

Explicit solutions of some Linear-Quadratic Mean Field Games

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Mean Field Games and Related Topics
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- L-Q games with N players and ergodic cost
- Symmetric and almost identical players
- The limit as $N \rightarrow +\infty$
- Other limiting cases
 - ▶ Vanishing viscosity
 - ▶ Cheap control
 - ▶ Discounted cost and small discount
- Models of population distribution

References

We look for L-Q analogues of the results for stationary periodic problems in

- J.-M. Lasry, P.-L. Lions: Jeux à champ moyen. I. Le cas stationnaire. C. R. A. S. 2006
- J.-M. Lasry, P.-L. Lions: Jpn. J. Math. 2007

Related discounted L-Q problems were studied by

- M. Huang, P.E. Caines, R.P. Malhamé: Proc. IEEE Conf. 2003.
- M. Huang, P.E. Caines, R.P. Malhamé: IEEE Trans. Automat. Control 2007

Related models of population distribution are in

- O. Guéant: J. Math. Pures Appl. 2009
- O. Guéant, J.-M. Lasry, P.-L. Lions: Mean field games and applications , to appear.

L-Q games with N players and ergodic payoff

Control systems

$$dX_t^i = (A^i X_t^i - \alpha_t^i) dt + \sigma^i dt W_t^i, \quad X_0^i = x^i \in \mathbf{R}, \quad i = 1, \dots, N$$

W_t^i independent Brownian motions, $\alpha_t^i =$ control of i -th player,
long-time-average cost functional:

$$J^i(X, \alpha^1, \dots, \alpha^N) := \lim_{T \rightarrow +\infty} \frac{1}{T} E \left[\int_0^T \frac{R_i}{2} (\alpha_t^i)^2 + F^i(X_t^1, \dots, X_t^N) dt \right],$$

with $R_i > 0$, quadratic running cost, for some reference position \bar{X}_i

$$\begin{aligned} F^i(x^1, \dots, x^N) &:= (x - \bar{X}_i)^T Q^i (x - \bar{X}_i) \\ &= \sum_{j,k=1}^N q_{jk}^i (x^j - \bar{x}_i^j)(x^k - \bar{x}_i^k), \quad q_{ii}^i > 0, \end{aligned}$$

An admissible strategy $\bar{\alpha}$ is a **Nash equilibrium** if

$$J^i(X, \bar{\alpha}) = \min_{\alpha^i} J^i(X, \bar{\alpha}^1, \dots, \bar{\alpha}^{i-1}, \alpha^i, \bar{\alpha}^{i+1}, \dots, \bar{\alpha}^N) \quad \forall i = 1, \dots, N.$$

Admissible strategies are adapted controls such that $E[X_t^i], E[(X_t^i)^2] \leq C, \forall t > 0$ and \exists prob. measure m_{α^i} :

$$\lim_{T \rightarrow +\infty} \frac{1}{T} E \left[\int_0^T g(X_t^i) dt \right] = \int_{\mathbf{R}} g(x) dm_{\alpha^i}(x).$$

for any **polynomial** g , $\deg(g) \leq 2$, locally uniformly in $x^i = X_0^i$.

Example: affine **feedbacks**

$$\alpha^i(x) = K^i x + c_i, \quad K^i > A^i,$$

whose corresponding diffusion process is **ergodic**

$$dX_t^i = [(A^i - K^i)X_t^i - c_i]dt + \sigma^i dW_t^i.$$

The HJB-K PDEs

As in Lasry-Lions we set

$$H^i(x, p) = \frac{p^2}{2R_i} - A^i x p, \quad \nu^i := \frac{(\sigma^i)^2}{2},$$

$$f^i(x; m_1, \dots, m_N) := \int_{\mathbf{R}^{N-1}} F^i(x^1, \dots, x^{i-1}, x, x^{i+1}, \dots, x^N) \prod_{j \neq i} dm_j(x^j),$$

$$\begin{cases} -\nu^i v_{xx}^i + \frac{(\nu_x^i)^2}{2R_i} - A^i x v_x^i + \lambda_i = f^i(x; m_1, \dots, m_N), & i = 1, \dots, N \\ -\nu^i (m_i)_{xx} - \left(\frac{\nu_x^i}{R_i} m_i - A^i x m_i \right)_x = 0, & i = 1, \dots, N \\ \int_{\mathbf{R}} m_i(x) dx = 1, \quad m_i > 0 \text{ in } \mathbf{R}, \end{cases}$$

Cannot normalize ν^i by $\int_{\mathbf{R}} \nu^i(x) dx = 0$ as in the periodic case!

