

## Research Article

# Representation Theorem for Generators of BSDEs Driven by $G$ -Brownian Motion and Its Applications

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We obtain a representation theorem for the generators of BSDEs driven by  $G$ -Brownian motions and then we use the representation theorem to get a converse comparison theorem for  $G$ -BSDEs and some equivalent results for nonlinear expectations generated by  $G$ -BSDEs.

## 1. Introduction

Let  $(\Omega, \mathcal{F}, P)$  be a probability space, and, for fixed  $T \in [0, +\infty)$ , let  $(B_t)_{0 \leq t \leq T}$  be a standard Brownian motion and let  $\mathcal{F}_t$  be the augmentation of  $\sigma\{B_s, 0 \leq s \leq t\}$ . Then Pardoux and Peng [1] introduced the backward stochastic differential equations (BSDEs) and proved the existence and uniqueness result of the BSDEs. In 1997, Peng [2] promoted  $g$ -expectations based on BSDEs. One of the important properties of  $g$ -expectations is comparison theorem or monotonicity. Chen [3] first considers a converse result of BSDEs under equal case. After that, Briand et al. [4] obtained a converse comparison theorem for BSDEs under general case. They also derived a representation theorem for the generator  $g$ . Following this paper, Jiang [5] discussed a more general representation theorem then, in his another paper [6], showed a more general converse comparison theorem. Here the representation theorem is an important method in solving the converse comparison problem and other problems (see Jiang [7]).

Peng [8–13] defined the  $G$ -expectations and  $G$ -Brownian motions ( $G$ -BMs) and proved the representation theorem of  $G$ -expectation by a set of singular probabilities, which differs from nonlinear  $g$ -expectations because  $g$ -expectations are equivalent with a group of absolutely continuous probabilities with respect to the probability measure  $P$ . Soner et al. [14] obtained an existence and uniqueness result of 2 BSDEs. Recently, Hu et al. [15] proved another existence and

uniqueness result on BSDEs driven by  $G$ -Brownian motions ( $G$ -BSDEs).

An important advantage of  $G$ -BSDEs is the easiness to define the nonlinear expectations. Hu et al. in [16] gave a comparison theorem for  $G$ -BSDEs and talked about the properties of corresponding nonlinear expectations. In this paper, we consider the representation theorem for generators of  $G$ -BSDEs and then consider the converse comparison theorem of  $G$ -BSDEs and some equivalent results for nonlinear expectations generated by  $G$ -BSDEs. In the following, in Section 2, we review some basic concepts and results about  $G$ -expectations. We give the representation theorem of  $G$ -BSDEs in Section 3. In Section 4, we consider the applications of representation theorem of  $G$ -BSDEs, which contain the converse comparison theorem and some equivalent results for nonlinear expectations generated by  $G$ -BSDEs.

## 2. Preliminaries

We review some basic notions and results of  $G$ -expectation, the related spaces of random variables, and the backward stochastic differential equations driven by a  $G$ -Brownian motion. The readers may refer to [10, 13, 15, 17–19] for more details.

*Definition 1.* Let  $\Omega$  be a given set and let  $\mathcal{H}$  be a vector lattice of real valued functions defined on  $\Omega$ , namely,  $c \in \mathcal{H}$  for each

constant  $c$  and  $|X| \in \mathcal{H}$  if  $X \in \mathcal{H}$ .  $\mathcal{H}$  is considered as the space of random variables. A sublinear expectation  $\widehat{\mathbb{E}}$  on  $\mathcal{H}$  is a functional  $\widehat{\mathbb{E}} : \mathcal{H} \rightarrow \mathbb{R}$  satisfying the following properties: for all  $X, Y \in \mathcal{H}$ , one has

- (a) monotonicity: if  $X \geq Y$ , then  $\widehat{\mathbb{E}}[X] \geq \widehat{\mathbb{E}}[Y]$ ;
- (b) constant preservation:  $\widehat{\mathbb{E}}[c] = c$ ;
- (c) subadditivity:  $\widehat{\mathbb{E}}[X + Y] \leq \widehat{\mathbb{E}}[X] + \widehat{\mathbb{E}}[Y]$ ;
- (d) positive homogeneity:  $\widehat{\mathbb{E}}[\lambda X] = \lambda \widehat{\mathbb{E}}[X]$  for each  $\lambda \geq 0$ .  $(\Omega, \mathcal{H}, \widehat{\mathbb{E}})$  is called a sublinear expectation space.

*Definition 2.* Let  $X_1$  and  $X_2$  be two  $n$ -dimensional random vectors defined, respectively, in sublinear expectation spaces  $(\Omega_1, \mathcal{H}_1, \widehat{\mathbb{E}}_1)$  and  $(\Omega_2, \mathcal{H}_2, \widehat{\mathbb{E}}_2)$ . They are called identically distributed, denoted by  $X_1 \stackrel{d}{=} X_2$ , if  $\widehat{\mathbb{E}}_1[\varphi(X_1)] = \widehat{\mathbb{E}}_2[\varphi(X_2)]$ , for all  $\varphi \in C_{b\text{-Lip}}(\mathbb{R}^n)$ , where  $C_{b\text{-Lip}}(\mathbb{R}^n)$  denotes the space of bounded and Lipschitz functions on  $\mathbb{R}^n$ .

*Definition 3.* In a sublinear expectation space  $(\Omega, \mathcal{H}, \widehat{\mathbb{E}})$ , a random vector  $Y = (Y_1, \dots, Y_n)$ ,  $Y_i \in \mathcal{H}$ , is said to be independent of another random vector  $X = (X_1, \dots, X_m)$ ,  $X_i \in \mathcal{H}$  under  $\widehat{\mathbb{E}}[\cdot]$ , denoted by  $Y \perp X$ , if for every test function  $\varphi \in C_{b\text{-Lip}}(\mathbb{R}^m \times \mathbb{R}^n)$  one has  $\widehat{\mathbb{E}}[\varphi(X, Y)] = \widehat{\mathbb{E}}[\widehat{\mathbb{E}}[\varphi(x, Y)]_{x=X}]$ .

*Definition 4 (G-normal distribution).* A  $d$ -dimensional random vector  $X = (X_1, \dots, X_d)$  in a sublinear expectation space  $(\Omega, \mathcal{H}, \widehat{\mathbb{E}})$  is called  $G$ -normally distributed if for each  $a, b \geq 0$  one has

$$aX + b\bar{X} \stackrel{d}{=} \sqrt{a^2 + b^2}X, \tag{1}$$

where  $\bar{X}$  is an independent copy of  $X$ ; that is,  $\bar{X} \stackrel{d}{=} X$  and  $\bar{X} \perp X$ . Here, the letter  $G$  denotes the function

$$G(A) := \frac{1}{2} \widehat{\mathbb{E}}[\langle AX, X \rangle] : \mathbb{S}_d \rightarrow \mathbb{R}, \tag{2}$$

where  $\mathbb{S}_d$  denotes the collection of  $d \times d$  symmetric matrices.

Peng [13] showed that  $X = (X_1, \dots, X_d)$  is  $G$ -normally distributed if and only if for each  $\varphi \in C_{b\text{-Lip}}(\mathbb{R}^d)$ ,  $u(t, x) := \widehat{\mathbb{E}}[\varphi(x + \sqrt{t}X)]$ ,  $(t, x) \in [0, \infty) \times \mathbb{R}^d$ , is the solution of the following  $G$ -heat equation:

$$\partial_t u - G(D_x^2 u) = 0, \quad u(0, x) = \varphi(x). \tag{3}$$

The function  $G(\cdot) : \mathbb{S}_d \rightarrow \mathbb{R}$  is a monotonic, sublinear mapping on  $\mathbb{S}_d$  and  $G(A) = (1/2)\widehat{\mathbb{E}}[\langle AX, X \rangle] \leq (1/2)|A|\widehat{\mathbb{E}}[|X|^2]$  implies that there exists a bounded, convex, and closed subset  $\Gamma \subset \mathbb{S}_d^+$  such that

$$G(A) = \frac{1}{2} \sup_{\gamma \in \Gamma} \text{tr}[\gamma A], \tag{4}$$

where  $\mathbb{S}_d^+$  denotes the collection of nonnegative elements in  $\mathbb{S}_d$ .

In this paper, we only consider nondegenerate  $G$ -normal distribution; that is, there exists some  $\underline{\sigma}^2 > 0$  such that  $G(A) - G(B) \geq \underline{\sigma}^2 \text{tr}[A - B]$  for any  $A \geq B$ .

