

Kolmogorov equations on spaces of measures associated to nonlinear filtering processes

Mattia Martini

Università degli Studi di Milano

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- Backward equation associated to the K.-S. equation

Kolmogorov equations on spaces of measures

We want to introduce and study a class of backward Kolmogorov equations on

- $\mathcal{M}_2^+(\mathbb{R}^d), \mathcal{P}_2(\mathbb{R}^d)$: positive and probability measures with finite second moment;
- $\langle \mu, \psi \rangle = \mu(\psi) = \int_{\mathbb{R}^d} \psi(x) \mu(\mathrm{d}x)$;

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SDEs for measure-valued processes arise naturally in the stochastic filtering framework.

- Many results when there is a density, using stochastic calculus on Hilbert spaces (e.g. Rozovsky [9], Pardoux [8]).
- New tools for calculus on spaces of (probability) measures (e.g. Ambrosio, Gigli & Savaré [1], P.-L. Lions [5], Carmona & Delarue [3]).
- Optimal control with partial observation (e.g. Gozzi & Świąch [4] in the Hilbert setting, or recently Bandini, Cosso, Fuhrman & Pham [2] on $\mathcal{P}_2(\mathbb{R}^d)$).

Signal process

$$dX_t = b(X_t) ds + \sigma(X_t) dW_t, \quad X_0 \in L^2(\Omega, \mathcal{F}_0), \quad t \in [0, T]. \quad (1)$$

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Observation process

For every $t \in [0, T]$,

$$\begin{aligned} dY_t &= h(X_t) dt + dB_t, \quad Y_0 = 0, \\ \mathcal{F}_t^Y &= \sigma(Y_s, 0 \leq s \leq t) \vee \mathcal{N}, \end{aligned} \quad (2)$$

where \mathcal{N} are \mathbb{P} -negligible sets.

Stochastic filtering The problem

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Goal

- The signal X is not directly observed;
- The available information is given by Y ;
- We want to provide an approximation of X given the observation Y .

- Given the information \mathcal{F}_t^Y , the best estimate for $\varphi(X_t)$ is

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- Let Π_t be the regular conditional probability distribution of X_t given \mathcal{F}_t^Y :
for any $A \in \mathcal{B}(\mathbb{R}^d)$

$$\Pi_t(A, \omega) = \mathbb{P} \left(X_t \in A | \mathcal{F}_t^Y \right) (\omega), \quad \text{a.e. } \omega.$$

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$\{\Pi_t = \text{Law}(X_t | \mathcal{F}_t^Y)\}_{t \in [0, T]}$ is a $\mathcal{P}(\mathbb{R}^d)$ -valued process called **filter**.

Stochastic filtering The unnormalized filter

Define \mathbb{Q} by $\frac{d\mathbb{Q}}{d\mathbb{P}}|_{\mathcal{F}_t} = M_t^{-1} = \exp \left\{ -\frac{1}{2} \int_0^t |h(X_s)|^2 ds - \int_0^t h(X_s) dB_s \right\}.$

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Theorem (Kallianpur-Striebel formula)

The filter Π can be represented as

$$\langle \Pi_t, \varphi \rangle = \frac{\langle \rho_t, \varphi \rangle}{\langle \rho_t, \mathbf{1} \rangle}, \quad t \in [0, T], \varphi \in C_b(\mathbb{R}^d), \quad (3)$$

where $\langle \rho_t, \varphi \rangle = \mathbb{E}^{\mathbb{Q}} [M_t \varphi(X_t) | \mathcal{F}_t^Y].$

$\{\rho_t\}_{t \in [0, T]}$ is a $\mathcal{M}^+(\mathbb{R}^d)$ -valued process called **unnormalized filter**.

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Y is a brownian motion under \mathbb{Q} . By Itô formula applied to $M_t \varphi(X)$ we obtain

The Zakai equation (Z)

The unnormalized filter satisfies, for every test φ ,

$$d\langle \rho_t, \varphi \rangle = \langle \rho_t, A\varphi \rangle dt + \langle \rho_t, h\varphi \rangle dY_t, \quad t \in (0, T], \quad (4)$$

where A is the infinitesimal generator of X .

