

# Price of Anarchy for Mean Field Games

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July 6, 2018

## Abstract

The price of anarchy, originally introduced to quantify the inefficiency of selfish behavior in routing games, is extended to mean field games. The price of anarchy is defined as the ratio of a worst case social cost computed for a mean field game equilibrium to the optimal social cost as computed by a central planner. We illustrate properties of such a price of anarchy on linear quadratic extended mean field games, for which explicit computations are possible. Various asymptotic behaviors of the price of anarchy are proved for limiting behaviors of the coefficients in the model and numerics are presented.

## 1 Introduction

The concept of the ‘price of anarchy’ was introduced to quantify the inefficiency of selfish behavior in finite player games [8][9][12][17][18][19]. In this report, we extend the notion of price of anarchy to mean field games (MFG). Mean field games were introduced by Lasry and Lions [13] and Caines and his collaborators [11] to describe the limiting regime of large symmetric games when the number of players,  $N$ , tends to infinity. A mean field game equilibrium characterizes the analogue of a Nash equilibrium in the  $N = \infty$  regime. Thus, as in the finite player case, it is possible that the mean field game equilibrium is inefficient. In fact, in the paper of Balandat and Tomlin [2], they present a numerical example that shows that mean field game equilibria are not efficient, in general. The suboptimality of a mean field game equilibrium is also illustrated numerically for a congestion model in a paper of Achdou and Laurière [1]. More recently Cardaliaguet and Rainer gave in [4] a partial differential equation based thorough analysis of the *(in)efficiency* of the mean field game equilibria.

In this report, the goal is to define the price of anarchy in the context of mean field games, and to compute it for a class of linear quadratic mean field game models, which can be solved explicitly. In fact, we consider an even more general class of games by allowing for interaction between the players through their controls, in addition to interaction through their states. This is often referred in the literature as extended mean field game, or mean field game of control. We

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\*Operations Research and Financial Engineering, Princeton University, Partially supported by NSF #DMS-1716673 and ARO #W911NF-17-1-0578

<sup>†</sup>Program in Applied and Computational Mathematics, Princeton University, Partially supported by NSF #DMS-1515753 and NSF GRFP

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compare the social cost of a mean field game equilibrium to the cost incurred when the players execute a strategy computed centrally.

We consider a system of  $N$  players whose private states are denoted at time  $t$  by  $X_t^1, X_t^2, \dots, X_t^N$ . To keep the presentations simple, we assume the state space is  $\mathbb{R}$ . We denote by  $\mu_t^N$  the empirical distribution of the states, namely

$$\mu_t^N = \frac{1}{N} \sum_{i=1}^N \delta_{X_t^i}.$$

We assume that these states evolve in continuous time under the influences of controls  $\alpha_t^1, \alpha_t^2, \dots, \alpha_t^N \in \mathbb{A}$ , where the set of admissible controls,  $\mathbb{A}$ , will be defined later. Let  $\nu_t^N$  denote the empirical measure of the controls:

$$\nu_t^N = \frac{1}{N} \sum_{i=1}^N \delta_{\alpha_t^i}.$$

We also assume that if and when interactions between these states and controls are present, they are of a mean field type, i.e. through  $\mu_t^N$  and  $\nu_t^N$ . The time evolution of the state for player  $i$  is given by the Itô dynamics:

$$dX_t^i = b(t, X_t^i, \mu_t^N, \alpha_t^i, \nu_t^N)dt + \sigma dW_t.$$

We work over the interval  $[0, T]$  limited by a finite time horizon  $T \in \mathbb{R}^+$ . We assume the drift function  $b : [0, T] \times \mathbb{R} \times \mathcal{P}(\mathbb{R}) \times \mathbb{A} \times \mathcal{P}(\mathbb{A}) \ni (t, x, \mu, \alpha, \nu) \rightarrow \mathbb{R}$  is Lipschitz in each of its inputs. For the sake of simplicity, we assume that the volatility,  $\sigma$ , is a positive constant.

## Cost Functionals

We assume that we are given two functions  $f : [0, T] \times \mathbb{R} \times \mathcal{P}(\mathbb{R}) \times \mathbb{A} \times \mathcal{P}(\mathbb{A}) \ni (t, x, \mu, \alpha, \nu) \rightarrow \mathbb{R}$  and  $g : \mathbb{R} \times \mathcal{P}(\mathbb{R}) \ni (x, \mu) \rightarrow \mathbb{R}$  which we call running and terminal cost functions, respectively. We assume  $f$  and  $g$  are Lipschitz in each of their arguments. The goal of player  $i$  is to minimize its expected cost as given by:

$$J^i(\alpha^1, \dots, \alpha^N) = \mathbb{E} \left[ \int_0^T f(t, X_t^i, \mu_t^N, \alpha_t^i, \nu_t^N) dt + g(X_T^i, \mu_T^N) \right].$$

## Social Cost

We restrict ourselves to Markovian control strategies  $\alpha = (\alpha_t)_{0 \leq t \leq T}$  given by feedback functions in the form  $\alpha_t = \phi(t, X_t)$  and we let  $\mathbb{A}$  denote the set of such controls. If the  $N$  players use distributed Markovian control strategies of the form  $\alpha_t^i = \phi(t, X_t^i)$ , we define the cost (per player) to the system as the quantity  $J_\phi^{(N)}$ :

$$J_\phi^{(N)} = \frac{1}{N} \sum_{i=1}^N J^i(\alpha^1, \dots, \alpha^N).$$

We shall compute this social cost in the limit  $N \rightarrow \infty$  when all the players use the distributed control strategies given by the same feedback function  $\phi$  identified by solving an optimization

problem in the limit  $N \rightarrow \infty$ . We take the social cost to be the limit as  $N \rightarrow \infty$  of  $J_\phi^{(N)}$ , namely

$$\begin{aligned} \lim_{N \rightarrow \infty} J_\phi^{(N)} &= \lim_{N \rightarrow \infty} \frac{1}{N} \sum_{i=1}^N J^i(\alpha^1, \dots, \alpha^N) \\ &= \lim_{N \rightarrow \infty} \frac{1}{N} \sum_{i=1}^N \mathbb{E} \left[ \int_0^T f(t, X_t^i, \mu_t^N, \phi(t, X_t^i), \nu_t^N) dt + g(X_T^i, \mu_T^N) \right], \\ &= \lim_{N \rightarrow \infty} \mathbb{E} \left[ \int_0^T \langle f(t, \cdot, \mu_{X_t}^N, \phi(t, \cdot), \nu_t^N), \mu_t^N \rangle dt + \langle g(\cdot, \mu_{X_T}^N), \mu_T^N \rangle \right], \end{aligned}$$

if we use the notation  $\langle \varphi, \rho \rangle$  for the integral  $\int \varphi(z) \rho(dz)$  of the function  $\varphi$  with respect to the measure  $\rho$ . Now if we assume that in the limit  $N \rightarrow \infty$  the empirical distributions  $\mu_t^N$  converge toward a measure  $\mu_t$ , and thus  $\nu_t^N = \frac{1}{N} \sum_{i=1}^N \delta_{\phi(t, X_t^i)}$  also converges toward a measure  $\nu_t$ , then the social cost of the feedback function  $\phi$  becomes:

$$SC(\phi) = \int_0^T \langle f(t, \cdot, \mu_t, \phi(t, \cdot), \nu_t), \mu_t \rangle dt + \langle g(\cdot, \mu_T), \mu_T \rangle$$

with the expectation,  $\mathbb{E}$ , disappearing when the limiting flows  $\mu = (\mu_t)_{0 \leq t \leq T}$  and  $\nu = (\nu_t)_{0 \leq t \leq T}$  are deterministic.

We would like to evaluate  $SC(\phi)$  in the  $N = \infty$  regime directly, without having to construct the deterministic measure flows  $\mu$  and  $\nu$  as limits of the finite player empirical measures. To do this, we assume that propagation of chaos holds and that the states of the  $N$  players become asymptotically independent in the limit as  $N \rightarrow \infty$ . We consider a representative agent whose state is given by  $X^\phi = (X_t^\phi)_{0 \leq t \leq T}$ , the continuous time solution of the stochastic differential equation of McKean-Vlasov type:

$$dX_t^\phi = b(t, X_t^\phi, \mathcal{L}(X_t^\phi), \phi(t, X_t^\phi), \mathcal{L}(\phi(t, X_t^\phi)))dt + \sigma dW_t \quad (1)$$

controlled by  $\phi$ . Then we can identify  $\mu$  as the law of a representative agent using the feedback function  $\phi$ , i.e.  $\mu_t = \mathcal{L}(X_t^\phi)$ , and similarly, we can identify  $\nu$  as the law of the control, such that  $\nu_t = \mathcal{L}(\phi(t, X_t^\phi))$ . Thus, in the  $N = \infty$  regime, we rewrite the social cost as

$$SC(\phi) = \int_0^T \langle f(t, \cdot, \mathcal{L}(X_t^\phi), \phi(t, \cdot), \mathcal{L}(\phi(t, X_t^\phi))), \mathcal{L}(X_t^\phi) \rangle dt + \langle g(\cdot, \mathcal{L}(X_T^\phi)), \mathcal{L}(X_T^\phi) \rangle$$

where  $X^\phi$  satisfies equation (1). For the remainder of the paper, we work in the  $N = \infty$  regime. As mentioned earlier,  $\phi$  should be identified by solving an optimal control problem. We consider two distinct problems:

- $\phi$  is a feedback function providing a mean field game equilibrium. We detail more precisely what is meant by  $\phi$  providing a mean field game equilibrium in Section 1.1.
- $\phi$  is the feedback function minimizing the social cost  $SC(\phi)$ , without having to be a mean field game equilibrium, in which case we use the notation  $SC^{MKV}$  for  $SC(\phi)$ . This is a control problem of McKean-Vlasov type, which is detailed more precisely in Section 1.2.

The two problems are detailed more precisely in Sections 1.1 and 1.2. In Section 1.3, we define the price of anarchy based on these two problem formulations. The class of linear quadratic models is explored in Section 2, where we provide some theoretical results on the price of anarchy for this class of games, show numerical results, and detail a particular example of flocking. We conclude in Section 3.

## 1.1 Nash Equilibrium: Mean Field Game Formulation

The goal of this subsection is to articulate what is meant by a feedback function providing a mean field game equilibrium. To begin, we define what we call the *mean field environment*. By symmetry of the players, we suppose all of the players in the mean field game use the same feedback function,  $\phi$ . Then the *mean field environment* specified by  $\phi$  is characterized by  $\mathcal{L}(X_t^\phi)_{0 \leq t \leq T}$  and  $\mathcal{L}(\phi(t, X_t^\phi))_{0 \leq t \leq T}$  where the dynamics of  $(X_t^\phi)_{0 \leq t \leq T}$  are given by equation (1). Since we search for a Nash equilibrium, we consider a representative agent who wishes to find their best response,  $\phi'$ , to the mean field environment specified by  $\phi$ , in which case their state is given by  $\mathbf{X}^{\phi', \phi} = (X_t^{\phi', \phi})_{0 \leq t \leq T}$  solving the standard stochastic differential equation:

$$dX_t^{\phi', \phi} = b(t, X_t^{\phi', \phi}, \mathcal{L}(X_t^\phi), \phi'(t, X_t^{\phi', \phi}), \mathcal{L}(\phi(t, X_t^\phi)))dt + \sigma dW_t.$$

Consider the function:

$$\mathcal{S}(\phi', \phi) = \left[ \int_0^T \langle f(t, \cdot, \mathcal{L}(X_t^\phi), \phi'(t, \cdot), \mathcal{L}(\phi(t, X_t^\phi))), \mathcal{L}(X_t^{\phi', \phi}) \rangle dt + \langle g(\cdot, \mathcal{L}(X_T^\phi)), \mathcal{L}(X_T^{\phi', \phi}) \rangle \right].$$

The best response for the representative agent in the mean field environment specified by  $\phi$  is the feedback function minimizing this cost, namely  $\phi^* = \arg \inf_{\phi'} \mathcal{S}(\phi', \phi)$ . Assuming the minimizer is unique (which will be the case for the models we consider), this defines a mapping  $\Phi : \phi \rightarrow \phi^*$ . If there is a  $\hat{\phi}$  such that  $\Phi(\hat{\phi}) = \hat{\phi}$ , then the players are in a mean field game equilibrium.

Thus, the search for a feedback function providing a mean field game equilibrium can be summarized as the following set of two successive steps:

1. For each feedback function  $\phi : [0, T] \times \mathbb{R} \ni (t, x) \rightarrow \mathbb{R}$ , solve the optimal control problem

$$\phi^* = \arg \inf_{\phi'} \mathcal{S}(\phi', \phi).$$

Define the mapping  $\Phi(\phi) := \phi^*$ .