Quadratic-Gaussian solutions

Look for solutions with each v^i **quadratic**

$$v^i(x) = \frac{(x - \mu_i)^2}{2s_i} + \frac{R_i A^i x^2}{2},$$

and measures m_i **Gaussian**

$$m_i(x) = c_i \exp\left(-\frac{v^i(x)}{\nu^i R_i} + \frac{R_i A^i x^2}{2}\right) = \frac{1}{\sqrt{2\pi s_i \nu^i R_i}} \exp\left(-\frac{(x - \mu_i)^2}{2s_i \nu^i R_i}\right)$$

Now the unknowns are the $3N$ constants μ_i, s_i, λ_i .

A useful auxiliary matrix B is

$$B_{ii} := 2q_{ii}^i + R_i(A^i)^2, \quad B_{ij} := 2q_{ij}^i \quad i \neq j.$$

Theorem: Q-G solution of the $2N$ HJB-K system

If $\det B \neq 0$ then

i) there exists unique μ_i, s_i, λ_i such that the Q-G solution v^i, m_i solves the $2N$ HJB-K system and

$$s_i = \left(2q_{ij}^i R_i + (R_i A^i)^2 \right)^{-1/2},$$

$$\mu = B^{-1} p, \quad p_i := 2q_{ij}^i \bar{x}_i^i + 2 \sum_{j \neq i} q_{ij}^i \bar{x}_i^j,$$

ii) the **affine feedback**

$$\bar{\alpha}^i(x) = \frac{x - \mu_i}{s_i R_i} + A^i x, \quad i = 1, \dots, N,$$

is a **Nash equilibrium** point for all initial positions $X \in \mathbf{R}^N$ among the admissible strategies and $J^i(X, \bar{\alpha}) = \lambda_i$ for all X and i .

Proof

i) Plug the candidate solutions in the equation,
find 2nd degree polynomial and equate the coefficients,
solve $3N$ algebraic equations for the $3N$ unknown parameters.

ii) It's a verification theorem, the idea of proof is classical,
here must be careful with the unbounded terms...

Remark:
the Nash equilibrium feedback does NOT depend on the noise intensities σ^i .

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A symmetry condition on the running cost

F^i symmetric with respect to the position of any two other players:

$$F^i(x^1, \dots, x^j, \dots, x^k, \dots, x^N) = F^i(x^1, \dots, x^k, \dots, x^j, \dots, x^N) \quad \forall j, k \neq i.$$

This is equivalent, $\forall j, k, l, m \neq i, l \neq j, k \neq l, k \neq m$,

$$q_{ij}^i = q_{ik}^i =: \frac{\beta_i}{2} = \text{primary cost of cross-displacement}$$

$$\bar{x}_i^j = \bar{x}_i^k =: r_i = \text{reference position of the other players}$$

$$q_{jj}^i = q_{kk}^i =: \eta_i = \text{secondary cost of self-displacement}$$

$$q_{jl}^i = q_{kl}^i = q_{km}^i =: \gamma_i = \text{secondary cost of cross-displacement.}$$

We also set

$$q_i := q_{ii}^i = \text{primary cost of self-displacement}$$

$$h_i := \bar{x}_i^i = \text{preferred own position (happy state)}$$

Now each F^i involves only 6 parameters and can be written as

$$(V_i) \quad F^i(x^1, \dots, x^N) = V_i \left[\frac{1}{N-1} \sum_{j \neq i} \delta_{x_j} \right] (x^i)$$

for $V_i : \{\text{prob. measures}\} \rightarrow \{\text{quadratic polynomials}\}$

$$\begin{aligned} V_i[m](x) &:= q_i(x - h_i)^2 \\ &+ \beta_i(x - h_i)(N - 1) \int_{\mathbf{R}} (y - r_i) dm(y) \\ &+ \gamma_i \left((N - 1) \int_{\mathbf{R}} (y - r_i) dm(y) \right)^2 \\ &+ (\eta_i - \gamma_i)(N - 1) \int_{\mathbf{R}} (y - r_i)^2 dm(y). \end{aligned}$$

Viceversa, F^i of the form (V_i) implies F^i symmetric.

