

Lévy processes, controlled time rate and mean field games

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NTNU, May 2023

Part I: Classical mean field games and their heuristic derivation

“Classical” mean field games

$$\begin{cases} -\partial_t u = \Delta u + H(\nabla u) + f(m) & \text{on } [0, T] \times \mathbb{R}^d, \\ u(T) = g(m(T)) & \text{on } \mathbb{R}^d, \\ \partial_t m = \Delta m + \operatorname{div}(H'(\nabla u) m) & \text{on } [0, T] \times \mathbb{R}^d, \\ m(0) = m_0 & \text{on } \mathbb{R}^d. \end{cases}$$

- Agents control (individually, but interchangeably) **the drift** of a Wiener process describing their positions.

Controlled Wiener process

- Controlled Wiener process $Y_s^{t,x,\gamma} = x + W_s^{t,x} + \gamma(s, \cdot)(s - t)$ — at each point (s, \cdot) we choose a direction γ , i.e. $\gamma : (s, \cdot) \mapsto \mathcal{A} \subset \mathbb{R}^d$.
- Y^γ is a Markov process associated with the families of operators P^γ and transition probabilities $p^\gamma(t, x, s, A) = \mathbb{P}(Y_s^{t,x,\gamma} \in A)$,

$$P_{t,s}^\gamma \phi(x) = \int_{\mathbb{R}^d} \phi(y) p^\gamma(t, x, s, dy) = E\phi(Y_s^{t,x,\gamma}), \quad \phi \in C_b(\mathbb{R}^d).$$

- We may compute the “generator”

$$\lim_{h \rightarrow 0} \frac{P_{t+h,t}^\gamma \phi(x) - \phi(x)}{h} = \Delta u + \gamma(t, x) \cdot \nabla u.$$

Dynamic programming

- Total gain functional

$$J(t, x, \gamma) = E \left(\int_t^T \ell(s, Y_s^{t,x,\gamma}, \gamma) ds + g(Y_T^{t,x,\gamma}) \right).$$

- Value function u (the optimal value of J) is given by

$$u(t, x) = \sup_{\gamma} J(t, x, \gamma).$$

- Dynamic programming principle — assume the “tail” is already optimized

$$u(t, x) = \sup_{\gamma} E \left(\int_t^{t+h} \ell(s, Y_s^{t,x,\gamma}, \gamma) ds + u(t+h, Y_{t+h}^{t,x,\gamma}) \right).$$

- In the limit we get the Bellman equation

$$-\partial_t u = \Delta u + \sup_{\gamma(t,x) \in \mathcal{A}} \left(\gamma \cdot \nabla u + \ell(t, x, \gamma) \right), \quad (1)$$

Hamilton–Jacobi–Bellman

- We now assume that

$$\ell(t, x, \gamma) = -L(\gamma) + f(t, x), \quad (2)$$

where $L : \mathbb{R}^d \rightarrow \mathbb{R} \cup \{\infty\}$ is a convex, lower-semicontinuous function.

- Legendre–Fenchel transform H of L (γ disappears, but we will need it later)

$$H(z) = \sup_{\zeta \in \mathbb{R}^d} (\zeta z - L(\zeta)),$$

- The Bellman equation becomes

$$\begin{cases} -\partial_t u = \Delta u + H(\nabla u) + f(t, x), \\ u(T, x) = Eg(Y_T^{T,x}) = g(x). \end{cases}$$

- Backward-in-time evolution equation. Because of Δ it has unique, smooth solutions.