2. Find a fixed point  $\hat{\phi}$  of  $\Phi$  such that  $\Phi(\hat{\phi}) = \hat{\phi}$ .

When these two steps can be taken successfully, we say that  $\hat{\phi}$  provides a mean field game equilibrium. Note that  $\mathbf{X}^{\hat{\phi}, \hat{\phi}} = \mathbf{X}^{\hat{\phi}}$  and therefore  $\mathcal{S}(\hat{\phi}, \hat{\phi}) = SC(\hat{\phi})$  gives the social cost for the mean field game equilibrium provided by  $\hat{\phi}$ . Notice that there could possibly be many feedback functions providing a mean field game equilibrium. Let  $\mathcal{N}$  denote the set of all such feedback functions providing mean field game equilibria, as detailed above, i.e.

$$\mathcal{N} = \{\phi : [0, T] \times \mathbb{R} \ni (t, x) \rightarrow \mathbb{R} \mid \Phi(\phi) = \phi\}.$$

## 1.2 Centralized Control: Optimal Control of McKean-Vlasov Type

The goal of this subsection is to articulate how to compute the cost associated with the control problem of McKean-Vlasov type,  $SC^{MKV}$ . The central planner considers the following control problem:

$$\begin{aligned} \hat{\phi} &= \arg \inf_{\phi} SC(\phi) \\ &= \arg \inf_{\phi} \left[ \int_0^T \langle f(t, \cdot, \mathcal{L}(X_t^\phi), \phi(t, \cdot), \mathcal{L}(\phi(t, X_t^\phi))), \mathcal{L}(X_t^\phi) \rangle dt + \langle g(\cdot, \mathcal{L}(X_T^\phi)), \mathcal{L}(X_T^\phi) \rangle \right]. \end{aligned}$$

Thus, the cost of the solution to the optimal control problem of McKean-Vlasov is given by

$$SC^{MKV} = SC(\hat{\phi}).$$

**Remark 1.** *We are not concerned with uniqueness for the control of McKean-Vlasov type problem, because  $SC^{MKV} = SC(\phi_1) = SC(\phi_2)$  is still well defined even if there are two different optimal feedback functions  $\phi_1$  and  $\phi_2$  minimizing  $SC(\phi)$ .*

### 1.3 Price of Anarchy

We have described two approaches to compute the optimal feedback function  $\phi$ . In the mean field game formulation, we require  $\phi \in \mathcal{N}$ , where  $\mathcal{N}$  denotes the set of feedback functions providing mean field game equilibria. In the optimal control of McKean-Vlasov type formulation, the optimal control to be adopted by all players is computed by a central planner, who optimizes the social cost function  $SC(\phi)$  directly. Thus, we necessarily have:

$$SC^{MKV} \leq SC(\phi), \quad \forall \phi \in \mathcal{N}.$$

In other words, there is a ‘price of anarchy’ associated with allowing players to choose their controls selfishly. We thus define the price of anarchy (denoted  $PoA$ ) as the ratio between the worst case cost for a mean field game equilibrium and the optimal cost computed by a central planner:

$$PoA = \frac{\sup_{\phi \in \mathcal{N}} SC(\phi)}{SC^{MKV}}.$$

## 2 Price of Anarchy for Linear Quadratic Extended Mean Field Games

The class of linear quadratic extended mean field games is a class of problems for which explicit solutions can be computed analytically, and thus, we can compute the price of anarchy explicitly. To the best of our knowledge, the case of linear quadratic extended mean field games has not been explored in the literature, as well as computing the price of anarchy for this class of games.

To begin, we need to describe in more detail the two problems that will be used to compute the price of anarchy: the linear quadratic extended mean field game, and the linear quadratic control problem of McKean-Vlasov type with dependence on the law of the control. To specify the problems, we only need to specify the drift and cost functions,  $b$ ,  $f$ , and  $g$  introduced in Section 1. For the linear quadratic models, we take the drift to be linear:

$$b(t, x, \mu, \alpha, \nu) = b_1(t)x + \bar{b}_1(t)\bar{\mu} + b_2(t)\alpha + \bar{b}_2(t)\bar{\nu},$$

where  $\bar{\mu}$  denotes the mean of the measure  $\mu$ , namely,  $\bar{\mu} = \int_{\mathbb{R}} x d\mu(x)$ , and similarly for  $\bar{\nu}$ . We take the running and terminal costs to be quadratic:

$$\begin{aligned} f(t, x, \mu, \alpha, \nu) &= \frac{1}{2} \left( q(t)x^2 + \bar{q}(t)(x - s(t)\bar{\mu})^2 + r(t)\alpha^2 + \bar{r}(t)(\alpha - \bar{s}(t)\bar{\nu})^2 \right), \\ g(x, \mu) &= \frac{1}{2} \left( q_T x^2 + \bar{q}_T (x - s_T \bar{\mu})^2 \right). \end{aligned}$$

**Remark 2.** *If  $\bar{b}_2(t) \equiv 0$  and  $\bar{r}(t) \equiv 0$ , then we have the standard mean field game or control problem of McKean-Vlasov type. (See Theorem 1 for assumptions that provide existence and uniqueness.)*

## 2.1 Linear Quadratic Extended Mean Field Games

To solve the linear quadratic extended mean field game (LQEMFG), we begin by considering the reduced Hamiltonian for this problem:

$$\begin{aligned} H(t, x, \bar{\mu}, \alpha, \bar{\nu}, y) = & (b_1(t)x + \bar{b}_1(t)\bar{\mu} + b_2(t)\alpha + \bar{b}_2(t)\bar{\nu})y \\ & + \frac{1}{2} (q(t)x^2 + \bar{q}(t)(x - s(t)\bar{\mu})^2 + r(t)\alpha^2 + \bar{r}(t)(\alpha - \bar{s}(t)\bar{\nu})^2), \end{aligned}$$

and whenever the flows  $\bar{\mu} = (\bar{\mu}_t)_{0 \leq t \leq T}$  and  $\bar{\nu} = (\bar{\nu}_t)_{0 \leq t \leq T}$  are fixed, we consider for each control process  $\alpha = (\alpha_t)_{0 \leq t \leq T}$  the adjoint equation:

$$\begin{aligned} dY_t &= -\partial_x H(t, X_t, \bar{\mu}_t, \alpha_t, \bar{\nu}_t, Y_t)dt + Z_t dW_t \\ Y_T &= \partial_x g(X_T, \mathcal{L}(X_T)). \end{aligned}$$

According to the Pontryagin stochastic maximum principle, a sufficient condition for optimality is  $\partial_\alpha H(t, X_t, \bar{\mu}_t, \hat{\alpha}_t, \bar{\nu}_t, y) = 0$ . We introduce the function:

$$\hat{\alpha}(t, x, \bar{\mu}, \bar{\nu}, y) = \frac{\bar{r}(t)\bar{s}(t)\bar{\nu} - b_2(t)y}{r(t) + \bar{r}(t)},$$

and use the control  $\hat{\alpha}_t = \hat{\alpha}(t, X_t, \bar{\mu}, \bar{\nu}, Y_t)$ . When solving the fixed point step, we identify  $\bar{\nu}_t = \mathbb{E}(\hat{\alpha}_t)$ . By taking the expectation, we find:

$$\mathbb{E}(\hat{\alpha}_t) = c^{MFG}(t)\mathbb{E}(Y_t)$$

with:

$$c^{MFG}(t) = -\frac{b_2(t)}{r(t) + \bar{r}(t)(1 - \bar{s}(t))}.$$

Thus, necessarily we must have:

$$\hat{\alpha}_t = a^{MFG}(t)Y_t + b^{MFG}(t)\mathbb{E}(Y_t)$$

with:

$$a^{MFG}(t) = -\frac{b_2(t)}{r(t) + \bar{r}(t)},$$

and:

$$b^{MFG}(t) = -\frac{\bar{r}(t)\bar{s}(t)b_2(t)}{(r(t) + \bar{r}(t))(r(t) + \bar{r}(t)(1 - \bar{s}(t)))}.$$

Note that  $c^{MFG}(t) = a^{MFG}(t) + b^{MFG}(t)$ . The solution of the mean field game equilibrium problem is given by the solution to the FBSDE system:

$$\begin{aligned} dX_t &= (b_1(t)X_t + \bar{b}_1(t)\mathbb{E}X_t + a^{MFG}(t)b_2(t)Y_t + (b^{MFG}(t)b_2(t) + c^{MFG}(t)\bar{b}_2(t))\mathbb{E}Y_t) dt + \sigma dW_t \\ dY_t &= -((q(t) + \bar{q}(t))X_t - \bar{q}(t)s(t)\mathbb{E}X_t + b_1(t)Y_t) dt + Z_t dW_t \end{aligned} \tag{2}$$

with initial condition  $X_0 = \xi$ , a random variable with finite mean and variance, and terminal condition  $Y_T = (q_T + \bar{q}_T)X_T - \bar{q}_T s_T \mathbb{E}X_T$ .

This is a linear FBSDE of McKean-Vlasov type, which can be solved explicitly under mild assumptions (or at least in the case of time-independent coefficients which we will consider later. See Appendix A). Let  $\bar{\eta}_t^{MFG}$ ,  $\eta_t^{MFG}$ ,  $\bar{x}_t^{MFG}$ , and  $v_t^{MFG}$  denote the solutions for this problem as described in the appendix so that  $Y_t = \eta_t^{MFG} X_t + (\bar{\eta}_t^{MFG} - \eta_t^{MFG}) \bar{x}_t^{MFG}$ ,  $\mathbb{E}(Y_t) = \bar{\eta}_t^{MFG} \bar{x}_t^{MFG}$ ,  $\mathbb{E}(X_t) = \bar{x}_t^{MFG}$ , and  $Var(X_t) = v_t^{MFG}$  provide a solution to the LQEMFG problem. Then from the appendix, we have:

$$\begin{aligned} \dot{\bar{\eta}}_t^{MFG} + c^{MFG}(t)(b_2(t) + \bar{b}_2(t))(\bar{\eta}_t^{MFG})^2 + (2b_1(t) + \bar{b}_1(t))\bar{\eta}_t^{MFG} + q(t) + \bar{q}(t)(1 - s(t)) &= 0, \\ \bar{\eta}_T^{MFG} - (q_T + \bar{q}_T(1 - s_T)) &= 0, \end{aligned} \quad (3)$$

$$\begin{aligned} \dot{\eta}_t^{MFG} + a^{MFG}(t)b_2(t)(\eta_t^{MFG})^2 + 2b_1(t)\eta_t^{MFG} + q(t) + \bar{q}(t) &= 0, \\ \eta_T^{MFG} - (q_T + \bar{q}_T) &= 0, \end{aligned} \quad (4)$$

$$\begin{aligned} \dot{\bar{x}}_t^{MFG} &= (b_1(t) + \bar{b}_1(t) + c^{MFG}(t)(b_2(t) + \bar{b}_2(t))\bar{\eta}_t^{MFG})\bar{x}_t^{MFG}, \\ \bar{x}_0^{MFG} &= \mathbb{E}(\xi). \end{aligned} \quad (5)$$

And thus,

$$\bar{x}_t^{MFG} = \mathbb{E}(\xi) e^{\int_0^t (b_1(u) + \bar{b}_1(u) + c^{MFG}(u)(b_2(u) + \bar{b}_2(u))\bar{\eta}_u^{MFG}) du}, \quad (6)$$

$$v_t^{MFG} = Var(\xi) e^{\int_0^t 2(b_1(s) + a^{MFG}(s)b_2(s)\eta_s^{MFG}) ds} + \sigma^2 \int_0^t e^{2 \int_s^t (b_1(u) + a^{MFG}(u)b_2(u)\eta_u^{MFG}) du} ds. \quad (7)$$

Let  $SC^{MFG} := SC(\phi)$  for the feedback function specified by this solution, i.e.

$$\phi(t, x) = a^{MFG}(t)\eta_t^{MFG}x + (a^{MFG}(t)(\bar{\eta}_t^{MFG} - \eta_t^{MFG}) + b^{MFG}(t)\bar{\eta}_t^{MFG})\bar{x}_t^{MFG}.$$

Then we can compute the social cost as described in Section 1.1:

$$\begin{aligned} SC^{MFG} &= \frac{1}{2}[(q_T + \bar{q}_T)v_T^{MFG} + (q_T + \bar{q}_T(1 - s_T)^2)(\bar{x}_T^{MFG})^2 \\ &\quad + \int_0^T (q(t) + \bar{q}(t) + (r(t) + \bar{r}(t))(a^{MFG}(t)\eta_t^{MFG})^2)v_t^{MFG} \\ &\quad + (q(t) + \bar{q}(t)(1 - s(t))^2 + (r(t) + \bar{r}(t)(1 - \bar{s}(t))^2)(c^{MFG}(t)\bar{\eta}_t^{MFG})^2)(\bar{x}_t^{MFG})^2 dt], \end{aligned}$$

where we have used the fact that:

$$\mathbb{E}(\phi(t, X_t)) = c^{MFG}(t)\bar{\eta}_t^{MFG}\bar{x}_t^{MFG},$$

and:

$$Var(\phi(t, X_t)) = (a^{MFG}(t)\eta_t^{MFG})^2 v_t^{MFG}.$$

## 2.2 Linear Quadratic Control of McKean-Vlasov Type Involving the Law of the Control

To solve the linear quadratic optimal control problem of McKean-Vlasov type involving the law of the control (LQEMKV), we begin with the reduced Hamiltonian, which is the same as in the LQEMFG problem:

$$\begin{aligned} H(t, x, \bar{\mu}, \alpha, \bar{\nu}, y) &= (b_1(t)x + \bar{b}_1(t)\bar{\mu} + b_2(t)\alpha + \bar{b}_2(t)\bar{\nu})y \\ &\quad + \frac{1}{2} (q(t)x^2 + \bar{q}(t)(x - s(t)\bar{\mu})^2 + r(t)\alpha^2 + \bar{r}(t)(\alpha - \bar{s}(t)\bar{\nu})^2). \end{aligned}$$

