

Weighted L^1 -semigroup approach for nonlinear Fokker–Planck equations and generalized Ornstein–Uhlenbeck processes

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FPEs, SDEs, Markov processes: Linear case

For $a = (a_{ij})_{i,j \leq d} : \mathbb{R}_+ \times \mathbb{R}^d \rightarrow \mathbb{R}^{d \times d}$, $b = (b_i)_{i \leq d} : \mathbb{R}_+ \times \mathbb{R}^d \rightarrow \mathbb{R}^d$ Borel measurable, consider **linear** Fokker–Planck equation (FPE)

$$\partial_t \mu_t = \partial_{ij}^2 (a_{ij}(t, x) \mu_t) - \partial_i (b_i(t, x) \mu_t), \quad t \geq s, \quad \mu_s = \zeta \in \mathcal{P}, \quad (\ell\text{FPE})$$

$\mathcal{P} =$ prob. measures on $\mathcal{B}(\mathbb{R}^d)$.

Corresponding SDE, with $\frac{1}{2} \sigma \sigma^T = a$, is

$$dX_t = b(t, X_t) dt + \sigma(t, X_t) dB_t, \quad t \geq s, \quad X_s \sim \zeta. \quad (\text{SDE})$$

Recall:

- X solution to (SDE) $\implies (\mu_t)_{t \geq s} := (\mathcal{L}_{X_t})_{t \geq s}$ solves (ℓ FPE).
- $(\mu_t)_{t \geq s}$ solution to (ℓ FPE) $\implies \exists$ solution X to (SDE) with $\mathcal{L}_{X_t} = \mu_t$.
- If (ℓ FPE) is well-posed, so is (SDE), and unique solution laws of the latter are a Markov process with one-dim. marginals equal to (ℓ FPE)-solutions.

This connection to Markov processes fails, if the FPE is **nonlinear**, i.e.

$$\partial_t \mu_t = \partial_{ij}^2 (a_{ij}(t, \mu_t, x) \mu_t) - \partial_i (b_i(t, \mu_t, x) \mu_t), \quad t \geq s, \quad \mu_s = \zeta, \quad (\text{nlfPE})$$

since the corresponding stochastic equation is distribution-dependent:

$$dX_t = b(t, \mathcal{L}_{X_t}, X_t)dt + \sigma(t, \mathcal{L}_{X_t}, X_t)dB_t, \quad t \geq s, \quad X_s \sim \zeta, \quad (\text{DDSDE})$$

and even in well-posedness cases, solutions to DDSDEs do NOT satisfy the usual Markov property.

To build a theory as in the linear case, in [R./Röckner22] we introduced and studied a notion of *nonlinear Markov processes*, and proved:

Given a semigroup of solutions to a nlfPE, there is a nonlinear Markov process, consisting of solution laws to the DDSDE with one-dim. marginals given by the semigroup.

This gives a *probabilistic representation* of nlfPE-solutions as marginals of nonlinear Markov processes.

Note: We do NOT need uniqueness of the nlfPE.

Nemytskii-type FPEs = interesting class of PDEs

Consider local, singular dependence of coefficients on measure:

$$a(t, \mu, x) = \tilde{a}\left(t, \frac{d\mu}{dx}(x), x\right), \quad b(t, \mu, x) = \tilde{b}\left(t, \frac{d\mu}{dx}(x), x\right),$$

for $\tilde{a}: \mathbb{R}_+ \times \mathbb{R} \times \mathbb{R}^d \rightarrow \mathbb{R}^{d \times d}$, $\tilde{b}: \mathbb{R}_+ \times \mathbb{R} \times \mathbb{R}^d \rightarrow \mathbb{R}^d$.