Let A be the generator of X : $A\varphi = b^\top (D_x\varphi) + \frac{1}{2} \text{tr}\{(D_x^2\varphi)\sigma\sigma^\top\}$.

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Stochastic filtering Kushner-Stratonovitch equation

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Using the Kallianpur-Striebel formula

The Kushner-Stratonovich equation (KS)

The filter satisfies, for every test φ ,

$$d\langle \Pi_t, \varphi \rangle = \langle \Pi_t, A\varphi \rangle dt + (\langle \Pi_t, h\varphi \rangle - \langle \Pi_t, \varphi \rangle \langle \Pi_t, h \rangle) dI_t, \quad t \in (0, T],$$

where $\{I_t\}_{t \in [0, T]}$ is called **innovation process** and is a Brownian motion under \mathbb{P} .

Stochastic filtering Example: Kalman-Bucy filter

Signal:

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Observation:

$$dY_t = h_t X_t dt + dB_t, \quad Y_0 = 0.$$

(X, Y) is a gaussian process.

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$$\iota(x) = x.$$

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$\iota(x) = x$. Moreover, for $\omega \in \Omega$ fixed, $\Pi_t(\omega)$ is gaussian with

- Mean \hat{X}_t that solves the SDE

$$d\hat{X}_t = b_t \hat{X}_t dt + \gamma_t h_t dl_t, \quad l_t = Y_t - \int_0^t h_s \hat{X}_s ds.$$

- Deterministic variance that solves the Riccati equation

$$\frac{d}{dt} \gamma_t = \gamma_t b_t^\top + b_t \gamma_t + a_t - \gamma_t (h^\top h) \gamma_t^\top.$$

Itô formula for the Zakai equation

Let $\{\rho_t\}_{t \in [0, T]}$ be a solution to (Z), i.e. for every test φ

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Hypotheses (H)

- a. b, σ, h are Borel-measurable and bounded, b, σ are Lipschitz;
- b. The matrix $\sigma\sigma^\top(x)$ is positive definite for every $x \in \mathbb{R}^d$.

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Proposition (M. [6])

Let u be in $C_L^2(\mathcal{M}_2^+(\mathbb{R}^d))$ and let us assume (H). Then, for every $t \in [0, T]$:

$$du(\rho_t) = \langle \rho_t, A\delta_\mu u(\rho_t) \rangle dt + \langle \rho_t, h\delta_\mu u(\rho_t) \rangle dY_t + \frac{1}{2} \langle \rho_t \otimes \rho_t, h^\top h \delta_\mu^2 u(\rho_t) \rangle dt.$$

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- $\delta_\mu u$ is a notions of derivatives for $u: \mathcal{M}^+(\mathbb{R}^d) \rightarrow \mathbb{R}$:

$$\delta_\mu u: \mathcal{M}^+(\mathbb{R}^d) \times \mathbb{R}^d \rightarrow \mathbb{R}, \quad \delta_\mu^2 u: \mathcal{M}^+(\mathbb{R}^d) \times \mathbb{R}^d \times \mathbb{R}^d \rightarrow \mathbb{R};$$

- Proof by cylindrical approximation: $u(\mu) := g(\langle \mu, \psi_1 \rangle, \dots, \langle \mu, \psi_n \rangle)$.

The infinitesimal generator of the Zakai equation

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Remark

- Formally $d\rho_t = A^* \rho_t dt + h^\top \rho_t dY_t$, so:

$$du(\rho_t) = \langle A^* \rho_t, \delta_\mu u(\rho_t) \rangle dt + \langle h^\top \rho_t, \delta_\mu u(\rho_t) \rangle dY_t + \frac{1}{2} \langle h^\top \rho_t \otimes h \rho_t, \delta_\mu^2 u(\rho_t) \rangle dt.$$

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- On \mathbb{R} , if $dX_t = bX_t dt + \sigma X_t dB_t$, then

$$du(X_t) = bX_t D_x u(X_t) dt + \sigma X_t D_x u(X_t) dB_t + \frac{1}{2} \sigma^2 X_t^2 D_x^2 u(X_t) dt.$$

The backward Kolmogorov equation Existence and uniqueness

Let

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Given $\Phi: \mathcal{M}_2^+(\mathbb{R}^d) \rightarrow \mathbb{R}$, the **Backward Kolmogorov equation** (BEZ) reads as

$$\begin{cases} \partial_s u(\mu, s) + \mathcal{L}u(\mu, s) = 0, & (\mu, s) \in \mathcal{M}_2^+(\mathbb{R}^d) \times [0, T], \\ u(\mu, T) = \Phi(\mu), & \mu \in \mathcal{M}_2^+(\mathbb{R}^d). \end{cases}$$

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Let $\{\rho_t^{s,\mu}\}_{t \in [s,T]}$ be a solution to (Z) starting at time s from $\mu \in \mathcal{M}_2^+(\mathbb{R}^d)$.