Since we require  $\bar{\nu}_t$  to be equal to  $\mathbb{E}(\alpha_t)$  throughout the optimization, it is not sufficient to minimize the Hamiltonian with respect to the  $\alpha$  input alone in order to guarantee optimality. A sufficient condition for control problems of McKean-Vlasov type involving the law of the control is derived in [5]. Since we consider a Hamiltonian that depends on the means of  $\bar{\mu}$  and  $\bar{\nu}$  instead of the full distributions, the sufficient condition is the following:  $\hat{\alpha}(t, X_t, \bar{\mu}, \bar{\nu}, Y_t)$  should satisfy:

$$\partial_\alpha H(t, X_t, \mathbb{E}(X_t), \hat{\alpha}_t, \mathbb{E}(\hat{\alpha}_t), Y_t) + \tilde{\mathbb{E}} \left[ \partial_{\bar{\nu}} H(t, \tilde{X}_t, \mathbb{E}(X_t), \hat{\alpha}_t, \mathbb{E}(\hat{\alpha}_t), \tilde{Y}_t) \right] = 0,$$

where the adjoint equation is given by:

$$\begin{aligned} dY_t &= - \left[ \partial_x H(t, X_t, \bar{\mu}_t, \alpha_t, \bar{\nu}_t, Y_t) + \tilde{\mathbb{E}} \left[ \partial_{\bar{\mu}} H(t, \tilde{X}_t, \bar{\mu}_t, \tilde{\alpha}_t, \bar{\nu}_t, \tilde{Y}_t) \right] \right] dt + Z_t dW_t \\ Y_T &= \partial_x g(X_T, \mathcal{L}(X_T)) + \tilde{\mathbb{E}} \left[ \partial_{\bar{\mu}} g(\tilde{X}_T, \mathcal{L}(X_T))(X_T) \right], \end{aligned}$$

and  $(\tilde{X}, \tilde{Y}, \tilde{\alpha})$  denotes an independent copy of  $(X, Y, \alpha)$ . In the present LQ case, the sufficient condition can be used to solve for:

$$\hat{\alpha}_t = a^{MKV}(t)Y_t + b^{MKV}(t)\mathbb{E}(Y_t),$$

and:

$$\mathbb{E}(\hat{\alpha}_t) = c^{MKV}(t)\mathbb{E}(Y_t),$$

with:

$$\begin{aligned} a^{MKV}(t) &= -\frac{b_2(t)}{r(t) + \bar{r}(t)} \\ b^{MKV}(t) &= -\frac{1}{r(t) + \bar{r}(t)} \left( \bar{b}_2(t) - \frac{\bar{r}(t)\bar{s}(t)(\bar{s}(t) - 2)(b_2(t) + \bar{b}_2(t))}{r(t) + \bar{r}(t)(1 - \bar{s}(t))^2} \right) \\ c^{MKV}(t) &= -\frac{b_2(t) + \bar{b}_2(t)}{r(t) + \bar{r}(t)(1 - \bar{s}(t))^2}. \end{aligned}$$

So the solution of the optimal control problem of McKean-Vlasov type is given by the solution to the FBSDE system:

$$\begin{aligned} dX_t &= (b_1(t)X_t + \bar{b}_1(t)\mathbb{E}X_t + a^{MKV}(t)b_2(t)Y_t + (b^{MKV}(t)b_2(t) + c^{MKV}(t)\bar{b}_2(t))\mathbb{E}Y_t) dt + \sigma dW_t \\ dY_t &= -((q(t) + \bar{q}(t))X_t + s(t)\bar{q}(t)(s(t) - 2)\mathbb{E}X_t + b_1(t)Y_t + \bar{b}_1(t)\mathbb{E}Y_t) dt + Z_t dW_t \end{aligned} \tag{8}$$

with initial condition  $X_0 = \xi$ , and terminal condition  $Y_T = (q_T + \bar{q}_T)X_T + s_T\bar{q}_T(s_T - 2)\mathbb{E}X_T$ .

As in the previous section, this is a linear FBSDE of McKean-Vlasov type, which can be solved explicitly under mild assumptions (or at least in the case of time-independent coefficients which we will consider later. See Appendix A). Let  $\bar{\eta}_t^{MKV}$ ,  $\eta_t^{MKV}$ ,  $\bar{x}_t^{MKV}$ , and  $v_t^{MKV}$  so that  $Y_t = \eta_t^{MKV}X_t + (\bar{\eta}_t^{MKV} - \eta_t^{MKV})\bar{x}_t^{MKV}$ ,  $\mathbb{E}(Y_t) = \bar{\eta}_t^{MKV}\bar{x}_t^{MKV}$ ,  $\mathbb{E}(X_t) = \bar{x}_t^{MKV}$ , and  $Var(X_t) = v_t^{MKV}$  provide a solution to the LQEMKV problem. Then from the appendix, we have:

$$\begin{aligned} \dot{\bar{\eta}}_t^{MKV} + c^{MKV}(t)(b_2(t) + \bar{b}_2(t))(\bar{\eta}_t^{MKV})^2 + 2(b_1(t) + \bar{b}_1(t))\bar{\eta}_t^{MKV} + q(t) + \bar{q}(t)(1 - s(t))^2 &= 0, \\ \bar{\eta}_T^{MKV} - (q_T + \bar{q}_T(1 - s_T))^2 &= 0, \end{aligned} \tag{9}$$



$$\begin{aligned}\dot{\eta}_t^{MKV} + a^{MKV}(t)b_2(t)(\eta_t^{MKV})^2 + 2b_1(t)\eta_t^{MKV} + q(t) + \bar{q}(t) &= 0, \\ \eta_T^{MKV} - (q_T + \bar{q}_T) &= 0,\end{aligned}\tag{10}$$

$$\begin{aligned}\dot{\bar{x}}_t^{MKV} &= (b_1(t) + \bar{b}_1(t) + c^{MKV}(t)(b_2(t) + \bar{b}_2(t))\bar{\eta}_t^{MKV})\bar{x}_t^{MKV}, \\ \bar{x}_0^{MKV} &= \mathbb{E}(\xi).\end{aligned}\tag{11}$$

And thus,

$$\begin{aligned}\bar{x}_t^{MKV} &= \mathbb{E}(\xi)e^{\int_0^t (b_1(u) + \bar{b}_1(u) + c^{MKV}(u)(b_2(u) + \bar{b}_2(u))\bar{\eta}_u^{MKV})du}, \\ v_t^{MKV} &= Var(\xi)e^{\int_0^t 2(b_1(s) + a^{MKV}(s)b_2(s)\eta_s^{MKV})ds} + \sigma^2 \int_0^t e^{2\int_s^t (b_1(u) + a^{MKV}(u)b_2(u)\eta_u^{MKV})du} ds.\end{aligned}$$

Then  $SC^{MKV} = SC(\phi)$  where  $\phi$  is the feedback function specified by this solution, i.e.

$$\phi(t, x) = a^{MKV}(t)\eta_t^{MKV}x + (a^{MKV}(t)(\bar{\eta}_t^{MKV} - \eta_t^{MKV}) + b^{MKV}(t)\bar{\eta}_t^{MKV})\bar{x}_t^{MKV}.$$

Then we can compute the social cost, denoted  $SC^{MKV}$ , as described in Section 1.2:

$$\begin{aligned}SC^{MKV} &= \frac{1}{2}[(q_T + \bar{q}_T)v_T^{MKV} + (q_T + \bar{q}_T(1 - s_T)^2)(\bar{x}_T^{MKV})^2 \\ &\quad + \int_0^T (q(t) + \bar{q}(t) + (r(t) + \bar{r}(t))(a^{MKV}(t)\eta_t^{MKV})^2)v_t^{MKV} dt \\ &\quad + \int_0^T (q(t) + \bar{q}(t)(1 - s(t))^2 + (r(t) + \bar{r}(t)(1 - \bar{s}(t))^2)(c^{MKV}(t)\bar{\eta}_t^{MKV})^2)(\bar{x}_t^{MKV})^2 dt],\end{aligned}$$

where we have used the fact that:

$$\mathbb{E}(\phi(t, X_t)) = c^{MKV}(t)\bar{\eta}_t^{MKV}\bar{x}_t^{MKV},$$

and:

$$Var(\phi(t, X_t)) = (a^{MKV}(t)\eta_t^{MKV})^2 v_t^{MKV}.$$

### 2.3 Theoretical Results

For the remainder of the paper, we assume the coefficients are independent of time and nonnegative:

$$(b_1(t), \bar{b}_1(t), b_2(t), \bar{b}_2(t), q(t), \bar{q}(t), r(t), \bar{r}(t), s(t), \bar{s}(t)) = (b_1, \bar{b}_1, b_2, \bar{b}_2, q, \bar{q}, r, \bar{r}, s, \bar{s}) \in (\mathbb{R}^+)^{10}$$

and therefore,

$$\begin{aligned}(a^{MFG}(t), b^{MFG}(t), c^{MFG}(t)) &= (a^{MFG}, b^{MFG}, c^{MFG}) \\ (a^{MKV}(t), b^{MKV}(t), c^{MKV}(t)) &= (a^{MKV}, b^{MKV}, c^{MKV}).\end{aligned}$$

**Theorem 1.** *Assume the following:*

$$\begin{array}{llll} b_2 & > & 0 & b_2 + \bar{b}_2 & > & 0 \\ r + \bar{r} & > & 0 & r + \bar{r}(1 - \bar{s}) & > & 0 & r + \bar{r}(1 - \bar{s})^2 & > & 0 \\ q + \bar{q} & > & 0 & q + \bar{q}(1 - s) & > & 0 & q + \bar{q}(1 - s)^2 & > & 0 \\ q_T + \bar{q}_T & \geq & 0 & q_T + \bar{q}_T(1 - s_T) & \geq & 0 & q_T + \bar{q}_T(1 - s_T)^2 & \geq & 0 \end{array}\tag{12}$$

then there exists a unique solution to the LQEMFG problem, and there exists a unique solution to the LQEMKV problem. And therefore,  $PoA = \frac{SC^{MFG}}{SC^{MKV}}$  where  $SC^{MFG} := SC(\phi)$  for  $\phi$  given by the explicit solution constructed in Appendix A.

**Remark 3.** Note that existence in Theorem 1 follows from the explicit construction in Appendix A, because the above conditions provide existence to the solutions of the Riccati equations. Uniqueness comes from the connection between LQEMFG or LQEMKV and deterministic LQ optimal control. (See Section 3.5.1 in [7]).

To compute the price of anarchy, it is useful to make the following observations:

$$\begin{aligned} a^{MFG} &= a^{MKV} =: a \\ \eta_t^{MFG} &= \eta_t^{MKV} =: \eta_t \\ v_t^{MFG} &= v_t^{MKV} =: v_t \end{aligned}$$

**Proposition 1.** Assuming (12), if furthermore,

$$\begin{aligned} (s\bar{q}(s-1) + \bar{b}_1\bar{\eta}_t^{MKV})\bar{x}_t^{MKV} &= 0, \quad \forall t \in [0, T] \\ \bar{\eta}_t^{MKV}\bar{x}_t^{MKV} [(b^{MFG} - b^{MKV})b_2 + (c^{MFG} - c^{MKV})\bar{b}_2] &= 0, \quad \forall t \in [0, T] \\ s_T\bar{q}_T(s_T-1)\bar{x}_T^{MKV} &= 0, \end{aligned}$$

then  $PoA = 1$ .

*Proof.* Comparing the FBSDE systems (2) and (8), the result is clear.  $\square$

**Corollary 1.** Assuming (12), if furthermore,  $\bar{b}_1 = 0$ ,  $s\bar{q}(s-1) = 0$ , and  $s_T\bar{q}_T(s_T-1) = 0$  and at least one of the following holds:  $b_2 = \bar{b}_2 = 0$  or  $\frac{b_2(r+\bar{r}(1-\bar{s})^2)}{(b_2+\bar{b}_2)(r+\bar{r}(1-\bar{s}))} = 1$  then  $PoA = 1$ .