Almost Identical players

We say the players are A. I. if F^i are symmetric and the players have the same

- control system, i.e., $A^i = A$ and $\sigma^i = \sigma$ (so $\nu^i = \nu > 0$) for all i ,
- cost of the control, i.e., $R_i = R > 0$ for all i ,
- reference positions, i.e., $h_i = h$ and $r_i = r$ for all i ,
- primary costs of displacement, i.e., $q_i = q > 0$ and $\beta_i = \beta$ for all i .

Secondary costs of displacement γ_i, η_i may still depend on i .

Theorem: identically distributed Q-G solutions

Almost Identical players and $2q + RA^2 \neq \beta(1 - N) \implies$
the unique μ , $s > 0$, λ_i , $i = 1, \dots, N$ such that

$$v^i(x) = v(x) := \frac{(x - \mu)^2}{2s} + \frac{RAx^2}{2}$$

$$m_i(x) = m(x) := \frac{1}{\sqrt{2\pi s\nu R}} \exp\left(-\frac{(x - \mu)^2}{2s\nu R}\right)$$

solve the $2N$ HJB-K equations are

$$s = \left(2qR + R^2A^2\right)^{-1/2}, \quad \mu = \frac{2qh + r\beta(N - 1)}{2q + \beta(N - 1) + RA^2},$$

$$\lambda_i = \frac{\nu}{s} + \nu RA - \frac{\mu^2}{2Rs^2} + qh^2 - h\beta(N - 1)(\mu - r) \\ + \gamma_i(N - 1)(N - 2)(\mu - r)^2 + \eta_i(N - 1)(s\nu R + (\mu - r)^2).$$

The limit as $N \rightarrow +\infty$

Assume the parameters A, ν, R, h, r independent of N
and the coefficients of $F^{i,N}$ scale, as $N \rightarrow +\infty$

$$q^N \rightarrow \bar{q}, \quad \beta^N \sim \frac{\bar{\beta}}{N}, \quad \eta_i^N \sim \frac{\bar{\eta}}{N}, \quad \gamma_i^N \sim \frac{\bar{\gamma}}{N^2}$$

Then for any prob. measure m on \mathbf{R} , $\forall i$,

$$V_i^N[m](x) \rightarrow \bar{V}[m](x) \quad \text{locally uniformly in } x,$$

$$\begin{aligned} \bar{V}[m](x) := & \bar{q}(x-h)^2 + \bar{\beta}(x-h) \int_{\mathbf{R}} (y-r) dm(y) \\ & + \bar{\gamma} \left(\int_{\mathbf{R}} (y-r) dm(y) \right)^2 + \bar{\eta} \int_{\mathbf{R}} (y-r)^2 dm(y). \end{aligned}$$

The Mean Field Equations

$$(MFE) \quad \left\{ \begin{array}{l} -\nu v_{xx} + \frac{(v_x)^2}{2R} - Axv_x + \lambda = \bar{V}[m](x) \quad \text{in } \mathbf{R}, \\ -\nu m_{xx} - \left(\frac{v_x}{R}m - Axm\right)_x = 0 \quad \text{in } \mathbf{R}, \\ \min \left[v(x) - \frac{RAx^2}{2} \right] = 0, \quad \int_{\mathbf{R}} m(x) dx = 1, \quad m > 0 \text{ in } \mathbf{R}. \end{array} \right.$$

Remark: the normalization of v is different from $\int v dx = 0$ of the periodic case.