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Theorem (M. [6])

Let $\Phi \in C_L^2(\mathcal{M}_2^+(\mathbb{R}^d))$. Let (H) holds and let us set

$$u(\mu, s) := \mathbb{E} [\Phi(\rho_T^{s,\mu})], \quad (\mu, s) \in \mathcal{M}_2^+(\mathbb{R}^d) \times [0, T]. \quad (6)$$

Then u is the unique classical solution to (BEZ).

Proof (key steps)

Uniqueness:

- By the Itô formula, every classical solution to (BEZ) has the form

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 - since $\Phi \in C_L^2(\mathcal{M}_2^+(\mathbb{R}^d))$ and by the previous point, we conclude by a chain rule.

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 - since $\Phi \in C_L^2(\mathcal{M}_2^+(\mathbb{R}^d))$ and by the previous point, we conclude by a chain rule.
- By Itô formula and Markov property

$$\begin{aligned} & \lim_{h \rightarrow 0} \frac{1}{h} [u(\mu, s+h) - u(\mu, s)] \\ &= - \lim_{h \rightarrow 0} \frac{1}{h} \mathbb{E} \left[\int_s^{s+h} \mathcal{L}u(\rho_\tau^{s, \mu}, s+h) d\tau \right] = -\mathcal{L}u(\mu, s). \end{aligned}$$

The Kushner-Stratonovich equation case

The operator $\mathcal{L}^{KS}: C_L^2(\mathcal{P}_2(\mathbb{R}^d)) \rightarrow C_b(\mathcal{P}_2(\mathbb{R}^d))$

$$\mathcal{L}^{KS}u(\pi) = \langle \pi, A\delta_\mu u(\pi) \rangle + \frac{1}{2} \langle \pi \otimes \pi, (h - \pi(h))^\top (h - \pi(h)) \delta_\mu^2 u(\pi) \rangle.$$

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Given $\Phi: \mathcal{P}_2(\mathbb{R}^d) \rightarrow \mathbb{R}$, **the Backward Kolmogorov equation (BEKS)** reads as

$$\begin{cases} \partial_s u(\pi, s) + \mathcal{L}^{KS}u(\pi, s) = 0, & (\pi, s) \in \mathcal{P}_2(\mathbb{R}^d) \times [0, T], \\ u(\pi, T) = \Phi(\pi), & \pi \in \mathcal{P}_2(\mathbb{R}^d). \end{cases}$$

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Let $\{\Pi_t^{s,\pi}\}_{t \in [s,T]}$ be a solution to (KS) starting at time s from $\pi \in \mathcal{P}_2(\mathbb{R}^d)$:

$$d\langle \Pi_t, \psi \rangle = \langle \Pi_t, A\psi \rangle dt + (\langle \Pi_t, h\psi \rangle - \langle \Pi_t, \psi \rangle \langle \Pi_t, h \rangle) \cdot dL_t, \quad t \in (0, T]. \quad (7)$$

Theorem (M. [6])

Let $\Phi \in C_L^2(\mathcal{P}_2(\mathbb{R}^d))$. Let (H) holds and let us set

$$u(\pi, s) = \mathbb{E} [\Phi(\Pi_T^{s,\pi})], \quad (\pi, s) \in \mathcal{P}_2(\mathbb{R}^d) \times [0, T].$$

Then u is the unique classical solution to (BEKS).

