MEAN-FIELD SDE DRIVEN BY FBM AND RELATED STOCHASTIC CONTROL PROBLEM

RAINER BUCKDAHN AND SHUAI JING

ABSTRACT. We study a class of mean-field stochastic differential equations driven by a fractional Brownian motion with Hurst parameter $H \in (1/2, 1)$ and a related stochastic control problem. We derive a Pontryagin type maximum principle and the associated adjoint mean-field backward stochastic differential equation driven by a classical Brownian motion, and we prove that under certain assumptions, which generalise the classical ones, the necessary condition for the optimality of an admissible control is also sufficient.

1. INTRODUCTION

In this paper we consider a class of mean-field stochastic control problem driven by a fractional Brownian motion with Hurst parameter $H \in (1/2, 1)$ given by

$$X_t^u = x + \int_0^t \sigma \left( \mathbb{P}(X_s^u) \right) dB_s^H + \int_0^t b \left( \mathbb{P}(X_s^u, u_s), X_s^u, u_s \right) ds,$$

where $x \in \mathbb{R}$, and $u \in \mathcal{U}([0, T])$ is an adapted control process taking values in a convex open set in $\mathbb{R}^m$, $\mathbb{P}(X_s^u)$ is the law of $X_s^u$ and $\mathbb{P}(X_s^u, u_s)$ is the joint law of $(X_s^u, u_s)$. Our aim is to characterise an optimal control $u^* \in \mathcal{U}([0, T])$ such that

$$J(u^*) = \inf_{u \in \mathcal{U}([0, T])} J(u),$$

where the cost functional has the form

$$J(u) = \mathbb{E} \left[ \int_0^T f \left( \mathbb{P}(X_t^u, u_t), X_t^u, u_t \right) dt + g \left( X_T^u, \mathbb{P}(X_T^u) \right) \right],$$

for some functions $f$ and $g$ specified later.

The mean-field (or McKean-Vlasov type) stochastic differential equation (SDE) driven by classical Brownian motion was introduced by Kac [14] [15] to study the Boltzman equation and the Vlasov kinetic equation. Later Lasry and Lions [16] worked on mean-field stochastic games. Henceforth the applications for mean-field problem attracted wide attention. Buckdahn et al. [4] [6] studied special mean-field games and derived a kind of mean-field BSDEs associated with non local PDEs. Carmona
and Delarue [8] studied the existence and uniqueness of a class of mean-field forward-backward SDEs by applying the continuation method proposed in Peng and Wu [19].

Stochastic control problems driven by a fractional Brownian motion also have been studied by several authors. However, compared with the vast literatures on stochastic control problems driven by classical Brownian motion, few has been done and there are a lot of open questions. The main reason is that fractional Brownian motion is neither a Markov process nor a semi-martingale, hence the classical methods cannot be applied directly here. Biagini et al. [1] obtained a maximum principle for a stochastic control problem driven by an \( m \)-dimensional fractional Brownian motion with Hurst parameter \( H \in (1/2, 1) \). For \( H \in (0, 1/2) \), Hu and Zhou [12] considered a linear stochastic optimal control problem and obtained a Riccati equation, a BSDE driven by the fractional Brownian motion and the underlying Brownian motion. Han et al. [10] obtained a stochastic maximum principle for a control problem driven by a fractional Brownian motion with \( H > 1/2 \) and their adjoint equations is a linear BSDE again driven by the fractional Brownian motion and the underlying Brownian motion. We emphasise that their results need strong assumptions, and in particular, Malliavin differentiability of the optimal control process, which are not easily fulfilled. By applying Girsanov transformation, in [5] we studied a stochastic control system involving both a standard and an independent fractional Brownian motion with Hurst parameter less than 1/2, and we obtained as adjoint equation a BSDE driven by the Brownian motion and an independent martingale.

In this paper, by applying Girsanov transformation, we first prove the existence and the uniqueness result for a mean-field SDE of the form

\[
X_t = \xi + \int_0^t \left( \gamma sX_s + \sigma(s, \mathbb{P}(X_s, \Theta_s)) \right) dB_s^H + \int_0^t b(s, \mathbb{P}(X_s, \Theta_s), X_s) ds,
\]

where \( \xi \) is a square integrable random variable, \( \Theta \) is a given square integrable process and \( \gamma \) is a deterministic function. Then we use these results to consider a stochastic control problem with dynamics \( X \) (for \( \gamma = 0 \) and \( \Theta = u \) an admissible control) and we derive the Pontryagin type maximum principle.

We give a necessary as well as a sufficient condition. The maximum principle leads to a coupled system involving a mean-field forward-backward SDE, where the forward equation is a mean-field SDE driven by the fractional Brownian motion, while the backward equation is a mean-field BSDE driven only by the underlying Brownian motion, with terminal condition depending on the fractional Brownian motion. We also show that, if the time interval is small enough, the mean-field FBSDE is solvable and allows us to get an optimal control and the associated dynamics. A more general discussion of such coupled FBSDEs is foreseen for a forthcoming paper. It is worth noting that our controls are not assumed to be Malliavin differentiable.

The paper is organised as follows: In Section 2 we give some preliminaries on fractional Brownian motion and differentiability for functions of measures. In Section 3 we study the existence and uniqueness of semi-linear mean-field stochastic differential equations driven by a fractional Brownian motion. Our main results on the Pontryagin’s maximum principle are stated in Section 4.

2. Preliminaries

2.1. Fractional Brownian Motion. Let \( T > 0 \) be a fixed horizon. We consider a complete probability space \( (\Omega, \mathcal{F}, \mathbb{P}) \). A fractional Brownian motion \( B^H = \{ B^H_t, t \in [0, T]\} \) with Hurst parameter
$H \in (0, 1)$ is a centred Gaussian process on $(\Omega, \mathcal{F}, \mathbb{P})$ with covariance function

$$R_H(t, s) = \mathbb{E} \left[ B^H_t B^H_s \right] = \frac{1}{2} \left( t^{2H} + s^{2H} - |t - s|^{2H} \right), s, t \in [0, T].$$

For $H \in (1/2, 1)$, it is well known that the fractional Brownian motion has the representation as follows:

$$B^H_t = \int_0^t K_H(t, s)dW_s,$$

where $W$ is a suitable Brownian motion on the space $(\Omega, \mathcal{F}, \mathbb{P})$. The kernel function is given by

$$K_H(t, s) = c_H s^{1/2-H} \int_0^t (u - s)^{H-3/2} u^{H-1/2} du, \quad t > s,$$

with the constant

$$c_H = [H(2H - 1)/\beta(2 - 2H, H - 1/2)]^{1/2},$$

where $\beta(\alpha, \gamma) = \Gamma(\alpha + \gamma)/\Gamma(\alpha)\Gamma(\gamma)$ is the Beta function and $\Gamma(\alpha) = \int_0^\infty x^{\alpha-1}e^{-x}dx$ is the Gamma function.

2.2. Fractional Calculus. For a detailed account on the fractional calculus theory, we refer, for instance, to Biagini et al. [2] and Samko et al. [20].

Let $f : [0, T] \to \mathbb{R}$ be a Lebesgue integrable function, and $\alpha \in (0, 1)$. The fractional Riemann Liouville integrals of $f$ are defined as follows:

The right-sided and left-sided fractional integrals $I_{T-}^\alpha(f)(x)$ and $I_{0+}^\alpha(f)(x)$ of $f$ of order $\alpha$ are given by

$$I_{T-}^\alpha(f)(x) = \frac{1}{\Gamma(\alpha)} \int_x^T \frac{f(u)}{(u - x)^{1-\alpha}} du, \quad \text{for almost all } x \in [0, T].$$

and

$$I_{0+}^\alpha(f)(x) = \frac{1}{\Gamma(\alpha)} \int_0^x \frac{f(u)}{(x - u)^{1-\alpha}} du, \quad \text{for almost all } x \in [0, T].$$

Note that $I_{T-}^\alpha(f)(x)$ and $I_{0+}^\alpha(f)(x)$ are well-defined because the Fubini theorem implies that they are functions in $L^p([0, T]), p \geq 1$, whenever $f \in L^p([0, T])$.

We denote by $I_{T-}^\alpha(L^p)$ (respectively, $I_{0+}^\alpha(L^p)$, $p \geq 1$, the families of all functions $f \in L^p([0, T])$ such that

$$f = I_{T-}^\alpha(\varphi), \text{ (respectively, } f = I_{0+}^\alpha(\varphi)), \text{ for some } \varphi \in L^p([0, T]).$$

Samko et al. [20] (Theorem 13.2) provide a characterization of the space $I_{T-}^\alpha(L^p), p > 1$. The function $\varphi$ satisfying (5) coincides with the right-sided fractional derivative

$$f(x) - f(u) + \alpha \int_x^T \frac{f(x) - f(u)}{(u - x)^{1+\alpha}} du,$$

respectively, the left-sided functional derivative

$$f(x) - f(u) + \alpha \int_0^x \frac{f(x) - f(u)}{(x - u)^{1+\alpha}} du,$$
when the integrals are well defined. Moreover, we have

\begin{equation}
D_{0+}^{\alpha} f = \frac{d}{dx} I_{0+}^{1-\alpha} f,
\end{equation}

and

\begin{equation}
D_{T-}^{\alpha} f = \frac{d}{dx} I_{T-}^{1-\alpha} f,
\end{equation}

if everything is well-defined.

Furthermore, we have the following integration by parts formula for the fractional integrals

\begin{equation}
\int_{0}^{T} I_{0+}^{\alpha} f(x) g(x) dx = \int_{0}^{T} f(x) I_{T-}^{\alpha} g(x) dx,
\end{equation}

if \( f \in L^p[0, T], \ g \in L^q[0, T], \ \frac{1}{p} + \frac{1}{q} \leq 1 + \alpha. \) The corresponding integration by parts formula for the fractional derivatives is

\begin{equation}
\int_{0}^{T} D_{0+}^{\alpha} f(x) g(x) dx = \int_{0}^{T} f(x) D_{T-}^{\alpha} g(x) dx,
\end{equation}

for \( f \in I_{0+}^{\alpha}(L^p[0, T]), \ g \in I_{T-}^{\alpha}(L^q[0, T]), \ \frac{1}{p} + \frac{1}{q} \leq 1 + \alpha. \)

2.3. **Stochastic integrals with respect to fractional Brownian motion.** Most of the results in this section can be found in Biagini et al. [2], Han et al. [10] and Hu [11].

For the kernel function \( K_H(t, s) \), let \( H \) be the set of functions \( f \) which can be represented as

\( f(t) = \int_{0}^{t} K_H(t, s) \hat{f}(s) ds \)

for some \( \hat{f} \in L^2([0, T]) \). We denote by \( \mathcal{E} \) be the space of step functions on \([0, T]\) and define \( \varphi(t, s) = H(2H - 1) |s - t|^{2H - 2} \). We consider the scalar product on \( L^2([0, T]) \):

\( \langle f, g \rangle_H := \int_{0}^{T} \int_{0}^{T} f(s)g(t) \varphi(t, s) dsdt, \)

and we define a linear map \( \mathcal{I} \) on the space \( \mathcal{E} \) by

\[ \mathcal{I} : (L^2([0, T]), \langle \cdot, \cdot \rangle_H) \rightarrow \mathcal{H}, \]

\[ \mathcal{I}[I_{[0, t]}] \mapsto R(t, \cdot). \]

Then the extension of this map to the closure of \( (L^2([0, T]), \langle \cdot, \cdot \rangle_H) \) is a representation of \( \mathcal{H} \). The map \( \mathcal{I} \) also induces the following isometry:

\[ \mathcal{J} : (L^2([0, T]), \langle \cdot, \cdot \rangle_H) \rightarrow L^2(\Omega), \]

\[ \mathcal{I}[I_{[0, t]}] \mapsto B^H_t. \]

This allows to define the Wiener integrals with respect to \( B^H \):

\[ B^H(\psi) := \mathcal{J}(\psi), \quad \psi \in \mathcal{H}. \]

We also use the notations \( B^H(\psi) = \int_{0}^{T} \psi(t) dB^H(t) \) and \( B^H(\psi I_{[0, t]}) = \int_{0}^{t} \psi(s) dB^H_s, t \in [0, T]. \)
We denote by $S$ the set of all polynomial functions of $B^H(\psi_j) = \int_0^T \psi_j(t)dB^H(t)$. For an element $F \in S$, having the form

$$F = g(B^H(\psi_1), \ldots, B^H(\psi_n)),$$

where $g$ is a polynomial of $n$ variables, we define its Malliavin derivative $D^H_s F$ by

$$D^H_s F := \sum_{i=1}^n \frac{\partial g}{\partial x_i}(B^H(\psi_1), \ldots, B^H(\psi_n))\psi_i(s), \quad 0 \leq s \leq T.$$

For any $F \in S$ as above and $p \in (0, \infty)$, we define the following norm

$$\|F\|_{H,1,p} := \|F\|_p + \sum_{i=1}^n \left[ \mathbb{E} \left( \int_0^T |D^H_t F|^2 \, dt \right)^{p/2} \right]^{1/p}.$$

We denote by $D_{H,1,p}$ the Banach space obtained by completing $S$ with respect to the norm $\| \cdot \|_{H,1,p}$.

The classical Malliavin derivative $D^W$ with respect to the underlying Brownian motion $W$ and the space $D^W_{1,p}$ can be defined in a similar and classical way, which we omit here.

We define an operator $K_H$ on $\mathcal{H}$ as:

$$(K_H \psi)(s) = c_H \Gamma(H - 1/2) s^{1/2-H} f^{H-1/2}_0 (u^{H-1/2} \psi(u))(s).$$

Then its adjoint operator $K_H^*$ on $\mathcal{H}$ is:

$$(K_H^* \psi)(s) = c_H \Gamma(H - 1/2) s^{1/2-H} f^{H-1/2}_T (u^{H-1/2} \psi(u))(s),$$

and its inverse operator $K_H^{-1}$ is:

$$(K_H^{-1} \psi)(s) = \frac{1}{c_H \Gamma(H - 1/2)} s^{1/2-H} (D^{H-1/2}_T u^{H-1/2} \psi(u))(s).$$

For $\psi \in \mathcal{H}$, the following relationship holds:

$$\int_0^T \psi(t)dB^H(t) = \int_0^T (K_H^* \psi)(t)dW(t)$$

and

$$\int_0^T \psi(t)dW(t) = \int_0^T (K_H^{-1} \psi)(t)dB^H(t).$$

Therefore, if we denote by $\mathcal{F} = \{\mathcal{F}_t, t \in [0, T]\}$ the filtration generated by the fractional Brownian motion $\{B^H_t\}_{t \in [0, T]}$, it coincides with the one generated by the underlying Brownian motion $\{W_t, t \in [0, T]\}$.

We have the following proposition:

**Proposition 2.1.** If $f \in D^W_{1,2} \cap D_{H,1,2}$, then

$$D^H_s F = K_H^{-1} D^W_s F.$$

However, it is more convenience for fractional Brownian motions to use another Malliavin derivative, which is defined as

$$D^H_s F = \int_0^T \varphi(s - r)D^H_r Fdr,$$

where $\varphi$ is a smooth kernel.
where
\[ \varphi(r) = H(2H - 1)|r|^{2H-2}, \quad 0 \leq r \leq T. \]
From Section 5.8 in [11] we know
\[ \mathbb{D}_t^H F = K_H K_t^* D_t^H F. \]

Now we can define by the following result the more general Skorohod type integral \( \int_0^T f(t) dB_t^H \) as the divergence operator related to \( \mathbb{D}_t^H \) (See, for example Theorem 6.23 in [11], or Proposition 2.3 in [10]).

**Definition 2.2.** Let \( f : ([0, T] \times \Omega, \mathcal{B}([0, T]) \otimes \mathcal{F}) \rightarrow (\mathbb{R}, \mathcal{B}(\mathbb{R})) \) be a jointly measurable square integrable process. We say that \( f \) is integrable with respect to \( B^H \) (\( f \in \text{Dom}(\delta_H) \)), if there is some \( \delta_H(f) \in L^2(\Omega, \mathcal{F}, \mathbb{P}) \) such that for all \( G \in \mathbb{D}_H,1,2 \),
\[
\mathbb{E}[G \delta_H(f)] = \int_0^T \mathbb{E}[f(t) \mathbb{D}_t^H G] \, dt.
\]

If \( fI_{[s,t]} \in \text{Dom}(\delta_H) \), we write \( \int_s^t f(r) dB_r^H := \delta_H(fI_{[s,t]}), \ s, t \in [0, T]. \)

From the classical Malliavin calculus theory (refer to, Nualart [18] and Buckdahn [3]), we have the following proposition.

