If B and \tilde{B} independent (Clark78, ClarkCrisan05, Davies 80/81,...) then

$$\exists! \theta_t^f : \left(C([0, t], \mathbb{R}^{d_Z}), \|\cdot\|_\infty \right) \rightarrow \mathbb{R}$$

continuous in supremums norm.

Correlated noise

$$\begin{aligned}dX_t &= \mu(X_t) dt + V(X_t) \circ dB_t + \sigma(X_t) \circ d\tilde{B}_t \text{ (signal)} \\dZ_t &= h(X_t) dt + d\tilde{B}_t \text{ (observation)}\end{aligned}$$

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Bad news!

$$\nexists \theta_t^f : C([0, t], \mathbb{R}^{dz}) \rightarrow \mathbb{R}$$

s.t.

- ▶ $\pi_t(f) = \theta_t^f(Z|_{[0,t]})$ \mathbb{P} -a.s.
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Our main result. $\exists ! \theta_t^f : C([0, t], G^2(\mathbb{R}^{dz})) \rightarrow \mathbb{R}$

- ▶ continuous in a rough path metric
- ▶ $\theta_t^f(1 + Z + \int Z \otimes dZ) = \pi_t(f)$ \mathbb{P} -a.s

II. ROBUSTNESS OF THE ZAKAI SPDE (joint with P. Friz)

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- ▶ Assume density $\rho_t(f) = \int_{\mathbb{R}^{d_X}} f(x) u_t(x) dx$
- ▶ Described via a linear, parabolic SPDE (*dual Zakai equation*)

$$\begin{aligned} du &= G^* u dt + \sum_i N_i^* u dZ^i \\ &= \left(G^* - \frac{1}{2} \sum_i N_i N_i^* \right) u dt + \sum_i N_i^* u \circ dZ_t^i \end{aligned}$$

with

$$\begin{aligned} G &\dots \text{ generator of } X \\ N_j^* u &= \sigma_j \cdot Du + h_j u. \end{aligned}$$

General SPDE

Find $u : [0, T] \times \mathbb{R}^e \rightarrow \mathbb{R}$ which solves

$$\begin{aligned} du + L(t, x, u, Du, D^2 u) dt &= \sum_{i=1}^d \Lambda_i(t, x, u, Du) \circ dZ_t^i, \quad (0, T) \times \mathbb{R}^e \\ u(0, \cdot) &= u_0(\cdot) \text{ on } \mathbb{R}^e \end{aligned}$$

where

$$L : [0, T] \times \mathbb{R}^e \times \mathbb{R} \times \mathbb{R}^e \times \mathbb{S}^e \rightarrow \mathbb{R}$$

$$\Lambda_i : [0, T] \times \mathbb{R}^e \times \mathbb{R} \times \mathbb{R}^e \rightarrow \mathbb{R}$$

are (affine) linear.

Question: Regularity of

$$Z \mapsto u$$

Heuristic explanation

Take $L \equiv 0$, only gradient noise (corresponds to correlation in the filtering set-up!!!), i.e.

$$\begin{aligned} du &= \langle Du, \sigma_1(x) \rangle \circ dZ_t^1 + \langle Du, \sigma_2(x) \rangle \circ dZ_t^2 \\ u(0, \cdot) &= u_0(\cdot) \end{aligned}$$

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If ϕ^Z denotes the SDE flow of

$$dY_t = \sigma_1(Y_t) \circ dZ_t^1 + \sigma_2(Y_t) \circ dZ_t^2$$

then (formally) $u(t, x) = u_0(\phi^Z(t, x))$.

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Question: Robustness of SDE solutions,

$$Z \mapsto \phi^Z.$$

Answer: Poor robustness in uniform norm (except for degenerate situations where vectorfields σ_1, σ_2 commute). Not even continuous!

INTERMEZZO: ROUGH PATH THEORY

Theorem

(Wong–Zakai). Let B^n be the piecewise linear approximation to B along the dyadics of $[0, T]$. Then the ODE solutions (Y^n) of

$$Y_t^n = V(Y_t^n) dB_t^n, Y_0^n = y$$

converge uniformly to Y , the solution of the Stratonovich SDE

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BUT there are approximations (B^n)

$$|B^n - B|_{\infty, [0, T]} \rightarrow_n 0 \text{ s.t. } |Y^n - \bar{Y}| \rightarrow_n 0$$

where

$$d\bar{Y} = V(\bar{Y}_t) \circ dB_t + c(\bar{Y}_t) dt$$

and c is any linear combination of

$$[V_{i_1}, [V_{i_2} \cdots [V_{i_{N-1}}, V_{i_N}] \cdots]] .$$

- **Lyons '98.** Let $(z^n) \subset C^1([0, T], \mathbb{R}^d)$ be Cauchy in rough path metric with limit z . Assume

$$\text{(ODE)} \quad y^n = V(y^n) \dot{z}^n, \quad y^n(0) = y_0 \in \mathbb{R}^e$$

then y^n converges uniformly to some $y = y^z \in C([0, T], \mathbb{R}^e)$ which is **independent** of the approximating sequence.