The Kushner-Stratonovich equation case Viscosity approach

$K \subset \mathbb{R}^d$ compact, $\Phi \in C_b(\mathcal{P}_2(K))$:

$$\begin{cases} \partial_s u(\pi, s) + \mathcal{L}^{KS} u(\pi, s) = 0, & (\pi, s) \in \mathcal{P}_2(K) \times (0, T], \\ u(\pi, T) = \Phi(\pi), & \pi \in \mathcal{P}_2(K). \end{cases}$$

Let $\{\Pi_t^{s,\pi}\}_{t \in [s,T]}$ be a solution to (KS) confined in $\mathcal{P}_2(K)$.

Theorem (M. [7])

Let $\Phi \in C_b(\mathcal{P}_2(K))$. Let (H) holds and let us set

$$u(\pi, s) = \mathbb{E} [\Phi(\Pi_T^{s,\pi})], \quad (\pi, s) \in \mathcal{P}_2(K) \times (0, T].$$

Then u is the unique viscosity solution to (BEKS).

Proof of the comparison principle (Key steps)

Let u_1 and u_2 be respectively a subsolution and a supersolution to (BEKS). Moreover, let $u(\pi, s) := \mathbb{E} [\Phi(\Pi_T^{s, \pi})]$. We want to show that $u_1 \leq u_2$.

- Show: $u_1 \leq u$ and $u \leq u_2$.
- Introduce a family of approximated problems:

$$\begin{cases} \partial_s u(\pi, s) + \mathcal{L}^{KS} u(\pi, s) = 0, & (\pi, s) \in \mathcal{P}_2(K) \times (0, T], \\ u(\pi, T) = \Phi_n(\pi) \in C_L^2(\mathcal{P}_2(K)), & \pi \in \mathcal{P}_2(K). \end{cases}$$

- $u^n(\pi, s) := \mathbb{E} [\Phi_n(\Pi_T^{s, \pi})]$ is a classical solution to the approximated problem which converges to u .
- Using the Borwein-Preiss variational principle with a suitable smooth gauge-type function, we introduce a suitable test function that allows us to conclude.

Thank you!

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Linear functional derivative

$u: \mathcal{M}^+(\mathbb{R}^d) \rightarrow \mathbb{R}$ is in $C_b^1(\mathcal{M}^+(\mathbb{R}^d))$ if it is continuous, bounded and if exists

$$\delta_\mu u: \mathcal{M}^+(\mathbb{R}^d) \times \mathbb{R}^d \ni (\mu, x) \mapsto \delta_\mu u(\mu, x) \in \mathbb{R},$$

bounded, continuous and such that for all μ and μ' in $\mathcal{M}^+(\mathbb{R}^d)$, it holds:

$$u(\mu') - u(\mu) = \int_0^1 \int_{\mathbb{R}^d} \delta_\mu u(t\mu' + (1-t)\mu, x) [\mu' - \mu](dx) dt. \quad (8)$$

Similarly we can define $C_b^k(\mathcal{M}^+(\mathbb{R}^d))$, $k \in \mathbb{N}$.

Example

Let $g \in C_b^2(\mathbb{R})$ and let $\psi \in C_b(\mathbb{R}^d)$. We define

$$u: \mathcal{M}^+(\mathbb{R}^d) \ni \mu \mapsto g(\langle \mu, \psi \rangle) \in \mathbb{R}.$$

Then $u \in C_b^2(\mathcal{M}^+(\mathbb{R}^d))$ and it holds:

$$\delta_\mu u(\mu, x) = g'(\langle \mu, \psi \rangle) \psi(x), \quad \delta_\mu^2 u(\mu, x, y) = g''(\langle \mu, \psi \rangle) \psi(x) \psi(y).$$

The space $C_L^2(\mathcal{M}^+(\mathbb{R}^d))$

$u: \mathcal{M}^+(\mathbb{R}^d) \rightarrow \mathbb{R}$ is in $C_L^2(\mathcal{M}^+(\mathbb{R}^d))$ if:

- u is in $C_b^2(\mathcal{M}^+(\mathbb{R}^d))$;
- $\mathbb{R}^d \ni x \mapsto \delta_\mu u(\mu, x) \in \mathbb{R}$ is twice differentiable, with continuous and bounded derivatives on $\mathcal{M}^+(\mathbb{R}^d) \times \mathbb{R}^d$;

We set

$$D_\mu u(\mu, x) := D_x \delta_\mu u(\mu, x) \in \mathbb{R}^d,$$

Remark

On $\mathcal{P}_2(\mathbb{R}^d)$, the derivative $D_\mu u$ coincides with the one introduced by P.-L. Lions through the lifting procedure in the context of mean field games ([5, 3]).