**Proposition 2.3.** If \( f \in \text{Dom}(\delta_H) \), it holds that:
\[
\mathbb{E}
\left[
\left|
\int_0^T f(t) dB_t^H
\right|^2
\right] = \mathbb{E}
\left[
\int_0^T |K_H^* f(t)|^2 dt
\right] + 2 \mathbb{E}
\left[
\int_0^T \int_0^t D_s^W (K_H^* f(r)) D_r^W (K_H^* f(s)) \, dr ds
\right].
\]

The Stratonovich integral with respect to fractional Brownian motion can be defined from the Skorohod integral as follows (see Theorem 3.9 in [9]).

**Proposition 2.4.** Let \( f : \Omega \times [0, T] \rightarrow \mathbb{R} \) be a stochastic process which is Malliavin differentiable such that the following holds:
\[
\mathbb{E}
\left[
\int_0^T \int_0^T |f(s)f(t)| \varphi(s-t) \, ds dt + \int_0^T \int_0^T |\mathbb{D}_s^H f(t)|^2 \, ds dt
\right] < \infty.
\]
Then the Stratonovich integral \( \int_0^T f(t) dB_t^H \) exists and
\[
\int_0^T f(t) dB_t^H = \int_0^T f(t) dB_t^H + \int_0^T \mathbb{D}_t^H f(t) \, dt.
\]

The following proposition can be derived from Remark 2.7.4 in Mishura [17].

**Proposition 2.5.** For \( t \in [0, T] \), let \( F_1(t) = \int_0^t f_1(s) ds + \int_0^t f_2(s) dB_s^H \) and \( G_1(t) = \int_0^t g_1(s) ds + \int_0^t g_2(s) dW_s \), where \( f_1, g_1 \) are integrable processes, \( f_2 \) satisfies the conditions in Proposition 2.4 and \( g_2 \) is continuous square integrable adapted process. Then we have
\[
df_1(t)G_1(t) = F_1(t)g_2(t) dW_t + G_1(t)f_2(t) dB_t^H + [F_1(t)g_1(t) + G_1(t)f_1(t)] dt.
\]
Combining Proposition 2.4 and 2.5, it is easy to deduce the following result.
Corollary 2.6. For \( t \in [0, T] \), let now \( F(t) = \int_0^t f_1(s) \, ds + \int_0^t f_1(s) \, dB_s^H \) and \( G(t) = \int_0^t g_1(s) \, ds + \int_0^t g_2(s) \, dW_s \), where \( f_1, g_1 \) are integrable processes, \( f_2 \) satisfies the conditions in Proposition 2.4 and \( g_2 \) is continuous square integrable adapted process. Then we have
\[
dF(t)G(t) = F(t)g_2(t) \, dW_t + G(t)f_2(t) \, dB_t^H + f_2(t) \, dH_t^s G(t) \, dt.
\]

2.4. Girsanov Transformation. Let \( \{\gamma(s), s \in [0, T]\} \) be a bounded function in \( \mathcal{H} \). For any \( \omega \in \Omega \), we define the following operators:
\[
T_t(\omega) = \omega + \int_0^{t\wedge} K_H^s(\gamma I_{[0,t]})(s) \, ds,
\]
and
\[
A_t(\omega) = \omega - \int_0^{t\wedge} K_H^s(\gamma I_{[0,t]})(s) \, ds, \quad t \in [0, T].
\]

It is clear that \( A_t T_t(\omega) = T_t A_t(\omega) = \omega \). Moreover, for any \( F \in \mathcal{S} \), we have from the Girsanov theorem (we refer to [3]),
\[
\mathbb{E}[F] = \mathbb{E}[F(T_t) \varepsilon_t^{-1}(T_t)] = \mathbb{E}[F(A_t) \varepsilon_t],
\]
where
\[
\varepsilon_t = \exp \left\{ - \int_0^t \gamma_s \, dB_s^H - \frac{1}{2} \int_0^t \left( K_H^s(\gamma I_{[0,t]}))^2 (s) \right) \, ds \right\}
\]
and hence
\[
\varepsilon_t^{-1}(T_t) = \exp \left\{ - \int_0^t \gamma_s \, dB_s^H - \frac{1}{2} \int_0^t \left( K_H^s(\gamma I_{[0,t]}))^2 (s) \right) \, ds \right\}.
\]

Following a similar argument in Lemma 2.4 in [13], we verify that
\[
\mathbb{E} \left[ \sup_{t \in [0, T]} \varepsilon_t^p \right] < +\infty \quad \text{and} \quad \mathbb{E} \left[ \sup_{t \in [0, T]} \varepsilon_t^p(T_t) \right] < +\infty, \quad \text{for all} \quad p \in \mathbb{R}.
\]

2.5. Differentiability of Functions of Measures. Let \( \mathcal{P}(\mathbb{R}^n) \) be the space of all probability measures on \( (\mathbb{R}^n, \mathcal{B}(\mathbb{R}^n)) \). We denote by \( \mathcal{P}_p(\mathbb{R}^n) \) the subspace of \( \mathcal{P}(\mathbb{R}^n) \) of order \( p \), which means
\[
\mathcal{P}_p(\mathbb{R}^n) = \{ \mu \in \mathcal{P}(\mathbb{R}^n) : \int_{\mathbb{R}^n} |x|^p \mu(dx) < +\infty \}.
\]

On \( \mathcal{P}_p(\mathbb{R}^n) \), the Wasserstein metric of order \( p \) is defined by
\[
W_p(\mu, \nu) = \inf \left\{ \left( \int_{\mathbb{R}^{2n}} |x-y|^p \rho(dx, dy) \right)^{\frac{1}{p}} \right\}, \quad \rho \in \mathcal{P}(\mathbb{R}^{2n}) \text{ such that } \rho(\cdot \times \mathbb{R}^n) = \mu \text{ and } \rho(\mathbb{R}^n \times \cdot) = \nu
\]

In this paper, we will use Wasserstein metrics of order 1 and 2: \( W_1 \) and \( W_2 \). Notice that if \( \xi \) and \( \eta \) are two \( p \)-integrable random variables with laws \( \mathbb{P}_\xi \) and \( \mathbb{P}_\eta \), then we have \( W_p(\mathbb{P}_\xi, \mathbb{P}_\eta) \leq (\mathbb{E} |\xi - \eta|^p)^{\frac{1}{p}} \).
since we can choose a special $\rho = \mathbb{P}(\xi, \eta)$ in the above definition. In this paper, the notion of differentiability for functions of measures we use is that introduced by P. L. Lions in his course at the Collège de France and summarized by Cardaliaguet [7]. We also refer to Carmona and Delarue [8].

Notice that, as $(\Omega, \mathcal{F}, \mathbb{P})$ carries a fractional Brownian motion, it is rich enough in the sense that $\mathcal{P}_2(\mathbb{R}^n) = \{P_\xi, \xi \in L^2(\Omega, \mathcal{F}, \mathbb{P}; \mathbb{R}^n)\}$, $n \geq 1$.

Given a function $\sigma : \mathcal{P}_2(\mathbb{R}) \to \mathbb{R}$, for any random variable $\xi \in L^2(\Omega, \mathcal{F}, \mathbb{P})$, we set $\hat{\sigma}(\xi) \triangleq \sigma(\mathbb{P}_\xi)$.

**Definition 2.7.** The function $\sigma$ is said to be differentiable in $\mu \in \mathcal{P}_2(\mathbb{R})$, if there exists a random variable $\tilde{\xi} \in L^2(\Omega, \mathcal{F}, \mathbb{P})$ with $\mathbb{P}_\xi = \mu$ such that $\hat{\sigma} : L^2(\Omega, \mathcal{F}, \mathbb{P}) \to \mathbb{R}$ is Fréchet differentiable in $\tilde{\xi}$.

For simplicity, we suppose that $\hat{\sigma} : L^2(\Omega, \mathcal{F}, \mathbb{P}) \to \mathbb{R}$ is Fréchet differentiable. We denote its Fréchet derivative at $\tilde{\xi}$ by $D\hat{\sigma}(\tilde{\xi})$. Notice that $D\hat{\sigma}(\tilde{\xi}) : L^2(\Omega, \mathcal{F}, \mathbb{P}) \to \mathbb{R}$ is a continuous linear mapping; we write $D\hat{\sigma}(\xi) \in L(L^2(\Omega, \mathcal{F}, \mathbb{P}), \mathbb{R})$. Hence,

$$
\sigma(\mathbb{P}_\xi) - \sigma(\mathbb{P}_\tilde{\xi}) = \hat{\sigma}(\xi) - \hat{\sigma}(\tilde{\xi}) = \langle (D\hat{\sigma})(\xi), (\xi - \tilde{\xi}) \rangle_{L^2} + o(|\xi - \tilde{\xi}|_{L^2}), \text{ as } |\xi - \tilde{\xi}|_{L^2} \to 0.
$$

According to Cardaliaguet [7], with Riesz representation theorem, $D\hat{\sigma}(\xi) \in L(L^2(\Omega, \mathcal{F}, \mathbb{P}), \mathbb{R}) \equiv L^2(\Omega, \mathcal{F}, \mathbb{P})$, i.e., there exists a random variable $\theta \in L^2(\Omega, \mathcal{F}, \mathbb{P})$ such that $D\hat{\sigma}(\xi)(\eta) = \mathbb{E}\{\theta|\eta\}$, for an $\eta \in L^2(\Omega, \mathcal{F}, \mathbb{P})$. Moreover, there is a Borel function $h_{\mathbb{P}_\xi} : \mathbb{R} \to \mathbb{R}$, such that $\theta = h_{\mathbb{P}_\xi}(\tilde{\xi})$, $\mathbb{P}$-a.s.

We define the derivative of $\sigma$ with respect to the measure at $\mathbb{P}_\xi$ by putting $\partial_\mu\sigma(\mathbb{P}_\xi, x) = h_{\mathbb{P}_\xi}(x)$. Notice that $\partial_\mu\sigma(\mathbb{P}_\xi, x)$ is defined only $\mathbb{P}_\xi(dx)$-a.e. uniquely. Therefore,

$$
\sigma(\mathbb{P}_\xi) - \sigma(\mathbb{P}_\tilde{\xi}) = \mathbb{E}[\partial_\mu\sigma(\mathbb{P}_\xi, \tilde{\xi}) - \tilde{\xi}] + o(|\xi - \tilde{\xi}|_{L^2}), \text{ as } |\xi - \tilde{\xi}|_{L^2} \to 0.
$$

For example, for $\xi \in L^2(\Omega, \mathcal{F}, \mathbb{P})$, we consider $\sigma(\mathbb{P}_\xi) = \sigma(\mathbb{E}[\varphi(\xi)])$. Hence, we have $\hat{\sigma}(\xi) = \sigma(\mathbb{E}[\varphi(\xi)])$, and a straight forward computation shows

$$
D\hat{\sigma}(\xi)(\eta) = \mathbb{E}[\sigma'(\mathbb{E}[\varphi(\xi)])\varphi'(\xi)|\varphi(\xi)\eta], \text{ for all } \eta \in L^2(\Omega, \mathcal{F}, \mathbb{P}).
$$

Therefore, $\partial_\mu\sigma(\mathbb{P}_\xi, x) = \sigma'(\mathbb{E}[\varphi(\xi)])\varphi'(x)$.

In the last part of this paper, we need the joint convexity of a function on $(\mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^d))$. A differentiable function $g$ defined on $(\mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^d))$ is convex, if for every $(x, \mu)$ and $(x', \mu') \in (\mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^d))$, we have

$$
(21) \quad g(x', \mu') - g(x, \mu) - \langle \partial_x g(x, \mu), (x' - x) \rangle \leq \mathbb{E} \left[ (\partial_\mu g(x, \mu)(X), X' - X) \right] \geq 0,
$$

where $X, X' \in L^2(\Omega, \mathcal{F}, \mathbb{P}; \mathbb{R}^d)$ with $\mathbb{P}_X = \mu$ and $\mathbb{P}_{X'} = \mu'$, and $\langle \cdot, \cdot \rangle$ stands for the scalar product.

Moreover, a differentiable function $g$ defined on $(\mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^d))$ is strictly convex, if there exists $\lambda > 0$, for every $(x, \mu)$ and $(x', \mu') \in (\mathbb{R}^n \times \mathcal{P}_2(\mathbb{R}^d))$, we have

$$
(22) \quad \lambda \left( |x - x'|^2 + \mathbb{E} \left[ |X - X'|^2 \right] \right) \leq g(x', \mu') - g(x, \mu) - \langle \partial_x g(x, \mu), (x' - x) \rangle \leq \mathbb{E} \left[ (\partial_\mu g(x, \mu)(X), X' - X) \right],
$$

where $X, X' \in L^2(\Omega, \mathcal{F}, \mathbb{P}; \mathbb{R}^d)$ with $\mathbb{P}_X = \mu$ and $\mathbb{P}_{X'} = \mu'$. 
3. MEAN FIELD SDE DRIVEN BY FRACTIONAL BROWNIAN MOTION

In this section, we will study a class of semi-linear stochastic differential equations driven by a fractional Brownian motion. In the following sections, the constant $C$ can vary from line to line.

Given an arbitrary square integrable process $\Theta = (\Theta_s)$ with values in $\mathbb{R}^m$, $m \geq 1$, let us consider the following equation:

$$X_t = \xi + \int_0^t (\gamma_s X_s + \sigma(s, \mathbb{P}(X_s, \Theta_s)) \, dB^H_s + \int_0^t b(s, \mathbb{P}(X_s, \Theta_s), X_s) \, ds,$$

where $\xi \in L^2(\Omega, \mathcal{F}_0, \mathbb{P}; \mathbb{R})$ and the coefficients $\sigma : [0, T] \times \mathcal{P}_2(\mathbb{R} \times U) \rightarrow \mathbb{R}$ and $b : \Omega \times [0, T] \times \mathcal{P}_2(\mathbb{R} \times U) \times \mathbb{R} \rightarrow \mathbb{R}$ satisfy the following conditions:

**H1** For any $s \in [0, T]$, $x, x' \in \mathbb{R}$, $\eta, \eta' \in L^2(\Omega, \mathcal{F}, \mathbb{F}; \mathbb{R})$ and $\Theta \in L^2(\Omega, \mathcal{F}, \mathbb{F}; \mathbb{R}^m)$, there exists a constant $C > 0$ such that

$$|\sigma(s, \mathbb{P}(\eta, \Theta))| \leq C,$$

$$|b(s, \mathbb{P}(\eta, \Theta), x)| \leq C (1 + \mathbb{E} \, |\eta| + |x|),$$

$$|\sigma(s, \mathbb{P}(\eta, \Theta)) - \sigma(s, \mathbb{P}(\eta', \Theta))| \leq C \mathbb{E} \, |\eta - \eta'|,$$

$$|b(s, \mathbb{P}(\eta, \Theta), x) - b(s, \mathbb{P}(\eta', \Theta), x')| \leq C (\mathbb{E} \, |\eta - \eta'| + |x - x'|).$$

We denote by $L^{2,*}([0, T]; \mathbb{R})$ the Banach space of $\mathbb{P}$-adapted process $\{\varphi(t), t \in [0, T]\}$ such that

$$\sup_{t \in [0, T]} \mathbb{E} \left[ |\varphi(t)|^2 \varepsilon_t^{-1} \right] < +\infty.$$

**Definition 3.1.** A solution of equation (23) is a stochastic process $X = (X_t)_{t \geq 0} \in L^{2,*}([0, T]; \mathbb{R})$ such that $XI_{[0,t]} \in \text{Dom}(\delta_H)$, $t \in [0, T]$ and equation (23) holds true $\mathbb{P}$-a.s.

**Remark 3.2.** Note that for $X \in L^{2,*}([0, T]; \mathbb{R})$, $(\sigma(s, \mathbb{P}(X_s, \Theta_s)))_{s \in [0, T]} \in L^\infty([0, T])$ is a bounded and hence, square integrable deterministic function, which implies that $\int_0^t \sigma(s, \mathbb{P}(X_s, \Theta_s)) \, dB^H_s$ is well defined.

To solve the equation (23), we first transform it to another one. Indeed, we have the following statement.