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- ▶ **Interpretation:** y is the solution of a *rough differential equation* driven by the *rough path* z . Write

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- ▶ What are *rough path metrics* and *rough paths*?
- ▶ First example (not applicable to Brownian motion): take

$$\rho_{\alpha\text{-Hol}}(z, \bar{z}) = \frac{|z_{s,t} - \bar{z}_{s,t}|}{(t-s)^\alpha} \text{ for } \alpha \in \left(\frac{1}{2}, 1\right]$$

and *rough paths* are just α -Hölder paths, RDEs are “Young” ODEs.

- Better example (applicable to Brownian paths): for $\alpha \in (\frac{1}{3}, \frac{1}{2}]$ take

$$\rho_{\alpha-Hoel}(z, \bar{z}) = \sup_{s < t} \frac{|z_{s,t}^1 - \bar{z}_{s,t}^1|}{|t-s|^\alpha} + \frac{|z_{s,t}^2 - \bar{z}_{s,t}^2|}{|t-s|^{2\alpha}}$$

where we introduced the generalized increments of $z \in C^1$ as

$$z_{s;t} := (z_{s,t}^1, z_{s,t}^2) := \left(\int_s^t dz, \int_s^t \int_s^{r_2} dz_{r_1} \otimes dz_{r_2} \right) \in \mathbb{R}^d \oplus (\mathbb{R}^d)^{\otimes 2}$$

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- ▶ The abstract completion of C^1 -paths wrt to $\rho_{\alpha-Hoel}$ leads to a *rough path space* which can be identified as a subset of

$$\left\{ z \in C \left([0, T], \mathbb{R}^d \oplus (\mathbb{R}^d)^{\otimes 2} \right) : \sup_{s \neq t} \frac{|z_{s,t}^1|}{(t-s)^\alpha} + \frac{|z_{s,t}^2|}{(t-s)^{2\alpha}} < \infty \right\}$$

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- ▶ From $d(z^i z^j) = z^i dz^j + z^j dz^i$ it follows $Sym(z^2) = \frac{1}{2} z^1 \otimes z^1$ and

$$(z_{s,t}^1; z_{s,t}^2) \leftrightarrow (z_{s,t}^1, \mathbf{a}_{s,t}) \text{ with } \mathbf{a}_{s,t} := Anti(z_{s,t}^2)$$

Advantages to a Probabilist

- ▶ RDE solution of $dy = V(y) dB(\omega)$ is solved for fixed ω ; depends continuously on $\mathbf{B} = (B, \int B \circ dB)$ and coincides with (Stratonovich) solution of $dY = V(Y) \circ dB$.

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For e.g. $V \in Lip^{3+\epsilon}$ can see that $\phi, D\phi, D^2\phi$ exist and depend continuously on z , also $\phi^{-1}, D\phi^{-1}, D^2\phi^{-1}$. Limit theorems on the level of stochastic flows!

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- ▶ No restriction to semimartingales as noise (as long as we can construct the higher levels)
- ▶ Continuity of solution map $z \mapsto y$ makes it easier to prove large deviations, Freidlin-Wentzell estimates, support theorems etc.

BACK TO ZAKAI SPDE

Viscosity solutions

Theorem (Friz-Caruana-O, Friz-O)

Let the coefficients in L and Λ fulfill (TC). Let $(z^n) \subset C^1([0, T], \mathbb{R}^d)$ and consider the viscosity solutions (u^n) of

$$\begin{aligned} du^n + L(t, x, u^n, Du^n, D^2 u^n) dt &= \Lambda(t, x, u^n, Du^n) dz^n \\ u^n(0, \cdot) &= u_0(\cdot) \in BUC(\mathbb{R}^e) \end{aligned}$$

If (z^n) converges to a geometric rough path z then $\exists!$
 $u^z \in BUC([0, T] \times \mathbb{R}^e, \mathbb{R})$ such that $u^n \rightarrow u^z$ (loc. uniformly on compacts). Further,