**Remark 4.** The only result similar to Proposition 1 that we are aware of is Remark 6.1 in [16].

Using the above observations, we can rewrite:

$$\begin{aligned} SC^{MFG} &= \frac{1}{2}(q_T + \bar{q}_T)v_T + \frac{1}{2}(q_T + \bar{q}_T(1-s_T)^2)(\bar{x}_T^{MFG})^2 + \frac{1}{2}\int_0^T (q + \bar{q} + (r + \bar{r})(a\eta_t)^2)v_t dt \\ &\quad + \frac{1}{2}\int_0^T [q + \bar{q}(1-s)^2 + (r + \bar{r}(1-\bar{s})^2)(c^{MFG}\bar{\eta}_t^{MFG})^2](\bar{x}_t^{MFG})^2 dt, \end{aligned} \quad (13)$$

and:

$$\begin{aligned} SC^{MKV} &= \frac{1}{2}(q_T + \bar{q}_T)v_T + \frac{1}{2}(q_T + \bar{q}_T(1-s_T)^2)(\bar{x}_T^{MKV})^2 + \frac{1}{2}\int_0^T (q + \bar{q} + (r + \bar{r})(a\eta_t)^2)v_t dt \\ &\quad + \frac{1}{2}\int_0^T [q + \bar{q}(1-s)^2 + (r + \bar{r}(1-\bar{s})^2)(c^{MKV}\bar{\eta}_t^{MKV})^2](\bar{x}_t^{MKV})^2 dt. \end{aligned} \quad (14)$$

In the following, we intend to simplify the explicit solutions (13) and (14) for the social costs in the LQEMFG and LQEMKV problems. First, consider the quantity  $\int_0^T (\bar{\eta}_t^{MFG})^2(\bar{x}_t^{MFG})^2 dt$ . Using equation (3), we have:

$$\begin{aligned} &\int_0^T (\bar{\eta}_t^{MFG})^2(\bar{x}_t^{MFG})^2 dt \\ &= -\frac{1}{c^{MFG}(b_2 + \bar{b}_2)} \left[ \int_0^T \dot{\bar{\eta}}_t^{MFG}(\bar{x}_t^{MFG})^2 dt + \int_0^T ((2b_1 + \bar{b}_1)\bar{\eta}_t^{MFG} + (q + \bar{q}(1-s)))(\bar{x}_t^{MFG})^2 dt \right] \end{aligned}$$

Then by integration by parts:

$$= -\frac{1}{c^{MFG}(b_2 + \bar{b}_2)} \left[ \bar{\eta}_T^{MFG} (\bar{x}_T^{MFG})^2 - \bar{\eta}_0^{MFG} (\bar{x}_0^{MFG})^2 - 2 \int_0^T \bar{\eta}_t^{MFG} \bar{x}_t^{MFG} \dot{\bar{x}}_t^{MFG} dt \right. \\ \left. + \int_0^T ((2b_1 + \bar{b}_1) \bar{\eta}_t^{MFG} + (q + \bar{q}(1-s)) (\bar{x}_t^{MFG})^2) dt \right]$$

Then using equation (5):

$$= 2 \int_0^T (\bar{\eta}_t^{MFG})^2 (\bar{x}_t^{MFG})^2 dt - \frac{1}{c^{MFG}(b_2 + \bar{b}_2)} \left[ \bar{\eta}_T^{MFG} (\bar{x}_T^{MFG})^2 - \bar{\eta}_0^{MFG} (\mathbb{E}(\xi))^2 \right. \\ \left. - \int_0^T (\bar{b}_1 \bar{\eta}_t^{MFG} - (q + \bar{q}(1-s))) (\bar{x}_t^{MFG})^2 dt \right].$$

Finally, we solve:

$$\int_0^T (\bar{\eta}_t^{MFG})^2 (\bar{x}_t^{MFG})^2 dt = \frac{1}{c^{MFG}(b_2 + \bar{b}_2)} \left[ \bar{\eta}_T^{MFG} (\bar{x}_T^{MFG})^2 - \bar{\eta}_0^{MFG} (\mathbb{E}(\xi))^2 \right. \\ \left. + \int_0^T (-\bar{b}_1 \bar{\eta}_t^{MFG} + (q + \bar{q}(1-s))) (\bar{x}_t^{MFG})^2 dt \right].$$

If we denote:

$$\lambda := \frac{c^{MFG}}{c^{MKV}} = \frac{b_2}{b_2 + \bar{b}_2} \cdot \frac{r + \bar{r}(1-s)^2}{r + \bar{r}(1-\bar{s})},$$

$$h_{var} := \frac{1}{2} \int_0^T (q + \bar{q} + (r + \bar{r})(a\eta_t)^2) v_t dt + \frac{1}{2} (q_T + \bar{q}_T) v_T,$$

and use the terminal condition for  $\bar{\eta}_T^{MFG}$ , then equation (13) can be rewritten as:

$$SC^{MFG} = h_{var} + \frac{1}{2} \int_0^T [\bar{b}_1 \lambda \bar{\eta}_t^{MFG} + (q + \bar{q}(1-s)^2) - \lambda(q + \bar{q}(1-s))] (\bar{x}_t^{MFG})^2 dt, \\ + \frac{1}{2} \lambda (\bar{\eta}_0^{MFG} (\mathbb{E}(\xi))^2 - (q_T + \bar{q}_T(1-s_T)) (\bar{x}_T^{MFG})^2) + \frac{1}{2} (q_T + \bar{q}_T(1-s_T)^2) (\bar{x}_T^{MFG})^2. \quad (15)$$

Repeating the calculation for  $\int_0^T (\bar{\eta}_t^{MFG})^2 (\bar{x}_t^{MFG})^2 dt$  (using equations (9) and (11)), we arrive at:

$$\int_0^T (\bar{\eta}_t^{MKV})^2 (\bar{x}_t^{MKV})^2 dt = \frac{1}{c^{MKV}(b_2 + \bar{b}_2)} \left[ \bar{\eta}_T^{MKV} (\bar{x}_T^{MKV})^2 - \bar{\eta}_0^{MKV} (\mathbb{E}(\xi))^2 \right. \\ \left. + \int_0^T (q + \bar{q}(1-s)^2) (\bar{x}_t^{MKV})^2 dt \right],$$

Using the terminal condition for  $\bar{\eta}_T^{MKV}$ , equation (14) can be rewritten as:

$$SC^{MKV} = h_{var} + \frac{1}{2} \bar{\eta}_0^{MKV} (\mathbb{E}(\xi))^2. \quad (16)$$

Let's denote the (weighted) difference between the solutions of the Riccati equations associated with  $\bar{\eta}_t^{MFG}$  and  $\bar{\eta}_t^{MKV}$  by:

$$\Delta\bar{\eta}_t = \lambda\bar{\eta}_t^{MFG} - \bar{\eta}_t^{MKV}. \quad (17)$$

**Proposition 2.** *Under assumption (12), the difference in the social costs in the LQEMFG and LQEMKV problems can be represented by*

$$\Delta SC := SC^{MFG} - SC^{MKV} = \frac{1}{2} \frac{(b_2 + \bar{b}_2)^2}{r + \bar{r}(1 - \bar{s})^2} \int_0^T (\Delta\bar{\eta}_t \cdot \bar{x}_t^{MFG})^2 dt.$$

*Proof.* The solutions  $\bar{\eta}_t^{MFG}$  and  $\bar{\eta}_t^{MKV}$  for the Riccati equations (3) and (9), respectively, are well defined under assumption (12) (see Appendix A). We notice that  $\Delta\bar{\eta}_t$  defined in (17) satisfies the following first-order differential equation:

$$\frac{d(\Delta\bar{\eta}_t)}{dt} = \gamma_t \Delta\bar{\eta}_t + \beta_t, \quad \Delta\bar{\eta}_T = \lambda\bar{\eta}_T^{MFG} - \bar{\eta}_T^{MKV}$$

with coefficients:

$$\begin{cases} \gamma_t &= -2b_1 - 2\bar{b}_1 + \frac{(b_2 + \bar{b}_2)^2}{r + \bar{r}(1 - \bar{s})^2} (\lambda\bar{\eta}_t^{MFG} + \bar{\eta}_t^{MKV}), \\ \beta_t &= \bar{b}_1 \lambda \bar{\eta}_t^{MFG} + (q + \bar{q}(1 - s)^2) - \lambda(q + \bar{q}(1 - s)). \end{cases}$$

Since  $q_T + \bar{q}_T(1 - s_T) = \bar{\eta}_T^{MFG}$ ,  $q_T + \bar{q}_T(1 - s_T)^2 = \bar{\eta}_T^{MKV}$  and  $\lambda\bar{\eta}_0^{MFG} - \bar{\eta}_0^{MKV} = \Delta\bar{\eta}_0$ , we deduce from equations (15) and (16) that:

$$\begin{aligned} & SC^{MFG} - SC^{MKV} \\ &= \frac{1}{2} \left[ \Delta\bar{\eta}_0 (\mathbb{E}(\xi))^2 - \Delta\bar{\eta}_T (\bar{x}_T^{MFG})^2 + \int_0^T \beta_t (\bar{x}_t^{MFG})^2 dt \right] \\ &= \frac{1}{2} \int_0^T \left[ -\frac{d(\Delta\bar{\eta}_t (\bar{x}_t^{MFG})^2)}{dt} + \left( \frac{d(\Delta\bar{\eta}_t)}{dt} - \gamma_t \Delta\bar{\eta}_t \right) (\bar{x}_t^{MFG})^2 \right] dt \\ &= \frac{1}{2} \int_0^T [-2\Delta\bar{\eta}_t \bar{x}_t^{MFG} \dot{\bar{x}}_t^{MFG} - \gamma_t \Delta\bar{\eta}_t (\bar{x}_t^{MFG})^2] dt \\ &= \frac{1}{2} \int_0^T \Delta\bar{\eta}_t (\bar{x}_t^{MFG})^2 \left[ -2 \left( b_1 + \bar{b}_1 - \frac{(b_2 + \bar{b}_2)^2}{r + \bar{r}(1 - \bar{s})^2} \lambda\bar{\eta}_t^{MFG} \right) - \gamma_t \right] dt \\ &= \frac{1}{2} \frac{(b_2 + \bar{b}_2)^2}{r + \bar{r}(1 - \bar{s})^2} \int_0^T (\Delta\bar{\eta}_t \cdot \bar{x}_t^{MFG})^2 dt, \end{aligned}$$

where we use equation (5) for the fourth equality. □

**Remark 5.** *We can see directly from Proposition 2 that the social cost in the LQEMFG problem is larger than (or possibly equal to) the social cost in the LQEMKV problem. This result is consistent with the definition of the price of anarchy in Section 1.3.*

Therefore, under assumption (12), the price of anarchy for the LQ model is given by

$$PoA = 1 + \frac{\Delta SC}{SC^{MKV}} = 1 + \frac{\frac{(b_2 + \bar{b}_2)^2}{r + \bar{r}(1 - \bar{s})^2} \int_0^T (\Delta \bar{\eta}_t \cdot \bar{x}_t^{MFG})^2 dt}{\int_0^T \left[ q + \bar{q} + \frac{b_2^2}{r + \bar{r}} \eta_t^2 \right] v_t dt + (q_T + \bar{q}_T) v_T + \bar{\eta}_0^{MKV} (\mathbb{E}(\xi))^2}. \quad (18)$$

**Corollary 2.** *Assuming (12), if the initial condition  $\xi$  is such that  $\mathbb{E}(\xi) = 0$ , then  $PoA = 1$ .*

*Proof.* By equation (6),  $\mathbb{E}(\xi) = 0$  implies  $\bar{x}_t^{MFG} = 0$ ,  $\forall t \in [0, T]$ . Therefore by Proposition 2,  $\Delta SC = 0$ . From equation (7),  $v_t > 0$  which implies  $h_{var} > 0$ . Therefore by equation (16),  $SC^{MKV} > 0$ . We conclude that  $PoA = 1$ .  $\square$

We study in the following the variation of  $PoA$  by letting one of the coefficients tend to 0 or to  $\infty$ . It will be useful for us to note here the scalar Riccati equations associated with  $u_t := \lambda \bar{\eta}_t^{MFG}$ ,  $w_t := \bar{\eta}_t^{MKV}$  and  $\eta_t$ :

$$\dot{u}_t - 2A^u u_t - B u_t^2 + C^u = 0 \quad u_T = D^u \quad (19)$$

$$\dot{w}_t - 2A^w w_t - B w_t^2 + C^w = 0 \quad w_T = D^w \quad (20)$$

$$\dot{\eta}_t - 2A^\eta \eta_t - B^\eta \eta_t^2 + C^\eta = 0 \quad \eta_T = D^\eta \quad (21)$$

where

$$\begin{aligned} A^u &= -\left(b_1 + \frac{\bar{b}_1}{2}\right) & A^w &= -(b_1 + \bar{b}_1) & A^\eta &= -b_1 \\ B^u &= \frac{(b_2 + \bar{b}_2)^2}{r + \bar{r}(1 - \bar{s})^2} & B^w &= \frac{(b_2 + \bar{b}_2)^2}{r + \bar{r}(1 - \bar{s})^2} & B^\eta &= \frac{b_2^2}{r + \bar{r}} \\ C^u &= \lambda(q + \bar{q}(1 - s)) & C^w &= q + \bar{q}(1 - s)^2 & C^\eta &= q + \bar{q} \\ D^u &= \lambda(q_T + \bar{q}_T(1 - s_T)) & D^w &= q_T + \bar{q}_T(1 - s_T)^2 & D^\eta &= q_T + \bar{q}_T \end{aligned}$$

If  $B^u \neq 0$ ,  $B^u D^u \geq 0$  and  $B^u C^u > 0$ , we have (see equation (28) in Appendix A) the existence and uniqueness for  $u_t$  which can be expressed by:

$$u_t = \frac{-(C^u + D^u \delta_u^+) + (C^u + D^u \delta_u^-) e^{-(\delta_u^+ - \delta_u^-)(T-t)}}{(\delta_u^- - D^u B^u) - (\delta_u^+ - D^u B^u) e^{-(\delta_u^+ - \delta_u^-)(T-t)}} \quad (22)$$

where  $\delta_u^\pm = -A^u \pm \sqrt{(A^u)^2 + B^u C^u}$ . We have analogous expressions for  $w_t$  and  $\eta_t$  in terms of  $\delta_w^\pm$  and  $\delta_\eta^\pm$ , respectively. Note that  $B^u = B^w =: B$ .