1. Prove the formula for functions of the form

$$u: \mathcal{M}_2^+(\mathbb{R}^d) \ni \mu \mapsto g(\langle \mu, \psi_1 \rangle, \dots, \langle \mu, \psi_n \rangle),$$

exploiting classical Itô formula and the the Zakai equation.

2. Prove the formula for functions of the form

$$u(\mu) = \left\langle \frac{\mu^r}{\mu(\mathbb{R}^d)^r}, \varphi(\cdot, \dots, \cdot, \mu(\mathbb{R}^d)) \right\rangle$$

by approximation, where $\varphi: \mathbb{R}^{d \times r+1} \rightarrow \mathbb{R}$ is symmetrical in the first r arguments.

3. Prove the formula for functions in $C_L^2(\mathcal{M}_2^+(\mathbb{R}^d))$ by approximation.

The backward Kolmogorov equation Existence and uniqueness

Theorem (M. [6])

Let us set

$$u(\mu, s) = \mathbb{E} [\Phi(\rho_T^{s,\mu})], \quad (9)$$

where $\rho_T^{s,\mu}$ is the weak solution to the Zakai equation starting at time s from $\mu \in \mathcal{M}_2^+(\mathbb{R}^d)$, $\Phi \in C_L^2(\mathcal{M}_2^+(\mathbb{R}^d))$ and let (H) hold. Then u is the unique classical solution to the backward Kolmogorov equation (BEZ).

Proof (uniqueness)

We show that if u is a classical solution to (BEZ), then $u(\mu, s) = \mathbb{E} [\Phi(\rho_T^{s,\mu})]$.

- By the Itô formula

$$u(\rho_T^{s,\mu}, T) - u(\rho_s^{s,\mu}, s) = \int_s^T \{ \partial_s u(\rho_\tau^{s,\mu}, \tau) + \mathcal{L}u(\rho_\tau^{s,\mu}, \tau) \} d\tau + \int_s^T \mathcal{G}u(\rho_\tau^{s,\mu}, \tau) \cdot dY_\tau.$$

- By taking the expectation and since u solves (BEZ)

$$\mathbb{E} [\Phi(\rho_T^{s,\mu})] - u(\mu, s) = \mathbb{E} \left[\int_s^T \mathcal{G}u(\rho_\tau^{s,\mu}, \tau) \cdot dY_\tau \right].$$

- The rhs is zero since the integral is a martingale, thus $u(\mu, s) = \mathbb{E} [\Phi(\rho_T^{s,\mu})]$.

The backward Kolmogorov equation Existence and uniqueness

Proof (existence)

Let $u(\mu, s) = \mathbb{E} [\Phi(\rho_T^{s, \mu})]$ be our candidate solution.

1. Prove that $\mu \mapsto u(\mu, s)$ is in $C_L^2(\mathcal{M}_2^+(\mathbb{R}^d))$:
 - given a suitable notion of derivative for functions from $C_L^2(\mathcal{M}_2^+(\mathbb{R}^d))$ to $C_L^2(\mathcal{M}_2^+(\mathbb{R}^d))$, we show that $\mu \mapsto \rho_T^{s, \mu}$ is twice differentiable;
 - since $\Phi \in C_L^2(\mathcal{M}_2^+(\mathbb{R}^d))$ and by the previous point, we conclude by a chain rule.
2. Prove the continuity of

$$[0, T] \ni s \mapsto \mathcal{L}u(\mu, s), \quad [s, T] \times [0, T] \ni (\tau, \sigma) \mapsto \mathcal{L}u(\rho_\tau^{s, \mu}, \sigma) \in L^2(\Omega).$$

3. By the Itô formula and the Markov property

$$\lim_{h \rightarrow 0} \frac{1}{h} [u(\mu, s+h) - u(\mu, s)] = - \lim_{h \rightarrow 0} \frac{1}{h} \mathbb{E} \left[\int_s^{s+h} \mathcal{L}u(\rho_\tau^{s, \mu}, s+h) d\tau \right] = -\mathcal{L}u(\mu, s).$$