**Theorem 3.3.** Assume $X \in L^{2,*}([0, T]; \mathbb{R})$. Then $X$ is a solution of (23) if and only if it solves the following equation:

$$X_t \varepsilon_t^{-1}(T_t) = \xi + \int_0^t \sigma(s, \mathbb{P}(X_s, \Theta_s)) s^{-1}(T_s) \, dB^H_s + \int_0^t b(s, T_s, \mathbb{P}(X_s, \Theta_s), X_s(T_s)) s^{-1}(T_s) \, ds.$$

**Remark 3.4.** We note that for any $X \in L^{2,*}([0, T]; \mathbb{R})$, $\int_0^t \sigma(s, \mathbb{P}(X_s, \Theta_s)) s^{-1}(T_s) \, dB^H_s$, $0 \leq t \leq T$, is well defined. Indeed, $\sigma(s, \mathbb{P}(X_s, \Theta_s))$ is a deterministic bounded function, and we have the following statement:

**Lemma 3.5.** For all $\Theta \in L^\infty([0, T])$, the process $(\Theta_s s^{-1}(T_s))_{s \in [0, T]}$ is in $\text{Dom}(\delta_H)$. 
Proof. (of Theorem 3.3). Suppose \( \{X_t, t \in [0, T]\} \in L^{2,\gamma}(\{0, T\}; \mathbb{R}) \) is a solution of equation (23), and that, in particular \( \gamma X_{I[0,t]} \in \text{Dom}(\delta_H) \), \( t \in [0, T] \). Then, for any \( F \in \mathcal{S} \), we have

\[
\mathbb{E} \left[ F X_t(T_t) \varepsilon_t^{-1}(T_t) - F\xi \right] = \mathbb{E}[F(A_t) X_t - F\xi]
\]

\[
= \mathbb{E}[F(A_t) \xi - F\xi] + \mathbb{E} \left[ F(A_t) \int_0^t (\gamma_s X_s + \sigma(s, \mathbb{P}(X_s, \Theta_s))) d\mathbb{B}_s^H \right]
\]

\[
+ \mathbb{E} \left[ F(A_t) \int_0^t b(s, \mathbb{P}(X_s, \Theta_s), X_s) ds \right],
\]

We remark that \( \frac{dF(A_s)}{ds} = -\gamma_s K_H K_H^* D_s^H F(A_s) = -\gamma_s \mathbb{D}_s F(A_s) \). Thus, from Proposition 2.2, we have

\[
\mathbb{E} \left[ F X_t(T_t) \varepsilon_t^{-1}(T_t) - F\xi \right]
\]

\[
= - \mathbb{E} \left[ \xi \int_0^t \gamma_s \mathbb{D}_s^H F(A_s) ds \right] + \mathbb{E} \left[ \int_0^t (\gamma_s X_s + \sigma(s, \mathbb{P}(X_s, \Theta_s))) \mathbb{D}_s^H F(A_t) ds \right]
\]

\[
+ \mathbb{E} \left[ \int_0^t b(s, \mathbb{P}(X_s, \Theta_s), X_s) F(A_t) ds \right].
\]

Using again that \( F(A_t) = F(A_s) - \int_s^t \gamma_r \mathbb{D}_r^H (F(A_r)) dr \), we see that

\[
\mathbb{D}_s^H F(A_t) = \mathbb{D}_s^H F(A_s) - \int_s^t \gamma_r \mathbb{D}_s^H (\mathbb{D}_r^H (F(A_r))) dr.
\]

Consequently,

\[
\mathbb{E} \left[ F X_t(T_t) \varepsilon_t^{-1}(T_t) - F\xi \right]
\]

\[
= - \mathbb{E} \left[ \xi \int_0^t \gamma_s \mathbb{D}_s^H F(A_s) ds \right] + \mathbb{E} \left[ \int_0^t (\gamma_s X_s + \sigma(s, \mathbb{P}(X_s, \Theta_s))) \mathbb{D}_s^H F(A_s) ds \right]
\]

\[
- \mathbb{E} \left[ \int_0^t \int_s^t \gamma_r \mathbb{D}_s^H (\mathbb{D}_r^H (F(A_s))) (\gamma_s X_s + \sigma(s, \mathbb{P}(X_s, \Theta_s))) dr ds \right]
\]

\[
+ \mathbb{E} \left[ \int_0^t b(s, \mathbb{P}(X_s, \Theta_s), X_s) F(A_s) ds \right] - \mathbb{E} \left[ \int_0^t \int_s^t b(s, \mathbb{P}(X_s, \Theta_s), X_s) \gamma_r \mathbb{D}_s^H F(A_r) dr ds \right],
\]

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and the Fubini theorem then yields

\[
\mathbb{E} \left[ FX_t(T_t) \xi_t^{-1}(T_t) - F\xi \right]
\]

\[
= -\mathbb{E} \left[ \xi \int_0^t \gamma_s \mathbb{D}_s^H F(A_s) ds \right] + \mathbb{E} \left[ \int_0^t (\gamma_s X_s + \sigma(s, \mathbb{P}(X_s, \Theta_s))) \mathbb{D}_s^H F(A_s) ds \right]
\]

\[
- \mathbb{E} \left[ \int_0^t \int_0^r \gamma_s \mathbb{D}_s^H (\mathbb{D}_r^H (F(A_r)))(\gamma_s X_s + \sigma(s, \mathbb{P}(X_s, \Theta_s))) ds dr \right]
\]

\[
+ \mathbb{E} \left[ \int_0^t b(s, \mathbb{P}(X_s, \Theta_s), X_s) F(A_s) ds \right] - \mathbb{E} \left[ \int_0^t \int_0^r b(s, \mathbb{P}(X_s, \Theta_s), X_s) \gamma_r \mathbb{D}_r^H F(A_r) ds dr \right].
\]

Applying Proposition 2.2 again, combined with the Fubini theorem, we have

\[
\mathbb{E} \left[ \int_0^t \int_0^r \gamma_r \mathbb{D}_s^H (\mathbb{D}_r^H (F(A_r)))(\gamma_s X_s + \sigma(s, \mathbb{P}(X_s, \Theta_s))) ds dr \right]
\]

\[
= \mathbb{E} \left[ \int_0^t \gamma_s \mathbb{D}_s^H (F(A_s)) \int_0^s (\gamma_r X_r + \sigma(r, \mathbb{P}(X_r, \Theta_r))) dB_r^H ds \right].
\]

Hence,

\[
\mathbb{E} \left[ FX_t(T_t) \xi_t^{-1}(T_t) - F\xi \right]
\]

\[
= -\mathbb{E} \left[ \xi \int_0^t \gamma_s \mathbb{D}_s^H F(A_s) ds \right] + \mathbb{E} \left[ \int_0^t (\gamma_s X_s + \sigma(s, \mathbb{P}(X_s, \Theta_s))) \mathbb{D}_s^H F(A_s) ds \right]
\]

\[
- \mathbb{E} \left[ \int_0^t \gamma_s \mathbb{D}_s^H (F(A_s)) \int_0^s (\gamma_r X_r + \sigma(r, \mathbb{P}(X_r, \Theta_r))) dB_r^H ds \right]
\]

\[
+ \mathbb{E} \left[ \int_0^t b(s, \mathbb{P}(X_s, \Theta_s), X_s) F(A_s) ds \right] - \mathbb{E} \left[ \int_0^t \int_0^r b(s, \mathbb{P}(X_s, \Theta_s), X_s) \gamma_r \mathbb{D}_r^H F(A_r) ds dr \right],
\]

where we have used that \(X\) solves (23). Thus, Girsanov transformation yields

\[
\mathbb{E} \left[ F \left( X_t(T_t) \xi_t^{-1}(T_t) - \xi \right) \right]
\]

\[
= \mathbb{E} \left[ \int_0^t \sigma(s, \mathbb{P}(X_s, \Theta_s)) e_s^{-1}(T_s) \mathbb{D}_s^H F ds + F \int_0^t b(s, T_s, \mathbb{P}(X_s, \Theta_s), X_s(T_s)) e_s^{-1}(T_s) ds \right],
\]

i.e.,

\[
\mathbb{E} \left[ \int_0^t \sigma(s, \mathbb{P}(X_s, \Theta_s)) e_s^{-1}(T_s) \mathbb{D}_s^H F ds \right]
\]

\[
= \mathbb{E} \left[ F \left( X_t(T_t) \xi_t^{-1}(T_t) - \xi - \int_0^t b(s, T_s, \mathbb{P}(X_s, \Theta_s), X_s(T_s)) e_s^{-1}(T_s) ds \right) \right].
\]
Then from the linear growth of $b$ as:

Proof. Given a process $X_t$, the same argument. But this is exactly equation (24). The proof that any solution of equation (24) solves also (23) uses Theorem 3.6.

Observing that $(X_t(T_t)\varepsilon_t^{-1}(T_t) - \xi - \int_0^t b(s, T_s, \mathbb{P}_t, X_s(T_s))\varepsilon_s^{-1}(T_s)ds$ is square integrable, we see from Definition 2.2 that $(\sigma(s, \mathbb{P}(X_s, \Theta_s))\varepsilon_s^{-1}(T_s))$, $s \in [0, T]$ belongs to Dom($\delta_H$), and

$$\int_0^t \sigma(s, \mathbb{P}_t, X_s(T_s))\varepsilon_s^{-1}(T_s)dB_t^H = X_t(T_t)\varepsilon_t^{-1}(T_t) - \xi - \int_0^t b(s, T_s, \mathbb{P}_t, X_s(T_s))\varepsilon_s^{-1}(T_s)ds.$$ 

But this is exactly equation (24). The proof that any solution of equation (24) solves also (23) uses the same argument.

Now let us focus on equation (23). We have the following existence and uniqueness result.

**Theorem 3.6.** Equation (23) admits a unique solution $X = \{X_t, t \in [0, T]\} \in L^2([0, T]; \mathbb{R})$.

**Proof.** Given a process $X^n$ such that $\sup_{t \in [0, T]} \mathbb{E}[|X^n_t|^2\varepsilon_t^{-1}] < +\infty$, we recursively define $X^{n+1}$ as: $X^0 = \xi$, and for $n \geq 0$, $X^{n+1}_t = Y^{n+1}_t(T_t)\varepsilon_t^{-1}(T_t)$, where

$$Y^{n+1}_t = \xi + \int_0^t \sigma(s, \mathbb{P}_t, X_s(T_s))\varepsilon_s^{-1}(T_s)dB_t^H + \int_0^t b(s, T_s, \mathbb{P}_t, X_s(T_s))\varepsilon_s^{-1}(T_s)ds.$$ 

Then from the linear growth of $b$ we have,

\[
\begin{align*}
\sup_{t \in [0, T]} \mathbb{E}[|X_t^{n+1}|^2\varepsilon_t^{-1}] &\leq 2\mathbb{E}[\xi^2] + 2 \sup_{t \in [0, T]} \mathbb{E} \left[ \left| \int_0^t \sigma(s, \mathbb{P}_t, X_s(T_s))\varepsilon_s^{-1}(T_s)dB_s^H \right|^2 \right] \\
&+ 2 \sup_{t \in [0, T]} \mathbb{E} \left[ \left| \int_0^t b(s, T_s, \mathbb{P}_t, X_s(T_s))\varepsilon_s^{-1}(T_s)ds \right|^2 \right] \\
&\leq 2\mathbb{E}[\xi^2] + 2 \sup_{t \in [0, T]} \mathbb{E} \left[ \left| \int_0^t (\sigma(s, \mathbb{P}_t, X_s(T_s))\varepsilon_s^{-1}(T_s)dB_s^H \right|^2 \right] \\
&+ 2 \sup_{t \in [0, T]} \mathbb{E} \left[ \left| \int_0^t b(s, T_s, \mathbb{P}_t, X_s(T_s))\varepsilon_s^{-1}(T_s)ds \right|^2 \right] \\
&\leq 2\mathbb{E}[\xi^2] + 2 \sup_{t \in [0, T]} \mathbb{E} \left[ \left| \int_0^t b(s, T_s, \mathbb{P}_t, X_s(T_s))\varepsilon_s^{-1}(T_s)ds \right|^2 \right] \\
&+ 2T \sup_{t \in [0, T]} \mathbb{E} \left[ \int_0^t \left( 1 + (\mathbb{E}[|X^n_s|]^2 + |X^n_s(T_s)|^2)\varepsilon_s^{-2}(T_s)ds \right) \right].
\end{align*}
\]
From the assumption that $X^n \in L^{2,\ast}([0, T]; \mathbb{R})$ and the results in Section 2, we get
\[
\sup_{t \in [0, T]} \mathbb{E} \left[ |X_t^{n+1}|^2 \epsilon_t^{-1} \right] \\
\leq C + 2 \sup_{t \in [0, T]} \mathbb{E} \left[ \int_0^t K_H^*(\sigma(\cdot, \mathbb{P}(X^n, \Theta_\tau))) \mathbb{E}^{-1}(\tau) (s) dW(s) \right]^2 \\
\leq C + 2 \sup_{t \in [0, T]} \mathbb{E} \left[ \int_0^t K_H^*(\sigma(\cdot, \mathbb{P}(X^n, \Theta_\tau))) \mathbb{E}^{-1}(\tau) (s) ds \right]^2 \\
+ 4 \sup_{t \in [0, T]} \mathbb{E} \left[ \int_0^t \int_0^s (K_H^*(\sigma(\cdot, \mathbb{P}(X^n, \Theta_\tau))) \mathbb{E}^{-1}(\tau))(r) \\
\times D_r^W (K_H^*(\sigma(\cdot, \mathbb{P}(X^n, \Theta_\tau))) \mathbb{E}^{-1}(\tau))(s) dr ds \right] \\
:= C + 2I_1 + 4I_2.
\]

Now for the term $I_1$, we have
\[
I_1 = \sup_{r \in [0, T]} \mathbb{E} \left[ C_H^2 \int_0^t \left| \int_s^t \sigma(r, \mathbb{P}(X^n, \Theta_\tau)) \mathbb{E}^{-1}(\tau) \left( \frac{r}{s} \right)^{H-\frac{1}{2}} (r-s)^{H-\frac{3}{2}} dr \right|^2 ds \right] \\
\leq C \mathbb{E} \left[ \sup_{r \in [0, T]} \mathbb{E}^{-2}(\tau) \right] \sup_{t \in [0, T]} \int_0^t |K_H(t, s)|^2 ds \\
\leq CT^{2H}.
\]

For the term $I_2$, we have
\[
I_2 = \sup_{t \in [0, T]} \mathbb{E} \left[ \int_0^t \int_0^s \int_s^t \sigma(v, \mathbb{P}(X^n, \Theta_\tau)) \mathbb{E}^{-1}(\tau) \left( K_H^*(\gamma[0, v)(\cdot)) \right) (s) \left( \frac{v}{r} \right)^{H-\frac{1}{2}} (v-r)^{H-\frac{3}{2}} dv \\
\times \int_0^t \sigma(u, \mathbb{P}(X^n, \Theta_\tau)) \mathbb{E}^{-1}(\tau) \left( K_H^*(\gamma[0, u](\cdot)) \right) (r) \left( \frac{u}{s} \right)^{H-\frac{1}{2}} (u-s)^{H-\frac{3}{2}} dudrds \right] \\
\leq C_H^2 \mathbb{E} \left[ \sup_{r \in [0, T]} \mathbb{E}^{-2}(\tau) \right] \sup_{t \in [0, T]} \int_0^t \int_0^s \int_s^t |\sigma(v, \mathbb{P}(X^n, \Theta_\tau))| |\sigma(u, \mathbb{P}(X^n, \Theta_\tau))| \\
\times (\frac{v}{s})^{2H-1} (v-r)^{H-\frac{3}{2}} (v-s)^{H-\frac{3}{2}} (\frac{u}{r})^{2H-1} (u-s)^{H-\frac{3}{2}} (u-r)^{H-\frac{3}{2}} dudrds \\
\leq C \sup_{t \in [0, T]} \int_0^t \int_0^s \int_s^t t^{2H-1} s^2 s^{1-2H} v^{H-\frac{1}{2}} (v-s)^{H-\frac{3}{2}} r^{2H-1} r^{1-2H} u^{H-\frac{1}{2}} (u-r)^{H-\frac{3}{2}} dudrds \\
\leq C \sup_{t \in [0, T]} \int_0^t \int_0^s \int_s^t s^{\frac{1}{2}-H} \left( \frac{v}{s} \right)^{H-\frac{1}{2}} (v-s)^{H-\frac{3}{2}} r^{\frac{1}{2}-H} \left( \frac{u}{r} \right)^{H-\frac{1}{2}} (u-r)^{H-\frac{3}{2}} dudrds \\
\leq CT^{2H}.
\]

Hence for $X^n \in L^{2,\ast}([0, T]; \mathbb{R})$, we deduce that $X^{n+1} \in L^{2,\ast}([0, T]; \mathbb{R})$. 

In the following we prove the convergence of $X^n \in L^{2,*}([0,T], \mathbb{R})$. We divide the proof into 4 steps.