Uniqueness

\bar{V} is monotone increasing in $L^2 \iff \bar{\beta} \geq 0$

so in this case there is at most one solution of MFEs.

Meaning: $\bar{\beta} > 0$ if imitation is costly, $\bar{\beta} < 0$ if imitation is rewarding.

Q-G solution of the MFEs

$2\bar{q} + RA^2 \neq -\bar{\beta} \implies$ the MFEs has exactly one solution
 v, m, λ of the Q-G form

$$v(x) := \frac{(x - \bar{\mu})^2}{2\bar{s}} + \frac{RAx^2}{2}, \quad m(x) := \frac{1}{\sqrt{2\pi\bar{s}\nu R}} \exp\left(-\frac{(x - \bar{\mu})^2}{2\bar{s}\nu R}\right),$$

given by

$$\bar{s} = \left(2\bar{q}R + R^2A^2\right)^{-1/2}, \quad \bar{\mu} = \frac{2\bar{q}h + r\bar{\beta}}{\bar{\beta} + 2\bar{q} + RA^2},$$

$$\lambda = \frac{\nu}{\bar{s}} + \nu RA - \frac{\bar{\mu}^2}{2R\bar{s}^2} + \bar{q}h^2 - h\bar{\beta}(\bar{\mu} - r) + (\bar{\gamma} + \bar{\eta})(\bar{\mu} - r)^2 + \bar{\eta}\bar{s}\nu R$$

Convergence Theorem

Call v^N, m^N, λ_i^N the identically distributed Q-G solutions of the $2N$ HJB-K equations for N players. Then $2\bar{q} + RA^2 \neq -\bar{\beta} \implies$
as $N \rightarrow +\infty,$

- $v^N \rightarrow v$ in $C_{loc}^2(\mathbf{R})$
- $m^N \rightarrow m$ in $C^k(\mathbf{R})$ for all $k,$
- $\lambda_i^N \rightarrow \lambda$ for all $i,$

where v, m, λ is the Q-G solution of the MFEs.

Remark: in the critical case $2\bar{q} + RA^2 = -\bar{\beta}$

- no Q-G solution if $2\bar{q}h \neq -\bar{\beta}r,$
- a continuum of Q-G solutions if $2\bar{q}h = -\bar{\beta}r.$

A similar phenomenon of infinitely many Q-G solutions occurs in the MF log-model of population distribution studied by Guéant.

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The vanishing viscosity limit for N players

In the system of $2N$ HJB-K equations let $\nu^j = (\sigma^j)^2/2 \rightarrow 0$ for the j -th player. From the explicit formulas we see

$$\nu^j, \bar{\alpha}^j, \mu_j, s_j \text{ independent of } \nu^j, \quad m_j \rightarrow \delta_{\mu_j}$$

and the j -th Kolmogorov equation $-\left(\frac{\nu_x^j}{R_j} m_j - A^j x m_j\right)_x = 0$ is satisfied in the sense of distributions.

If **all** $\nu^i = (\sigma^i)^2/2 \rightarrow 0$, $i = 1, \dots, N$, the limit $2N$ HJB-K equations are of first order and the **Quadratic-Dirac** solutions give the value of a **deterministic** game for the **same** feedback Nash equilibrium

$$\bar{\alpha}^i(x) = \frac{x - \mu_i}{s_i R_i} + A^i x.$$

The vanishing viscosity limit in the MFEs

Letting $\nu \rightarrow 0$ in the Mean Field equations we have $v, \bar{s}, \bar{\mu}$ independent of ν , $m \rightarrow \delta_{\bar{\mu}}$ and get a **Quadratic-Dirac** solution of the **first order MFEs**

$$\left\{ \begin{array}{l} \frac{(v_x)^2}{2R} - Axv_x + \lambda = \bar{V}[m](x) \\ - \left(\frac{v_x}{R} m - Axm \right)_x = 0 \\ \min \left[v(x) - \frac{RAx^2}{2} \right] = 0, \quad \int_{\mathbf{R}} m(x) dx = 1, \quad m > 0. \end{array} \right.$$

Rmk: the **limits** $\nu \rightarrow 0$ and $N \rightarrow \infty$ **commute**.