*Definition 5.* (i) Let  $\Omega = C_0^d(\mathbb{R}^+)$  denote the space of  $\mathbb{R}^d$ -valued continuous functions on  $[0, \infty)$  with  $\omega_0 = 0$  and let  $B_t(\omega) = \omega_t$  be the canonical process. Set

$$L_{ip}(\Omega) := \left\{ \varphi(B_{t_1}, \dots, B_{t_n}) : n \geq 1, t_1, \dots, t_n \in [0, \infty), \right. \\ \left. \varphi \in C_{b\text{-Lip}}(\mathbb{R}^{d \times n}) \right\}. \tag{5}$$

Let  $G : \mathbb{S}_d \rightarrow \mathbb{R}$  be a given monotonic and sublinear function.  $G$ -expectation is a sublinear expectation defined by

$$\widehat{\mathbb{E}}[X] = \widehat{\mathbb{E}}\left[\varphi\left(\sqrt{t_1 - t_0}\xi_1, \dots, \sqrt{t_m - t_{m-1}}\xi_m\right)\right], \tag{6}$$

for all  $X = \varphi(B_{t_1} - B_{t_0}, B_{t_2} - B_{t_1}, \dots, B_{t_m} - B_{t_{m-1}})$ , where  $\xi_1, \dots, \xi_m$  are identically distributed  $d$ -dimensional  $G$ -normally distributed random vectors in a sublinear expectation space  $(\widetilde{\Omega}, \widetilde{\mathcal{H}}, \widetilde{\mathbb{E}})$  such that  $\xi_{i+1}$  is independent of  $(\xi_1, \dots, \xi_i)$  for every  $i = 1, \dots, m - 1$ . The corresponding canonical process  $B_t = (B_t^i)_{i=1}^d$  is called a  $G$ -Brownian motion.

(ii) For each fixed  $t \in [0, \infty)$ , the conditional  $G$ -expectation  $\widehat{\mathbb{E}}_t$  for  $\xi = \varphi(B_{t_1} - B_{t_0}, B_{t_2} - B_{t_1}, \dots, B_{t_m} - B_{t_{m-1}}) \in L_{ip}(\Omega)$ , where without loss of generality we suppose  $t_i = t$ , is defined by

$$\widehat{\mathbb{E}}_t\left[\varphi\left(B_{t_1} - B_{t_0}, B_{t_2} - B_{t_1}, \dots, B_{t_m} - B_{t_{m-1}}\right)\right] \\ = \psi\left(B_{t_1} - B_{t_0}, B_{t_2} - B_{t_1}, \dots, B_{t_i} - B_{t_{i-1}}\right), \tag{7}$$

where

$$\psi(x_1, \dots, x_i) \\ = \widehat{\mathbb{E}}\left[\varphi\left(x_1, \dots, x_i, B_{t_{i+1}} - B_{t_i}, \dots, B_{t_m} - B_{t_{m-1}}\right)\right]. \tag{8}$$

For each fixed  $T > 0$ , we set

$$L_{ip}(\Omega_T) := \left\{ \varphi(B_{t_1}, \dots, B_{t_n}) : n \geq 1, t_1, \dots, t_n \in [0, T], \right. \\ \left. \varphi \in C_{b\text{-Lip}}(\mathbb{R}^{d \times n}) \right\}. \tag{9}$$

For each  $p \geq 1$ , we denote by  $L_G^p(\Omega)$  (resp.,  $L_G^p(\Omega_T)$ ) the completion of  $L_{ip}(\Omega)$  (resp.,  $L_{ip}(\Omega_T)$ ) under the norm  $\|\xi\|_{p,G} = (\widehat{\mathbb{E}}[|\xi|^p])^{1/p}$ . It is easy to check that  $L_G^q(\Omega) \subset L_G^p(\Omega)$  for  $1 \leq p \leq q$  and  $\widehat{\mathbb{E}}_t[\cdot]$  can be extended continuously to  $L_G^1(\Omega)$ .

For each fixed  $\mathbf{a} \in \mathbb{R}^d$ ,  $B_t^{\mathbf{a}} = \langle \mathbf{a}, B_t \rangle$  is a 1-dimensional  $G_{\mathbf{a}}$ -Brownian motion, where  $G_{\mathbf{a}}(\alpha) = (1/2)(\sigma_{\mathbf{a}\mathbf{a}^T}^2 \alpha^+ - \sigma_{-\mathbf{a}\mathbf{a}^T}^2 \alpha^-)$ ,  $\sigma_{\mathbf{a}\mathbf{a}^T}^2 = 2G(\mathbf{a}\mathbf{a}^T)$  and  $\sigma_{-\mathbf{a}\mathbf{a}^T}^2 = -2G(-\mathbf{a}\mathbf{a}^T)$ . Let  $\pi_t^N = \{t_0^N, \dots, t_N^N\}$ ,  $N = 1, 2, \dots$ , be a sequence of partitions of  $[0, t]$  such that  $\mu(\pi_t^N) = \max\{|t_{i+1}^N - t_i^N| : i = 0, \dots, N - 1\} \rightarrow 0$ ; the quadratic variation process of  $B^{\mathbf{a}}$  is defined by

$$\langle B^{\mathbf{a}} \rangle_t = \lim_{\mu(\pi_t^N) \rightarrow 0} \sum_{j=0}^{N-1} \left( B_{t_{j+1}^N}^{\mathbf{a}} - B_{t_j^N}^{\mathbf{a}} \right)^2. \tag{10}$$

For each fixed  $\mathbf{a}, \bar{\mathbf{a}} \in \mathbb{R}^d$ , the mutual variation process of  $B^{\mathbf{a}}$  and  $B^{\bar{\mathbf{a}}}$  is defined by

$$\langle B^{\mathbf{a}}, B^{\bar{\mathbf{a}}} \rangle_t = \frac{1}{4} [\langle B^{\mathbf{a}+\bar{\mathbf{a}}} \rangle_t - \langle B^{\mathbf{a}-\bar{\mathbf{a}}} \rangle_t]. \quad (11)$$

**Definition 6.** For fixed  $T > 0$ , let  $M_G^0(0, T)$  be the collection of processes in the following form: for a given partition  $\{t_0, \dots, t_N\} = \pi_T$  of  $[0, T]$ ,

$$\eta_t(\omega) = \sum_{j=0}^{N-1} \xi_j I_{[t_j, t_{j+1})}(t), \quad (12)$$

where  $\xi_j \in L_{ip}(\Omega_{t_j})$ ,  $j = 0, 1, 2, \dots, N - 1$ . For  $p \geq 1$ , one denotes by  $H_G^p(0, T)$ ,  $M_G^p(0, T)$  the completion of  $M_G^0(0, T)$  under the norms  $\|\eta\|_{H_G^p} = \{\widehat{\mathbb{E}}[(\int_0^T |\eta_s|^2 ds)^{p/2}]\}^{1/p}$ ,  $\|\eta\|_{M_G^p} = \{\widehat{\mathbb{E}}[\int_0^T |\eta_s|^p ds]\}^{1/p}$ , respectively.

For each  $\eta \in M_G^1(0, T)$ , we can define the integrals  $\int_0^T \eta_t dt$  and  $\int_0^T \eta_t d\langle B^{\mathbf{a}}, B^{\bar{\mathbf{a}}} \rangle_t$  for each  $\mathbf{a}, \bar{\mathbf{a}} \in \mathbb{R}^d$ . For each  $\eta \in H_G^p(0, T; \mathbb{R}^d)$  with  $p \geq 1$ , we can define Itô's integral  $\int_0^T \eta_t dB_t$ .

Let  $S_G^0(0, T) = \{h(t, B_{t_1 \wedge t}, \dots, B_{t_n \wedge t}) : t_1, \dots, t_n \in [0, T], h \in C_{b, \text{Lip}}(\mathbb{R}^{n+1})\}$ . For  $p \geq 1$  and  $\eta \in S_G^0(0, T)$ , set  $\|\eta\|_{S_G^p} = \{\widehat{\mathbb{E}}[\sup_{t \in [0, T]} |\eta_t|^p]\}^{1/p}$ . Denote by  $S_G^p(0, T)$  the completion of  $S_G^0(0, T)$  under the norm  $\|\cdot\|_{S_G^p}$ .

We consider the following type of G-BSDEs (in this paper, we always use Einstein convention):

$$Y_t = \xi + \int_t^T f(s, Y_s, Z_s) ds + \int_t^T g_{ij}(s, Y_s, Z_s) d\langle B^i, B^j \rangle_s - \int_t^T Z_s dB_s - (K_T - K_t), \quad (13)$$

where

$$f(t, \omega, y, z), g_{ij}(t, \omega, y, z) : [0, T] \times \Omega_T \times \mathbb{R} \times \mathbb{R}^d \rightarrow \mathbb{R}, \quad (14)$$

satisfy the following properties.

(H1) There exists some  $\beta > 1$  such that for any  $y, z, f(\cdot, \cdot, y, z), g_{ij}(\cdot, \cdot, y, z) \in M_G^\beta(0, T)$ .