**Proposition 3.** *Assuming (12),*

$$\lim_{r \rightarrow \infty} PoA = 1 \quad \text{and} \quad \lim_{\bar{r} \rightarrow \infty} PoA = 1.$$

*Proof.* When assumption (12) is satisfied, we have the existence and uniqueness of solutions for the scalar Riccati equations associated with  $u_t$ ,  $w_t$  and  $\eta_t$ .

When  $r \rightarrow \infty$  we have

$$\lambda \rightarrow \lambda^{\infty,r} := \frac{b_2}{b_2 + \bar{b}_2}$$

and

$$B \rightarrow 0, \quad B^\eta \rightarrow 0, \quad C^u \rightarrow C_u^{\infty,r} := \lambda^{\infty,r}(q + \bar{q}(1-s)), \quad D^u \rightarrow D_u^{\infty,r} := \lambda^{\infty,r}(q_T + \bar{q}_T(1-s_T)).$$

Let's  $u^{\infty,r} : [0, T] \rightarrow \mathbb{R}$  be the solution to a first-order differential equation

$$\dot{u}_t^{\infty,r} - 2A^u u_t^{\infty,r} + C_u^{\infty,r} = 0, \quad u_T^{\infty,r} = D_u^{\infty,r} \quad (23)$$

when  $A^u \neq 0$ , or the solution to the equation

$$\dot{u}_t^{\infty,r} + C_u^{\infty,r} = 0, \quad u_T^{\infty,r} = D_u^{\infty,r} \quad (24)$$

when  $A^u = 0$ .

Since assumption (12) holds for arbitrary large  $r$ , with the explicit formula of the solution to the scalar Riccati equation, it is easy to show that the solution  $u$  to equation (19) converges pointwisely to the solution  $u^{\infty,r}$  for any fixed time  $t \in [0, T]$  when  $r$  tends towards infinity (see Appendix A). This limiting solution  $u^{\infty,r}$  can be given by the following expressions: for every  $t \in [0, T]$

$$\begin{aligned} \text{when } A^u \neq 0, \quad u_t^{\infty,r} &= \left( D_u^{\infty,r} - \frac{C_u^{\infty,r}}{2A^u} \right) e^{-2A^u(T-t)} + \frac{C_u^{\infty,r}}{2A^u}. \\ \text{when } A^u = 0, \quad u_t^{\infty,r} &= D_u^{\infty,r} + C_u^{\infty,r}(T-t) \end{aligned}$$

It is also clear that  $u^{\infty,r}$  is bonded over  $[0, T]$ .

Similarly, for any fixed time  $t \in [0, T]$ , the solution  $w_t$  and  $\eta_t$  to the equations (20) and (21) at time  $t$  converge towards the solution  $w_t^{\infty,r}$  and  $\eta_t^{\infty,r}$  of some first-order differential equations. Namely, for every  $t \in [0, T]$ ,

$$w_t \xrightarrow{r \rightarrow \infty} w_t^{\infty,r}, \quad \eta_t \xrightarrow{r \rightarrow \infty} \eta_t^{\infty,r}$$

and the function  $w^{\infty,r}$  and  $\eta^{\infty,r}$  are bounded over  $[0, T]$ .

Moreover, since  $\bar{x}_t^{MFG} = \bar{x}_0 \exp(\int_0^t (b_1 + \bar{b}_1 - B \cdot u_s) ds)$  and  $v_t = \text{Var}(\xi) \exp(2 \int_0^t (b_1 - B^\eta \eta_s) ds) + \sigma^2 \int_0^t \exp(2 \int_s^t (b_1 - B^\eta \eta_u) du) ds$ , by the bounded convergence theorem we have

$$\bar{x}_t^{MFG} \xrightarrow{r \rightarrow \infty} \bar{x}_0 e^{(b_1 + \bar{b}_1)t} =: \bar{x}_t^{\infty,r,MFG}, \quad v_t \xrightarrow{r \rightarrow \infty} \text{Var}(\xi) e^{2b_1 t} + \sigma^2 \int_0^t e^{2b_1(t-s)} ds =: v_t^{\infty,r}$$

The limits  $\bar{x}_t^{\infty,r,MFG}$  as well as  $v_t^{\infty,r}$  are bounded over  $[0, T]$ , then by the bounded convergence theorem again we obtain

$$\lim_{r \rightarrow \infty} \int_0^T (\Delta \bar{\eta}_t \cdot \bar{x}_t^{MFG})^2 dt = \int_0^T (u_t^{\infty,r} - w_t^{\infty,r})^2 (\bar{x}_t^{\infty,r,MFG})^2 dt < \infty.$$

It can be shown that  $w_0^{\infty,r} \geq 0$  and  $v_t^{\infty,r} > 0$  for any  $t > 0$ . Hence,

$$\lim_{r \rightarrow \infty} SC^{MKV} = \int_0^T (q + \bar{q}) v_t^{\infty,r} dt + (q_T + \bar{q}_T) v_T^{\infty,r} + w_0^{\infty,r} \bar{x}_0^2 > 0.$$

Therefore, from Proposition 2 and equation (18), when  $r$  tends towards infinity we have  $\frac{(b_2 + \bar{b}_2)^2}{r + \bar{r}(1 - \bar{s})^2} \rightarrow 0$  and then  $\lim_{r \rightarrow \infty} PoA = 1$ .

Idem, by replacing  $\lambda^{\infty, r}$  with  $\lambda^{\infty, \bar{r}} := \frac{b_2}{b_2 + \bar{b}_2}(1 - \bar{s})$ , we obtain  $\lim_{\bar{r} \rightarrow \infty} PoA = 1$ .

□

**Proposition 4.** *Assuming (12), if initial condition  $\xi$  satisfies  $\mathbb{E}(\xi) \neq 0$ , and if furthermore,  $\bar{b}_2 = 0$ , then:*

$$\lim_{b_2 \rightarrow 0} PoA = 1,$$

whereas if  $\bar{b}_2 \neq 0$ , then:

$$\lim_{b_2 \rightarrow 0} PoA =: PoA^{0, b_2} > 1.$$

Moreover, if

$$\frac{r + \bar{r}(1 - \bar{s})^2}{r + \bar{r}(1 - \bar{s})} \neq \frac{q + \bar{q}(1 - s)^2}{q + \bar{q}(1 - s)},$$

then

$$\lim_{b_2 \rightarrow \infty} PoA =: PoA^{\infty, b_2} > 1,$$

otherwise

$$\lim_{b_2 \rightarrow \infty} PoA = 1.$$

*Proof.* First, we consider  $b_2 \rightarrow 0$ . If  $\bar{b}_2 = 0$ , we can use the same technique in Proposition 3 to show that  $\lim_{b_2 \rightarrow 0} PoA = 1$ . Now assume  $\bar{b}_2 > 0$ . As  $b_2 \rightarrow 0$ , we have  $\lambda \rightarrow 0$  and

$$B \rightarrow \frac{\bar{b}_2^2}{r + \bar{r}(1 - \bar{s})^2} =: B^{0, b_2} > 0, \quad B^\eta \rightarrow 0, \quad C^u \rightarrow 0, \quad D^u \rightarrow 0.$$

We also have  $\delta_u^\pm \rightarrow -A^u \pm |A^u|$  and therefore,  $\delta_u^+ - \delta_u^- \rightarrow 2|A^u|$ . Thus, from equation (22) we deduce

$$u_t \xrightarrow{b_2 \rightarrow 0} 0, \quad \forall t \in [0, T].$$

From equation (6), we have  $\bar{x}_t^{MFG} = \mathbb{E}(\xi) e^{\int_0^t (b_1 + \bar{b}_1 - Bu_s) ds}$ , and therefore,  $\bar{x}_t^{MFG} \rightarrow \mathbb{E}(\xi) \exp((b_1 + \bar{b}_1)t) =: \bar{x}_t^{0, b_2, MFG}$ , which is bounded over  $[0, T]$ .

Since  $B^{0, b_2} > 0, B^{0, b_2} C^w > 0, B^{0, b_2} D^w \geq 0$ , we still have the unique solution for  $w_t^{0, b_2}$  when  $b_2 \rightarrow 0$ . Since  $B^\eta \rightarrow 0$ , using the same argument shown in Proposition 3, we get:

$$\begin{aligned} \text{when } A^\eta \neq 0, \quad \eta_t &\rightarrow \eta_t^{0, b_2} := \left( D^\eta - \frac{C^\eta}{2A^\eta} \right) e^{-2A^\eta(T-t)} + \frac{C^\eta}{2A^\eta}, \\ \text{when } A^\eta = 0, \quad \eta_t &\rightarrow \eta_t^{0, b_2} := D^\eta - C^\eta(T-t), \end{aligned}$$

which are bounded over  $[0, T]$ . Thus, the limiting variance function  $v_t \rightarrow v_t^{0, b_2} := Var(\xi) e^{2b_1 t} + \sigma^2 \int_0^t e^{2b_1(t-s)} ds > 0$  for  $t > 0$  and  $v_t^{0, b_2}$  is bounded. Hence, from Proposition 2, the bounded convergence theorem, and the assumption on  $\xi$ , we deduce

$$\begin{aligned} \lim_{b_2 \rightarrow 0} \Delta SC &= \frac{1}{2} B^{0, b_2} \int_0^T (w_t^{0, b_2} \cdot \bar{x}_t^{0, b_2, MFG})^2 dt, \\ 0 &< \lim_{b_2 \rightarrow 0} \Delta SC < \infty, \end{aligned}$$

and using the fact that  $w_t^{0,b_2} \not\equiv 0$  and using equation (16), we have  $0 < \lim_{b_2 \rightarrow 0} SC^{MKV} < \infty$  (ToDo: need to justify why it is also finite!). Therefore, combined with equation (18), we obtain

$$\lim_{b_2 \rightarrow 0} PoA = 1 + \frac{\lim_{b_2 \rightarrow 0} \Delta SC}{\lim_{b_2 \rightarrow 0} SC^{MKV}} =: PoA^{0,b_2} > 1.$$

Now, we consider the case when  $b_2 \rightarrow \infty$ . We have  $\lambda \rightarrow \lambda^{\infty,b_2} := \frac{r + \bar{r}(1 - \bar{s})^2}{r + \bar{r}(1 - \bar{s})}$  and

$$B \rightarrow \infty, B^\eta \rightarrow \infty, C^u \rightarrow C_u^{\infty,b_2} := \lambda^{\infty,b_2}(q + \bar{q}(1 - s)), D^u \rightarrow D_u^{\infty,b_2} := \lambda^{\infty,b_2}(q_T + \bar{q}_T(1 - s_T)).$$

Since  $\frac{\delta_u^\pm}{b_2 + \bar{b}_2} \xrightarrow{b_2 \rightarrow \infty} \pm \sqrt{\frac{C_u^{\infty,b_2}}{r + \bar{r}(1 - \bar{s})^2}}$  and therefore,  $e^{-(\delta_u^+ - \delta_u^-)(T-t)} \rightarrow 0$ , for  $t \in [0, T)$ , with equation (22) we have for  $t \in [0, T)$ :

$$(b_2 + \bar{b}_2)u_t = \frac{-\left(\frac{C^u}{b_2 + \bar{b}_2} + D^u \cdot \frac{\delta_u^+}{b_2 + \bar{b}_2}\right) + \left(\frac{C^u}{b_2 + \bar{b}_2} + D^u \cdot \frac{\delta_u^-}{b_2 + \bar{b}_2}\right) e^{-(\delta_u^+ - \delta_u^-)(T-t)}}{\left(\frac{\delta_u^-}{(b_2 + \bar{b}_2)^2} - \frac{D^u}{r + \bar{r}(1 - \bar{s})^2}\right) - \left(\frac{\delta_u^+}{(b_2 + \bar{b}_2)^2} - \frac{D^u}{r + \bar{r}(1 - \bar{s})^2}\right) e^{-(\delta_u^+ - \delta_u^-)(T-t)}} \rightarrow \sqrt{(r + \bar{r}(1 - \bar{s})^2)C_u^{\infty,b_2}}.$$

Similarly, we have  $(b_2 + \bar{b}_2)w_t \rightarrow \sqrt{(r + \bar{r}(1 - \bar{s})^2)C^w}$  and  $b_2\eta_t \rightarrow \sqrt{(r + \bar{r})C^\eta}$  for  $t \in [0, T)$ .