Rewriting the FPE for densities $t \mapsto u_t = \frac{d\mu_t}{dx}$ gives a nonlinear PDE:

$$du_t = \partial_{ij}^2(a_{ij}(t, u_t, x)u_t) - \partial_i(b_i(t, u_t, x)u_t), \quad t \geq s, \quad u_t(x)dx \xrightarrow{t \rightarrow s} \zeta.$$

Interesting PDEs are of this type: e.g. Burgers, classical and generalized PME, 2D vorticity Navier–Stokes. For all of these:

There is a nonlinear Markov process with one-dim. marginals equal to solutions to this equation, see [R./Röckner22], [BarbuRöcknerZhang23].

Next goals:

- Develop theory of nonlinear Markov processes and apply it to nonlinear PDEs and DDSDE to prove new results.
- Solve new classes of nIFPE and link them to nonlinear Markov processes. **This we do today!**

Good method to construct solution semigroups for Nemytskii-type FPE: Crandall–Liggett nonlinear semigroup approach in $L^1(\mathbb{R}^d, \mathbb{R}; dx)$. This way, [BarbuRöckner21] solved

$$\partial_t u(t) = \Delta \beta(u(t)) - \operatorname{div}(D(x)b(u(t))u(t))$$

for curves of probability densities $t \mapsto u(t)$. Then: Connection to DDSDE, nonlinear Markov processes...

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Our goal: Solve

$$\partial_t u = \Delta \beta(u) - \nabla \Phi \cdot \nabla \beta(u) - \operatorname{div}_\rho(D(x)b(u)u) \quad (\text{FP}_\rho)$$

by semigroup method in $L^1(\mathbb{R}^d; \rho dx)$, where

$\rho : \mathbb{R}^d \rightarrow \mathbb{R}$ L^1 -(probability) density, $\Phi = -\log \rho$ (i.e. $\rho = e^{-\Phi}$),

$\operatorname{div}_\rho =$ dual of ∇ in $L^2(\mathbb{R}^d; \rho dx)$.

Example: $\rho(x) = e^{-\frac{|x|^2}{2}}$, $\Phi = \frac{|x|^2}{2}$ (Ornstein–Uhlenbeck processes, see later).

Theorem (R.23)

Assume the following hypothesis.

(H1) $0 \leq \Phi \in C^2$ is convex with $\lim_{|x| \rightarrow \infty} \Phi(x) = \infty$, and $\nabla \Phi \in L^1(\rho)$.

(H2) $\beta \in C^2(\mathbb{R})$, $\beta'(r) > 0$ for $r \neq 0$, and $\beta(0) = 0$.

(H3) $b \geq 0$, $b \in C_b(\mathbb{R})$.

(H4) $D \in L^\infty(\mathbb{R}^d, \mathbb{R}^d)$, $\operatorname{div}_\rho D \in L^2_{loc}$, $(\operatorname{div}_\rho D)^- \in L^\infty$.

Then, for each $u_0 \in L^1(\rho dx) \cap L^\infty$, there is a mild and distributional solution $u = u(u_0) \in C([0, T]; L^1(\rho dx))$ with

$$|u(t)|_\infty \leq \exp(|(\operatorname{div}_\rho D)^- + \|D\|_\infty^{\frac{1}{2}} t|) |u_0|_\infty, \quad \forall t \geq 0.$$

Moreover, $S(t) : u_0 \mapsto u(t)$, is a semigroup of $L^1(\rho)$ -contractions, i.e.

$$S(t+s)u_0 = S(t)S(s)u_0, \quad S(0) = Id,$$

$$|S(t)u_0 - S(t)v_0|_{L^1(\rho)} \leq |u_0 - v_0|_{L^1(\rho)}.$$

If $u_0(x)\rho(x)$ is a probability density, so is each $u(t)\rho(x)$.

If in addition β is Lipschitz, everything holds for $u_0 \in L^1(\rho)$ instead of $u_0 \in L^1(\rho) \cap L^\infty$.