**Step 1.** Define $X^n_t = X^n_t - X^{n-1}_t$ and $\rho^n(t) = \sigma\left(t, \mathbb{P}(X^n_t, \Theta_t)\right) - \sigma\left(t, \mathbb{P}(X^{n-1}_t, \Theta_t)\right)$. Notice that $\rho^n$ is a deterministic function. Then we have

\[
\mathbb{E} \left[ \left| X^n_{t+1} \right|^2 \varepsilon_t^{-1} \right] = \mathbb{E} \left[ \left| X^{n+1}_t - (T_t) \varepsilon_t^{-1}(T_t) \right|^2 \right]
\]

\[
\leq 2 \mathbb{E} \left[ \left| \int_0^t (\sigma(s, \mathbb{P}(X^n_s, \Theta_s)) - \sigma(s, \mathbb{P}(X^{n-1}_s, \Theta_s))) \varepsilon_s^{-1}(T_s) dB^H_s \right|^2 \right]
\]

\[
+ 2 \mathbb{E} \left[ \left| \int_0^t (b(s, T_s, \mathbb{P}(X^n_s, \Theta_s), X^n_s(T_s)) - b(s, T_s, \mathbb{P}(X^{n-1}_s, \Theta_s), X^{n-1}_s(T_s))) \varepsilon_s^{-1}(T_s) ds \right|^2 \right]
\]

\[
= 2 \mathbb{E} \int_0^t \left| K_H^*(\rho^n(s, 0, t)(\cdot) \varepsilon_s^{-1}(T_s))(s) \right|^2 ds
\]

\[
+ 4 \mathbb{E} \int_0^t \int_0^s D^W_s \left( K_H^*(\rho^n(s, 0, t)(\cdot) \varepsilon_s^{-1}(T_s)) \right)(r) D^W_r \left( K_H^*(\rho^n(s, 0, t)(\cdot) \varepsilon_s^{-1}(T_s)) \right)(s) dr ds
\]

\[
+ 2 \mathbb{E} \left[ \left| \int_0^t (b(s, T_s, \mathbb{P}(X^n_s, \Theta_s), X^n_s(T_s)) - b(s, T_s, \mathbb{P}(X^{n-1}_s, \Theta_s), X^{n-1}_s(T_s))) \varepsilon_s^{-1}(T_s) ds \right|^2 \right]
\]

\[
= 2I_3(t) + 4I_4(t) + 2I_5(t).
\]

Now we deal with $I_3(t)$, $I_4(t)$ and $I_5(t)$ separately.

**Step 2. The term $I_3(t)$.**

From the definition of operator $K_H^*$, we have

\[
I_3(t) = \mathbb{E} \left[ \int_0^t \left| K_H^*(\rho^n(s, 0, t)(\cdot) \varepsilon_s^{-1}(T_s))(s) \right|^2 ds \right]
\]

\[
\leq \mathbb{E} \sup_{r \in [0,T]} \varepsilon_r^{-2}(T_r) C_H^2 \int_0^t \left( \int_s^t |\rho^n(r)| \left( \frac{r}{s} \right)^{H-\frac{1}{2}} (r-s)^{H-\frac{3}{2}} dr \right)^2 ds
\]

\[
\leq C \int_0^t \left| K_H^*(\rho^n(s, 0, t)(\cdot))(s) \right|^2 ds.
\]

Let $q > 1$ be adjoint to $p > \frac{1}{H-\frac{3}{2}}$: $\frac{1}{p} + \frac{1}{q} = 1$. Then, $1 < q < \frac{1}{2-H} < 2$. Observe that

\[
\int_0^t \left| K_H^*(\rho^n(s, 0, t)(\cdot))(s) \right|^2 ds
\]

\[
\leq C \int_0^t \left( \int_s^t |\rho^n(r)|^p dr \right)^2 \left( \int_s^t \left[ \left( \frac{r}{s} \right)^{H-\frac{1}{2}} (r-s)^{H-\frac{3}{2}} \right]^q dr \right)^{\frac{2}{q}} ds
\]

\[
\leq C \left( \int_0^t |\rho^n(r)|^p dr \right)^\frac{2}{p} \int_0^t \left( \frac{t}{s} \right)^{2H-1} \left( \int_s^t (r-s)^{q(H-\frac{3}{2})} dr \right)^{\frac{2}{q}} ds.
\]
we obtain

\[ \int_0^t |K_H^*(\rho^n(r)1_{[0,\varepsilon]}(\cdot))(s)|^2 \, ds \]
\[ \leq C \left( \int_0^t |\rho^n(r)|^p \, dr \right)^{\frac{2}{p}} \int_0^t \left( \frac{t}{s} \right)^{2H-1} \left( \frac{1}{q \left( H - \frac{3}{2} \right) + 1} \right)^{\frac{2}{q}} \left( t - s \right)^{2H-3+\frac{2}{q}} \, ds \]
\[ \leq C \left( \frac{1}{q \left( H - \frac{3}{2} \right) + 1} \right)^{\frac{2}{q}} \left( \int_0^t |\rho^n(r)|^p \, dr \right)^{\frac{2}{p}}. \]

Since on the other hand from the Lipschitz continuity of \( \sigma \) with respect to the 1-Wasserstein metric \( W_1 \) we have

\[ \left( \int_0^t |\rho^n(r)|^p \, dr \right)^{\frac{2}{p}} = \left( \int_0^t |\sigma(r, \mathbb{P}(X^n_r, \Theta_r)) - \sigma(r, \mathbb{P}(X^n_{r-}, \Theta_r))|^p \, dr \right)^{\frac{2}{p}} \]
\[ \leq C \left( \int_0^t \left( \mathbb{E} [ |X^n_r|^p ] \right)^{\frac{2}{p}} \, dr \right)^{\frac{2}{p}} \leq C \left( \int_0^t \left( \mathbb{E} [ |X^n_{r-}|^2 ] \right)^{\frac{p}{2}} \, dr \right)^{\frac{2}{p}}, \]

we obtain

\[ I_3(t) \leq C \left( \frac{1}{q \left( H - \frac{3}{2} \right) + 1} \right)^{\frac{2}{q}} \left( \int_0^t \left( \mathbb{E} [ |X^n_{r-}|^2 ] \right)^{\frac{p}{2}} \, dr \right)^{\frac{2}{p}}. \]

**Step 3. The term \( I_4(t) \).**

Now we deal with the term \( I_4(t) \), which can be written as

\[ I_4(t) = \mathbb{E} \left[ \int_0^t \int_0^s D_s^W \left( K_H^*(\rho^n(r)1_{[0,\varepsilon]}(\cdot)\varepsilon^{-1}(\mathcal{T}_r))(r) \right) D_r^W \left( K_H^*(\rho^n(\cdot)1_{[0,\varepsilon]}(\cdot)\varepsilon^{-1}(\mathcal{T}_r))(s) \right) \varepsilon \, dr \, ds \right] \]
\[ = \mathbb{E} \left[ \int_0^t \int_s^t C_H^2 \int_s^t \rho^n(v)\varepsilon^{-1}(\mathcal{T}_v) \left( K_H^*(\gamma_{[0,v]}(\cdot))(s) \right) \left( \frac{v}{s} \right)^{H-\frac{1}{2}} \left( s - r \right)^{H-\frac{3}{2}} \, dv \, dr \right. \]
\[ \times \int_s^t \rho^n(u)\varepsilon_{u-}^{-1}(\mathcal{T}_u) \left( K_H^*(\gamma_{[0,u]}(\cdot))(r) \right) \left( \frac{u}{s} \right)^{H-\frac{1}{2}} \left( u - s \right)^{H-\frac{3}{2}} \, du \, dr \, ds \right]. \]
Following a similar argument to the first part of this proof, we have

\[
I_4(t) \leq C \mathbb{E} \left[ \sup_{t \in [0,T]} \varepsilon_r^{-2}(T_r) \right] \int_0^t \int_s^t \int_s^t |\rho^n(v)| |\rho^n(u)| \left( \frac{v}{s} \right)^{2H-1} (v - r)^{H - \frac{3}{2}} (v - s)^{H - \frac{3}{2}} \times \left( \frac{u}{r} \right)^{2H-1} (u - s)^{H - \frac{3}{2}} (u - r)^{H - \frac{3}{2}} dvdu drds
\leq C \int_0^t \int_s^t \int_s^t |\rho^n(v)| |\rho^n(u)| \left[ \int_0^{s+2H-1} (v-r)^{H - \frac{3}{2}} (v-s)^{H - \frac{3}{2}} \right] \times r^{2H-2} \left( \frac{u}{r} \right)^{H - \frac{3}{2}} (u - r)^{H - \frac{3}{2}} dvdu drds
\leq C t^{4H-2} \int_0^t \int_s^{s+2H-2} \left( \frac{u}{r} \right)^{H - \frac{3}{2}} (u - r)^{H - \frac{3}{2}} dvdu drds
\leq C t^{4H-2} \int_0^t \left( K_H(\rho^n(\cdot)1_{[0,t]}(\cdot))(s) \right)^2 drds
\leq C t^{2H} \int_0^t \left( K_H(\rho^n(\cdot)1_{[0,t]}(\cdot))(s) \right)^2 ds.
\]

From the computations (26) and (27) of term \( I_3(t) \), we have

\[
I_4(t) \leq C t^{4H-2+2\frac{2}{p}} \left( \int_0^t \mathbb{E} \left[ \left| \bar{X}_r^n \right|^2 \varepsilon_r^{-1} \right] dr \right)^{\frac{p}{2}}.
\]

**Step 4. The term \( I_5(t) \).**

From the Lipschitz continuity of function \( b \), for \( p > 2 \), we have

\[
I_5(t) = \mathbb{E} \left[ \left\| \int_0^t (b(s, T_s, \mathbb{P}(X^n_{s-1}, \Theta_s), X^n_s(T_s)) - b(s, T_s, \mathbb{P}(X^n_{s-1}, \Theta_s), X^n_{s-1}(T_s))) \varepsilon_s^{-1}(T_s) ds \right\|^2 \right]
\leq t \mathbb{E} \left[ \left\| \int_0^t (b(s, T_s, \mathbb{P}(X^n_{s-1}, \Theta_s), X^n_s(T_s)) - b(s, T_s, \mathbb{P}(X^n_{s-1}, \Theta_s), X^n_{s-1}(T_s))) \varepsilon_s^{-1}(T_s) ds \right\|^2 \right]
\leq C t \mathbb{E} \left[ \int_0^t \left( \mathbb{E} \left[ \left| X^n_{s-1}(T_s) \right| \right]^2 + \left| X^n_{s-1}(T_s) \right|^2 \right) \varepsilon_s^{-2}(T_s) ds \right]
\leq C t \mathbb{E} \left[ \sup_{s \in [0,T]} \varepsilon_s^{-2}(T_s) \right] \int_0^t \mathbb{E} \left[ \left| X^n_s(T_s) \right|^2 \varepsilon_s^{-2}(T_s) ds + C t \mathbb{E} \left[ \int_0^t \left| X^n_s(T_s) \right|^2 \varepsilon_s^{-2}(T_s) ds \right] \right]
\leq C t \int_0^t \mathbb{E} \left[ \left| X^n_s(T_s) \right|^2 \varepsilon_s^{-1} \right] ds
\leq C t^{1+\frac{p}{2}} \left( \int_0^t \mathbb{E} \left[ \left| X^n_s \right|^2 \varepsilon_s^{-1} \right] ds \right)^{\frac{p}{2}}.
\]
Moreover, if $\gamma$ motion $B$ as follows:

$$\sigma$$

we consider the equation with $\sigma$ in the next section), then we only need the following assumption on $\sigma$

One can see in the Step 4 of the above proof, assumption $(H1)$ is essential. However, if $\tau$.

**Remark 3.7.** One can see in the Step 4 of the above proof, assumption $(H1)$ is essential. However, if we consider the equation with $\gamma_s \equiv 0$, i.e., $\varepsilon_s \equiv 1$ in equation $(23)$ (which is the case that we consider in the next section), then we only need the following assumption on $\sigma$ and $b$:

$(H1')$ For any $s \in [0, T]$, $x, x' \in \mathbb{R}, \eta, \eta' \in L^2(\Omega, \mathcal{F}, \mathbb{P}; \mathbb{R})$ and $\Theta \in L^2(\Omega, \mathcal{F}, \mathbb{P}; \mathbb{R}^m)$, there exists a constant $C > 0$ such that

$$|\sigma(s, \mathbb{P}(\eta, \Theta))| \leq C \left(1 + (\mathbb{E}[|\eta|^2])^{\frac{1}{2}} \right),$$

$$|b(s, \mathbb{P}(\eta, \Theta), x)| \leq C \left(1 + (\mathbb{E}[|\eta|^2])^{\frac{1}{2}} + |x| \right),$$

$$|\sigma(s, \mathbb{P}(\eta, \Theta)) - \sigma(s, \mathbb{P}(\eta', \Theta))| \leq CW_2(\eta, \eta'),$$

$$|b(s, \mathbb{P}(\eta, \Theta), x) - b(s, \mathbb{P}(\eta', \Theta), x')| \leq C (W_2(\eta, \eta') + |x - x'|).$$

Moreover, if $\gamma_s \equiv 0$, the space $L^{2, *}(\mathbb{R}^m)$ becomes the classical space $L^2(\mathbb{R}^m)$, the space of $\mathbb{F}$-adapted square integrable processes.

4. **Mean-field stochastic control problem driven by a fractional Brownian motion with $H > 1/2$**

In this section, we study a mean-field stochastic control problem driven by a fractional Brownian motion $B^H$ with $H > 1/2$.

Let $U$ be a nonempty bounded convex subset of $\mathbb{R}^m$. We define the space of admissible controls as follows:

$$\mathcal{U}([0, T]) := \{ u : [0, T] \times \Omega \to U \mid u \text{ is an } \mathbb{F}\text{-adapted process} \}. $$
We consider the following dynamics for our mean-field controlled system:

\[
X^u_t = x + \int_0^t \sigma(\mathbb{P}_{X^u_s})dB^H_s + \int_0^t b(\mathbb{P}_{(X^u_s, u_s)}), X^u_s, u_s)ds,
\]

where \( x \in \mathbb{R} \), and \( u \in \mathcal{U}([0, T]) \) is an admissible control process. For any given \( u \in \mathcal{U}([0, T]) \), we know from Theorem 3.3 that there exists a unique solution to the controlled system (31). In fact, (31) constitute a particular case of equation (23) with \( \gamma \equiv 0 \) (See Remark 3.7).

The cost functional is assumed to be depend on a running cost function \( f : [0, T] \times \mathcal{P}_2(\mathbb{R} \times U) \times \mathbb{R} \times U \to \mathbb{R} \) and a terminal cost function \( g : \mathbb{R} \times \mathcal{P}_2(\mathbb{R}) \to \mathbb{R} \):

\[
J(u) = \mathbb{E} \left[ \int_0^T f \left( \mathbb{P}_{(X^u_t, u_t)}, X^u_t, u_t \right) dt + g \left( X^u_T, \mathbb{P}_{X^u_T} \right) \right].
\]

Our aim is to characterise an optimal control \( u^* \in \mathcal{U}([0, T]) \) such that

\[
J(u^*) = \inf_{u \in \mathcal{U}([0, T])} J(u).
\]

If there exists such optimal control \( u^* \), we call the corresponding pair \( (X^*, u^*) \) optimal for the control problem. Here \( X^* = X^{u^*} \) denotes the solution of (31) associated with the control process \( u^* \).

The main purpose of this section is to find a necessary condition under which the pair \( (X^*, u^*) \) is optimal. This condition will be based on Pontryagin’s maximum principle.