(H2) There exists some  $L > 0$  such that

$$|f(t, y, z) - f(t, y', z')| + \sum_{i,j=1}^d |g_{ij}(t, y, z) - g_{ij}(t, y', z')| \leq L(|y - y'| + |z - z'|). \quad (15)$$

For simplicity, we denote by  $\mathfrak{S}_G^\alpha(0, T)$  the collection of processes  $(Y, Z, K)$  such that  $Y \in S_G^\alpha(0, T)$ ,  $Z \in H_G^\alpha(0, T; \mathbb{R}^d)$ ,  $K$  is a decreasing G-martingale with  $K_0 = 0$  and  $K_T \in L_G^\alpha(\Omega_T)$ .

**Definition 7.** Let  $\xi \in L_G^\beta(\Omega_T)$  and  $f$  and  $g_{ij}$  satisfy (H1) and (H2) for some  $\beta > 1$ . A triplet of processes  $(Y, Z, K)$  is called a solution of (13) if for some  $1 < \alpha \leq \beta$  the following properties hold:

(a)  $(Y, Z, K) \in \mathfrak{S}_G^\alpha(0, T)$ ;

(b)  $Y_t = \xi + \int_t^T f(s, Y_s, Z_s) ds + \int_t^T g_{ij}(s, Y_s, Z_s) d\langle B^i, B^j \rangle_s - \int_t^T Z_s dB_s - (K_T - K_t)$ .

**Theorem 8** (see [15]). Assume that  $\xi \in L_G^\beta(\Omega_T)$  and  $f$  and  $g_{ij}$  satisfy (H1) and (H2) for some  $\beta > 1$ . Then, (13) has a unique solution  $(Y, Z, K)$ . Moreover, for any  $1 < \alpha < \beta$ , one has  $Y \in S_G^\alpha(0, T)$ ,  $Z \in H_G^\alpha(0, T; \mathbb{R}^d)$ , and  $K_T \in L_G^\alpha(\Omega_T)$ .

We have the following estimates.

**Proposition 9** (see [15]). Let  $\xi \in L_G^\beta(\Omega_T)$  and  $f, g_{ij}$  satisfy (H1) and (H2) for some  $\beta > 1$ . Assume that  $(Y, Z, K) \in \mathfrak{S}_G^\alpha(0, T)$  for some  $1 < \alpha < \beta$  is a solution of (13). Then, there exists a constant  $C_\alpha > 0$  depending on  $\alpha, T, G, L$  such that

$$\begin{aligned} |Y_t|^\alpha &\leq C_\alpha \widehat{\mathbb{E}}_t \left[ |\xi|^\alpha + \left( \int_t^T |h_s^0| ds \right)^\alpha \right], \\ \widehat{\mathbb{E}} \left[ \left( \int_0^T |Z_s|^2 ds \right)^{\alpha/2} \right] &\leq C_\alpha \left\{ \widehat{\mathbb{E}} \left[ \sup_{t \in [0, T]} |Y_t|^\alpha \right] + \left( \widehat{\mathbb{E}} \left[ \sup_{t \in [0, T]} |Y_t|^\alpha \right] \right)^{1/2} \right. \\ &\quad \left. \times \left( \widehat{\mathbb{E}} \left[ \left( \int_0^T |h_s^0| ds \right)^\alpha \right] \right)^{1/2} \right\}, \\ \widehat{\mathbb{E}} [ |K_t|^\alpha ] &\leq C_\alpha \left\{ \widehat{\mathbb{E}} \left[ \sup_{t \in [0, T]} |Y_t|^\alpha \right] + \widehat{\mathbb{E}} \left[ \left( \int_0^T |h_s^0| ds \right)^\alpha \right] \right\}, \end{aligned} \quad (16)$$

where  $h_s^0 = |f(s, 0, 0)| + \sum_{i,j=1}^d |g_{ij}(s, 0, 0)|$ .

**Proposition 10** (see [15, 20]). Let  $\alpha \geq 1$  and  $\delta > 0$  be fixed. Then, there exists a constant  $C$  depending on  $\alpha$  and  $\delta$  such that

$$\begin{aligned} \widehat{\mathbb{E}} \left[ \sup_{t \in [0, T]} \widehat{\mathbb{E}}_t [ |\xi|^\alpha ] \right] &\leq C \left\{ \left( \widehat{\mathbb{E}} [ |\xi|^{\alpha+\delta} ] \right)^{\alpha/(\alpha+\delta)} + \widehat{\mathbb{E}} [ |\xi|^{\alpha+\delta} ] \right\}, \end{aligned} \quad (17)$$

$$\forall \xi \in L_G^{\alpha+\delta}(\Omega_T).$$

**Theorem 11** (see [16]). Let  $(Y^l, Z^l, K^l)$ ,  $l = 1, 2$ , be the solutions of the following G-BSDEs:

$$\begin{aligned} Y_t^l &= \xi + \int_t^T f(s, Y_s^l, Z_s^l) ds + \int_t^T g_{ij}(s, Y_s^l, Z_s^l) d\langle B^i, B^j \rangle_s \\ &\quad + V_T^l - V_t^l - \int_t^T Z_s^l dB_s - (K_T^l - K_t^l), \end{aligned} \quad (18)$$

where  $\xi \in L_G^\beta(\Omega_T)$ ,  $f$  and  $g_{ij}$  satisfy (H1) and (H2) for some  $\beta > 1$  and  $(V_t^l)_{t \leq T}$  are RCLL processes in  $M_G^\beta(0, T)$  such that  $\widehat{\mathbb{E}}[\sup_{t \in [0, T]} |V_t^l|^\beta] < \infty$ . If  $V_t^1 - V_t^2$  is an increasing process, then  $Y_t^1 \geq Y_t^2$  for  $t \in [0, T]$ .

In this paper, we also need the following assumptions for G-BSDE (13).

(H3) For each fixed  $(\omega, y, z) \in \Omega_T \times \mathbb{R} \times \mathbb{R}^d$ ,  $t \rightarrow f(t, \omega, y, z)$  and  $t \rightarrow g_{ij}(t, \omega, y, z)$  are continuous.

(H4) For each fixed  $(t, y, z) \in [0, T] \times \mathbb{R} \times \mathbb{R}^d$ ,  $f(t, y, z)$ ,  $g_{ij}(t, y, z) \in L_G^\beta(\Omega_t)$ , and

$$\begin{aligned} \lim_{\varepsilon \rightarrow 0^+} \frac{1}{\varepsilon} \widehat{\mathbb{E}} \left[ \int_t^{t+\varepsilon} \left( |f(u, y, z) - f(t, y, z)|^\beta \right. \right. \\ \left. \left. + \sum_{i,j=1}^d |g_{ij}(u, y, z) - g_{ij}(t, y, z)|^\beta \right) du \right] = 0. \end{aligned} \quad (19)$$

(H5) For each  $(t, \omega, y) \in [0, T] \times \Omega_T \times \mathbb{R}$ ,  $f(t, \omega, y, 0) = g_{ij}(t, \omega, y, 0) = 0$ .

Assume that  $\xi \in L_G^\beta(\Omega_T)$ ;  $f$  and  $g_{ij}$  satisfy (H1), (H2), and (H5) for some  $\beta > 1$ . Let  $(Y^{T,\xi}, Z^{T,\xi}, K^{T,\xi})$  be the solution of G-BSDE (13) corresponding to  $\xi$ ,  $f$ , and  $g_{ij}$  on  $[0, T]$ . It is easy to check that  $Y^{T,\xi} = Y^{T',\xi}$  on  $[0, T]$  for  $T' > T$ . Following [16], we can define consistent nonlinear expectation

$$\widehat{\mathbb{E}}_t[\xi] = Y_t^{T,\xi} \quad \text{for } t \in [0, T] \quad (20)$$

and set  $\widehat{\mathbb{E}}[\xi] = \widehat{\mathbb{E}}_0[\xi] = Y_0^{T,\xi}$ .

### 3. Representation Theorem of Generators of G-BSDEs

We consider the following type of G-FBSDEs:

$$\begin{aligned} X_s^{t,x} &= x + \int_t^s b(X_u^{t,x}) du \\ &\quad + \int_t^s h_{ij}(X_u^{t,x}) d\langle B^i, B^j \rangle_u + \int_t^s \sigma(X_u^{t,x}) dB_u, \end{aligned} \quad (21)$$

$$\begin{aligned} {}^\varepsilon Y_s^{t,x,y,p} &= y + \langle p, X_{t+\varepsilon}^{t,x} - x \rangle \\ &\quad + \int_s^{t+\varepsilon} f(u, {}^\varepsilon Y_u^{t,x,y,p}, {}^\varepsilon Z_u^{t,x,y,p}) du \\ &\quad + \int_s^{t+\varepsilon} g_{ij}(u, {}^\varepsilon Y_u^{t,x,y,p}, {}^\varepsilon Z_u^{t,x,y,p}) d\langle B^i, B^j \rangle_u \\ &\quad - \int_s^{t+\varepsilon} {}^\varepsilon Z_u^{t,x,y,p} dB_u - ({}^\varepsilon K_{t+\varepsilon}^{t,x,y,p} - {}^\varepsilon K_s^{t,x,y,p}), \end{aligned} \quad (22)$$

where  $h_{ij} = h_{ji}$  and  $g_{ij} = g_{ji}$ ,  $1 \leq i, j \leq d$ .