Other singular limits

- **Cheap control** : can take the limit as $R_i \rightarrow 0$ of solutions to $2N$ HJB-K equations for N players. Again
 - ▶ $m_i \rightarrow$ Dirac mass,
 - ▶ limits $R \rightarrow 0$ and $N \rightarrow \infty$ **commute**.
- **Large cost of cross-displacement** : if

$$\lim_{N \rightarrow \infty} N|\beta^N| = +\infty$$

instead of $\bar{\beta}$, the term

$$\beta^N(N-1)(x-h) \int_{\mathbf{R}} (y-r) dm(y)$$

is singular but the solutions have a Q-G limit with $\bar{\mu} = r$.

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Discounted cost functional and vanishing discount

For $\rho > 0$, infinite horizon and discounted running cost:

$$J^i = E \left[\int_0^{+\infty} \left(\frac{R_i}{2} (\alpha_t^i)^2 + F^i(X_t^1, \dots, X_t^N) \right) e^{-\rho t} dt \right].$$

For quadratic F^i find again Q-G solutions of $2N$ HJB-K equations where λ_j is replaced by ρv^j . Can show that

- as $\rho \rightarrow 0$ $\rho v_\rho^i \rightarrow \lambda_i$, $m_i^\rho \rightarrow m_i$,

$v_\rho^i(x) - v_\rho^i(0) + (\mu_i^\rho)^2 / 2s_i^\rho \rightarrow v^i(x)$ and the limit solves the $2N$ HJB-K equations of the ergodic games,

- as $N \rightarrow +\infty$ and $\rho > 0$ fixed the Q-G solutions converge to the MFEs with 1st equation

$$-\nu v_{xx} + (v_x)^2 / 2R - Axv_x + \rho v = \bar{V}[m](x)$$

- limits $\rho \rightarrow 0$ and $N \rightarrow \infty$ commute.

Models of population distribution

Guéant studied a model with discounted cost functional

$$E \left[\int_0^{+\infty} \left(\frac{|\alpha_t^i|^2}{2} + \bar{q} |X_t^i - h|^2 - \log m(X_t^i) \right) e^{-\rho t} dt \right], \quad \bar{q} \geq 0,$$

and therefore the first MF equation is ($A = 0, R = 1$)

$$-\nu v_{xx} + (v_x)^2 / 2 + \rho v = \bar{q} |x - h|^2 - \log m(x)$$

- V is NOT integral operator, not deriving from limit of N players;
- V decreasing in m models a population whose agents wish to resemble their peers, imitation is rewarding;
- MFEs have one Q-G solution of mean $h \forall \bar{q} > 0$,
- a continuum of Q-G solutions with any mean μ if $\bar{q} = 0$.

L-Q model with $A = 0, R = 1, r = h$ has first MF equation

$$\begin{aligned}
 -\nu v_{xx} + (v_x)^2 / 2 + \lambda = & \bar{q}(x - h)^2 + \bar{\beta}(x - h) \int_{\mathbf{R}} (y - r) dm(y) \\
 & + \bar{\gamma} \left(\int_{\mathbf{R}} (y - r) dm(y) \right)^2 + \bar{\eta} \int_{\mathbf{R}} (y - r)^2 dm(y)
 \end{aligned}$$

- if $\beta > 0$ imitation is costly, uniqueness of solution,
- if $\beta < 0$ imitation is rewarding as in the LOG model,
- MFEs have one Q-G solution of mean $h \quad \forall \bar{q} > 0$ and $\bar{q} \neq -\beta/2$,
- a continuum of Q-G solutions with any mean μ if $\bar{q} = -\beta/2$:
same bifurcation phenomenon at different critical value,
- $Var(m) \rightarrow +\infty$ as $\bar{q} \rightarrow 0$, m tends to uniform distribution.

Developments

- d -dimensional control system for each player, $d \geq 1$: need to solve matrix Riccati equations; the MFEs are PDEs instead of ODEs.
- Interacting populations with L-Q costs, to be compared with the log-model of Guéant.

Thanks for your attention !

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