The next step is to look at the order of magnitude in terms of  $b_2$  for the mean  $\bar{x}_t^{MFG}$  and the variance  $v_t$ . For large  $b_2$  and  $t \in [0, T)$ , we have:

$$B \cdot u_t = \frac{b_2 + \bar{b}_2}{r + \bar{r}(1 - \bar{s})^2} (b_2 + \bar{b}_2)u_t \sim \sqrt{\frac{q + \bar{q}(1 - s)}{r + \bar{r}(1 - \bar{s})}} (b_2 + \bar{b}_2) =: \kappa_u \cdot (b_2 + \bar{b}_2).$$

Thus, from equation (6) we have the approximation:

$$\bar{x}_t^{MFG} = \mathbb{E}(\xi) e^{\int_0^t (b_1 + \bar{b}_1 - B u_s) ds} \sim \mathbb{E}(\xi) e^{(b_1 + \bar{b}_1 - \kappa_u(b_2 + \bar{b}_2))t}$$

Similarly, we have for large  $b_2$  and  $t \in [0, T)$ :

$$B^\eta \cdot \eta_t = \frac{b_2}{r + \bar{r}} b_2 \eta_t \sim \sqrt{\frac{q + \bar{q}}{r + \bar{r}}} b_2 =: \kappa_v b_2,$$

and:

$$\begin{aligned} v_t &= Var(\xi) e^{2 \int_0^t (b_1 - B^\eta \eta_u) du} + \sigma^2 \int_0^t e^{2 \int_s^t (b_1 - B^\eta \eta_u) du} ds \\ &\sim Var(\xi) e^{(2b_1 - 2\kappa_v b_2)t} + \frac{\sigma^2}{2b_1 - 2\kappa_v b_2} \left( e^{(2b_1 - 2\kappa_v b_2)t} - 1 \right). \end{aligned}$$

Hence, using Proposition 2, for large  $b_2$ , we deduce:

$$\begin{aligned} \Delta SC &\sim \frac{1}{2} (\sqrt{C^{\infty,b_2}} - \sqrt{C^w})^2 \int_0^T (\bar{x}_t^{MFG})^2 dt \\ &= \frac{(\sqrt{C^{\infty,b_2}} - \sqrt{C^w})^2 \cdot (\mathbb{E}(\xi))^2}{2(2b_1 + 2\bar{b}_1 - 2\kappa_u(b_2 + \bar{b}_2))} \left( e^{(2b_1 + 2\bar{b}_1 - 2\kappa_u(b_2 + \bar{b}_2))T} - 1 \right) \\ &\sim \frac{\zeta_1}{b_2} e^{-2\kappa_u T b_2} + \frac{\zeta_2}{b_2}, \end{aligned}$$



and using equation (16), we deduce:

$$\begin{aligned}
SC^{MKV} &\sim \frac{1}{2}(q + \bar{q}) \int_0^T v_t dt + \frac{1}{2}(q_T + \bar{q}_T)v_T + \frac{1}{2}w_0(\mathbb{E}(\xi))^2 \\
&\sim \frac{(q + \bar{q})}{2} \left[ \left( \frac{Var(\xi)}{2b_1 - 2\kappa_v b_2} + \frac{\sigma^2}{(2b_1 - 2\kappa_v b_2)^2} \right) e^{2b_1 T - 2\kappa_v T b_2} - \left( \frac{Var(\xi) + \sigma^2 T}{2b_1 - 2\kappa_v b_2} + \frac{\sigma^2}{(2b_1 - 2\kappa_v b_2)^2} \right) \right] \\
&\quad + \frac{(q_T + \bar{q}_T)}{2} \left[ Var(\xi) e^{2b_1 T - 2\kappa_v T b_2} + \frac{\sigma^2}{2b_1 - 2\kappa_v b_2} (e^{2b_1 T - 2\kappa_v T b_2} - 1) \right] \\
&\quad + \frac{(\mathbb{E}(\xi))^2 \sqrt{(r + \bar{r}(1 - \bar{s})^2) C^w}}{2(b_2 + \bar{b}_2)} \\
&\sim \left( \frac{\zeta_3}{b_2^2} + \frac{\zeta_4}{b_2} + \zeta_5 \right) e^{-2\kappa_v T b_2} + \left( \frac{\zeta_6}{b_2^2} + \frac{\zeta_7}{b_2} \right),
\end{aligned}$$

where  $(\zeta_i)_{i=1,\dots,7}$  are constant w.r.t.  $b_2$ . In particular

$$\zeta_2 = \frac{\left( \sqrt{\frac{(r + \bar{r}(1 - \bar{s})^2)(q + \bar{q}(1 - s))}{r + \bar{r}(1 - \bar{s})}} - \sqrt{q + \bar{q}(1 - s)^2} \right)^2 \cdot (\mathbb{E}(\xi))^2}{4\kappa_u} \geq 0,$$

and

$$\zeta_7 = \frac{(q + \bar{q})(Var(\xi) + \sigma^2 T) + \sigma^2(q_T + \bar{q}_T)}{4\kappa_v} + \frac{1}{2}(\mathbb{E}(\xi))^2 \sqrt{(r + \bar{r}(1 - \bar{s})^2)(q + \bar{q}(1 - s)^2)} > 0.$$

Hence, for large  $b_2$  we have:

$$\frac{\Delta SC}{SC^{MKV}} \sim \zeta_2 \zeta_7.$$

Therefore, if  $\zeta_2 \neq 0$ , namely:

$$\frac{r + \bar{r}(1 - \bar{s})^2}{r + \bar{r}(1 - \bar{s})} \neq \frac{q + \bar{q}(1 - s)^2}{q + \bar{q}(1 - s)},$$

then we deduce:

$$\lim_{b_2 \rightarrow \infty} PoA =: PoA^{\infty, b_2} > 1;$$

otherwise

$$\lim_{b_2 \rightarrow \infty} PoA = 1.$$

□

**Proposition 5.** Assuming (12), the initial condition  $\xi$  satisfies  $\mathbb{E}(\xi) \neq 0$ , and assuming  $\bar{b}_1 \neq 0$ , then

$$\lim_{\bar{b}_2 \rightarrow 0} PoA =: PoA^{0, \bar{b}_2} > 1 \quad \text{and} \quad \lim_{\bar{b}_2 \rightarrow \infty} PoA = 1.$$

*Proof.* For the case  $\bar{b}_2 \rightarrow 0$ , we have  $\lambda \rightarrow \lambda^{0, \bar{b}_2} := \frac{r + \bar{r}(1 - \bar{s})^2}{r + \bar{r}(1 - \bar{s})}$  and

$$B \rightarrow B^{0, \bar{b}_2} := \frac{b_2^2}{r + \bar{r}(1 - \bar{s})^2} \neq 0, \quad C^u \rightarrow C_u^{0, \bar{b}_2} > 0, \quad D^u \rightarrow D_u^{0, \bar{b}_2} \geq 0.$$

Thus, the existence and uniqueness for solutions of the Riccati equations (19) and (20) associated with  $u_t$  and  $w_t$ , respectively, in the limiting region  $\bar{b}_2 \rightarrow 0$  still hold. Let's denote the limiting solutions by  $u_t^{0,\bar{b}_2}, w_t^{0,\bar{b}_2}$ . By the bounded convergence theorem, the mean function  $\bar{x}_t^{MFG}$  has the limit:

$$\lim_{\bar{b}_2 \rightarrow 0} \bar{x}_t^{MFG} = \mathbb{E}(\xi) e^{\int_0^t (b_1 + \bar{b}_1 - B^{0,\bar{b}_2} \cdot u_s^{0,\bar{b}_2}) ds} =: \bar{x}_t^{0,\bar{b}_2,MFG}.$$

By our assumption on  $\bar{b}_1$ , we have  $A^u \neq A^w$  and thus,  $u_t^{0,\bar{b}_2} \neq w_t^{0,\bar{b}_2}$  on a set of positive Lebesgue measure. Thus, by the bounded convergence theorem and the assumption on  $\xi$ , we deduce

$$\lim_{\bar{b}_2 \rightarrow 0} \Delta SC = \frac{1}{2} B^{0,\bar{b}_2} \int_0^T \left( (u_t^{0,\bar{b}_2} - w_t^{0,\bar{b}_2}) \cdot \bar{x}_t^{0,\bar{b}_2,MFG} \right)^2 dt > 0.$$

Meanwhile,  $\eta_t$  does not depend on  $\bar{b}_2$ , and therefore the variance  $v_t$  also does not depend on  $\bar{b}_2$ . By the explicit formula,  $0 < v_t < \infty$  for  $t > 0$  and  $0 \leq w_0^{0,\bar{b}_2} < \infty$ , and thus,

$$\begin{aligned} \lim_{\bar{b}_2 \rightarrow 0} SC^{MKV} &= h_{var} + w_0^{0,\bar{b}_2} (\mathbb{E}(\xi))^2, \\ \lim_{\bar{b}_2 \rightarrow 0} SC^{MKV} &< \infty, \\ \lim_{\bar{b}_2 \rightarrow 0} SC^{MKV} &> 0. \end{aligned}$$

Hence, we deduce:

$$\lim_{\bar{b}_2 \rightarrow 0} PoA =: PoA^{0,\bar{b}_2} > 1.$$

Now, for the case when  $\bar{b}_2 \rightarrow \infty$ , we have  $\lambda \rightarrow 0$  and  $B \rightarrow \infty$ ,  $C^u \rightarrow 0$ ,  $D^u \rightarrow 0$ . Moreover, if we consider the magnitude w.r.t.  $\bar{b}_2$ , we have:

$$\lambda = \mathcal{O}(\bar{b}_2^{-1}), \quad C^u = \mathcal{O}(\bar{b}_2^{-1}), \quad D^u = \mathcal{O}(\bar{b}_2^{-1}), \quad B = \mathcal{O}(\bar{b}_2^2),$$

and

$$\frac{\delta_u^\pm}{\sqrt{b_2 + \bar{b}_2}} \xrightarrow{\bar{b}_2 \rightarrow \infty} \pm \sqrt{\frac{b_2(q + \bar{q}(1-s))}{r + \bar{r}(1-\bar{s})}} =: \pm \delta^{\infty, \bar{b}_2}.$$

Thus, from equation (22), for  $t \in [0, T]$  we can deduce:

$$\begin{aligned} (b_2 + \bar{b}_2)^{\frac{3}{2}} \cdot u_t &= \frac{- \left( C^u \sqrt{b_2 + \bar{b}_2} + \zeta \frac{\delta_u^+}{\sqrt{b_2 + \bar{b}_2}} \right) + \left( C^u \sqrt{b_2 + \bar{b}_2} + \zeta \frac{\delta_u^-}{\sqrt{b_2 + \bar{b}_2}} \right) e^{-(\delta_u^+ - \delta_u^-)(T-t)}}{\left( \frac{\delta_u^-}{b_2 + b_2} - \frac{b_2(q_T + \bar{q}_T(1-s_T))}{r + \bar{r}(1-\bar{s})} \right) - \left( \frac{\delta_u^+}{b_2 + b_2} - \frac{b_2(q_T + \bar{q}_T(1-s_T))}{r + \bar{r}(1-\bar{s})} \right) e^{-(\delta_u^+ - \delta_u^-)(T-t)}}} \\ &\xrightarrow{\bar{b}_2 \rightarrow \infty} \delta^{\infty, \bar{b}_2} (r + \bar{r}(1-\bar{s})^2) \end{aligned}$$

where  $\zeta = \frac{b_2(r + \bar{r}(1-\bar{s})^2)}{r + \bar{r}(1-\bar{s})}(q_T + \bar{q}_T(1-s_T))$ . Thus, we have  $B \cdot u_t = \mathcal{O}(\bar{b}_2^{1/2})$  for  $t \in [0, T]$  and

$$\bar{x}_t^{MFG} = \mathbb{E}(\xi) e^{\int_0^t (b_1 + \bar{b}_1 - B \cdot u_s) ds} \xrightarrow{\bar{b}_2 \rightarrow \infty} 0, \quad \forall t \in (0, T].$$

As in the proof of Proposition 4, for  $t \in [0, T]$ :

$$(b_2 + \bar{b}_2)w_t \rightarrow \sqrt{(r + \bar{r}(1 - \bar{s})^2)C^w}$$

so that  $w_t \rightarrow 0, \forall t \in [0, T]$  as  $\bar{b}_2 \rightarrow \infty$ . Hence, by the bounded convergence theorem, we deduce:

$$\lim_{\bar{b}_2 \rightarrow \infty} \Delta SC = \frac{1}{2(r + \bar{r}(1 - \bar{s})^2)} \int_0^T \lim_{\bar{b}_2 \rightarrow \infty} \left( \frac{(b_2 + \bar{b}_2)^{\frac{3}{2}} u_t}{\sqrt{b_2 + \bar{b}_2}} - (b_2 + \bar{b}_2)w_t \right)^2 \cdot (\bar{x}_t^{MFG})^2 dt \rightarrow 0.$$