We now give the main result in this section.

**Theorem 12.** Let  $b : \mathbb{R}^n \rightarrow \mathbb{R}^n$ ,  $h_{ij} : \mathbb{R}^n \rightarrow \mathbb{R}^n$ , and  $\sigma : \mathbb{R}^n \rightarrow \mathbb{R}^{n \times d}$  be Lipschitz functions and let  $f$  and  $g_{ij}$  satisfy (H1), (H2), (H3), and (H4) for some  $\beta > 1$ . Then, for each  $(t, x, y, p) \in [0, T] \times \mathbb{R}^n \times \mathbb{R} \times \mathbb{R}^n$  and  $\alpha \in (1, \beta)$ , one has

$$\begin{aligned} L_G^\alpha - \lim_{\varepsilon \rightarrow 0^+} \frac{1}{\varepsilon} \{ {}^\varepsilon Y_t^{t,x,y,p} - y \} \\ = f(t, y, \sigma^T(x)p) + \langle p, b(x) \rangle \\ + 2G \left( \left( g_{ij}(t, y, \sigma^T(x)p) + \langle p, h_{ij}(x) \rangle \right)_{i,j=1}^d \right). \end{aligned} \quad (23)$$

*Proof.* For each fixed  $(t, x, y, p) \in [0, T] \times \mathbb{R}^n \times \mathbb{R} \times \mathbb{R}^n$ , we write  $(Y^\varepsilon, Z^\varepsilon, K^\varepsilon)$  instead of  $({}^\varepsilon Y^{t,x,y,p}, {}^\varepsilon Z^{t,x,y,p}, {}^\varepsilon K^{t,x,y,p})$  for simplicity. We have  $\widehat{\mathbb{E}}[|X_{t+\varepsilon}^{t,x}|^\beta] < \infty$  for each  $\gamma \geq 1$  (see [16, 19]). Thus, by Theorem 8, G-BSDE (22) has a unique solution  $(Y^\varepsilon, Z^\varepsilon, K^\varepsilon)$  and  $Y_t^\varepsilon \in L_G^\alpha(\Omega_t)$ . We set, for  $s \in [t, t+\varepsilon]$ ,

$$\begin{aligned} \widetilde{Y}_s^\varepsilon &= Y_s^\varepsilon - \left( y + \langle p, X_s^{t,x} - x \rangle \right), \\ \widetilde{Z}_s^\varepsilon &= Z_s^\varepsilon - \sigma^T(X_s^{t,x})p, \quad \widetilde{K}_s^\varepsilon = K_s^\varepsilon. \end{aligned} \quad (24)$$

Applying Itô's formula to  $\widetilde{Y}_s^\varepsilon$  on  $[t, t+\varepsilon]$ , it is easy to verify that  $(\widetilde{Y}^\varepsilon, \widetilde{Z}^\varepsilon, \widetilde{K}^\varepsilon)$  solves the following G-BSDE:

$$\begin{aligned} \widetilde{Y}_s^\varepsilon &= \int_s^{t+\varepsilon} f(u, \widetilde{Y}_u^\varepsilon + y + \langle p, X_u^{t,x} - x \rangle, \\ &\quad \widetilde{Z}_u^\varepsilon + \sigma^T(X_u^{t,x})p) du \\ &\quad + \int_s^{t+\varepsilon} \langle p, b(X_u^{t,x}) \rangle du \\ &\quad + \int_s^{t+\varepsilon} g_{ij}(u, \widetilde{Y}_u^\varepsilon + y + \langle p, X_u^{t,x} - x \rangle, \\ &\quad \widetilde{Z}_u^\varepsilon + \sigma^T(X_u^{t,x})p) d\langle B^i, B^j \rangle_u \\ &\quad + \int_s^{t+\varepsilon} \langle p, h_{ij}(X_u^{t,x}) \rangle d\langle B^i, B^j \rangle_u \\ &\quad - \int_s^{t+\varepsilon} \widetilde{Z}_u^\varepsilon dB_u - (\widetilde{K}_{t+\varepsilon}^\varepsilon - \widetilde{K}_s^\varepsilon). \end{aligned} \quad (25)$$

From Proposition 9,

$$\begin{aligned}
 |\tilde{Y}_s^\varepsilon|^\alpha &\leq C_\alpha \widehat{\mathbb{E}}_s \left[ \left( \int_s^{t+\varepsilon} (|f(u, y + \langle p, X_u^{t,x} - x \rangle, \sigma^T(X_u^{t,x})p)| \right. \right. \\
 &\quad \left. \left. + |\langle p, b(X_u^{t,x}) \rangle| \right. \right. \\
 &\quad \left. \left. + \sum_{i,j=1}^d |g_{ij}(u, y + \langle p, X_u^{t,x} - x \rangle, \sigma^T(X_u^{t,x})p)| \right. \right. \\
 &\quad \left. \left. + |\langle p, h_{ij}(X_u^{t,x}) \rangle|) du \right)^\alpha \right], \\
 \widehat{\mathbb{E}} \left[ \left( \int_t^{t+\varepsilon} |\tilde{Z}_u^\varepsilon|^2 du \right)^{\alpha/2} \right] \\
 &\leq C_\alpha \left\{ \widehat{\mathbb{E}} \left[ \left( \int_t^{t+\varepsilon} (|f(u, y + \langle p, X_u^{t,x} - x \rangle, \sigma^T(X_u^{t,x})p)| \right. \right. \right. \\
 &\quad \left. \left. + |\langle p, b(X_u^{t,x}) \rangle| + |\langle p, h_{ij}(X_u^{t,x}) \rangle| \right. \right. \\
 &\quad \left. \left. + \sum_{i,j=1}^d |g_{ij}(u, y + \langle p, X_u^{t,x} - x \rangle, \right. \right. \\
 &\quad \left. \left. \left. \sigma^T(X_u^{t,x})p \right)|) du \right)^\alpha \right] \right. \\
 &\quad \left. + \widehat{\mathbb{E}} \left[ \sup_{s \in [t, t+\varepsilon]} |\tilde{Y}_s^\varepsilon|^\alpha \right] \right\}, \tag{26}
 \end{aligned}$$

hold for some constant  $C_\alpha > 0$ , only depending on  $\alpha, T, G$ , and  $L$ . By Proposition 10 and the Lipschitz assumption, we obtain

$$\begin{aligned}
 \widehat{\mathbb{E}} \left[ \sup_{s \in [t, t+\varepsilon]} |\tilde{Y}_s^\varepsilon|^\alpha + \left( \int_t^{t+\varepsilon} |\tilde{Z}_u^\varepsilon|^2 du \right)^{\alpha/2} \right] \\
 \leq C_1 \varepsilon^\alpha \widehat{\mathbb{E}} \left[ 1 + \left( \frac{1}{\varepsilon} \int_t^{t+\varepsilon} (|f(u, 0, 0)|^\beta \right. \right. \\
 \left. \left. + \sum_{i,j=1}^d |g_{ij}(u, 0, 0)|^\beta) du \right)^{\alpha/\beta} \right. \\
 \left. + \sup_{s \in [t, t+\varepsilon]} |X_s^{t,x}|^\beta \right], \tag{27}
 \end{aligned}$$

where  $C_1$  is a constant depending on  $x, y, p, \alpha, \beta, T, G$ , and  $L$ . Noting that  $\widehat{\mathbb{E}}[\sup_{s \in [t, t+\varepsilon]} |X_s^{t,x}|^\beta] \leq C_2(1 + |x|^\beta)$  (see [16, 19]), where  $C_2$  depends on  $T$  and  $L$ , and the following inequality holds:

$$\begin{aligned}
 \int_t^{t+\varepsilon} (|f(u, 0, 0)|^\beta + \sum_{i,j=1}^d |g_{ij}(u, 0, 0)|^\beta) du \\
 \leq 2^{\beta-1} \left\{ \varepsilon \left( |f(t, 0, 0)|^\beta + \sum_{i,j=1}^d |g_{ij}(t, 0, 0)|^\beta \right) \right. \\
 \left. + \int_t^{t+\varepsilon} (|f(u, 0, 0) - f(t, 0, 0)|^\beta \right. \\
 \left. + \sum_{i,j=1}^d |g_{ij}(u, 0, 0) - g_{ij}(t, 0, 0)|^\beta) du \right\}. \tag{28}
 \end{aligned}$$