To achieve this goal, we make first the following assumptions on the coefficients \( \sigma : \mathcal{P}_2(\mathbb{R}) \to \mathbb{R} \), \( b : \mathcal{P}_2(\mathbb{R} \times U) \times \mathbb{R} \times U \to \mathbb{R} \), \( g : \mathbb{R} \times \mathcal{P}_2(\mathbb{R}) \to \mathbb{R} \) and \( f : \mathcal{P}_2(\mathbb{R} \times U) \times \mathbb{R} \times U \to \mathbb{R} \):

\textbf{(H2)} \( \sigma, b, g, f \) are Lipschitz continuous, i.e., there exists a constant \( C > 0 \) such that

(i) \( |\sigma(\mu) - \sigma(\mu')| \leq C W_2(\mu, \mu') \), for any \( \mu, \mu' \in \mathcal{P}_2(\mathbb{R}) \);

(ii) \( |b(\mu, x, u) - b(\mu', x', u')| \leq C (W_2(\mu, \mu') + |x - x'| + |u - u'|) \), for any \( (\mu, x, u) \), \( (\mu', x', u') \in \mathcal{P}_2(\mathbb{R} \times U) \times \mathbb{R} \times U \);

(iii) \( |g(x, \mu) - g(x', \mu')| \leq C (|x - x'| + W_2(\mu, \mu')) \), for any \( (x, \mu) \), \( (x', \mu') \in \mathbb{R} \times \mathcal{P}_2(\mathbb{R}) \);

(iv) \( |f(\mu, x, u) - f(\mu', x', u')| \leq C (W_2(\mu, \mu') + |x - x'| + |u - u'|) \), for any \( (\mu, x, u) \), \( (\mu', x', u') \in \mathcal{P}_2(\mathbb{R} \times U) \times \mathbb{R} \times U \);

\textbf{(H3)} \( \sigma \) is differentiable in \( (\mu, x) \in \mathcal{P}_2(\mathbb{R}) \times \mathbb{R} \), and the derivative \( \partial_\mu \sigma : \mathcal{P}_2(\mathbb{R}) \times \mathbb{R} \to \mathbb{R} \) is bounded and Lipschitz continuous, i.e., there exists a constant \( C > 0 \) such that

(i) \( |\partial_\mu \sigma(\mu, y)| \leq C \), for any \( (\mu, y) \in \mathcal{P}_2(\mathbb{R}) \times \mathbb{R} \);

(ii) \( |\partial_\mu \sigma(\mu, y) - \partial_\mu \sigma(\mu', y')| \leq C (W_2(\mu, \mu') + |y - y'|) \), for any \( (\mu, y) \), \( (\mu', y') \in \mathcal{P}_2(\mathbb{R}) \times \mathbb{R} \).

\textbf{(H4)} \( w = b, f \) is differentiable in \( (\mu, x, u) \in \mathcal{P}_2(\mathbb{R} \times U) \times \mathbb{R} \times U \) and the derivatives \( \partial_\mu w : \mathcal{P}_2(\mathbb{R} \times U) \times \mathbb{R} \times U \to \mathbb{R} \times U \times \mathcal{P}_2(\mathbb{R}) \times \mathbb{R} \times U \), \( \partial_x w : \mathcal{P}_2(\mathbb{R} \times U) \times \mathbb{R} \times U \to \mathbb{R} \times U \), \( \partial_u w : \mathcal{P}_2(\mathbb{R} \times U) \times \mathbb{R} \times U \to \mathbb{R} \) are bounded and Lipschitz continuous, i.e., there exists a constant \( C > 0 \) such that

(i) \( |\partial_\mu w(\mu, x, u, y) + \partial_x w(\mu, x, u) + \partial_u w(\mu, x, u)| \leq C \), for any \( (\mu, x, u, y) \in \mathcal{P}_2(\mathbb{R} \times U) \times \mathbb{R} \times U \times \mathcal{P}_2(\mathbb{R}) \times \mathbb{R} \times U \);

(ii) \( |\partial_\mu w(\mu, x, u, y) - \partial_\mu w(\mu', x', u', y')| \leq C (W_2(\mu, \mu') + |x - x'| + |u - u'| + |y - y'|) \), for any \( (\mu, x, u, y) \), \( (\mu', x', u', y') \in \mathcal{P}_2(\mathbb{R} \times U) \times \mathbb{R} \times U \times \mathcal{P}_2(\mathbb{R}) \times \mathbb{R} \times U \);

(iii) \( |\partial_x w(\mu, x, u) - \partial_x w(\mu', x', u')| \leq C (W_2(\mu, \mu') + |x - x'| + |u - u'|) \), for any \( (\mu, x, u) \), \( (\mu', x', u') \in \mathcal{P}_2(\mathbb{R} \times U) \times \mathbb{R} \times U \);
The following SDE obtained by formal differentiation of \( U \) has a unique solution \( Y \). In the above equation, \((35)\)
\[
\lim_{\varepsilon \to 0} \sup_{t \in [0,T]} \mathbb{E} \left[ \left| Y_t - \frac{X_t - X_t^\varepsilon}{\varepsilon} \right|^2 \right] = 0.
\]
In the above equation, \((\bar{X}^\varepsilon, \bar{Y}, \bar{u}, \bar{u}^\varepsilon)\) is an independent copy of \((X^*, Y, u, u^*)\) defined on a probability space \((\Omega, \bar{F}, \bar{P})\). The expectation \(\bar{E}[\cdot]\) under \(\bar{P}\) only concerns \((\bar{X}^\varepsilon, \bar{Y}, \bar{u}, \bar{u}^\varepsilon)\) but not \((X^*, Y, u, u^*)\).

**Lemma 4.1.** The following SDE obtained by formal differentiation of \((31)\) for \( u = u^\varepsilon \) with respect to \( \varepsilon \) at \( \varepsilon = 0 \),
\[
Y_t = \int_0^t \mathbb{E} \left[ \mu(Y_s - X^\varepsilon_s) \right] dH_s
\]
\[
+ \int_0^t \mathbb{E} \left[ \left( \partial_u b \left( \mathbb{P}(X^\varepsilon_s, u^\varepsilon_s), X^s_s, u^\varepsilon_s, \bar{X}^\varepsilon_s, \bar{u}^\varepsilon_s \right) \right) Y_s \right] ds
\]
\[
+ \int_0^t \partial_u b \left( \mathbb{P}(X^\varepsilon_s, u^\varepsilon_s), X^s_s, u^\varepsilon_s \right) (u_s - u^\varepsilon_s) ds, \quad t \in [0,T],
\]
has a unique solution \( Y = (Y_t)_{t \in [0,T]} \in L^2([0,T]; \mathbb{R}) \). Moreover,
\[
\lim_{\varepsilon \to 0} \sup_{t \in [0,T]} \mathbb{E} \left[ \left| Y_t - \frac{X_t - X_t^\varepsilon}{\varepsilon} \right|^2 \right] = 0.
\]

**Remark 4.2.** With the above convention concerning \(\bar{E}[\cdot]\) we have,
\[
\bar{E} \left[ \partial_u \sigma \left( \mathbb{P}(X^\varepsilon_s, \bar{X}^\varepsilon_s) \bar{Y}^\varepsilon_s \right) \right] = \mathbb{E} \left[ \partial_u \sigma \left( \mathbb{P}(X^s_s, X^s_s) Y_s \right) \right],
\]
and
\[
\bar{E} \left[ \partial_u b \left( \mathbb{P}(X^\varepsilon_s, u^\varepsilon_s), X^s_s, u^\varepsilon_s, \bar{X}^\varepsilon_s, \bar{u}^\varepsilon_s \right) \left( \bar{Y}^\varepsilon_s, \bar{u}_s \right) \right]
\]
\[
= \mathbb{E} \left[ \left| \partial_u b \left( \mathbb{P}(X^\varepsilon_s, u^\varepsilon_s), x, u^\varepsilon_s, \bar{u}^\varepsilon_s \right) \left( Y_s, u_s - u^\varepsilon_s \right) \right| \right] \bigg|_{x=X^\varepsilon_s, u=u^\varepsilon_s}.
\]

**Proof.** (of Lemma 4.1). The existence and uniqueness of the solution \( Y \in L^2([0,T]; \mathbb{R}) \) is a special case of \((23)\). Indeed, for \( \Theta_s = (X^*_s, u^*_s, u_s), s \in [0,T], \eta \in L^2(\Omega, F, \mathbb{P}), \omega \in \Omega \), we can choose the coefficients in equation \((23)\) as follows:
\[
\gamma_s := 0;
\]
\[
\sigma(s, \mathbb{P}(\eta, \Theta_s)) := \bar{E} \left[ \partial_u \sigma \left( \mathbb{P}(X^s_s, \bar{X}^s_s) \bar{Y}^s_s \right) \right];
\]
\[
b(s, \mathbb{P}(\eta, \Theta_s), \eta, \omega) := \bar{E} \left[ \left( \partial_u b \left( \mathbb{P}(X^s_s, u^s_s), \bar{X}^s_s, \bar{u}^s_s \right) \left( \bar{Y}^s_s, \bar{u}_s \right) \right) \left( \bar{Y}^s_s + \bar{u}_s \right) \right]
\]
\[
+ \partial_x b \left( \mathbb{P}(X^s_s, u^s_s), \bar{X}^s_s, \bar{u}^s_s \right) \eta
\]
\[
+ \partial_u b \left( \mathbb{P}(X^s_s, u^s_s), \bar{X}^s_s, \bar{u}^s_s \right) (u_s - u^s_s(\omega)),
\]
for which we have
\[ |\bar{\sigma}(s, \mathbb{P}_{(\eta, \Theta_s)})| \leq C(1 + \mathbb{E}[|\eta|]); \]
\[ |\bar{b}(s, \mathbb{P}_{(\eta, \Theta_s)}, x, \omega) - \bar{b}(s, \mathbb{P}_{(\eta', \Theta_s)}, x', \omega)| \leq C \left( \mathbb{E}[|\eta - \eta'|] + |x - x'| \right), \]
for all \( s \in [0, T], \omega \in \Omega, \eta, \eta' \in L^2(\Omega, \mathcal{F}, \mathbb{P}) \) and \( x, x' \in \mathbb{R} \). Then the result follows.

The proof of (35) is split into 5 steps.

**Step 1.** Following the same method in the proof of Theorem 3.6, the only difference in the argument consists in Step 4 of the proof, where we have to take into account that we have now different \( \Theta_s \)'s. Recall also that, as \( \gamma = 0 \) here, \( \varepsilon_\gamma = 1, t \in [0, T] \). Thus, we obtain
\[
\left( \mathbb{E} \left[ |X_t^\varepsilon - X_t^\varepsilon|^2 \right] \right)^{\frac{p}{2}} \leq C \int_0^t \left( \mathbb{E} \left[ |X_s^\varepsilon - X_s^\varepsilon|^2 + |u_s^\varepsilon - u_s^\varepsilon|^2 \right] \right)^{\frac{p}{2}} ds
\]
\[
\leq C \int_0^t \left( \mathbb{E} \left[ |X_s^\varepsilon - X_s^\varepsilon|^2 \right] \right)^{\frac{p}{2}} ds + C \int_0^t \left( \mathbb{E} \left[ |u_s^\varepsilon - u_s|^2 \right] \right)^{\frac{p}{2}} ds
\]
\[
\leq C \int_0^t \left( \mathbb{E} \left[ |X_s^\varepsilon - X_s^\varepsilon|^2 \right] \right)^{\frac{p}{2}} ds + C |\varepsilon|^p.
\]
From Gronwall’s inequality, we get
\[
\left( \mathbb{E} \left[ |X_t^\varepsilon - X_t^\varepsilon|^2 \right] \right)^{\frac{p}{2}} \leq C |\varepsilon|^p.
\]
Hence, we deduce that
\[
\mathbb{E} \left[ |X_t^\varepsilon - X_t^\varepsilon|^2 \right] \leq C |\varepsilon|^2, \quad t \in [0, T].
\]
This yields, as \( \varepsilon \to 0 \), \( X_t^\varepsilon \) converges to \( X_t^\varepsilon \) in \( L^2 \), whence \( X_t^\varepsilon \) also converges to \( X_t^\varepsilon \) in probability.

**Step 2.** Let \( Y_t \) be the solution of equation (34), we want to prove that
\[
\lim_{\varepsilon \downarrow 0} \sup_{t \in [0, T]} \mathbb{E} \left[ \left| \frac{X_t^\varepsilon - X_t^\varepsilon}{\varepsilon} - Y_t \right|^2 \right] = 0.
\]
Indeed, we have
\[
\frac{X_t^\varepsilon - X_t^\varepsilon}{\varepsilon} - Y_t = \int_0^t \left\{ \frac{1}{\varepsilon} \left[ \sigma(\mathbb{P}_{X_s^\varepsilon}) - \sigma(\mathbb{P}_{X_s^\varepsilon}) \right] - \mathbb{E} \left[ \partial_\mu \sigma \left( \mathbb{P}_{X_s^\varepsilon}, \tilde{X}_s^\varepsilon, \tilde{Y}_s \right) \right] \right\} dB_s^H
\]
\[
+ \int_0^t \left\{ \frac{1}{\varepsilon} \left[ b(\mathbb{P}_{X_s^\varepsilon, u_s^\varepsilon}) - b(\mathbb{P}_{X_s^\varepsilon, u_s^\varepsilon}, X_s^\varepsilon, u_s^\varepsilon) \right] - \mathbb{E} \left[ \partial_\mu b \left( \mathbb{P}_{X_s^\varepsilon, u_s^\varepsilon}, X_s^\varepsilon, u_s^\varepsilon, \tilde{X}_s^\varepsilon, \tilde{Y}_s \right) \right] \right\} ds
\]
\[
- \mathbb{E} \left[ \left( \tilde{X}_s^\varepsilon, \tilde{Y}_s \right) \right] ds \]
In what follows, for simplicity of notations and for \( \theta \in [0, 1] \), we denote \( X_\varepsilon^* + \theta(X_s^\varepsilon - X_s^*) \) by \( X_\theta^\varepsilon \), and \( u_s^* + \theta(u_s^\varepsilon - u_s^*) \) by \( u_\theta^\varepsilon \). Then we can write

\[
\frac{1}{\varepsilon} \left[ \sigma(\mathbb{P}_{X_\varepsilon^*}) - \sigma(\mathbb{P}_{X_s^*}) \right] = \frac{1}{\varepsilon} \int_0^1 \partial_\theta \left[ \sigma \left( \mathbb{P}_{X_\theta^\varepsilon} \right) \right] d\theta = \int_0^1 \mathbb{E} \left[ \partial_\mu \sigma \left( \mathbb{P}_{X_\theta^\varepsilon}, \tilde{X}_s^\varepsilon - \tilde{X}_s^* \right) \right] d\theta,
\]

and

\[
\frac{1}{\varepsilon} \left[ b(\mathbb{P}_{(X_\varepsilon^*, u_\varepsilon^*)}, X_s^\varepsilon, u_s^*) - b(\mathbb{P}_{(X_s^*, u_s^*)}, X_s^*, u_s^*) \right]
= \frac{1}{\varepsilon} \int_0^1 \partial_\theta \left[ b \left( \mathbb{P}_{(X_\theta^\varepsilon, u_\theta^\varepsilon)} \right) \right] d\theta
= \int_0^1 \left\{ \mathbb{E} \left[ \left( \partial_\mu b \left( \mathbb{P}_{(X_\theta^\varepsilon, u_\theta^\varepsilon)}, X_\theta^\varepsilon, u_\theta^\varepsilon, \tilde{X}_s^\varepsilon, \tilde{u}_s^\varepsilon \right) \right) \right] + \partial_\theta b(\mathbb{P}_{(X_\theta^\varepsilon, u_\theta^\varepsilon)}) (u_s^* - u_s^*) \right\} d\theta.
\]

Substituting equations (38) and (39) into equation (37), we get

\[
\frac{X_\varepsilon^* - X_t^*}{\varepsilon} = Y_t
= \int_0^t \left[ \mathbb{E} \left[ \partial_\mu \sigma \left( \mathbb{P}_{X_\theta^\varepsilon}, \tilde{X}_s^\varepsilon - \tilde{X}_s^* \right) \right] \right] d\theta dB_H^s
+ \int_0^t \mathbb{E} \left[ \left( \partial_\mu b \left( \mathbb{P}_{(X_\theta^\varepsilon, u_\theta^\varepsilon)}, X_\theta^\varepsilon, u_\theta^\varepsilon, \tilde{X}_s^\varepsilon, \tilde{u}_s^\varepsilon \right) \right) \right] d\theta ds
+ \partial_\theta b(\mathbb{P}_{(X_\theta^\varepsilon, u_\theta^\varepsilon)}) (u_s^* - u_s^*)
- \mathbb{E} \left[ \left( \partial_\mu b \left( \mathbb{P}_{(X_s^*, u_s^*)}, X_s^*, u_s^*, \tilde{X}_s^*, \tilde{u}_s^* \right) \right) \right]
- \partial_\theta b(\mathbb{P}_{(X_s^*, u_s^*)}) (u_s^* - u_s^*)
\]

\( = I_6(t) + I_7(t) \).