Fokker–Planck–Kolmogorov

- Under reasonable assumptions on L , by the properties of LF transform, we have (the optimal control) $\gamma^* = \nabla H(\nabla u)$ for every $(t, x) \in [0, \infty) \times \mathbb{R}^d$

- For initial condition $m(0) = m_0 \in \mathcal{P}(\mathbb{R}^d)$, input distribution m of Y satisfies

$$\int_{\mathbb{R}^d} \varphi(x) m(t+h, dx) = \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} \varphi(y) p^{\gamma^*}(t, x, t+h, dy) m(t, dx),$$

- This leads to

$$\partial_t \int_{\mathbb{R}^d} \varphi(t, x) m(t, dx) = \int_{\mathbb{R}^d} \left(\Delta \varphi + \nabla H(\nabla u) \nabla \varphi + \partial_t \varphi(t, x) \right) m(t, dx).$$

- By duality m is a very weak solution of

$$\partial_t m = \Delta u + \operatorname{div}(\nabla H(\nabla u) m), \quad m(0) = m_0,$$

- m describes the joint distribution of all players, each of whom moves according to their own copy of Y – this leads to the **mean field game**.

Part II: Fully nonlinear, nonlocal mean field games

Fully nonlinear (parabolic, local/nonlocal) MFG

$$\begin{cases} -\partial_t u = F(\mathcal{L}u) + f(m) & \text{on } [0, T] \times \mathbb{R}^d, \\ u(T) = g(m(T)) & \text{on } \mathbb{R}^d, \\ \partial_t m = \mathcal{L}^*(F'(\mathcal{L}u) m) & \text{on } [0, T] \times \mathbb{R}^d, \\ m(0) = m_0 & \text{on } \mathbb{R}^d. \end{cases}$$

- Agents control **the time rate** θ of any Lévy process (\mathcal{L})
- θ is a stochastic process such that $\theta(t)$ is a stopping time
- “Local-in-time generator” $\theta'(t)\mathcal{L}$ — not Lévy, but Markov (inhomog.)
- Same for any number of Lévy processes
- To get the classical model: $\Delta, dx_1, \dots, dx_d, -dx_1, \dots, -dx_d$

Definition

A stochastic process $X = \{X_t : t \geq 0\}$ with law P on $\mathbb{R}^{[0, \infty)}$ is a Lévy process if

- X has P -almost surely right-continuous paths with left-limits.
- $P(X_0 = 0) = 1$
- $X_t - X_s = X_{t-s}$ in distribution
- $X_t - X_s$ is independent of $\{X_r : r \leq s\}$

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- A Lévy process is a Wiener process if it has P -almost surely continuous paths.
 - Lévy–Khintchine–Carrère formula for the generator

$$\mathcal{L}\phi(x) = c \cdot \nabla \phi(x) + \text{tr}(aa^T D^2 \phi(x)) + \int_{\mathbb{R}^d} \left(\phi(x+z) - \phi(x) - \mathbb{1}_{B_1}(z) z \cdot \nabla \phi(x) \right) \nu(dz).$$

- Order $2\sigma \Leftrightarrow \mathcal{L} : C^{2\sigma+\alpha} \rightarrow C^\alpha$
- Non-degenerate $\Leftrightarrow \nu \asymp |z|^{-d-2\sigma} dz$
- Degenerate $\Leftrightarrow \nu \leq |z|^{-d-2\sigma} dz$ (or analogue if singular)

Controlled Lévy process

- Controlled Lévy process $Y_s^{t,x,\gamma} = x + X_{\theta(s)}^{t,x}$
- $\theta(s)$ is an **random time change**, i.e. a stochastic process which is almost surely non-negative, non-decreasing, and is a finite stopping time for each fixed s .
- We assume θ is **absolutely continuous**, i.e. there exists an \mathcal{F}_s -adapted process θ' such that $\theta(s) - \theta(0) = \int_0^s \theta'(\tau) d\tau$.
- Then (with technical assumptions on θ), Y^γ is Markov
- Operators P^γ and transition probabilities $p^\gamma(t, x, s, A) = \mathbb{P}(Y_s^{t,x,\gamma} \in A)$

$$P_{t,s}^\gamma \phi(x) = \int_{\mathbb{R}^d} \phi(y) p^\gamma(t, x, s, dy) = E\phi(Y_s^{t,x,\gamma}), \quad \phi \in C_b(\mathbb{R}^d).$$