Together with assumption (H4), we get

$$\widehat{\mathbb{E}} \left[ \sup_{s \in [t, t+\varepsilon]} |\tilde{Y}_s^\varepsilon|^\alpha + \left( \int_t^{t+\varepsilon} |\tilde{Z}_u^\varepsilon|^2 du \right)^{\alpha/2} \right] \leq C_3 \varepsilon^\alpha, \tag{29}$$

where  $C_3$  depends on  $x, y, p, \alpha, \beta, T, G$ , and  $L$ . Now, we prove (23). Let us consider

$$\begin{aligned}
 \frac{1}{\varepsilon} \{Y_t^\varepsilon - y\} &= \frac{1}{\varepsilon} \tilde{Y}_t^\varepsilon = \frac{1}{\varepsilon} \widehat{\mathbb{E}}_t [\tilde{Y}_t^\varepsilon + \tilde{K}_{t+\varepsilon}^\varepsilon - \tilde{K}_t^\varepsilon] \\
 &= \frac{1}{\varepsilon} \widehat{\mathbb{E}}_t \left[ \int_t^{t+\varepsilon} f(u, y + \langle p, X_u^{t,x} - x \rangle, \sigma^T(X_u^{t,x})p) du \right. \\
 &\quad \left. + \int_t^{t+\varepsilon} \langle p, b(X_u^{t,x}) \rangle du \right. \\
 &\quad \left. + \int_t^{t+\varepsilon} g_{ij}(u, y + \langle p, X_u^{t,x} - x \rangle, \sigma^T(X_u^{t,x})p) \right. \\
 &\quad \left. \times d\langle B^i, B^j \rangle_u \right. \\
 &\quad \left. + \int_t^{t+\varepsilon} \langle p, h_{ij}(X_u^{t,x}) \rangle d\langle B^i, B^j \rangle_u \right] + L_\varepsilon, \tag{30}
 \end{aligned}$$

where

$$\begin{aligned}
 L_\varepsilon &= \frac{1}{\varepsilon} \left\{ \widehat{\mathbb{E}}_t \left[ \int_t^{t+\varepsilon} f(u, \tilde{Y}_u^\varepsilon + y + \langle p, X_u^{t,x} - x \rangle, \right. \right. \\
 &\quad \left. \left. \tilde{Z}_u^\varepsilon + \sigma^T(X_u^{t,x})p) du \right. \right. \\
 &\quad \left. \left. + \int_t^{t+\varepsilon} \langle p, b(X_u^{t,x}) \rangle du \right. \right. \\
 &\quad \left. \left. + \int_t^{t+\varepsilon} g_{ij}(u, \tilde{Y}_u^\varepsilon + y + \langle p, X_u^{t,x} - x \rangle, \right. \right. \\
 &\quad \left. \left. \tilde{Z}_u^\varepsilon + \sigma^T(X_u^{t,x})p) d\langle B^i, B^j \rangle_u \right. \right. \\
 &\quad \left. \left. + \int_t^{t+\varepsilon} \langle p, h_{ij}(X_u^{t,x}) \rangle d\langle B^i, B^j \rangle_u \right] \right. \\
 &\quad \left. - \widehat{\mathbb{E}}_t \left[ \int_t^{t+\varepsilon} f(u, y + \langle p, X_u^{t,x} - x \rangle, \sigma^T(X_u^{t,x})p) du \right. \right. \\
 &\quad \left. \left. + \int_t^{t+\varepsilon} \langle p, b(X_u^{t,x}) \rangle du \right. \right. \\
 &\quad \left. \left. + \int_t^{t+\varepsilon} g_{ij}(u, y + \langle p, X_u^{t,x} - x \rangle, \sigma^T(X_u^{t,x})p) \right. \right. \\
 &\quad \left. \left. \times d\langle B^i, B^j \rangle_u \right. \right. \\
 &\quad \left. \left. + \int_t^{t+\varepsilon} \langle p, h_{ij}(X_u^{t,x}) \rangle d\langle B^i, B^j \rangle_u \right] \right\}. \tag{31}
 \end{aligned}$$

It is easy to check that  $|L_\varepsilon| \leq (C_4/\varepsilon)\widehat{\mathbb{E}}_t[\int_t^{t+\varepsilon} (|\widetilde{Y}_u^\varepsilon| + |\widetilde{Z}_u^\varepsilon|)du]$ , where  $C_4$  depends on  $G, L$ , and  $T$ . Thus, by (29), we get

$$\begin{aligned} & \widehat{\mathbb{E}}[|L_\varepsilon|^\alpha] \\ & \leq \frac{C_4^\alpha}{\varepsilon^\alpha} \widehat{\mathbb{E}}\left[\left(\int_t^{t+\varepsilon} (|\widetilde{Y}_u^\varepsilon| + |\widetilde{Z}_u^\varepsilon|) du\right)^\alpha\right] \\ & \leq \frac{2^{\alpha-1}C_4^\alpha}{\varepsilon^\alpha} \widehat{\mathbb{E}}\left[\left(\int_t^{t+\varepsilon} |\widetilde{Y}_u^\varepsilon| du\right)^\alpha + \left(\int_t^{t+\varepsilon} |\widetilde{Z}_u^\varepsilon| du\right)^\alpha\right] \\ & \leq 2^{\alpha-1}C_4^\alpha \left\{ \widehat{\mathbb{E}}\left[\sup_{s \in [t, t+\varepsilon]} |\widetilde{Y}_s^\varepsilon|^\alpha\right] + \varepsilon^{-\alpha/2} \widehat{\mathbb{E}}\left[\left(\int_t^{t+\varepsilon} |\widetilde{Z}_u^\varepsilon|^2 du\right)^{\alpha/2}\right] \right\} \\ & \leq 2^{\alpha-1}C_4^\alpha C_3 (\varepsilon^\alpha + \varepsilon^{\alpha/2}), \end{aligned} \tag{32}$$

which implies  $L_G^\alpha - \lim_{\varepsilon \rightarrow 0+} L_\varepsilon = 0$ . We set

$$\begin{aligned} M_\varepsilon &= \frac{1}{\varepsilon} \left\{ \widehat{\mathbb{E}}_t \left[ \int_t^{t+\varepsilon} f(u, y + \langle p, X_u^{t,x} - x \rangle, \sigma^T(X_u^{t,x}) p) du \right. \right. \\ & \quad + \int_t^{t+\varepsilon} \langle p, b(X_u^{t,x}) \rangle du \\ & \quad + \int_t^{t+\varepsilon} g_{ij}(u, y + \langle p, X_u^{t,x} - x \rangle, \sigma^T(X_u^{t,x}) p) \\ & \quad \quad \times d\langle B^i, B^j \rangle_u \\ & \quad \left. + \int_t^{t+\varepsilon} \langle p, h_{ij}(X_u^{t,x}) \rangle d\langle B^i, B^j \rangle_u \right] \\ & \quad - \widehat{\mathbb{E}}_t \left[ \int_t^{t+\varepsilon} f(u, y, \sigma^T(x) p) du + \langle p, b(x) \rangle \varepsilon \right. \\ & \quad + \int_t^{t+\varepsilon} g_{ij}(u, y, \sigma^T(x) p) d\langle B^i, B^j \rangle_u \\ & \quad \left. + \int_t^{t+\varepsilon} \langle p, h_{ij}(x) \rangle d\langle B^i, B^j \rangle_u \right] \Big\}. \end{aligned} \tag{33}$$

By the Lipschitz condition, we can get  $|M_\varepsilon| \leq (C_5/\varepsilon)\widehat{\mathbb{E}}_t[\int_t^{t+\varepsilon} |X_u^{t,x} - x|du]$ , where  $C_5$  depends on  $p, G, L$ , and  $T$ . Noting that  $\widehat{\mathbb{E}}[\sup_{s \in [t, t+\varepsilon]} |X_s^{t,x} - x|^\alpha] \leq C_6(1 + |x|^\alpha)\varepsilon^{\alpha/2}$  (see [16, 19]), where  $C_6$  depends on  $L, G$ , and  $\alpha$ , we obtain

$$\begin{aligned} \widehat{\mathbb{E}}[|M_\varepsilon|^\alpha] & \leq C_5^\alpha \widehat{\mathbb{E}} \left[ \sup_{s \in [t, t+\varepsilon]} |X_s^{t,x} - x|^\alpha \right] \\ & \leq C_5^\alpha C_6 (1 + |x|^\alpha) \varepsilon^{\alpha/2}, \end{aligned} \tag{34}$$

which implies  $L_G^\alpha - \lim_{\varepsilon \rightarrow 0+} M_\varepsilon = 0$ . Now, we set

$$\begin{aligned} N_\varepsilon &= \frac{1}{\varepsilon} \left\{ \widehat{\mathbb{E}}_t \left[ \int_t^{t+\varepsilon} f(u, y, \sigma^T(x) p) du + \langle p, b(x) \rangle \varepsilon \right. \right. \\ & \quad + \int_t^{t+\varepsilon} g_{ij}(u, y, \sigma^T(x) p) d\langle B^i, B^j \rangle_u \\ & \quad \left. + \int_t^{t+\varepsilon} \langle p, h_{ij}(x) \rangle d\langle B^i, B^j \rangle_u \right] \\ & \quad - \widehat{\mathbb{E}}_t \left[ \int_t^{t+\varepsilon} f(t, y, \sigma^T(x) p) du + \langle p, b(x) \rangle \varepsilon \right. \\ & \quad + \int_t^{t+\varepsilon} g_{ij}(t, y, \sigma^T(x) p) d\langle B^i, B^j \rangle_u \\ & \quad \left. + \int_t^{t+\varepsilon} \langle p, h_{ij}(x) \rangle d\langle B^i, B^j \rangle_u \right] \Big\}. \end{aligned} \tag{35}$$