**Step 3. The term** \( I_6(t) \). Recall that

\[
I_6(t) = \int_0^t \int_0^1 \mathbb{E} \left[ \left( \partial_\mu \sigma \left( \mathbb{P}_{X_\theta^\varepsilon}, \tilde{X}_s^\varepsilon \right) - \partial_\mu \sigma \left( \mathbb{P}_{X_s^*, \tilde{X}_s^*} \right) \right) \right] d\theta dB_H^s
+ \int_0^t \int_0^1 \mathbb{E} \left[ \partial_\mu \sigma \left( \mathbb{P}_{X_s^*, \tilde{X}_s^*} \right) \right] d\theta dB_H^s.
\]
By applying the same method and the same $p$ as in the proof of Theorem 3.6 (with $\gamma \equiv 0$), observing that the integrands with respect to $B^H$ are deterministic, we get

$$
\mathbb{E}[I_6^2(t)] \leq C \left\{ \int_0^t \mathbb{E}\left[ \left\| \int_0^1 \left( \partial_\mu \sigma \left( \mathbb{P}_{X_s^\theta}, X_s^\theta \right) - \partial_\mu \sigma \left( \mathbb{P}_{X_s^\theta}, X_s^\theta \right) \right) \frac{X_s^\varepsilon - X_s^*: \varepsilon}{d\theta} \right\|^p \right\}^{\frac{2}{p}}
+ C \left\{ \int_0^t \mathbb{E}\left[ \left\| \int_0^1 \partial_\mu \sigma \left( \mathbb{P}_{X_s^\theta}, X_s^\theta \right) \left( \frac{X_s^\varepsilon - X_s^*}{\varepsilon} - Y_s \right) d\theta \right\|^p \right\}^{\frac{2}{p}}
\right\}^{\frac{2}{p}}
$$

(41)

Hence, from the results of Step 1, we deduce from the bounded convergence theorem that the first term of right hand side in the above inequality converges to 0 as $\varepsilon \to 0$.

**Step 4. The term $I_7(t)$.** We can write $I_7(t)$ as follows:

$$
I_7(t) = \int_0^t \mathbb{E}\left[ \left( \partial_\mu b \left( \mathbb{P}_{(X_s^\theta, u_s^\theta)}, X_s^\theta, u_s^\theta, \tilde{X}_s^\theta, \tilde{u}_s^\theta \right) - \partial_\mu b \left( \mathbb{P}_{(X_s^\theta, u_s^\theta)}, X_s^\theta, u_s^\theta, \tilde{X}_s^\theta, \tilde{u}_s^\theta \right) \right) \right]
+ \mathbb{E}\left[ \left( \partial_\sigma b \left( \mathbb{P}_{(X_s^\theta, u_s^\theta)}, X_s^\theta, u_s^\theta, \tilde{X}_s^\theta, \tilde{u}_s^\theta \right) \left( \frac{\tilde{X}_s^\varepsilon - \tilde{X}_s^*}{\varepsilon} \right) - \tilde{Y}_s - 0 \right) \right]
+ \mathbb{E}\left[ \left( \partial_\sigma b \left( \mathbb{P}_{(X_s^\theta, u_s^\theta)}, X_s^\theta, u_s^\theta, \tilde{X}_s^\theta, \tilde{u}_s^\theta \right) \left( \frac{X_s^\varepsilon - X_s^*}{\varepsilon} - Y_s \right) \right]
+ \mathbb{E}\left[ \left( \partial_\sigma b \left( \mathbb{P}_{(X_s^\theta, u_s^\theta)}, X_s^\theta, u_s^\theta, \tilde{X}_s^\theta, \tilde{u}_s^\theta \right) \left( u_s - u_s^* \right) \right) \right] d\theta ds.
$$

From our assumptions, we have

$$
\mathbb{E}[I_7^2(t)] \leq C \left\{ \int_0^t \mathbb{E}\left[ \left( \left( \mathbb{E} \left[ |X_s^\varepsilon - X_s^*|^2 \right] \right)^{\frac{1}{2}} + |X_s^\varepsilon - X_s^*| + \mathbb{E} \left[ |u_s - u_s^*|^2 \right] \right)^{\frac{1}{2}} + |u_s - u_s^*| \right) \right\]^{\frac{2}{p}}
\times \left( \mathbb{E} \left[ \left( \left( \frac{X_s^\varepsilon - X_s^*}{\varepsilon} \right) + |u_s - u_s^*| \right) \right] \right)^{\frac{2}{p}}
\right\}^{\frac{2}{p}}
$$

(42)

$$
+ C \left( \int_0^t \mathbb{E}\left[ \left( \left( \frac{X_s^\varepsilon - X_s^*}{\varepsilon} \right) - Y_s \right) \right]^p \right) \right\}^{\frac{2}{p}}.
$$
From Step 1, we get by the bounded convergence theorem that the first term of (42) converges to zero when \( \varepsilon \to 0 \).

**Step 5.** Combining the relations (40), (41) and (42), we deduce that

\[
\mathbb{E} \left[ \frac{X_t^\varepsilon - X_t^*}{\varepsilon} - Y_t \right]^2 \leq C_\varepsilon + C \int_0^t \left( \mathbb{E} \left[ \frac{X_s^\varepsilon - X_s^*}{\varepsilon} - Y_s \right]^2 \right)^{\frac{2}{p}} ds, \quad t \in [0, T],
\]

where \( C_\varepsilon \to 0 \) as \( \varepsilon \to 0 \). Consequently,

\[
\sup_{r \in [0,t]} \left( \mathbb{E} \left[ \frac{X_r^\varepsilon - X_r^*}{\varepsilon} - Y_r \right]^2 \right)^{\frac{2}{p}} \leq C_\varepsilon + C \int_0^t \left( \mathbb{E} \left[ \frac{X_s^\varepsilon - X_s^*}{\varepsilon} - Y_s \right]^2 \right)^{\frac{2}{p}} ds, \quad t \in [0, T],
\]

and Gronwall’s inequality allows to conclude that

\[
\sup_{t \in [0,T]} \left( \mathbb{E} \left[ \frac{X_t^\varepsilon - X_t^*}{\varepsilon} - Y_t \right]^2 \right)^{\frac{2}{p}} \leq C_\varepsilon \to 0, \quad \text{as} \quad \varepsilon \to 0.
\]

This proves that \( \frac{X_t^\varepsilon - X_t^*}{\varepsilon} \) converges to \( Y_t \) in \( L^2 \), uniformly in \( t \in [0, T] \). Our proof is completed. \( \square \)

As \( (X^*, u^*) \) is an optimal pair, \( J(u^\varepsilon) \geq J(u^*), \varepsilon \in [0, 1], \) and therefore,

\[
\frac{d}{d\varepsilon} J(u^\varepsilon) \bigg|_{\varepsilon=0} = \lim_{0<\varepsilon \to 0} \frac{1}{\varepsilon} (J(u^\varepsilon) - J(u^*)) \geq 0.
\]

That is, due to Lemma 4.1 and the computations on \( f \) and \( g \),

\[
0 \leq \frac{d}{d\varepsilon} J(u^\varepsilon) \bigg|_{\varepsilon=0} = \mathbb{E} \left[ \partial_x g(X_T^\varepsilon, \mathbb{P}X_T^\varepsilon) Y_T \right] + \mathbb{E} \left[ \tilde{\mathbb{E}} \left[ \partial_\mu g(X_T^\varepsilon, \mathbb{P}X_T^\varepsilon, \tilde{X}_T^\varepsilon, \tilde{Y}_T) \right] \right] + \mathbb{E} \left[ \int_0^T \partial_x f(\mathbb{P}(X_t^\varepsilon, u_t^\varepsilon), X_t^\varepsilon, u_t^\varepsilon) Y_t dt \right] + \mathbb{E} \left[ \int_0^T \tilde{\mathbb{E}} \left[ (\partial_\mu f)1(\mathbb{P}(X_t^\varepsilon, u_t^\varepsilon), X_t^\varepsilon, u_t^\varepsilon, \tilde{X}_t^\varepsilon, \tilde{u}_t^\varepsilon) \tilde{Y}_t \right] dt \right] + \mathbb{E} \left[ \int_0^T \partial_u f(\mathbb{P}(X_t^\varepsilon, u_t^\varepsilon), X_t^\varepsilon, u_t^\varepsilon) (u_t - u_t^\varepsilon) dt \right].
\]

We define an adjoint process \( P_t \) by

\[
P_t = P_T - \int_t^T \alpha_s ds - \int_t^T \beta_s dW_s,
\]

where \( \{\alpha_s, s \in [0, T]\} \) is a square integrable progressively measurable process, which will be specified later, \( P_T = \partial_x g(X_T^*, \mathbb{P}X_T^*) + \mathbb{E} \left[ \partial_\mu g(X_T^*, \mathbb{P}X_T^*, X_T^*) \right] \), and \( (P_t)_{t \in [0,T]}, \beta = (\beta_t)_{t \in [0,T]} \) is the solution of the backward equation (45). In particular, \( P = (P_t)_{t \in [0,T]} \) and \( \beta = (\beta_t)_{t \in [0,T]} \) are square integrable, progressively measurable processes.
We apply Ito’s formula (Corollary 2.6) to $Y_t P_t$ and get
\[
\begin{align*}
\,dY_t P_t &= P_t \,dY_t + Y_t \,dP_t + \mathbb{E} \left[ \partial_t \sigma \left( \mathbb{P}_{X_t^*}, \bar{X}_t^* \right) \bar{Y}_t \right] \,dH^t P_t \,dt \\
&= P_t \mathbb{E} \left[ \partial_t \sigma \left( \mathbb{P}_{X_t^*}, \bar{X}_t^* \right) \bar{Y}_t \right] \,dB_t^H \\
&\quad + P_t \mathbb{E} \left[ \left( \partial_t b \left( \mathbb{P}_{\left( X_t^*, u_t^* \right)}, X_t^*, u_t^*, \bar{X}_t^*, \bar{u}_t^* \right), \left( \bar{Y}_t, \bar{u}_t - \bar{u}_t^* \right) \right) \right] \,dt \\
&\quad + P_t \partial_s b \left( \mathbb{P}_{\left( X_t^*, u_t^* \right)}, X_t^*, u_t^* \right) Y_t \,dt + P_t \partial_u b \left( \mathbb{P}_{\left( X_t^*, u_t^* \right)}, X_t^*, u_t^* \right) \left( u_t - u_t^* \right) \,dt \\
&\quad + Y_t \alpha \,dt + Y_t \beta \,dW_t + \mathbb{E} \left[ \partial_t \sigma \left( \mathbb{P}_{X_t^*}, \bar{X}_t^* \right) \bar{Y}_t \right] \,dH^t P_t \,dt.
\end{align*}
\]
\[(46)\]

Therefore, by integrating over the interval $[0, T]$ and considering that $Y_0 = 0$, we have
\[
\begin{align*}
Y_T P_T &= \int_0^T P_t \mathbb{E} \left[ \partial_t \sigma \left( \mathbb{P}_{X_t^*}, \bar{X}_t^* \right) \bar{Y}_t \right] \,dB_t^H + \int_0^T Y_s \alpha \,ds \\
&\quad + \int_0^T P_t \mathbb{E} \left[ \left( \partial_t b \left( \mathbb{P}_{\left( X_t^*, u_t^* \right)}, X_t^*, u_t^*, \bar{X}_t^*, \bar{u}_t^* \right), \left( \bar{Y}_s, \bar{u}_s - \bar{u}_s^* \right) \right) \right] \,ds \\
&\quad + \int_0^T P_t \partial_s b \left( \mathbb{P}_{\left( X_t^*, u_t^* \right)}, X_t^*, u_t^* \right) Y_s \,ds + \int_0^T P_t \partial_u b \left( \mathbb{P}_{\left( X_t^*, u_t^* \right)}, X_t^*, u_t^* \right) \left( u_s - u_s^* \right) \,ds \\
&\quad + \int_0^T Y_s \alpha \,ds + \int_0^T \mathbb{E} \left[ \partial_t \sigma \left( \mathbb{P}_{X_t^*}, \bar{X}_t^* \right) \bar{Y}_s \right] \,dH^t P_s \,ds.
\end{align*}
\[(47)\]

By taking expectations with respect to $\mathbb{P}$, we get
\[
\mathbb{E} \left[ Y_T P_T \right] = \int_0^T \mathbb{E} \left[ P_t \mathbb{E} \left[ \left( \partial_t b \left( \mathbb{P}_{\left( X_t^*, u_t^* \right)}, X_t^*, u_t^*, \bar{X}_t^*, \bar{u}_t^* \right), \left( \bar{Y}_s, \bar{u}_s - \bar{u}_s^* \right) \right) \right] \,ds \\
&\quad + \int_0^T \mathbb{E} \left[ P_t \partial_s b \left( \mathbb{P}_{\left( X_t^*, u_t^* \right)}, X_t^*, u_t^* \right) Y_s \right] \,ds \\
&\quad + \int_0^T \mathbb{E} \left[ P_t \partial_u b \left( \mathbb{P}_{\left( X_t^*, u_t^* \right)}, X_t^*, u_t^* \right) \left( u_s - u_s^* \right) \right] \,ds \\
&\quad + \int_0^T \mathbb{E} \left[ Y_s \alpha \,ds \right] + \int_0^T \mathbb{E} \left[ \partial_t \sigma \left( \mathbb{P}_{X_t^*}, \bar{X}_t^* \right) \bar{Y}_s \right] \,dH^t P_s \,ds.
\]
\[(48)\]

From the definition of $\left( \tilde{\Omega}, \tilde{F}, \tilde{\mathbb{P}} \right)$, it follows that
\[
\begin{align*}
\int_0^T \mathbb{E} \left[ P_s \mathbb{E} \left[ \left( \partial_t b \left( \mathbb{P}_{\left( X_t^*, u_t^* \right)}, X_t^*, u_t^*, \bar{X}_t^*, \bar{u}_t^* \right), \left( \bar{Y}_s, \bar{u}_s - \bar{u}_s^* \right) \right) \right] \right] \,ds \\
&= \int_0^T \mathbb{E} \left[ \mathbb{E} \left[ \mathbb{E} \left[ \left( \partial_t b \left( \mathbb{P}_{\left( X_t^*, u_t^* \right)}, X_t^*, u_t^*, \bar{X}_t^*, \bar{u}_t^* \right), \left( \bar{Y}_s, \bar{u}_s - \bar{u}_s^* \right) \right) \right] \right] \,ds \\
&= \int_0^T \mathbb{E} \left[ \mathbb{E} \left[ \partial_t b \left( \mathbb{P}_{\left( X_t^*, u_t^* \right)}, X_t^*, u_t^*, \bar{X}_t^*, \bar{u}_t^* \right) \left( \bar{Y}_s, \bar{u}_s - \bar{u}_s^* \right) \right] \,ds \\
&\quad + \int_0^T \mathbb{E} \left[ \mathbb{E} \left[ \partial_t b \left( \mathbb{P}_{\left( X_t^*, u_t^* \right)}, X_t^*, u_t^*, \bar{X}_t^*, \bar{u}_t^* \right) \left( \bar{Y}_s, \bar{u}_s - \bar{u}_s^* \right) \right] \,ds.
\end{align*}
\[(49)\]
Moreover, for the latter term of equation (48), we have

\[
E \left[ Y_T P_T \right] = \int_0^T E \left[ \left( \tilde{P}_s (\partial_\mu b)_{1} \left( P_{(X^*_s, u^*_s)}, \tilde{X}_s^*, \tilde{u}_s^*, X^*_s \right) \right) + P_s \partial_\mu b \left( P_{(X^*_s, u^*_s)}, X^*_s, u^*_s \right) + \alpha_s + \partial_\mu \sigma \left( P_{X^*_s}, X^*_s \right) E \left[ D_H^s P_s \right] \right] ds
\]

Substituting equations (49) and (50) in equation (48), we obtain

\[
\int_0^T E \left[ \left( \tilde{P}_s (\partial_\mu b)_{1} \left( P_{(X^*_s, u^*_s)}, \tilde{X}_s^*, \tilde{u}_s^*, X^*_s \right) \right) + P_s \partial_\mu b \left( P_{(X^*_s, u^*_s)}, X^*_s, u^*_s \right) + \alpha_s + \partial_\mu \sigma \left( P_{X^*_s}, X^*_s \right) E \left[ D_H^s P_s \right] \right] ds
\]

Now we substitute equation (44) in (51), and get

\[
0 \leq \int_0^T E \left[ Y_s \left( \tilde{P}_s (\partial_\mu b)_{1} \left( P_{(X^*_s, u^*_s)}, \tilde{X}_s^*, \tilde{u}_s^*, X^*_s \right) \right) + P_s \partial_\mu b \left( P_{(X^*_s, u^*_s)}, X^*_s, u^*_s \right) + \alpha_s + \partial_\mu \sigma \left( P_{X^*_s}, X^*_s \right) E \left[ D_H^s P_s \right] + \partial_x f \left( P_{(X^*_s, u^*_s)}, X^*_s, u^*_s \right) \right] ds
\]

Letting the first integral, which integrand contains \( Y_s \), equal to zero, we get

\[
\alpha_s = - E \left[ \tilde{P}_s (\partial_\mu b)_{1} \left( P_{(X^*_s, u^*_s)}, \tilde{X}_s^*, \tilde{u}_s^*, X^*_s \right) \right] - P_s \partial_\mu b \left( P_{(X^*_s, u^*_s)}, X^*_s, u^*_s \right)
\]

This gives the following form of the BSDE for \( P_t = P_T - \int_t^T \alpha_s ds - \int_t^T \beta_s dW_s \):

\[
P_t = P_T + \int_t^T \left\{ E \left[ \tilde{P}_s (\partial_\mu b)_{1} \left( P_{(X^*_s, u^*_s)}, \tilde{X}_s^*, \tilde{u}_s^*, X^*_s, u^*_s \right) \right] + P_s \partial_\mu b \left( P_{(X^*_s, u^*_s)}, X^*_s, u^*_s \right) + \partial_\mu \sigma \left( P_{X^*_s}, X^*_s \right) E \left[ D_H^s P_s \right] + \partial_x f \left( P_{(X^*_s, u^*_s)}, X^*_s, u^*_s \right) \right\] ds - \int_t^T \beta_s dW_s
\]

which is a mean-field BSDE driven by the standard Brownian motion \( W \). Such kind of mean-field BSDE (without the term of Malliavin derivative) was studied firstly by Buckdahn et al. \[4\] \[5\]. We recall again that in the above BSDE, the expectation \( \tilde{E} \) only concerns the processes with tildes.
We suppose that there exists such a solution \((P, \beta)\) of (54); its existence and uniqueness will be discussed later for a special case.