Moreover, as shown in the case  $\bar{b}_2 \rightarrow 0$ ,  $\eta_t$  as well as the variance  $v_t$  are invariant w.r.t.  $\bar{b}_2$  and  $v_t > 0$  for  $t > 0$ . Thus,

$$\lim_{\bar{b}_2 \rightarrow \infty} SC^{MKV} = h_{var} > 0.$$

Therefore, we conclude that:

$$\lim_{\bar{b}_2 \rightarrow \infty} PoA = 1.$$

□

**Proposition 6.** *Assuming (12), the initial condition  $\xi$  satisfies  $\mathbb{E}(\xi) \neq 0$ , and  $\lambda \neq 1$ , then*

$$\begin{aligned} \lim_{b_1 \rightarrow 0} PoA =: PoA^{0,b_1} > 1 \quad \text{and} \quad \lim_{\bar{b}_1 \rightarrow 0} PoA =: PoA^{0,\bar{b}_1} > 1, \\ \lim_{b_1 \rightarrow \infty} PoA = 1 \quad \text{and} \quad \lim_{\bar{b}_1 \rightarrow \infty} PoA = \infty. \end{aligned}$$

*Proof.* Clearly we have  $\lim_{b_1 \rightarrow 0} u_t =: u_t^{0,b_1}$ ,  $\lim_{b_1 \rightarrow 0} w_t =: w_t^{0,b_1}$ , and  $\lim_{b_1 \rightarrow 0} \eta_t =: \eta_t^{0,b_1}$ , where all three limiting processes are bounded. By the assumption on  $\lambda$ ,  $u_t \neq w_t$  on a set of positive Lebesgue measure. By the assumption on  $\xi$ , we have  $\lim_{b_1 \rightarrow 0} \bar{x}_t^{MFG} =: \bar{x}_t^{0,b_1,MFG} > 0, \forall t \in [0, T]$  which is also bounded. Therefore, by Proposition 2, and the bounded convergence theorem,

$$\lim_{b_1 \rightarrow 0} \Delta SC =: \Delta SC^{0,b_1}, \Delta SC^{0,b_1} < \infty, \text{ and } \Delta SC^{0,b_1} > 0.$$

Since  $\eta_t^{0,b_1}$  is bounded,  $\lim_{b_1 \rightarrow 0} v_t =: v_t^{0,b_1} > 0, \forall t > 0$  which is also bounded. Using equations (15) and (16), the fact that  $0 \leq \eta_0^{0,b_1} < \infty$ , and the bounded convergence theorem,  $\lim_{b_1 \rightarrow 0} SC^{MKV} > 0$  and  $\lim_{b_1 \rightarrow 0} SC^{MKV} < \infty$ . Therefore,  $\lim_{b_1 \rightarrow 0} PoA =: PoA^{0,b_1} > 1$ . The proof can be repeated to show  $\lim_{\bar{b}_1 \rightarrow 0} PoA =: PoA^{0,\bar{b}_1} > 1$ .

Now we consider  $b_1 \rightarrow \infty$ . We see that:

$$\delta_u^+ \rightarrow 0, \frac{\delta_u^-}{b_1 + \frac{\bar{b}_1}{2}} \rightarrow -2, \delta_w^+ \rightarrow 0, \frac{\delta_w^-}{b_1 + \bar{b}_1} \rightarrow -2, \delta_\eta^+ \rightarrow 0, \frac{\delta_\eta^-}{b_1} \rightarrow -2.$$

and therefore we find for  $t \in [0, T]$ :

$$\begin{aligned} \left( b_1 + \frac{\bar{b}_1}{2} \right) u_t &\rightarrow 2C^u \\ (b_1 + \bar{b}_1) w_t &\rightarrow 2C^w \\ b_1 \eta_t &\rightarrow 2C^\eta. \end{aligned}$$

Thus,  $\Delta\bar{\eta}_t = \mathcal{O}(b_1^{-1})$  for  $t \in [0, T]$ . We find  $\bar{x}_t^{MFG} = \mathcal{O}(e^{b_1})$  for  $t \in (0, T]$ . Thus, using Proposition 2, we have  $\Delta SC = \mathcal{O}\left(\frac{e^{b_1}}{b_1}\right)$ . We also find  $v_t = \mathcal{O}(e^{2b_1})$ . Therefore with equation (16) we have:

$$\frac{\Delta SC}{SC^{MKV}} = \mathcal{O}\left(\frac{e^{-b_1}}{b_1}\right) \rightarrow 0.$$

We conclude that  $PoA \rightarrow 1$  as  $b_1 \rightarrow \infty$ .

Similarly, for the case  $\bar{b}_1 \rightarrow \infty$ , we find that  $\Delta SC = \mathcal{O}\left(\frac{e^{\bar{b}_1}}{\bar{b}_1}\right)$  and  $SC^{MKV} = \mathcal{O}(1)$ . We conclude that  $PoA \rightarrow \infty$  as  $\bar{b}_1 \rightarrow \infty$ .  $\square$

## 2.4 Numerical Results

The solutions to the problems we have considered are given by the formulas derived in Appendix A, which are explicit up to evaluating integrals. Using the simple rectangle rule to evaluate integrals, we numerically compute the price of anarchy when the coefficients are time-independent, nonnegative, and satisfy assumption (12). In particular, when we allow for full interaction (i.e. through the states and the controls), we choose the following default values:

$$\begin{aligned} \xi &\equiv 1, \quad T = 1 \\ b_1 &= 1, \quad \bar{b}_1 = 1, \quad b_2 = 1, \quad \bar{b}_2 = 1, \quad \sigma = 1 \\ q &= 1, \quad \bar{q} = 1, \quad s = 0.5, \quad r = 1, \quad \bar{r} = 1, \quad \bar{s} = 0.5 \\ q_T &= 1, \quad \bar{q}_T = 1, \quad s_T = 0.5. \end{aligned}$$

Unless otherwise stated, the parameters stay at these default values. For results involving only interaction through the states, we set  $\bar{b}_2 = 0$  and  $\bar{r} = 0$ . For results involving only interaction through the controls, we set  $\bar{b}_1 = 0$ ,  $\bar{q} = 0$ , and  $\bar{q}_T = 0$ . Figures 1-5 show the price of anarchy as we vary one parameter at a time for each of three cases: full interaction (i.e. through the states and the controls), interaction only through the states, and interaction only through the controls.

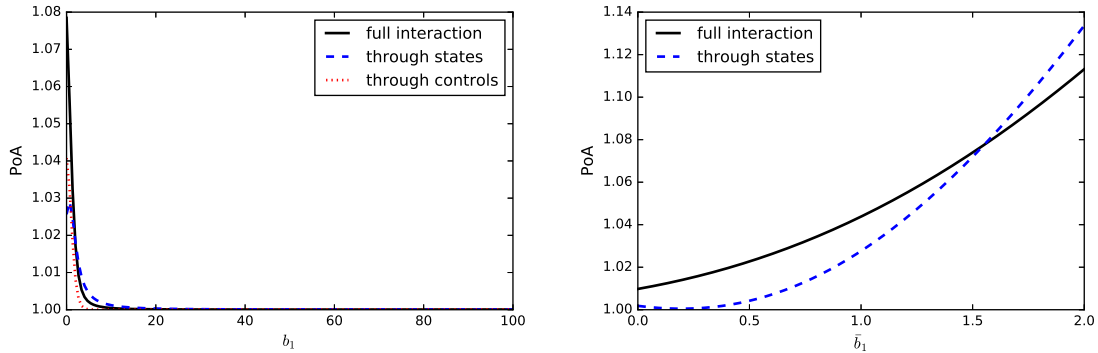


Figure 1: PoA as we vary  $b_1$  (left) and  $\bar{b}_1$  (right).

The results show various limiting behaviors, such as some of the cases proved in the previous section. In Figure 1, we note that Proposition 6 is confirmed. For all three cases, as  $b_1 \rightarrow 0$ , we see that  $PoA \rightarrow PoA^{0,b_1} > 1$  and as  $\bar{b}_1 \rightarrow 0$ , we see that  $PoA \rightarrow PoA^{0,\bar{b}_1} > 1$ . We also see

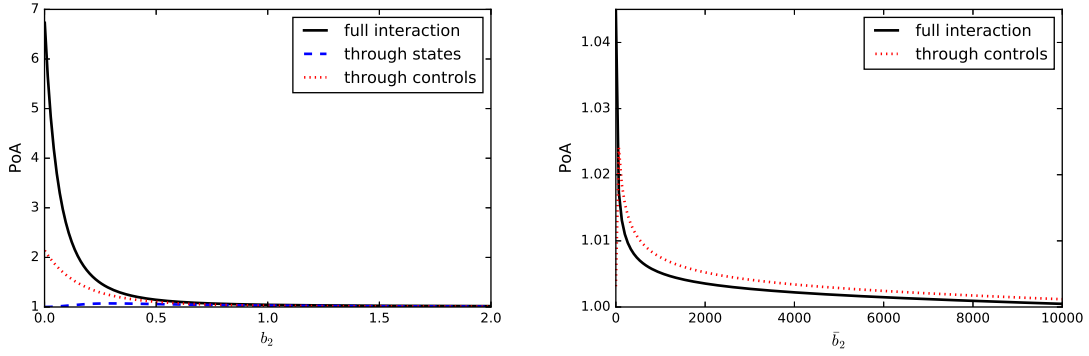


Figure 2: PoA as we vary  $b_2$  (left) and  $\bar{b}_2$  (right).

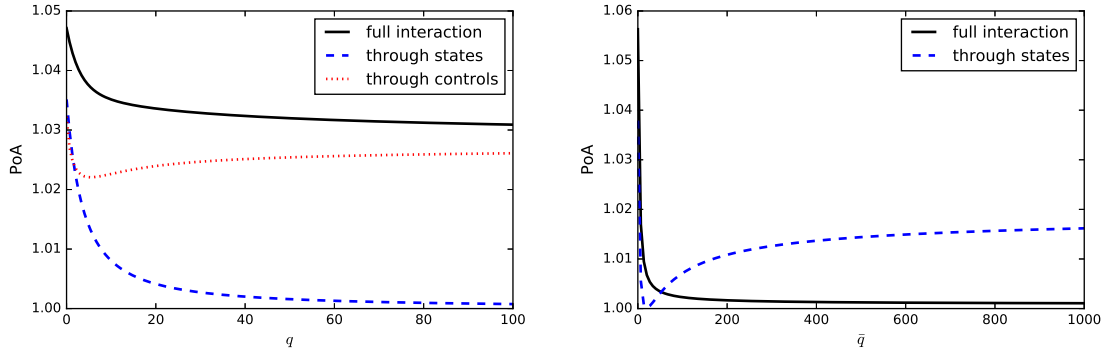


Figure 3: PoA as we vary  $q$  (left) and  $q_T$  (right).

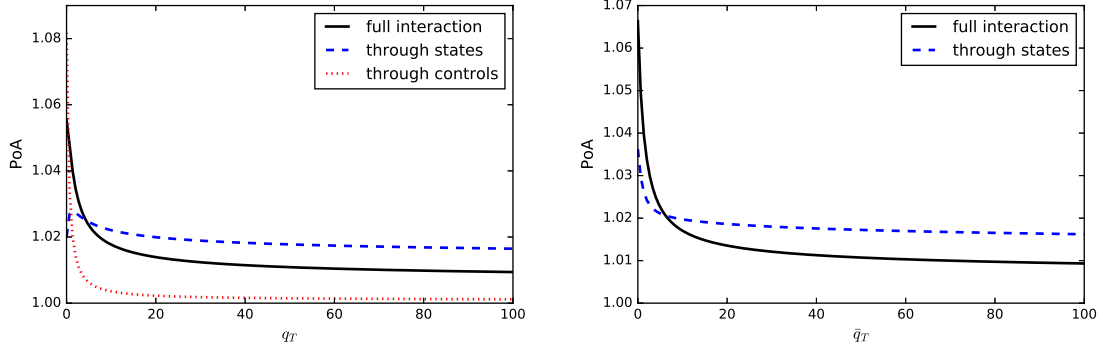


Figure 4: PoA as we vary  $\bar{q}$  (left) and  $\bar{q}_T$  (right).

that  $PoA \rightarrow 1$  as  $b_1 \rightarrow \infty$  and  $PoA \rightarrow \infty$  as  $\bar{b}_1 \rightarrow \infty$ . Proposition 4 is confirmed in Figure 2. When there is only interaction through the states, then  $\bar{b}_2 = 0$  and we see that  $PoA \rightarrow 1$  as  $b_2 \rightarrow 0$ . When there is full interaction or only interaction through the controls, then  $\bar{b}_2 \neq 0$  and we see that  $PoA \rightarrow PoA^{0,b_2} > 1$  as  $b_2 \rightarrow 0$ . For all three cases, we note that the condition  $\frac{r+\bar{r}(1-\bar{s})^2}{r+\bar{r}(1-\bar{s})} = \frac{q+\bar{q}(1-s)^2}{q+\bar{q}(1-s)}$  is satisfied, and thus,  $PoA \rightarrow 1$  as  $b_2 \rightarrow \infty$ . For Proposition 5, Figure 2



Figure 5: PoA as we vary  $r$  (left) and  $\bar{r}$  (right).

confirms that  $PoA \rightarrow PoA^{0, \bar{b}_2} > 1$  as  $\bar{b}_2 \rightarrow 0$  and  $PoA \rightarrow 1$  as  $\bar{b}_2 \rightarrow \infty$ . Figure 5 confirms as in Proposition 3 that  $PoA \rightarrow 1$  as  $r \rightarrow \infty$  or  $\bar{r} \rightarrow \infty$ .