It is easy to deduce that  $|N_\varepsilon| \leq (C_7/\varepsilon)\widehat{\mathbb{E}}_t[\int_t^{t+\varepsilon} (|f(u, y, \sigma^T(x) p) - f(t, y, \sigma^T(x) p)| + \sum_{i,j=1}^d |g_{ij}(u, y, \sigma^T(x) p) - g_{ij}(t, y, \sigma^T(x) p)|)du]$ , where  $C_7$  depends on  $G$ . Then,

$$\begin{aligned} & \widehat{\mathbb{E}}[|N_\varepsilon|^\alpha] \\ & \leq C_7^\alpha \frac{1}{\varepsilon} \widehat{\mathbb{E}} \left[ \int_t^{t+\varepsilon} \left( |f(u, y, \sigma^T(x) p) - f(t, y, \sigma^T(x) p)| \right. \right. \\ & \quad \left. \left. + \sum_{i,j=1}^d |g_{ij}(u, y, \sigma^T(x) p) - g_{ij}(t, y, \sigma^T(x) p)| \right)^\alpha du \right] \\ & \leq C_7^\alpha \left( \frac{1}{\varepsilon} \widehat{\mathbb{E}} \left[ \int_t^{t+\varepsilon} \left( |f(u, y, \sigma^T(x) p) - f(t, y, \sigma^T(x) p)| \right. \right. \right. \\ & \quad \left. \left. + \sum_{i,j=1}^d |g_{ij}(u, y, \sigma^T(x) p) - g_{ij}(t, y, \sigma^T(x) p)| \right)^\beta du \right] \right)^{\alpha/\beta}. \end{aligned} \tag{36}$$

Take limit on both sides of the above inequality and use assumption (H4); then, we have

$$L_G^\alpha - \lim_{\varepsilon \rightarrow 0+} N_\varepsilon = 0. \tag{37}$$

On the other hand,

$$\begin{aligned}
 & \widehat{\mathbb{E}}_t \left[ \int_t^{t+\varepsilon} f(t, y, \sigma^T(x)p) du + \langle p, b(x) \rangle \varepsilon \right. \\
 & \quad + \int_t^{t+\varepsilon} g_{ij}(t, y, \sigma^T(x)p) d\langle B^i, B^j \rangle_u \\
 & \quad \left. + \int_t^{t+\varepsilon} \langle p, h_{ij}(x) \rangle d\langle B^i, B^j \rangle_u \right] \\
 & = f(t, y, \sigma^T(x)p) \varepsilon + \langle p, b(x) \rangle \varepsilon \\
 & \quad + \widehat{\mathbb{E}}_t \left[ (g_{ij}(t, y, \sigma^T(x)p) + \langle p, h_{ij}(x) \rangle) \right. \\
 & \quad \quad \left. \times (\langle B^i, B^j \rangle_{t+\varepsilon} - \langle B^i, B^j \rangle_t) \right] \\
 & = \left( f(t, y, \sigma^T(x)p) + \langle p, b(x) \rangle \right. \\
 & \quad \left. + 2G \left( (g_{ij}(t, y, \sigma^T(x)p) + \langle p, h_{ij}(x) \rangle)_{i,j=1}^d \right) \right) \varepsilon.
 \end{aligned} \tag{38}$$

Then, we have

$$\begin{aligned}
 & L_G^\alpha - \lim_{\varepsilon \rightarrow 0^+} \frac{1}{\varepsilon} \{Y_t^\varepsilon - y\} \\
 & = f(t, y, \sigma^T(x)p) + \langle p, b(x) \rangle \\
 & \quad + 2G \left( (g_{ij}(t, y, \sigma^T(x)p) + \langle p, h_{ij}(x) \rangle)_{i,j=1}^d \right).
 \end{aligned} \tag{39}$$

The proof is complete.  $\square$

### 4. Some Applications

4.1. *Converse Comparison Theorem for G-BSDEs.* We consider the following G-BSDEs:

$$\begin{aligned}
 Y_t^{l,\xi} & = \xi + \int_t^T f^l(s, Y_s^{l,\xi}, Z_s^{l,\xi}) ds \\
 & \quad + \int_t^T g_{ij}^l(s, Y_s^{l,\xi}, Z_s^{l,\xi}) d\langle B^i, B^j \rangle_s \\
 & \quad - \int_t^T Z_s^{l,\xi} dB_s - (K_T^{l,\xi} - K_t^{l,\xi}), \quad l = 1, 2,
 \end{aligned} \tag{40}$$

where  $g_{ij}^l = g_{ji}^l$ .

We first generalized the comparison theorem in [16].

**Proposition 13.** *Let  $f^l$  and  $g_{ij}^l$  satisfy (H1) and (H2) for some  $\beta > 1, l = 1, 2$ . If  $f^2 - f^1 + 2G((g_{ij}^2 - g_{ij}^1)_{i,j=1}^d) \leq 0$ , then, for each  $\xi \in L_G^\beta(\Omega_T)$ , one has  $Y_t^{1,\xi} \geq Y_t^{2,\xi}$  for  $t \in [0, T]$ .*

*Proof.* From the above G-BSDEs, we have

$$\begin{aligned}
 Y_t^{2,\xi} & = \xi + \int_t^T f^2(s, Y_s^{2,\xi}, Z_s^{2,\xi}) ds \\
 & \quad + \int_t^T g_{ij}^2(s, Y_s^{2,\xi}, Z_s^{2,\xi}) d\langle B^i, B^j \rangle_s \\
 & \quad - \int_t^T Z_s^{2,\xi} dB_s - (K_T^{2,\xi} - K_t^{2,\xi}) \\
 & = \xi + \int_t^T f^1(s, Y_s^{2,\xi}, Z_s^{2,\xi}) ds \\
 & \quad + \int_t^T g_{ij}^1(s, Y_s^{2,\xi}, Z_s^{2,\xi}) d\langle B^i, B^j \rangle_s \\
 & \quad + V_T - V_t - \int_t^T Z_s^{2,\xi} dB_s - (K_T^{2,\xi} - K_t^{2,\xi}),
 \end{aligned} \tag{41}$$

where

$$\begin{aligned}
 V_t & = \int_0^t (f^2 - f^1)(s, Y_s^{2,\xi}, Z_s^{2,\xi}) ds \\
 & \quad + \int_0^t (g_{ij}^2 - g_{ij}^1)(s, Y_s^{2,\xi}, Z_s^{2,\xi}) d\langle B^i, B^j \rangle_s \\
 & = \int_0^t \left( f^2 - f^1 + 2G \left( (g_{ij}^2 - g_{ij}^1)_{i,j=1}^d \right) \right) (s, Y_s^{2,\xi}, Z_s^{2,\xi}) ds \\
 & \quad + \int_0^t (g_{ij}^2 - g_{ij}^1)(s, Y_s^{2,\xi}, Z_s^{2,\xi}) d\langle B^i, B^j \rangle_s \\
 & \quad - \int_0^t 2G \left( (g_{ij}^2 - g_{ij}^1)_{i,j=1}^d \right) (s, Y_s^{2,\xi}, Z_s^{2,\xi}) ds.
 \end{aligned} \tag{42}$$

By the assumption, it is easy to check that  $(V_t)_{t \leq T}$  is a decreasing process. Thus, using Theorem 11, we obtain  $Y_t^{1,\xi} \geq Y_t^{2,\xi}$  for  $t \in [0, T]$ .  $\square$

*Remark 14.* Suppose  $d = 1$  and let  $f^1 = 10|z|, f^2 = |z|, g^1 = |z|$ , and  $g^2 = 2|z|$ . It is easy to check that  $f^2 - f^1 + 2G(g^2 - g^1) \leq 0$ . Thus,  $f^2 - f^1 + 2G((g_{ij}^2 - g_{ij}^1)_{i,j=1}^d) \leq 0$  does not imply  $f^2 \leq f^1$  and  $(g_{ij}^2)_{i,j=1}^d \leq (g_{ij}^1)_{i,j=1}^d$ .

Now, we give the converse comparison theorem.