With this choice of \(P_t\), equation (52) now becomes

\[
0 \leq \int_0^T \mathbb{E} \left( u_s - u^*_s \right) \left\{ \mathbb{E} \left[ (\partial_\mu f)_2 (\mathbb{P}(X^*_s, u^*_s), X^*_s, u^*_s) \right] + \partial_u f (\mathbb{P}(X^*_s, u^*_s), X^*_s, u^*_s) \right\} \right) ds.
\]

(55)

From the fact that \(U\) is open and from the arbitrariness of \(u \in \mathcal{U}([0, T])\), we have

\[
0 = \mathbb{E} \left[ (\partial_\mu f)_2 (\mathbb{P}(X^*_t, u^*_t), X^*_t, u^*_t) \right] + \partial_u f (\mathbb{P}(X^*_t, u^*_t), X^*_t, u^*_t)
+ \mathbb{E} \left[ \tilde{P}_t (\partial_\mu b)_2 \left( \mathbb{P}(X^*_t, u^*_t), X^*_t, u^*_t \right) \right] + P_t \partial_u b \left( \mathbb{P}(X^*_t, u^*_t), X^*_t, u^*_t \right), \quad \text{dP-a.s., dt-a.e.}
\]

(56)

Now we can conclude the above calculations in the following necessary conditions of Pontryagin-type maximum principle, which is our main result.

**Theorem 4.3.** If \((X^*, u^*)\) is an optimal pair of mean-field stochastic control problem (31) – (33), then \((X^*, u^*)\) satisfies the following system:

\[
\begin{align*}
X^*_t = & x + \int_0^t \sigma (\mathbb{P}(X^*_s, u^*_s)) dB^H_s + \int_0^t b(\mathbb{P}(X^*_s, u^*_s), X^*_s, u^*_s) ds, \\
P_T = & \partial_x g(X^*_T, \mathbb{P}_{X^*_T}) + \mathbb{E} \left[ \partial_\mu g(\mathbb{X}^*_T, \mathbb{P}_{X^*_T}, X^*_T) \right], \\
P_t = & P_T - \int_t^T \beta_s dW_s + \int_t^T \left\{ \mathbb{E} \left[ \tilde{P}_s (\partial_\mu b)_1 \left( \mathbb{P}(X^*_s, u^*_s), \mathbb{X}^*_s, \mathbb{u}^*_s, X^*_s, u^*_s \right) \right] \\
& + P_s \partial_x f(\mathbb{P}(X^*_s, u^*_s), X^*_s, u^*_s) + \mathbb{E} \left[ (\partial_\mu f)_1 (\mathbb{P}(X^*_s, u^*_s), \mathbb{X}^*_s, \mathbb{u}^*_s, X^*_s, u^*_s) \right] \right\} ds,
\end{align*}
\]

(57)

\[
0 = \mathbb{E} \left[ \tilde{P}_t (\partial_\mu b)_2 \left( \mathbb{P}(X^*_t, u^*_t), X^*_t, u^*_t \right) \right] + P_t \partial_u b \left( \mathbb{P}(X^*_t, u^*_t), X^*_t, u^*_t \right), \quad \text{dP-a.s., dt-a.e.}
\]

We can also give a sufficient condition for optimality under some more assumptions. In the following we define our Hamiltonian, for \((\mu, x, u, y, z) \in \mathcal{P}_2(\mathbb{R} \times U) \times \mathbb{R} \times U \times \mathbb{R} \times \mathbb{R},

\[H(\mu, x, u, y, z) := f(\mu, x, u) + b(\mu, x, u)y + \sigma (\mu)z.\]

For the following assumption, we recall the definition of joint convexity (21).

**(H6)** \(g : \mathbb{R} \times \mathcal{P}_2(\mathbb{R}) \to \mathbb{R}\) is jointly convex in \((x, \mu)\). The Hamiltonian \(H(\mu, x, u, y, z)\) is jointly convex in \((\mu, x, u)\).

**Theorem 4.4.** Suppose (H2)-(H6) hold. Let \((u^*_t, X^*_t)_{t \in [0, T]}\) satisfy system (57). Then \((u^*_t, X^*_t)_{t \in [0, T]}\) is optimal and \(J(u^*) = \inf_{u \in \mathcal{U}([0, T])} J(u)\).
Proof. Suppose \((u_t, X^u_t)\) is an arbitrary control and the corresponding state. Then from the definition of cost functional \(J(u)\), we have

\[
J(u^*) - J(u) = \mathbb{E} \left[ g(X^*_T, \mathbb{P}_{X^*_T}) - g(X^u_T, \mathbb{P}_{X^u_T}) \right] + \mathbb{E} \left[ \int_0^T \left( f(\mathbb{P}_{X^*_t,u^*_t}, X^*_t, u^*_t) - f(\mathbb{P}_{X^u_t,u^*_t}, X^u_t, u_t) \right) \, dt \right].
\]

(58)

Hence from the joint convexity of \(g\), we have

\[
J(u^*) - J(u) \leq \mathbb{E} \left[ \left( \partial_x g(X^*_T, \mathbb{P}_{X^*_T}) + \bar{\mathbb{E}} \left[ \partial_\mu g(\bar{X}^*_T, \mathbb{P}_{\bar{X}^*_T}, X^*_T) \right] \right) (X^*_T - X^u_T) \right] + \mathbb{E} \left[ \int_0^T \left( f(\mathbb{P}_{X^*_t,u^*_t}, X^*_t, u^*_t) - f(\mathbb{P}_{X^u_t,u^*_t}, X^u_t, u_t) \right) \, dt \right].
\]

(59)

By applying Ito’s formula (equation (17)) to \(P_t(X^*_t - X^u_t)\) and taking the expectation on both sides, we get:

\[
\mathbb{E} \left[ P_T(X^*_T - X^u_T) \right] + \mathbb{E} \left[ \int_0^T \left( f(\mathbb{P}_{X^*_s,u^*_s}, X^*_s, u^*_s) - f(\mathbb{P}_{X^u_s,u^*_s}, X^u_s, u_s) \right) \, ds \right]
\]

\[
= \int_0^T \mathbb{E} \left[ P_s \left( b(\mathbb{P}_{X^*_s,u^*_s}, X^*_s, u^*_s) - b(\mathbb{P}_{X^u_s,u^*_s}, X^u_s, u_s) \right) \right] \, ds
\]

\[
- \int_0^T \mathbb{E} \left[ \partial_\mu b (\mathbb{P}_{X^*_s,u^*_s}, X^*_s, u^*_s) (X^*_s - X^u_s) \right] \, ds
\]

\[
- \int_0^T \mathbb{E} \left[ \partial_\mu f (\mathbb{P}_{X^*_s,u^*_s}, X^*_s, u^*_s) (X^*_s - X^u_s) \right] \, ds
\]

\[
+ \int_0^T \mathbb{E} \left[ \partial_\mu \sigma (\mathbb{P}_{X^*_s}) - \sigma (\mathbb{P}_{X^u_s}) \right] \mathbb{E} \left[ D_s^H P_s \right] \, ds
\]

\[
- \int_0^T \mathbb{E} \left[ \partial_\mu \sigma (\mathbb{P}_{X^*_s}) (X^*_s - X^u_s) \right] \mathbb{E} \left[ D_s^H P_s \right] \, ds
\]

\[
- \int_0^T \mathbb{E} \left[ \partial_x f (\mathbb{P}_{X^*_s,u^*_s}, X^*_s, u^*_s) (X^*_s - X^u_s) \right] \, ds
\]

\[
- \int_0^T \mathbb{E} \left[ \partial_x g (\mathbb{P}_{X^*_s,u^*_s}, X^*_s, u^*_s, \tilde{X}^*_s, \tilde{u}_s) (\tilde{X}^*_s - \tilde{X}^u_s) \right] \, ds.
\]

(60)
From the joint convexity of Hamiltonian $H$ we get
\begin{equation}
\begin{aligned}
P_s(b(\mathbb{P}(X^*_s,u^*_s),X^*_s,u^*_s))-b(\mathbb{P}(X^*_s,u_s),X^*_s,u^*_s)) + \left(\sigma(\mathbb{P}X^*_s) - \sigma(\mathbb{P}X^*_s)\right) \mathbb{E} \left[ \mathbb{D}^H P_s \right] \\
+ f(\mathbb{P}(X^*_s,u^*_s),X^*_s,u^*_s) - f(\mathbb{P}(X^*_s,u_s),X^*_s,u_s)
\end{aligned}
\leq P_s \mathbb{E} \left[ (\partial_\mu b) \right] \mathbb{P}(X^*_s,u^*_s),X^*_s,u^*_s,\bar{X}^*_s,\bar{u}^*_s] \left( \bar{X}^*_s - \bar{X}^*_s \right) + P_s \partial_x b \mathbb{P}(X^*_s,u^*_s),X^*_s,u^*_s \left( X^*_s - X^*_s \right) \\
+ P_s \mathbb{E} \left[ (\partial_\mu b) \right] \mathbb{P}(X^*_s,u^*_s),X^*_s,u^*_s,\bar{X}^*_s,\bar{u}^*_s] \left( \bar{u}^*_s - \bar{u}^*_s \right) + P_s \partial_u b \mathbb{P}(X^*_s,u^*_s),X^*_s,u^*_s \left( u^*_s - u_s \right) \\
+ \mathbb{E} \left[ \partial_\mu \sigma \mathbb{P}(X^*_s,X^*_s) \left( X^*_s - X^*_s \right) \right] \mathbb{E} \left[ \mathbb{D}^H P_s \right] \\
+ \partial_x f \mathbb{P}(X^*_s,u^*_s),X^*_s,u^*_s \left( X^*_s - X^*_s \right) + \mathbb{E} \left[ (\partial_\mu f) \right] \mathbb{P}(X^*_s,u^*_s),X^*_s,u^*_s,\bar{X}^*_s,\bar{u}^*_s] \left( \bar{X}^*_s - \bar{X}^*_s \right) \\
+ \mathbb{E} \left[ (\partial_\mu f) \right] \mathbb{P}(X^*_s,u^*_s),X^*_s,u^*_s,\bar{X}^*_s,\bar{u}^*_s] \left( \bar{u}^*_s - \bar{u}^*_s \right) + \partial_u f \mathbb{P}(X^*_s,u^*_s),X^*_s,u^*_s \left( u^*_s - u_s \right),
\end{aligned}
\end{equation}
where due to equation (57),
\begin{equation}
\begin{aligned}
\mathbb{E} \left[ (\partial_\mu f) \right] \mathbb{P}(X^*_s,u^*_s),X^*_s,u^*_s,\bar{X}^*_s,\bar{u}^*_s] \left( \bar{u}^*_s - \bar{u}^*_s \right) + \mathbb{E} \left[ \partial_u f \mathbb{P}(X^*_s,u^*_s),X^*_s,u^*_s \left( u^*_s - u_s \right) \\
+ \mathbb{E} \left[ \partial_\mu \sigma \mathbb{P}(X^*_s,X^*_s) \left( X^*_s - X^*_s \right) \right] \mathbb{E} \left[ \mathbb{D}^H P_s \right] \\
+ \partial_x f \mathbb{P}(X^*_s,u^*_s),X^*_s,u^*_s \left( X^*_s - X^*_s \right) + \mathbb{E} \left[ (\partial_\mu f) \right] \mathbb{P}(X^*_s,u^*_s),X^*_s,u^*_s,\bar{X}^*_s,\bar{u}^*_s] \left( \bar{X}^*_s - \bar{X}^*_s \right) \\
+ \mathbb{E} \left[ (\partial_\mu f) \right] \mathbb{P}(X^*_s,u^*_s),X^*_s,u^*_s,\bar{X}^*_s,\bar{u}^*_s] \left( \bar{u}^*_s - \bar{u}^*_s \right) + \partial_u f \mathbb{P}(X^*_s,u^*_s),X^*_s,u^*_s \left( u^*_s - u_s \right)
\end{aligned}
\end{equation}
\begin{equation}
\begin{aligned}
= \mathbb{E} \left[ \mathbb{E} \left[ (\partial_\mu f) \right] \mathbb{P}(X^*_s,u^*_s),\bar{X}^*_s,\bar{u}^*_s,\bar{X}^*_s,\bar{u}^*_s] \right] + \partial_u f \mathbb{P}(X^*_s,u^*_s),X^*_s,u^*_s \\
+ \mathbb{E} \left[ \mathbb{E} \left[ (\partial_\mu f) \right] \mathbb{P}(X^*_s,u^*_s),\bar{X}^*_s,\bar{u}^*_s,\bar{X}^*_s,\bar{u}^*_s] \right] + P_s \partial_u b \mathbb{P}(X^*_s,u^*_s),X^*_s,u^*_s \left( u^*_s - u_s \right)
\end{aligned}
\end{equation}
\begin{equation}
\begin{aligned}
= 0.
\end{aligned}
\end{equation}
Therefore we get from the equations (58)-(61) that
\begin{equation}
J(u^*) - J(u) \leq 0,
\end{equation}
which means $(u^*,X^*)$ is an optimal pair. \qed

**Remark 4.5.** We emphasise that in (H6), similarly to Carmona and Delarue [8], we assume the convexity of the Hamiltonian $H$. If there is no running cost function $f$, supposing convexity is in some sense equivalent to assuming linearity, because of the multiplications with $P_s$ and $\mathbb{E} \left[ \mathbb{D}^H P_t \right]$, respectively, which sign can change. With the assumption of linearity, the inequality in (61) become equality.

In the following we give another sufficient condition which allows to have more general coefficients which are not necessarily linear. For this we need the following assumption and we recall the definition of strict convexity [22].

(H7). Then function $g : \mathbb{R} \times \mathcal{P}_2(\mathbb{R}) \to \mathbb{R}$ is jointly convex in $(x,\mu)$, with $\partial_x g \geq 0$ and $\partial_\mu g \geq 0$, and $b(\eta, x, u) : \mathcal{P}_2(\mathbb{R} \times U) \times \mathbb{R} \times U \to \mathbb{R}$ is jointly convex in $(\eta, x, u)$ with $(\partial_\eta b) \geq 0$ and strictly convex in $(\mu, u)$. Moreover, $f(\eta, x, u) : \mathcal{P}_2(\mathbb{R} \times U) \times \mathbb{R} \times U \to \mathbb{R}$ is jointly convex in $(\eta, x, u)$ and strictly convex in $(\mu, u)$, with $(\partial_\mu f) \geq 0$, and $\sigma(\mu) \equiv \sigma \in \mathbb{R}$.

**Theorem 4.6.** Suppose (H2)-(H5) and (H7) hold. Let $(u^*_t,X^*_t)_{t \in [0,T]}$ satisfy system (57). Then $(u^*_t,X^*_t)_{t \in [0,T]}$ is optimal and $J(u^*) = \inf_{u \in U_t} J(u)$. 

Remark 4.7. The existence of such functions can be easily verified. For instance, we can choose $g$ as an increasing convex function with bounded derivatives. For an increasing joint convex function $\bar{b} : \mathbb{R} \times U \times \mathbb{R} \times U$ with bounded derivatives, we can choose $b \left( P^t(X^*_s, u^*_s), X^*_s, u^*_s \right) := \tilde{E} \left[ \bar{b} \left( \tilde{X}^*_s, \tilde{u}^*_s, X^*_s, u^*_s \right) \right]$.