- We may compute the “generator” using Dynkin’s formula

$$\begin{aligned} \frac{P_{t+h,t}^\gamma \phi(x) - \phi(x)}{h} &= \frac{E\phi(Y_{t+h}^{t,x,\gamma}) - \phi(x)}{h} = E\left(\frac{1}{h} \int_t^{\theta_s} \mathcal{L}\phi(X_\tau^{t,x}) d\tau\right) \\ &= E\left(\frac{1}{h} \int_t^{t+h} \mathcal{L}\phi(X_\tau^{t,x}) \theta'(\tau) d\tau\right) \rightarrow \theta'(t) \mathcal{L}\phi(x) \end{aligned}$$

Mean field game

- In the same way as before we obtain the pairs of equations

$$\begin{cases} -\partial_t u = F(\mathcal{L}u) + f(t, x), \\ u(T, x) = g(x). \end{cases} \quad \begin{cases} \partial_t m = \mathcal{L}^*(F'(\mathcal{L}u) m), \\ m(0) = m_0, \end{cases}$$

- \mathcal{L}^* is the formal adjoint of \mathcal{L}
- Since the process is one-dimensional, ∇H is replaced by F' (Legendre–Fenchel transform of L). F is **convex**.
- Since the time control has non-negative values, F is also **non-decreasing**.
- Mean field game: the cost functions f and g depend on m – individual players move according to the joint distribution m of all players.
- Each player may perceive the distribution as \hat{m} , but in the equilibrium for all of them it should overlap with m .
- We put $f = \mathfrak{f}(m)$ and $g = \mathfrak{g}(m(T))$ and we require $\mathfrak{f}, \mathfrak{g}$ to be **continuous**, **monotone** operators with values in continuous functions.

Part III: Well-posedness

MFG – uniqueness

- Take (m_1, u_1) , (m_2, u_2) and test m 's against u 's

$$\begin{aligned} & (m_1(T) - m_2(T)) [u_1(T) - u_2(T)] - (m_1(0) - m_2(0)) [u_1(0) - u_2(0)] \\ &= \int_0^T \left(m_1 [\partial_t u + F'(\mathcal{L}u_1)\mathcal{L}u] - m_2 [\partial_t u + F'(\mathcal{L}u_2)\mathcal{L}u] \right) (\tau) d\tau = \dots = 0 \end{aligned}$$

- F —convex, non-decreasing, $C^{1+\gamma}(\mathbb{R})$, f, g — monotone
- Then

$$m_1 = \mathcal{L}^*(b m_1) \text{ and } m_2 = \mathcal{L}^*(b m_2), \quad m_1(0) = m_2(0) = m_0,$$

where

$$b(t, x) = \begin{cases} \frac{F(\mathcal{L}u_1(t, x)) - F(\mathcal{L}u_2(t, x))}{\mathcal{L}u_1(t, x) - \mathcal{L}u_2(t, x)}, & \text{if } \mathcal{L}u_1(t, x) \neq \mathcal{L}u_2(t, x), \\ F'(\mathcal{L}u_1(t, x)), & \text{if } \mathcal{L}u_1(t, x) = \mathcal{L}u_2(t, x) \end{cases}$$

- We need: uniqueness of FPK, regularity of HJB.