**Theorem 15.** *Let  $f^l$  and  $g_{ij}^l$  satisfy (H1), (H2), (H3), (H4), and (H5) for some  $\beta > 1, l = 1, 2$ . If  $Y_t^{1,\xi} \geq Y_t^{2,\xi}$  for each  $t \in [0, T]$  and  $\xi \in L_G^\beta(\Omega_T)$ , then  $f^2 - f^1 + 2G((g_{ij}^2 - g_{ij}^1)_{i,j=1}^d) \leq 0$  q.s..*

*Proof.* For simplicity, we take the notation  $\widehat{\mathbb{E}}_t^l[\xi] = Y_t^{l,\xi}, l = 1, 2$ . For each fixed  $(t, y, z) \in [0, T] \times \mathbb{R} \times \mathbb{R}^d$ , let us consider

$$\eta_\varepsilon = y + \langle z, h_{ij} \rangle (\langle B^i, B^j \rangle_{t+\varepsilon} - \langle B^i, B^j \rangle_t) + \langle z, B_{t+\varepsilon} - B_t \rangle, \tag{43}$$

where  $h_{ij} = h_{ji} \in \mathbb{R}^d$ . By Theorem 12, we have, for each  $\alpha \in (1, \beta)$ ,

$$\begin{aligned} L_G^\alpha - \lim_{\varepsilon \rightarrow 0^+} \frac{1}{\varepsilon} (\tilde{\mathbb{E}}_t^l[\eta_\varepsilon] - y) \\ = f^l(t, y, z) + 2G\left(\left(g_{ij}^l(t, y, z) + \langle z, h_{ij} \rangle\right)_{i,j=1}^d\right). \end{aligned} \quad (44)$$

Since  $\tilde{\mathbb{E}}_t^1[\eta_\varepsilon] \geq \tilde{\mathbb{E}}_t^2[\eta_\varepsilon]$ ,

$$\begin{aligned} f^1(t, y, z) + 2G\left(\left(g_{ij}^1(t, y, z) + \langle z, h_{ij} \rangle\right)_{i,j=1}^d\right) \\ \geq f^2(t, y, z) + 2G\left(\left(g_{ij}^2(t, y, z) + \langle z, h_{ij} \rangle\right)_{i,j=1}^d\right) \text{ q.s.} \end{aligned} \quad (45)$$

Take a  $h_{ij}$  such that  $\langle z, h_{ij} \rangle = -g_{ij}^1(t, y, z)$ . Therefore,  $\{f^2 - f^1 + 2G((g_{ij}^2 - g_{ij}^1)_{i,j=1}^d)\}(t, y, z) \leq 0$  q.s. By the assumptions (H2) and (H3), it is easy to deduce that  $f^2 - f^1 + 2G((g_{ij}^2 - g_{ij}^1)_{i,j=1}^d) \leq 0$  q.s.  $\square$

In the following, we use the notation  $\tilde{\mathbb{E}}_t^l[\xi] = Y_t^{l,\xi}$ ,  $l = 1, 2$ .

**Corollary 16.** Let  $f^l$  and  $g_{ij}^l$  be deterministic functions and satisfy (H1), (H2), (H3), and (H5) for some  $\beta > 1$ ,  $l = 1, 2$ . If  $\tilde{\mathbb{E}}^1[\xi] \geq \tilde{\mathbb{E}}^2[\xi]$  for each  $\xi \in L_G^\beta(\Omega_T)$ , then  $f^2 - f^1 + 2G((g_{ij}^2 - g_{ij}^1)_{i,j=1}^d) \leq 0$ .

*Proof.* Taking  $\eta_\varepsilon$  as in Theorem 15, since  $f^l$  and  $g_{ij}^l$  are deterministic, we could get  $\tilde{\mathbb{E}}_t^l[\eta_\varepsilon] = \tilde{\mathbb{E}}^l[\eta_\varepsilon]$ , for  $l = 1, 2$ . And the proof in Theorem 15 still holds true.  $\square$

**4.2. Some Equivalent Relations.** We consider the following G-BSDE:

$$\begin{aligned} Y_t = \xi + \int_t^T f(s, Y_s, Z_s) ds + \int_t^T g_{ij}(s, Y_s, Z_s) d\langle B^i, B^j \rangle_s \\ - \int_t^T Z_s dB_s - (K_T - K_t), \end{aligned} \quad (46)$$

where  $g_{ij} = g_{ji}$ . We use the notation  $\tilde{\mathbb{E}}_t[\xi] = Y_t$ .

**Proposition 17.** Let  $f$  and  $g_{ij}$  satisfy (H1), (H2), (H3), (H4), and (H5) for some  $\beta > 1$  and fix  $\alpha \in (1, \beta)$ . Then, one has

$$(1) \tilde{\mathbb{E}}_t[\xi + \eta] = \tilde{\mathbb{E}}_t[\xi] + \eta \text{ for } t \in [0, T], \xi \in L_G^\alpha(\Omega_T), \text{ and } \eta \in L_G^\alpha(\Omega_t) \text{ if and only if for each } t \in [0, T], y, y' \in \mathbb{R}, z \in \mathbb{R}^d,$$

$$\begin{aligned} f(t, y, z) - f(t, y', z) \\ + 2G\left(\left(g_{ij}(t, y, z) - g_{ij}(t, y', z)\right)_{i,j=1}^d\right) = 0; \end{aligned} \quad (47)$$

$$(2) \tilde{\mathbb{E}}_t[\xi + \eta] \leq \tilde{\mathbb{E}}_t[\xi] + \tilde{\mathbb{E}}_t[\eta] \text{ for } t \in [0, T], \xi \in L_G^\alpha(\Omega_T), \text{ and } \eta \in L_G^\alpha(\Omega_T) \text{ if and only if for each } t \in [0, T], y, y' \in \mathbb{R}, z, z' \in \mathbb{R}^d,$$

$$\begin{aligned} 0 \geq f(t, y + y', z + z') - f(t, y, z) - f(t, y', z') \\ + 2G\left(\left(g_{ij}(t, y + y', z + z') - g_{ij}(t, y, z) \right. \right. \\ \left. \left. - g_{ij}(t, y', z')\right)_{i,j=1}^d\right); \end{aligned} \quad (48)$$

$$(3) \tilde{\mathbb{E}}_t[\lambda\xi + (1-\lambda)\eta] \leq \lambda\tilde{\mathbb{E}}_t[\xi] + (1-\lambda)\tilde{\mathbb{E}}_t[\eta] \text{ for } t \in [0, T], \lambda \in [0, 1], \xi \in L_G^\alpha(\Omega_T), \text{ and } \eta \in L_G^\alpha(\Omega_T) \text{ if and only if for each } t \in [0, T], y, y' \in \mathbb{R}, z, z' \in \mathbb{R}^d, \lambda \in [0, 1],$$

$$\begin{aligned} 0 \geq f(t, \lambda y + (1-\lambda)y', \lambda z + (1-\lambda)z') \\ - \lambda f(t, y, z) - (1-\lambda)f(t, y', z') \\ + 2G\left(\left(g_{ij}(t, \lambda y + (1-\lambda)y', \lambda z + (1-\lambda)z') \right. \right. \\ \left. \left. - \lambda g_{ij}(t, y, z) - (1-\lambda)g_{ij}(t, y', z')\right)_{i,j=1}^d\right); \end{aligned} \quad (49)$$

$$(4) \tilde{\mathbb{E}}_t[\lambda\xi] = \lambda\tilde{\mathbb{E}}_t[\xi] \text{ for } t \in [0, T], \lambda \geq 0, \text{ and } \xi \in L_G^\alpha(\Omega_T) \text{ if and only if for each } t \in [0, T], y \in \mathbb{R}, z \in \mathbb{R}^d, \lambda \geq 0,$$

$$\begin{aligned} f(t, \lambda y, \lambda z) - \lambda f(t, y, z) \\ = 2G\left(\left(\lambda g_{ij}(t, y, z) - g_{ij}(t, \lambda y, \lambda z)\right)_{i,j=1}^d\right) \\ = -2G\left(\left(g_{ij}(t, \lambda y, \lambda z) - \lambda g_{ij}(t, y, z)\right)_{i,j=1}^d\right). \end{aligned} \quad (50)$$

*Proof.* (1) “ $\Rightarrow$ ” part. For each fixed  $t \in [0, T]$ ,  $y, y' \in \mathbb{R}, z \in \mathbb{R}^d$ , we take

$$\begin{aligned} \xi_\varepsilon = y + \langle z, h_{ij} \rangle (\langle B^i, B^j \rangle_{t+\varepsilon} - \langle B^i, B^j \rangle_t) \\ + \langle z, B_{t+\varepsilon} - B_t \rangle, \quad \eta = y' - y, \end{aligned} \quad (51)$$

where  $h_{ij} = h_{ji} \in \mathbb{R}^d$ . Then, by Theorem 12 and  $\tilde{\mathbb{E}}_t[\xi_\varepsilon + \eta] = \tilde{\mathbb{E}}_t[\xi_\varepsilon] + \eta$ , we can obtain

$$\begin{aligned} f(t, y', z) + 2G\left(\left(g_{ij}(t, y', z) + \langle z, h_{ij} \rangle\right)_{i,j=1}^d\right) \\ = f(t, y, z) + 2G\left(\left(g_{ij}(t, y, z) + \langle z, h_{ij} \rangle\right)_{i,j=1}^d\right). \end{aligned} \quad (52)$$

We choose  $h_{ij}$  such that  $g_{ij}(t, y', z) + \langle z, h_{ij} \rangle = 0$ , which implies (47).