## 2.5 A Particular Example: Flocking

Mean field game models of flocking have been proposed in the literature by Nourian et al [14][15]. Here we consider a slightly different formulation as described in Section 3.6.1 of the book [7]. A representative bird in the flock controls their velocity,  $X_t$ , through the drift:

$$b(t, x, \mu, \alpha, \nu) = \alpha.$$

They choose the control with two goals in mind: to minimize their kinetic energy put into their control, and to align their velocity with the average velocity of the group. Thus, they consider the cost functions given by:

$$f(t, x, \mu, \alpha, \nu) = \frac{1}{2} (\bar{q}|x - s\bar{\mu}|^2 + \alpha^2),$$

$$g(x, \mu) = 0.$$

In our general linear quadratic formulation, this is equivalent to taking  $b_1 = 0$ ,  $\bar{b}_1 = 0$ ,  $b_2 = 1$ ,  $\bar{b}_2 = 0$ ,  $q = 0$ ,  $s = 1$ ,  $r = 1$ ,  $\bar{r} = 0$ , and  $\bar{s} = 0$ . Note that for these values of the parameters, the assumptions of Theorem 1 and Corollary 1 are satisfied. Therefore,  $PoA = 1$ . In fact, if we took the state space to be  $\mathbb{R}^d$  instead of  $\mathbb{R}$ , the result would still hold.

## 3 Conclusion

We defined the price of anarchy ( $PoA$ ) in the context of extended mean field games as the ratio of the worst case social cost when the players are in a mean field game equilibrium to the social cost as computed by a central planner. Since the central planner does not require that the players be in a mean field game equilibrium, the central planner will realize a social cost that is no worse than that of a mean field game equilibrium. Thus,  $PoA \geq 1$ .

We computed the price of anarchy for linear quadratic extended mean field games, for which explicit computations are possible. We identify a large class of models for which  $PoA = 1$  (see

Proposition 1 and Corollary 1), as well as some limiting cases where  $PoA \rightarrow 1$  as certain parameters tend to zero or to infinity. The numerics support our theoretical results.

## 4 Acknowledgments

The authors would like to thank the organizers of CEMRACS 2017 for putting together such a successful summer school.

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## Appendices

### A Solving Linear FBSDEs of McKean-Vlasov Type

Consider a linear FBSDE system of McKean-Vlasov type:

$$\begin{aligned}
dX_t &= (a_t^x X_t + a_t^{\bar{x}} \mathbb{E}X_t + a_t^y Y_t + a_t^{\bar{y}} \mathbb{E}Y_t) dt + \sigma dW_t \\
X_0 &= \xi \\
dY_t &= (b_t^x X_t + b_t^{\bar{x}} \mathbb{E}X_t + b_t^y Y_t + b_t^{\bar{y}} \mathbb{E}Y_t) dt + Z_t dW_t \\
Y_T &= c^x X_T + c^{\bar{x}} \mathbb{E}X_T.
\end{aligned} \tag{25}$$

For the LQEMFG model considered in Section 2.1, the FBSDE system in equation (2) is of the form of equation (25) if we set:

$$\begin{aligned}
a_t^x &= b_1(t), \quad a_t^{\bar{x}} = \bar{b}_1(t), \quad a_t^y = a^{MFG}(t)b_2(t), \quad a_t^{\bar{y}} = b^{MFG}(t)b_2(t) + c^{MFG}(t)\bar{b}_2(t) \\
b_t^x &= -(q(t) + \bar{q}(t)), \quad b_t^{\bar{x}} = \bar{q}(t)s(t), \quad b_t^y = -b_1(t), \quad b_t^{\bar{y}} = 0 \\
c_t^x &= q_T + \bar{q}_T, \quad c_t^{\bar{x}} = -\bar{q}_T s_T.
\end{aligned}$$

For the LQEMKV model considered in Section 2.2, the FBSDE system in equation (8) is of the form of equation (25) if we set:

$$\begin{aligned}
a_t^x &= b_1(t), \quad a_t^{\bar{x}} = \bar{b}_1(t), \quad a_t^y = a^{MKV}(t)b_2(t), \quad a_t^{\bar{y}} = b^{MKV}(t)b_2(t) + c^{MKV}(t)\bar{b}_2(t) \\
b_t^x &= -(q(t) + \bar{q}(t)), \quad b_t^{\bar{x}} = -s(t)\bar{q}(t)(s(t) - 2), \quad b_t^y = -b_1(t), \quad b_t^{\bar{y}} = -\bar{b}_1 \\
c_t^x &= q_T + \bar{q}_T, \quad c_t^{\bar{x}} = s_T \bar{q}_T (s_T - 2).
\end{aligned}$$



Now we return to the general FBSDE system (25). By taking expectations in equation (25), and letting  $\bar{x}_t$  and  $\bar{y}_t$  denote  $\mathbb{E}X_t$  and  $\mathbb{E}Y_t$ , respectively, we get:

$$\begin{aligned}\dot{\bar{x}}_t &= (a_t^x + a_t^{\bar{x}})\bar{x}_t + (a_t^y + a_t^{\bar{y}})\bar{y}_t \\ \bar{x}_0 &= \mathbb{E}(\xi) \\ \dot{\bar{y}}_t &= (b_t^x + b_t^{\bar{x}})\bar{x}_t + (b_t^y + b_t^{\bar{y}})\bar{y}_t \\ \bar{y}_T &= (c^x + c^{\bar{x}})\bar{x}_T\end{aligned}\tag{26}$$

where the dot is the standard ODE notation for a derivative. We then make the ansatz  $\bar{y}_t = \bar{\eta}_t \bar{x}_t + \bar{\chi}_t$  for deterministic functions  $[0, T] \ni t \mapsto \bar{\eta}_t \in \mathbb{R}$  and  $[0, T] \ni t \mapsto \bar{\chi}_t \in \mathbb{R}$ . By plugging in the ansatz, the system in equation (26) is equivalent to the ODE system:

$$\begin{aligned}\dot{\bar{\eta}}_t + (a_t^y + a_t^{\bar{y}})\bar{\eta}_t^2 + (a_t^x + a_t^{\bar{x}} - b_t^y - b_t^{\bar{y}})\bar{\eta}_t - b_t^x - b_t^{\bar{x}} &= 0 \\ \bar{\eta}_T - c^x - c^{\bar{x}} &= 0 \\ \dot{\bar{\chi}}_t + (\bar{\eta}_t(a_t^y + a_t^{\bar{y}}) - b_t^y - b_t^{\bar{y}})\bar{\chi}_t &= 0 \\ \bar{\chi}_T &= 0.\end{aligned}$$

The first equation is a Riccati equation. Note that  $\bar{\chi}_t$  solves a first order homogeneous linear equation. Thus  $\bar{\chi}_t = 0, \forall t \in [0, T]$ . Once the equation for  $\bar{\eta}_t$  is solved, we can compute  $\bar{x}_t$  by solving the linear ODE:

$$\begin{aligned}\dot{\bar{x}}_t &= (a_t^x + a_t^{\bar{x}} + (a_t^y + a_t^{\bar{y}})\bar{\eta}_t)\bar{x}_t \\ \bar{x}_0 &= \mathbb{E}(\xi)\end{aligned}$$

and thus,

$$\bar{x}_t = \mathbb{E}(\xi) e^{\int_0^t (a_u^x + a_u^{\bar{x}} + (a_u^y + a_u^{\bar{y}})\bar{\eta}_u) du}.$$

Once we have computed  $(\bar{x}_t)_{0 \leq t \leq T}$ , we can rewrite the original FBSDE system:

$$\begin{aligned}dX_t &= (a_t^x X_t + a_t^y Y_t + a_t^0) dt + \sigma dW_t \\ X_0 &= \xi \\ dY_t &= (b_t^x X_t + b_t^y Y_t + b_t^0) dt + Z_t dW_t \\ Y_T &= c^x X_T + c^0\end{aligned}$$

with:

$$\begin{aligned}a_t^0 &= (a_t^{\bar{x}} + a_t^{\bar{y}}\bar{\eta}_t)\bar{x}_t \\ b_t^0 &= (b_t^{\bar{x}} + b_t^{\bar{y}}\bar{\eta}_t)\bar{x}_t \\ c^0 &= c^{\bar{x}}\bar{x}_T.\end{aligned}$$

Now we make the ansatz:  $Y_t = \eta_t X_t + \chi_t$ , which reduces the problem to the ODE system:

$$\begin{aligned}\dot{\eta}_t + a_t^y \eta_t^2 + (a_t^x - b_t^y)\eta_t - b_t^x &= 0 \\ \eta_T &= c^x, \\ \dot{\chi}_t + (-b_t^y + a_t^y \eta_t)\chi_t + a_t^0 \eta_t - b_t^0 &= 0, \\ \chi_T &= c^0, \\ Z_t &= \sigma \eta_t.\end{aligned}$$

Again, the first equation is a Riccati equation. Note that it is not necessary to solve for  $\chi_t$  because of the relationship:

$$\bar{\eta}_t \bar{x}_t = \bar{y}_t = \mathbb{E}(Y_t) = \mathbb{E}(\eta_t X_t + \chi_t) = \eta_t \bar{x}_t + \chi_t.$$

Thus,

$$\chi_t = (\bar{\eta}_t - \eta_t) \bar{x}_t.$$

In summary, the solution to the linear FBSDE of McKean-Vlasov type is reduced to solving linear ODEs and Riccati equations. It will also be useful to compute  $Var(X_t)$ , which we denote by  $v_t$ . After we have solved the above equations, we have

$$\begin{aligned} dX_t &= ((a_t^x + a_t^y \eta_t) X_t + a_t^y \chi_t + a_t^0) dt + \sigma dW_t \\ X_0 &= \xi. \end{aligned}$$

Thus,

$$v_t = Var(X_t) = Var(\xi) e^{\int_0^t 2(a_s^x + a_s^y \eta_s) ds} + \sigma^2 \int_0^t e^{2 \int_s^t (a_u^x + a_u^y \eta_u) du} ds.$$

In the case where the coefficients are time-independent, the Riccati equations for  $\bar{\eta}_t$  and  $\eta_t$  can be solved explicitly.

## Scalar Riccati Equation

If the scalar Riccati equation

$$\dot{\rho}_t - B\rho_t^2 - 2A\rho_t + C = 0$$

with terminal condition  $\rho_T = D$  satisfies:

$$B \neq 0, BD \geq 0, BC > 0, \tag{27}$$

then it has a unique solution:

$$\rho_t = \frac{-(C + D\delta^+) + (C + D\delta^-)e^{-(\delta^+ - \delta^-)(T-t)}}{(\delta^- - DB) - (\delta^+ - DB)e^{-(\delta^+ - \delta^-)(T-t)}} \tag{28}$$

where  $\delta^\pm = -A \pm \sqrt{(A)^2 + BC}$ .

Furthermore, if  $B \rightarrow 0$  and  $A \neq 0$ , we can deduce that the limiting solution of the scalar Riccati equation coincides with the first-order differential equation:

$$\dot{\rho}_t - 2A\rho_t + C = 0$$

with terminal condition  $\rho_T = D$ , namely:

$$\rho_t = \left(D - \frac{C}{2A}\right) e^{-2A(T-t)} + \frac{C}{2A}.$$

If  $B \rightarrow 0$  and  $A = 0$ , the limiting solution of the scalar Riccati equation coincides with the first-order differential equation:

$$\dot{\rho}_t + C = 0$$

with terminal condition  $\rho_T = D$ , namely:

$$\rho_t = D + C(T - t).$$

Hence, returning to the linear FBSDE (25), for  $\bar{\eta}_t$ , we use:

$$\begin{cases} A = -\frac{1}{2}(a^x + a^{\bar{x}} - b^y - b^{\bar{y}}) \\ B = -(a^y + a^{\bar{y}}) \\ C = -(b^x + b^{\bar{x}}) \\ D = c^x + c^{\bar{x}} \end{cases}$$

The conditions (27) are satisfied if:

$$\begin{cases} -(a^y + a^{\bar{y}}) > 0 \\ -(b^x + b^{\bar{x}}) > 0 \\ c^x + c^{\bar{x}} \geq 0 \end{cases}$$

For  $\eta_t$ , we use:

$$\begin{cases} A = -\frac{1}{2}(a^x - b^y) \\ B = -a^y \\ C = -b^x \\ D = c^x. \end{cases}$$

The conditions (27) are satisfied if:

$$\begin{cases} -a^y > 0 \\ -b^x > 0 \\ c^x \geq 0. \end{cases}$$

Returning to the LGEMFG and LGEMKV problems, if we assume the coefficients are nonnegative, we see that these conditions are exactly assumption (12).