Definition

(i) Let $A: D(A) \subseteq X \rightarrow X$ be a nonlinear operator in a Banach space X .
A **mild solution** to

$$\frac{d}{dt}u(t) = -A(u(t)), \quad u(0) = u_0, \quad t \geq 0,$$

is $u \in C(\mathbb{R}_+, X)$ such that $u = \lim_{h \rightarrow 0}^X u_h$ locally uniformly in time,

$$u_h(0) := u_h^0 := u_0,$$

$$u_h(t) := u_h^i, \forall t \in ((i-1)h, ih], i \in \mathbb{N},$$

$$u_h^i \in D(A), u_h^i + hAu_h^i = u_h^{i-1}.$$

(ii) A **mild solution** to (FP_ρ) is a mild solution in $X = L^1(\rho dx)$ with

$$A_0(u) = -\Delta\beta(u) - \nabla\Phi \cdot \nabla\beta(u) - \operatorname{div}_\rho(D(x)b(u)u).$$

Idea of proof: Apply Crandall–Liggett theorem, i.e.

A m -accretive $\implies \forall u_0 \in \overline{D(A)}^X : \exists!$ mild sol. to $\frac{d}{dt}u(t) = -Au(t), u(0) = u_0$, forming a semigroup of $|\cdot|_X$ -contractions.

A is m -accretive, if

$$R(\text{Id} + \lambda A) = X, \forall \lambda > 0 \text{ and } |(\text{Id} + \lambda A)^{-1}f - (\text{Id} + \lambda A)^{-1}g|_X \leq |f - g|_X.$$

For A_0 as above, one can prove the full-range-condition in $(X, |\cdot|_X) = (L^1(\rho dx) \cap L^\infty, |\cdot|_{L^1(\rho)})$, and there exists an m -accretive restriction $A \subseteq A_0$. Then, by small relaxations of the original C.–L.-proof the existence claim follows.

Price to pay: Since range- and contraction property only in $L^1(\rho dx) \cap L^\infty$, cannot conclude uniqueness of the mild solution, only uniqueness among solutions approximated by step functions with values in $L^1(\rho dx) \cap L^\infty$.

Solving

$$v + \lambda Av = f, \quad f \in L^1(\rho dx) \cap L^\infty$$

similar to [BarbuRöckner21]: Approximate A by smooth A_ε , $\varepsilon > 0$, and use

symmetry of $Lv := \Delta v - \nabla \Phi \cdot \nabla v$ in $L^2(\rho dx)$,

i.e. $(L, L^2(\rho dx))$ replaces $(\Delta, L^2(dx))$ in [BarbuRöckner21].

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Superposition principle: For any weakly continuous distributional probability density solution $t \mapsto u_t$ to the Nemytskii-type nIFPE

$$du_t = \partial_{ij}^2(a_{ij}(t, u_t(x), x)u_t) - \partial_i(b_i(t, u_t(x), x)u_t), \quad t \geq s, \quad (1)$$

there exists a weak solution X to

$$dX_t = b(t, u_t(X_t), X_t)dt + \sigma(t, u_t(X_t), X_t)dB_t, \quad t \geq s$$

with $\mathcal{L}_{X_t} = u_t(x)dx$.

Note:

$$\partial_t u = \Delta \beta(u) - \nabla \Phi \cdot \nabla \beta(u) - \operatorname{div}_\rho(D(x)b(u)u)$$

appears not of the above type. **But:** due to symmetry of $\Delta - \nabla \Phi$ in $L^2(\rho dx)$: Distributional formulation (i.e. integrating with test fct., putting derivatives to test fct.) is of same type as (1).