“ $\Leftarrow$ ” part. Let  $(Y, Z, K)$  be the solution of G-BSDE (46) corresponding to terminal condition  $\xi$ . We claim that  $(Y_s + \eta, Z_s, K_s)_{s \in [t, T]}$  is the solution of G-BSDE (46) corresponding



to terminal condition  $\xi + \eta$  on  $[t, T]$ . For this, we only need to check that, for  $s \in [t, T]$ ,

$$\begin{aligned} & \int_s^T f(u, Y_u, Z_u) du + \int_s^T g_{ij}(u, Y_u, Z_u) d\langle B^i, B^j \rangle_u \\ &= \int_s^T f(u, Y_u + \eta, Z_u) du \\ &+ \int_s^T g_{ij}(u, Y_u + \eta, Z_u) d\langle B^i, B^j \rangle_u. \end{aligned} \tag{53}$$

By (47) we can get

$$\begin{aligned} & \int_s^T (g_{ij}(u, Y_u, Z_u) - g_{ij}(u, Y_u + \eta, Z_u)) d\langle B^i, B^j \rangle_u \\ & - 2 \int_s^T G\left(\left(g_{ij}(u, Y_u, Z_u) - g_{ij}(u, Y_u + \eta, Z_u)\right)_{i,j=1}^d\right) du \\ &= \int_s^T (g_{ij}(u, Y_u, Z_u) - g_{ij}(u, Y_u + \eta, Z_u)) d\langle B^i, B^j \rangle_u \\ &+ \int_s^T (f(u, Y_u, Z_u) - f(u, Y_u + \eta, Z_u)) du \leq 0, \\ & \int_s^T (g_{ij}(u, Y_u + \eta, Z_u) - g_{ij}(u, Y_u, Z_u)) d\langle B^i, B^j \rangle_u \\ & - 2 \int_s^T G\left(\left(g_{ij}(u, Y_u + \eta, Z_u) - g_{ij}(u, Y_u, Z_u)\right)_{i,j=1}^d\right) du \\ &= \int_s^T (g_{ij}(u, Y_u + \eta, Z_u) - g_{ij}(u, Y_u, Z_u)) d\langle B^i, B^j \rangle_u \\ &+ \int_s^T (f(u, Y_u + \eta, Z_u) - f(u, Y_u, Z_u)) du \leq 0, \end{aligned} \tag{54}$$

which implies (53). The proof of (1) is complete.

(2) “ $\Rightarrow$ ” part. For each fixed  $t \in [0, T)$ ,  $y, y' \in \mathbb{R}, z, z' \in \mathbb{R}^d$ , we consider  $\xi_\varepsilon = y + \langle z, h_{ij} \rangle (\langle B^i, B^j \rangle_{t+\varepsilon} - \langle B^i, B^j \rangle_t) + \langle z, B_{t+\varepsilon} - B_t \rangle$  and  $\eta_\varepsilon = y' + \langle z', h'_{ij} \rangle (\langle B^i, B^j \rangle_{t+\varepsilon} - \langle B^i, B^j \rangle_t) + \langle z', B_{t+\varepsilon} - B_t \rangle$ , where  $h_{ij} = h_{ji} \in \mathbb{R}^d$  and  $h'_{ij} = h'_{ji} \in \mathbb{R}^d$ . Then, by Theorem 12 and  $\tilde{\mathbb{E}}_t[\xi_\varepsilon + \eta_\varepsilon] = \tilde{\mathbb{E}}_t[\xi_\varepsilon] + \tilde{\mathbb{E}}_t[\eta_\varepsilon]$ , we obtain

$$\begin{aligned} & f(t, y + y', z + z') \\ &+ 2G\left(\left(g_{ij}(t, y + y', z + z')\right.\right. \\ &\quad \left.\left.+ \langle z, h_{ij} \rangle + \langle z', h'_{ij} \rangle\right)_{i,j=1}^d\right) \\ &\leq f(t, y, z) + f(t, y', z') \\ &+ 2G\left(\left(g_{ij}(t, y, z) + \langle z, h_{ij} \rangle\right)_{i,j=1}^d\right) \\ &+ 2G\left(\left(g_{ij}(t, y', z') + \langle z', h'_{ij} \rangle\right)_{i,j=1}^d\right). \end{aligned} \tag{55}$$

We choose  $h_{ij}, h'_{ij}$  such that  $g_{ij}(t, y, z) + \langle z, h_{ij} \rangle = 0$  and  $g_{ij}(t, y', z') + \langle z', h'_{ij} \rangle = 0$ , which implies (48).

“ $\Leftarrow$ ” part. Let  $(Y, Z, K)$  and  $(Y', Z', K')$  be the solutions of G-BSDE (46) corresponding to terminal condition  $\xi$  and  $\eta$ , respectively. Then,  $(Y + Y', Z + Z', K)$  solves the following G-BSDE:

$$\begin{aligned} Y_t + Y'_t &= \xi + \eta + \int_t^T f(s, Y_s + Y'_s, Z_s + Z'_s) ds \\ &+ \int_t^T g_{ij}(s, Y_s + Y'_s, Z_s + Z'_s) d\langle B^i, B^j \rangle_s \\ &+ V_T - V_t - \int_t^T (Z_s + Z'_s) dB_s - (K_T - K_t), \end{aligned} \tag{56}$$

where

$$\begin{aligned} V_t &= -K'_t - \int_0^t (f(s, Y_s + Y'_s, Z_s + Z'_s) \\ &\quad - f(s, Y_s, Z_s) - f(s, Y'_s, Z'_s)) ds \\ &- \int_0^t (g_{ij}(s, Y_s + Y'_s, Z_s + Z'_s) - g_{ij}(s, Y_s, Z_s) \\ &\quad - g_{ij}(s, Y'_s, Z'_s)) d\langle B^i, B^j \rangle_s \\ &= -K'_t - \left\{ \int_0^t (g_{ij}(s, Y_s + Y'_s, Z_s + Z'_s) \right. \\ &\quad - g_{ij}(s, Y_s, Z_s) - g_{ij}(s, Y'_s, Z'_s)) d\langle B^i, B^j \rangle_s \\ &\quad - 2 \int_0^t G\left(\left(g_{ij}(s, Y_s + Y'_s, Z_s + Z'_s) \right. \right. \\ &\quad \left. \left. - g_{ij}(s, Y_s, Z_s) \right. \right. \\ &\quad \left. \left. - g_{ij}(s, Y'_s, Z'_s)\right)_{i,j=1}^d\right) ds \Big\} \\ &- \int_0^t \left\{ f(s, Y_s + Y'_s, Z_s + Z'_s) \right. \\ &\quad - f(s, Y_s, Z_s) - f(s, Y'_s, Z'_s) \\ &\quad + 2G\left(\left(g_{ij}(s, Y_s + Y'_s, Z_s + Z'_s) - g_{ij}(s, Y_s, Z_s) \right. \right. \\ &\quad \left. \left. - g_{ij}(s, Y'_s, Z'_s)\right)_{i,j=1}^d\right) \Big\} ds. \end{aligned} \tag{57}$$

By (48), it is easy to check that  $V_t$  is an increasing process. Then, by Theorem 11, we can get  $\tilde{\mathbb{E}}_t[\xi + \eta] \leq \tilde{\mathbb{E}}_t[\xi] + \tilde{\mathbb{E}}_t[\eta]$ . The proof of (2) is complete.

Finally, we could prove (3) as in (2) and (4) as in (1).  $\square$

**Proposition 18.** *One has the following.*

- (1) If  $G(A) + G(-A) > 0$  for any  $A \in \mathbb{S}_d$  and  $A \neq 0$ , then (47) holds if and only if  $f$  and  $g_{ij}$  are independent of  $y$ .

- (2) If there exists an  $A \in \mathbb{S}_d$  with  $A \neq 0$  such that  $G(A) + G(-A) = 0$  and  $G(A) \neq 0$ , then, for any fixed  $g(t, y, z)$  satisfying (H1)–(H5), one has  $f(t, y, z) = -2G(A)g(t, y, z)$  and  $(g_{ij}(t, y, z))_{i,j=1}^d = g(t, y, z)A$  satisfying (47).

*Proof.* It is easy to verify (2), and we only need to prove (1). If (47) holds, it is easy to check that  $G((g_{ij}(t, y, z) - g_{ij}(t, 0, z))_{i,j=1}^d) + G((g_{ij}(t, 0, z) - g_{ij}(t, y, z))_{i,j=1}^d) = 0$  holds. Then, from the assumption, we get  $g_{ij}(t, y, z) = g_{ij}(t, 0, z)$ . Therefore, by (47), we have  $f(t, y, z) = f(t, 0, z)$ , which implies that  $f$  and  $g_{ij}$  are independent of  $y$ . The converse part is obvious.  $\square$

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