Fokker–Planck–Kolmogorov

$$\begin{cases} \partial_t m = \mathcal{L}^*(bm) & \text{on } [0, T] \times \mathbb{R}^d, \\ m(0) = m_0 & \text{on } \mathbb{R}^d. \end{cases} \quad (\text{FPK})$$

$$b = F'(\mathcal{L}u)$$

- $b \in C([0, T] \times \mathbb{R}^d)$ and $b \geq 0$
- Natural space to look for solutions: $m \in C([0, T], \mathcal{P}(\mathbb{R}^d))$:

$$m(t)[\phi(t)] = m_0[\phi(0)] + \int_0^t m(\tau)[\partial_t \phi(\tau) + b(\tau)(\mathcal{L}\phi)(\tau)] d\tau.$$

- Existence: “easy” – set of solutions is convex, **compact** and non-empty.
- Uniqueness by Holmgren: existence of classical solutions to the dual equation

$$\partial_t w = -b \mathcal{L}w, \quad w(t) = \psi \in C_c^\infty(\mathbb{R}^d)$$

- **Non-deg**: $b \in C^\alpha$, $b \geq \kappa > 0$, Mikulevičius & Pragarauskas PotAn14
- **Deg**: $b \in C^\alpha$, $b \geq 0$, \mathcal{L} of order at most $2\sigma < \frac{7-\sqrt{33}}{4}$
- If $b_n \rightarrow b$ locally uniformly, then $\mathcal{M}_n \rightarrow \mathcal{M}$ as closed sets (“ K – lim sup”)

Hamilton–Jacobi–Bellman

$$\begin{cases} -\partial_t u = F(\mathcal{L}u) + f(t, x) & \text{on } [0, T] \times \mathbb{R}^d, \\ u(T, x) = g(x) & \text{on } \mathbb{R}^d. \end{cases} \quad (\text{HJB})$$

$f = \mathfrak{f}(m), \quad g = \mathfrak{g}(m(T))$

- Fully nonlinear equation \rightarrow viscosity solutions.
- Comparison principle (VS uniquely exist): Chasseigne & Jakobsen JDE17
- But we need classical solutions and **a bit more**
- **Deg**: for $2\sigma < 1$ the comparison principle is enough; no regularization

$$f, g \in C^{2\sigma+\alpha} \quad \Rightarrow \quad \partial_t u, \mathcal{L}u \in C^\alpha$$

- **Non-deg, local**: Schauder–Caccioppoli estimates (interior regularity)

$$f \in C^{\alpha/2, \alpha}(\mathbb{R}^d) \quad \Rightarrow \quad \partial_t u, D^2 u \in C^\alpha(B_1) \quad (\text{Wang CPAM92})$$

- **Non-deg, non-local**: **Conjecture**: Schauder estimates as above (works under additional assumptions).
- (we end up assuming $f \in C^{1, \alpha}$ to get global boundedness, but this is bad)

MFG – existence

- We use Kakutani–Glicksberg–Fan fixed point theorem (i.e. Schauder, but for set-valued maps; solutions to FPK are **compact, convex, non-empty sets**)
- Take $\mu \in C([0, T], \mathcal{P}(\mathbb{R}^d))$, solve HJB: $\mathcal{K}_1(\mu) = u$.
- Take u and solve FPK: $\mathcal{K}_2(u) = m$
- Look for a fixed point of $\mathcal{K}(\mu) = \mathcal{K}_2(\mathcal{K}_1(\mu))$.
- Compactness of the map is easy (Prohorov theorem)
- For (semi-)continuity:

$$\begin{array}{ccccccccc}
 \mu_n & \xrightarrow{\mathcal{K}_1} & \mathfrak{f}(\mu_n), \mathfrak{g}(\mu_n) & \longmapsto & \mathcal{L}u_n & \longmapsto & b_n = F'(\mathcal{L}u_n) & \xrightarrow{\mathcal{K}_2} & \mathcal{M}_n \\
 \downarrow \text{weak} & & \downarrow \text{uniform} & & \downarrow \text{loc unif} & & \downarrow \text{loc unif} & & \downarrow K\text{-lim sup} \\
 \mu & \xrightarrow{\mathcal{K}_1} & \mathfrak{f}(\mu), \mathfrak{g}(\mu) & \longmapsto & \mathcal{L}u & \longmapsto & b = F'(\mathcal{L}u) & \xrightarrow{\mathcal{K}_2} & \mathcal{M}
 \end{array}$$

Thank you!