Proof. With these assumptions, the BSDE in system (57) now becomes
\begin{align*}
\left(63\right) 
P_t &= \partial_x g(X^*_T, P_X^T) + \tilde{E} \left[ \partial_\mu g(X^*_T, P_X^T, X^*_T) \right] \\
&+ \int_t^T \tilde{E} \left[ \tilde{P}_s(\partial_\mu b)_1 \left( P(X^*_s, u^*_s), \tilde{X}^*_s, \tilde{u}^*_s, X^*_s, u^*_s \right) \right] ds \\
&+ \int_t^T \left[ \partial_x f(P(X^*_s, u^*_s), X^*_s, u^*_s) + \tilde{E} \left[ (\partial_\mu f)_1(P(X^*_s, u^*_s), \tilde{X}^*_s, \tilde{u}^*_s, X^*_s, u^*_s) \right] \right] ds - \int_t^T \beta_s dW_s,
\end{align*}

which is a mean-field BSDE in the classical sense which was studied by Buckdahn et. al. [6]. We compare it with the following BSDE
\begin{align*}
\left(64\right) 
Q_t &= 0 + \int_t^T \tilde{E} \left[ \tilde{Q}_s(\partial_\mu b)_1 \left( P(X^*_s, u^*_s), \tilde{X}^*_s, \tilde{u}^*_s, X^*_s, u^*_s \right) \right] ds \\
&+ \int_t^T Q_s \partial_2 b \left( P(X^*_s, u^*_s), X^*_s, u^*_s \right) ds - \int_t^T Z_s dW_s,
\end{align*}

which has a unique solution $(Q_t, Z_t) = (0, 0)$. From the comparison result (Theorem 3.2) in [6], we see that $P_t \geq Q_t = 0$, for all $t \in [0, T]$. On the other hand, due to $(H3)-(H5)$ and $(H7)$, we have $0 \leq \partial_x g \leq C$, $0 \leq \partial_\mu g \leq C$, $0 \leq (\partial_\mu b)_1 \leq C$ and $\vert \partial_x b \vert \leq C$. We compare equation (63) again with the following BSDE
\begin{align*}
\left(65\right) 
Q'_t &= C + C \int_t^T \tilde{E} \left[ Q'_s \right] ds + C \int_t^T Q'_s ds - \int_t^T Z'_s dW_s,
\end{align*}
which has a unique solution \((Q_t', Z_t') = (C \exp\{2C(T-t)\}, 0)\), \(t \in [0, T]\). From the comparison result again, we get \(P_t \leq Q_t'\), hence \(P_t\) is uniformly bounded. The equation \((60)\) becomes

\[
\mathbb{E} [P_T(X^*_T - X^*_T)] + \mathbb{E} \left[ \int_0^T \left( f(\mathbb{P}(X^*_s, u^*_s), X^*_s, u^*_s) - f(\mathbb{P}(X^*_s, u^*_s), X^*_s, u^*_s) \right) ds \right]
\]

\[
= \mathbb{E} \left[ \int_0^T \left( f(\mathbb{P}(X^*_s, u^*_s), X^*_s, u^*_s) - f(\mathbb{P}(X^*_s, u^*_s), X^*_s, u^*_s) \right) ds \right]
\]

\[
+ \int_0^T \mathbb{E} \left[ P_s (\mathbb{P}(X^*_s, u^*_s), X^*_s, u^*_s) - b(\mathbb{P}(X^*_s, u^*_s), X^*_s, u^*_s)) \right] ds
\]

\[
- \int_0^T \mathbb{E} \left[ P_s \partial_x f(\mathbb{P}(X^*_s, u^*_s), X^*_s, u^*_s)) (X^*_s - X^*_s) \right] ds
\]

\[
- \int_0^T \mathbb{E} \left[ \partial_x (\mathbb{P}(X^*_s, u^*_s), X^*_s, u^*_s)) (X^*_s - X^*_s) \right] ds
\]

\[
- \int_0^T \mathbb{E} \left[ (\partial_x f)(\mathbb{P}(X^*_s, u^*_s), X^*_s, u^*_s)) (X^*_s - X^*_s) \right] ds
\]

\[
\leq \partial_x f(\mathbb{P}(X^*_s, u^*_s), X^*_s, u^*_s)) (X^*_s - X^*_s) + \mathbb{E} \left[ (\partial_x f)(\mathbb{P}(X^*_s, u^*_s), X^*_s, u^*_s)) (X^*_s - X^*_s) \right]
\]

\[
+ \mathbb{E} \left[ (\partial_x f)(\mathbb{P}(X^*_s, u^*_s), X^*_s, u^*_s)) (u^*_s - u^*_s) \right] + P_s \partial_x f(\mathbb{P}(X^*_s, u^*_s), X^*_s, u^*_s)) (X^*_s - X^*_s)
\]

\[
+ P_s \mathbb{E} \left[ (\partial_x f)(\mathbb{P}(X^*_s, u^*_s), X^*_s, u^*_s)) (u^*_s - u^*_s) \right] + P_s \partial_x b(\mathbb{P}(X^*_s, u^*_s), X^*_s, u^*_s)) (u^*_s - u^*_s),
\]

where again due to equation \((57)\), for almost all \(s \in [0, T]\), we have equation \((62)\). Hence from equations \((59)\), \((66)\) and \((67)\), we get

\[
J(u^*) - J(u) \leq 0.
\]

Therefore, the optimality of \((u^*, X^*)\) follows. \(\square\)

Concerning the solvability of system \((57)\) under the conditions of Theorem 4.6, we proceed as follows. For any given \((P, \xi) \in L^2(\mathcal{F}_T) \times L^2(\mathcal{F}_T)\), we suppose that there is some \(\eta \in L^2(\mathcal{F}_T; U)\) such that:

\[
0 = \mathbb{E} \left[ (\partial_x f)(\mathbb{P}(\xi, \eta), \tilde{\xi}, \tilde{\eta}, \xi, \eta) \right] + \partial_u f(\mathbb{P}(\xi, \eta), \xi, \eta)
\]

\[
+ \mathbb{E} \left[ (\partial_x b)(\mathbb{P}(\xi, \eta), \tilde{\xi}, \tilde{\eta}, \xi, \eta) \right] + P(\partial_u b)(\mathbb{P}(\xi, \eta), \xi, \eta).
\]

**Lemma 4.8.** The mapping \((P, \xi) \rightarrow \eta\) is Lipschitz under \(L^2\)-norm.
Proof. Given \((P, \xi) \in L^2(\mathcal{F}_t) \times L^2(\mathcal{F}_t)\) and \((\hat{P}, \hat{\xi}) \in L^2(\mathcal{F}_t) \times L^2(\mathcal{F}_t)\), let \(\eta \in L^2(\mathcal{F}_t, U)\) be the solution of (68) associated with \((P, \xi)\) and \(\hat{\eta} \in L^2(\mathcal{F}_t, U)\) be that for \((\hat{P}, \hat{\xi})\). Then the strict convexity of \(b\) and \(f\) allows us to show, there exists \(\lambda > 0\), such that

\[
\lambda \mathbb{E} [\|\hat{\eta} - \eta\|^2] \leq \mathbb{E} \left[ \left( P(\partial_\mu b)_{2}(\mathbb{P}(\xi, \eta)), \xi, \hat{\xi}, \hat{\eta} - P(\partial_\mu b)_{2}(\mathbb{P}(\xi, \eta)), \xi, \xi, \eta, \hat{\eta} \right) (\hat{\eta} - \eta) \right] \\
+ \mathbb{E} \left[ (\partial_\mu f)_{2}(\mathbb{P}(\xi, \eta)), \xi, \hat{\xi}, \hat{\eta} - (\partial_\mu f)_{2}(\mathbb{P}(\xi, \eta)), \xi, \xi, \eta, \hat{\eta} \right] (\hat{\eta} - \eta) \\
+ \mathbb{E} \left[ (\partial_\mu f)_{2}(\mathbb{P}(\xi, \eta)), \xi, \hat{\xi}, \hat{\eta} - (\partial_\mu f)_{2}(\mathbb{P}(\xi, \eta)), \xi, \xi, \eta, \hat{\eta} \right] (\hat{\eta} - \eta) \\
+ \mathbb{E} \left[ (\partial_\mu f)_{2}(\mathbb{P}(\xi, \eta)), \xi, \hat{\xi}, \hat{\eta} - (\partial_\mu f)_{2}(\mathbb{P}(\xi, \eta)), \xi, \xi, \eta, \hat{\eta} \right] (\hat{\eta} - \eta).
\]

Then we get from the boundedness of \(P\) (due to the proof of Theorem 4.6) and the Lipschitz continuity of the derivatives of \(b\) and \(f\) (see (H4)) that

\[
\lambda \mathbb{E} [\|\hat{\eta} - \eta\|^2] \leq C \left( \mathbb{E} [\|\hat{\eta} - \eta\|^2] \right)^{1/2} \left( \left[ \mathbb{E} [\|\hat{P} - P\|^2] \right]^{1/2} + \left[ \mathbb{E} [\|\xi - \xi\|^2] \right]^{1/2} \right).
\]

Thus, we have

\[
\mathbb{E} [\|\hat{\eta} - \eta\|^2] \leq C \left( \mathbb{E} [\|\xi - \xi\|^2 + |\hat{P} - P|^2] \right).
\]

Hence, the mapping \((P, \xi) \to \eta\) is Lipschitz in the \(L^2\)-norm, i.e., there exists a Lipschitz function \(\eta : L^2(\mathcal{F}_t) \times L^2(\mathcal{F}_t) \to L^2(\mathcal{F}_t, U)\) such that \(\eta = \eta(P, \xi)\) solves (68).

With the above lemma, we know, in particular, that the solution of (68) is unique. Thus, the optimal control \(u^*_t\), if it exists, must satisfy the relation: \(u^*_t = \eta(P_t, X^*_t)\). Therefore, with such \(u^*_t\), under the conditions of Theorem 4.6, the system (57) becomes

\[
\begin{align*}
X^*_t &= x + \sigma B^H_t + \int_0^t b(\mathbb{P}(X^*_s, \eta(P_s, X^*_s)), X^*_s, \eta(P_s, X^*_s))ds, \\
P^*_t &= \partial_x g(X^*_t, \mathbb{P}X^*_t) + \tilde{\mathbb{E}} \left[ \partial_\mu g(X^*_t, \mathbb{P}X^*_t) \right] - \int_t^T \beta_s dW_s \\
&\quad + \int_t^T \left( \tilde{\mathbb{E}} \left[ \partial_\mu b(\mathbb{P}(X^*_s, \eta(P_s, X^*_s)), X^*_s, \eta(P_s, X^*_s)) \right] \\
&\quad + P_s \partial_\mu b(\mathbb{P}(X^*_s, \eta(P_s, X^*_s)), X^*_s, \eta(P_s, X^*_s)) \\
&\quad + \partial_\mu f(\mathbb{P}(X^*_s, \eta(P_s, X^*_s)), X^*_s, \eta(P_s, X^*_s)) \right) ds,
\end{align*}
\]

(69)
which is a coupled mean-field FBSDE, with the forward SDE driven by a fractional Brownian motion.

**Theorem 4.9.** Under the same conditions as in Theorem 4.6, there exists a unique solution \((X_t^*, P_t) \in S_{\mathbb{P}}^2([0, T] ; \mathbb{R}) \times S_{\mathbb{P}}^2([0, T] ; \mathbb{R})\) of mean-field FBSDE (69) for a small enough \(T > 0\). Here \(S_{\mathbb{P}}^2([0, T] ; \mathbb{R})\) denotes the space of real-valued \(\mathbb{P}\)-adapted continuous uniformly square integrable processes under the norm \(\|\varphi\|_{S_{\mathbb{P}}^2([0, T] ; \mathbb{R})} = \mathbb{E} \left[ \sup_{0 \leq s \leq T} |\varphi_s|^2 \right]\).

**Proof.** For any given \((x_t, p_t) \in S_{\mathbb{P}}^2([0, T] ; \mathbb{R}) \times S_{\mathbb{P}}^2([0, T] ; \mathbb{R})\), we construct the following map \((X_t, P_t) = I(x_t, p_t)\):

\[
\begin{align*}
X_t &= x + \sigma B_t^H + \int_0^t b(\mathbb{P}(x_s, \eta(p_s, x_s)), x_s, \eta(p_s, x_s)) \, ds, \\
P_t &= \partial_x g(x_T, \mathbb{P}(x_T, \mathbb{P}(x_T, x_T))) - \int_t^T \beta_s \, dW_s \\
&+ \int_t^T \left\{ \mathbb{E} \left[ \tilde{P}_s (\partial_x b)_1 (\mathbb{P}(x_s, \eta(p_s, x_s)), x_s, \eta(p_s, x_s)) \right] + P_s \partial_x b (\mathbb{P}(x_s, \eta(p_s, x_s)), x_s, \eta(p_s, x_s)) + \partial_x f (\mathbb{P}(x_s, \eta(p_s, x_s)), x_s, \eta(p_s, x_s)) \right\} \, ds.
\end{align*}
\]

From the proof of Theorem 4.6 we know that \(0 \leq P \leq C\) and the square integrability of \(X\) can be derived from the linear growth of \(b\) as in the proof of Theorem 3.6. Hence \(I\) maps from \(S_{\mathbb{P}}^2([0, T] ; \mathbb{R}) \times S_{\mathbb{P}}^2([0, T] ; \mathbb{R})\) to itself. Now we prove that it is a contracting map. For \((x^1, p^1)\) and \((x^2, p^2)\) in \(S_{\mathbb{P}}^2([0, T] ; \mathbb{R}) \times S_{\mathbb{P}}^2([0, T] ; \mathbb{R})\), from the Lipschitzianity of the functions \(b, \partial_x g, \partial_\mu g, \partial_\mu b, \partial_\mu f\), and the boundedness of \(P^1\) and \(P^2\), there exist a constant \(C\) which does not depend on \(T\), that

\[
\mathbb{E} \left[ \sup_{0 \leq s \leq t} |X_s^1 - X_s^2|^2 \right] \leq C \int_0^t \left( \mathbb{E} \left[ |x_s^1 - x_s^2|^2 \right] + \mathbb{E} \left[ |p_s^1 - p_s^2|^2 \right] \right) \, ds
\]

(71)

\[
\leq Ct \left( \mathbb{E} \left[ \sup_{0 \leq s \leq t} |x_s^1 - x_s^2|^2 \right] + \mathbb{E} \left[ \sup_{0 \leq s \leq t} |p_s^1 - p_s^2|^2 \right] \right),
\]

and

\[
\mathbb{E} \left[ \sup_{1 \leq s \leq T} |P_s^1 - P_s^2|^2 \right] \leq C \mathbb{E} \left[ |x_T^1 - x_T^2|^2 \right] + C \int_1^T \left( \mathbb{E} \left[ |x_s^1 - x_s^2|^2 \right] + \mathbb{E} \left[ |p_s^1 - p_s^2|^2 \right] \right) \, ds
\]

(72)

\[
\leq C(T - 1) \left( \mathbb{E} \left[ \sup_{1 \leq s \leq T} |x_s^1 - x_s^2|^2 \right] + \mathbb{E} \left[ \sup_{1 \leq s \leq T} |p_s^1 - p_s^2|^2 \right] \right).
\]

Hence for a small enough \(T\) such that \(CT < \alpha < 1\), we have

\[
\mathbb{E} \left[ \sup_{0 \leq s \leq T} |X_s^1 - X_s^2|^2 \right] + \mathbb{E} \left[ \sup_{0 \leq s \leq T} |P_s^1 - P_s^2|^2 \right] \leq \alpha \mathbb{E} \left[ \sup_{0 \leq s \leq T} |x_s^1 - x_s^2|^2 + \sup_{0 \leq s \leq T} |p_s^1 - p_s^2|^2 \right],
\]

(73)

which means \(I\) is a contracting map. Therefore, there exists a unique fixed point \((X^*, P)\), which is the solution for equation (69). \(\Box\)
With this solution \((X^*, P)\), we substitute into \(\eta(X^*_t, P_t)\) and get \(u^*_t\), then due to Theorem 4.6 \((u^*_t, X^*_t)_{t\in[0,T]}\) is the optimal pair for the stochastic control system.

**Remark 4.10.** One can see that the coupled mean-field FBSDE (69) is only in a special form. The general form of the fully coupled mean-field FBSDE driven by both a fractional Brownian motion and a classical Brownian motion is foreseen for a forthcoming paper.

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DÉPARTEMENT DE MATHEMATIQUES, UNIVERSITÉ DE BRETAGNE OCCIDENTALE, 29285, BREST, FRANCE
School of Mathematics, Shandong University, 250100, Jinan, P.R.China
E-mail address: rainer.buckdahn@univ-brest.fr

School of Management Science and Engineering, Central University of Finance and Economics, 100081, Beijing, P.R.China
E-mail address: jing@cufe.edu.cn