1. u^z **independent** of the choice of (z^n) ,
2. $(z, u_0) \mapsto u^z$ is continuous,
3. $|u^z - v^z|_{\infty; [0, T] \times \mathbb{R}^n} \leq e^{cT} |u_0 - v_0|_{\infty, \mathbb{R}^n}$

Remark

Motivated by Lions-Souganidis theory of viscosity SPDEs

Conditions **(TC)**:

$$L(t, x, r, p, M) = -\text{Tr} \left[a(t, x) \cdot a^T(t, x) M \right] + b(t, x) \cdot p + c(t, x, r)$$

$$\Lambda = (\Lambda_1, \dots, \Lambda_d)$$

$$\Lambda_k(t, x, r, p) = \langle p, \sigma_k(t, x) \rangle + r \cdot \nu_k(t, x) + g_k(t, x)$$

- ▶ a, b bounded, continuous in t , Lipschitz in x (uniformly in t)
- ▶ c continuous and bounded for bounded r with a lower Lipschitz constant
- ▶ All coefficients in Λ are Lip^γ for $\gamma > \frac{1}{\alpha} + 2$

- ▶ L can be semilinear and degenerate elliptic (i.e. first order case no problem)
- ▶ If $z = (B, \int B \circ dB)$ get approximations theorems, support results, large deviations for SPDEs
- ▶ SPDEs with non-Brownian or non-semimartingale (e.g. fBM) noise
- ▶ etc.

L^2 solutions

Apply with $z = \mathbf{B}(\omega) = (B(\omega), \int B \circ dB(\omega))$.

Proposition (Friz-O)

If L is uniformly elliptic, (TC) and \tilde{L}^* exists, then $u^{\mathbf{B}}$ is “the” unique $L^2(\mathbb{R}^n)$ -solution: $\forall \varphi \in C_c^\infty(\mathbb{R}^n)$

$$\langle u_t, \varphi \rangle_{L^2} - \langle u_0, \varphi \rangle_{L^2} = \int_0^t \langle u_r, \tilde{L}^* \varphi \rangle_{L^2} dr + \int_0^t \langle u_r, \Lambda_k^* \varphi \rangle_{L^2} dB_r^k$$

with $\tilde{L}\varphi = \left(L + \frac{1}{2} \sum_{k=1}^d \Lambda_k \Lambda_k^* \right) \varphi$.

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Remark

- ▶ Connects RPDEs to classic L^2 -theory
- ▶ $u^{\mathbf{B}}$ is a robust version (in the equivalence class) of unique L^2 -solution
- ▶ no Sobolev embedding needed

III. ROBUSTNESS VIA THE KALLIANPUR-STRIEBEL FUNCTIONAL (joint with Crisan, Diehl, Friz)

Pathwise filtering

$$dX_t = \mu(X_t) dt + V(X_t) \circ dB_t \text{ (signal)}$$

$$dZ_t = h(X_t) dt + d\tilde{B}_t \text{ (observation)}$$

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- ▶ $B = (B^i)_{i=1}^d$ and $\tilde{B} = (\tilde{B}^j)_{j=1}^e$ standard BM
- ▶ $\langle B^i, \tilde{B}^j \rangle = \rho_{ij} t$

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Goal: Find a robust version of

$$\pi_t(f) = \mathbb{E}[f(X_t) | \sigma(Z_s, s \in [0, t])].$$

Transformation

(For simplicity of presentation X, Z, B, \tilde{B} 1-dimensional)

- ▶ Define \mathbb{P}_0 via

$$\frac{d\mathbb{P}_0}{d\mathbb{P}} = \exp\left(-\int_0^T h(X_r) d\tilde{B} - \frac{1}{2} \int_0^T |h(X_r)|^2 dr\right)$$

- ▶ Set $v_t := \rho h(X_t) + B_t$ and $w_t := v - \rho Z_t$

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- ▶ Set $v_t := \rho h(X_t) + B_t$ and $w_t := v - \rho Z_t$

Then under \mathbb{P}_0 ,

1. Z and v are standard BM and $\langle v_t, Z_t \rangle = \rho t$
2. $W_t := \frac{1}{\sqrt{1-\rho^2}} w_t$ is standard BM, independent of Z

Transformation

Using this, the signal X becomes

$$dX_t = L_0(X_t) dt + L(X_t) \circ dZ_t + M(X_t) \circ dW_t$$

- ▶ W, Z independent BM under \mathbb{P}_0
- ▶ $M = \sqrt{1 - \rho^2} V, L_0 = \mu - \rho h V, L = \rho V$