Hence: Superposition principle applies: For any weakly continuous solution $t \mapsto u(t, x)$ such that $u(t, x)\rho(x)dx \in \mathcal{P}$ to

$$\partial_t u = \Delta \beta(u) - \nabla \Phi \cdot \nabla \beta(u) - \operatorname{div}_\rho(D(x)b(u)u),$$

there exists a weak solution X to

$$dX_t = \left[D(X_t)b(v(t, X_t)\rho^{-1}(X_t)) - \frac{\beta(v(t, X_t)\rho^{-1}(X_t))}{v(t, X_t)\rho^{-1}(X_t)} \nabla \Phi(X_t) \right] dt + \sqrt{2 \frac{\beta(v(t, X_t)\rho^{-1}(X_t))}{v(t, X_t)\rho^{-1}(X_t)}} dB_t,$$

$$\mathcal{L}_{X_t} = v(t, x)dx := u(t, x)\rho(x)dx.$$

\implies We can solve this SDE for initial data

$$\mathcal{L}_{X_0} = u(0, x)\rho(x)dx, \quad u(0, \cdot) \in L^1(\rho dx) \cap L^\infty.$$

This SDE is a model for *nonlinear generalized perturbed Ornstein–Uhlenbeck processes*. Indeed: For $\beta = \operatorname{Id}$, $\Phi = \frac{|x|^2}{2}$ and $D = 0$, it reduces to

$$dX_t = -X_t dt + dB_t.$$

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We want to show: The family of solution path laws \mathbb{P}_{u_0} , $u_0 \in L^1(\rho dx) \cap L^\infty$, to the above SDE with initial datum $u_0 \rho dx \in \mathcal{P}$ and one-dim. marginals

$$\mathbb{P}_{u_0} \circ \pi_t^{-1} = S(t)u_0, \quad t \geq 0,$$

is a *nonlinear Markov process*, i.e. satisfies the following definition. Set $\mathcal{F}_r := \sigma(\pi_s, 0 \leq s \leq r) \subseteq \mathcal{B}(C(\mathbb{R}_+, \mathbb{R}^d))$.

Definition

$\{\mathbb{P}_{u_0}, u_0 \in L^1(\rho dx) \cap L^\infty\}$ is a *nonlinear Markov process*, if $\forall A \in \mathcal{B}(\mathbb{R}^d)$, $0 \leq r \leq t, u_0 \in L^1(\rho dx) \cap L^\infty$

$$\mathbb{P}_{u_0}(\pi_t \in A | \mathcal{F}_r)(\cdot) = \mathbb{P}_{S(r)u_0}(\pi_{t-r} \in A | \pi_0 = \pi_r(\cdot)) \mathbb{P}_{u_0} - \text{a.s.}$$

We can apply the following result:

Theorem (R./Röckner22)

If $\mathcal{P}_0 \subseteq \mathcal{P}$ and $(S(t))_{t \geq 0}, S(t) : \mathcal{P}_0 \rightarrow \mathcal{P}_0$, is a semigroup of distributional solutions to the FPE

$$\partial_t \mu_t = \partial_{ij}^2 (a_{ij}(t, \mu_t, x) \mu_t) - \partial_i (b_i(t, \mu_t, x) \mu_t),$$

and if $t \mapsto S(t)\mu_0, \mu_0 \in \mathcal{P}_0$, is the unique solution to the *linearized* FPE

$$\partial_t \nu_t = \partial_{ij}^2 (a_{ij}(t, S(t)\mu_0, x) \nu_t) - \partial_i (b_i(t, S(t)\mu_0, x) \nu_t), \nu_0 = S(0)\mu_0,$$

then the superposition solutions $\{\mathbb{P}_{\mu_0}\}_{\mu_0 \in \mathcal{P}_0}$ to the corresponding DDSDE are a nonlinear Markov process with marginals $S(t)\mu_0, t \geq 0, \mu_0 \in \mathcal{P}_0$.

We can prove the linearized uniqueness claim, and hence obtain:

The solution path laws $\{\mathbb{P}_{u_0}\}_{u_0 \in L^1(\rho dx) \cap L^\infty}$ to the nl-gen-per-OU-equation, obtained as superposition solutions of the semigroup solution to the weighted FPE, are a nonlinear Markov process.

Thank you for your attention!