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- ▶ W, Z independent BM under \mathbb{P}_0
- ▶ $M = \sqrt{1 - \rho^2} V, L_0 = \mu - \rho h V, L = \rho V$
- ▶ **This is the key formula to robust filtering**

Kallianpur-Striebel

KS-formula

$$\pi_t(f) = \frac{\rho_t(f)}{\rho_t(1)}$$

with

$$\rho_t(f) = \mathbb{E}_{\mathbb{P}_0} \left[\underbrace{f(X_t)}_{(i)} \underbrace{\exp \left(\int_0^t h(X_r) dZ_r - \frac{1}{2} \int_0^t |h(X_r)|^2 dr \right)}_{(ii)} \middle| Z \right].$$

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Express (i) and (ii) as functionals of Z and show regularity of

$$Z \mapsto (i) \text{ and } Z \mapsto (ii)$$

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A rough path result

Theorem (Crisan, Diehl, Friz, O)

Let $(z^n) \subset C^1([0, T], \mathbb{R}^d)$ and $z^n \rightarrow z$ in rough path metric. If W is a standard BM, then for a.s. the solutions of the SDE

$$dX_t^n = L_0(X_t^n) dt + L(X_t^n) dz_t^n + M(X_t^n) \circ dW_t.$$

converge uniformly to a continuous path $X(\omega)$. We write formally

$$dX_t = L_0(X_t) dt + L(X_t) dz_t + M(X_t) \circ dW_t.$$

Further, $z \mapsto X$ is continuous wrt $|X|_{S^q} := \mathbb{E} \left[\sup_{t \in [0, T]} |X_t|^q \right]^{1/q}$ any $q \geq 1$.

Proof (sketch)

Two possible approaches:

1. Use a “Kunita flow decomposition”
2. Construct a joint rough path of

$$z = (z^1, z^2) \text{ and } \mathbf{W} = \left(W, \int dW \otimes dW \right)$$

use rough path continuity

1. Flow decomposition

Lemma

Take $z \in C^1([0, T], \mathbb{R}^e)$, W a standard d -dimensional BM. Let X be the unique SDE solution of

$$dX_t = L_0(X_t) dt + L(X_t) dz + M(X_t) \circ dW.$$

Consider the transformation

$$\phi(t, x) = x + \int_0^t L(\phi(t, x)) dz.$$

Then $\bar{X}_t := \phi^{-1}(t, X_t)$ solves the SDE

$$d\bar{X}_t = \bar{L}_0(\bar{X}_t) dt + \bar{M}(\bar{X}_t) \circ dW_t$$

with $\bar{M}_{ij} := \sum_k \partial_k \phi_i^{-1}(t, \phi(t, x)) M_{k,j}(t, \phi(t, x))$, $\bar{L}_0 := \dots$

Proof.

Ito!

1. Flow decomposition

- ▶ Construct the “rough path flow”

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Note: stable under smooth approximations to z in rough metric

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$$d\bar{X}_t^z = \bar{L}_0^z(\bar{X}_t) dt + \bar{M}^z(\bar{X}_t) \circ dW_t$$

and set

$$X_t^z := \phi^z(t, \bar{X}_t^z)$$

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- ▶ By construction

$$z \mapsto X^z \text{ is continuous wrt } |X|_{S^q} := \mathbb{E} \left[\sup_{t \in [0, T]} |X_t|^q \right]^{1/q}$$

A Rough & Stochastic DE

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Executive summary

$$dX_t = \mu(X_t) dt + V(X_t) \circ dB_t \text{ (signal)}$$

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Fact: If B and \tilde{B} are correlated then

$$\nexists \theta_t^f : C([0, t], \mathbb{R}^e) \rightarrow \mathbb{R}$$

s.t.

- ▶ $\pi_t(f) = \theta_t^f(Z|_{[0,t]})$ \mathbb{P} -a.s.
- ▶ θ_t^f is continuous in uniform norm

Executive summary

$$dX_t = \mu(X_t) dt + V(X_t) \circ dB_t \text{ (signal)}$$

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Fact: If B and \tilde{B} are correlated then








$$\exists! \theta_t^f : C([0, t], G^2(\mathbb{R}^e)) \rightarrow \mathbb{R}$$

s.t.

- ▶ $\pi_t(f) = \theta_t^f(Z|_{[0,t]})$ \mathbb{P} -a.s. with

$$Z = 1 + \int dZ + \int dZ \otimes dZ = \exp(Z, \text{Area}(Z))$$

- ▶ θ_t^f is continuous in rough path norm (even locally Lipschitz!)

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THANK YOU!