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## CONVERSE COMPARISON THEOREMS FOR BACKWARD DOUBLY STOCHASTIC DIFFERENTIAL EQUATIONS

MOHAMED EL OTMANI AND NAOUAL MRHARDY

ABSTRACT. Converse and general converse comparison theorems are proved for backward doubly stochastic differential equations.

### 1. Introduction

A new kind of backward stochastic differential equations (BSDEs in short) was introduced by Pardoux and Peng [6] in 1994, that is a class of backward doubly stochastic differential equations (BDSDEs in short) with two different directions of stochastic integrals, i.e., equations involving both a standard (forward) stochastic integral and a backward stochastic integral. That is, BDSDEs are stochastic differential equations of the form

$$Y_t = \xi + \int_t^T g(s, Y_s, Z_s) ds + \int_t^T h(s, Y_s, Z_s) d\bar{B}_s - \int_t^T Z_s dW_s, \quad 0 \leq t \leq T, \quad (1.1)$$

where  $\xi$  is the terminal value,  $g$  the generator and  $h$  is a drift function.

The comparison theorem, which is an important and effective technique in the theory of BSDE, was first established for BDSDEs by [8]. It allows one to compare the solutions of two real-valued BDSDEs whenever we can compare the terminal conditions and the generators. An inverse problem is interesting: namely, if we can compare the solution of two BDSDEs with the same terminal condition, can we compare the generators?

To put our result in context, we note that if  $h = 0$ , the result of Chen [2] can be thought as the first step in solving this theorem and then it was further developed by Briand et al. [1], Coquet et al. [3] and Jiang [4].

On the other context, using BSDEs, Peng introduced in [7] the notion of  $g$ -expectation; he considers the function  $\mathcal{E}_g$  defined on  $\mathbb{L}^2(\mathcal{F}_T^W)$  with values in  $\mathbb{R}$  by simply setting  $\mathcal{E}_g(\xi) = Y_0$  where  $(Y, Z)$  is the unique solution of the BSDE

$$Y_t = \xi + \int_t^T g(s, Y_s, Z_s) ds - \int_t^T Z_s dW_s, \quad 0 \leq t \leq T.$$

$\mathcal{E}_g(\xi)$  is called the  $g$ -expectation of  $\xi$ . Similarly, the conditional  $g$ -expectation is introduced by setting, for any stopping time  $\tau$ ,  $\mathcal{E}_g[\xi | \mathcal{F}_\tau^W] = Y_\tau$  which is the unique

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$\mathcal{F}_t$ -measurable and square integrable random variable  $\eta$  such that

$$\forall A \in \mathcal{F}_t, \quad \mathcal{E}_g(\mathbb{I}_A \eta) = \mathcal{E}_g(\mathbb{I}_A \xi).$$

In this paper, we are concerned with converse and general converse comparison theorems for BDSDEs. Let  $(Y^i, Z^i)$  be the unique square integrable and adapted solution of the BDSDE (1.1) with data  $(\xi, g_i, h)$ . Under suitable conditions, we prove that for each terminal condition  $\xi$ , if  $\forall t Y_t^1 \leq Y_t^2$  then  $g_1 \leq g_2$ . More in general, the result remain true if we suppose only that  $Y_0^1 \leq Y_0^2$ .

The rest of the paper is organized as follows: in Subsection 1.1, we introduce some notations and we make our main assumptions. In Section 2, we provide a priori estimate and we introduce the notion of  $g$ -expectation for BDSDEs. In Section 3, we discuss converse comparison theorem for BDSDEs whereas general comparison theorem will be proved in Section 4.

**1.1. Backgrounds.** More precisely, we consider two independent  $d$ -dimensional Brownian motions ( $d \geq 1$ ),  $\{W_t, 0 \leq t \leq T\}$  and  $\{B_t, 0 \leq t \leq T\}$  defined on the complete probability spaces  $(\Omega_1, \mathcal{F}_1, \mathbb{P}_1)$  and  $(\Omega_2, \mathcal{F}_2, \mathbb{P}_2)$  respectively.

We denote

$$\mathcal{F}_{s,t}^B := \sigma \{B_r - B_s, s \leq r \leq t\}, \quad \mathcal{F}_t^W := \sigma \{W_r, 0 \leq r \leq t\}.$$

Moreover, we define  $\Omega \triangleq \Omega_1 \times \Omega_2$ ,  $\mathcal{F} \triangleq \mathcal{F}_1 \otimes \mathcal{F}_2$  and  $\mathbb{P} \triangleq \mathbb{P}_1 \otimes \mathbb{P}_2$ , and we put  $\mathcal{F}_t \triangleq \mathcal{F}_t^W \otimes \mathcal{F}_{t,T}^B \vee \mathcal{N}$ , where  $\mathcal{N}$  is the collection of  $\mathbb{P}$ -null-sets. We notice that the family of  $\sigma$ -algebras  $\mathbb{F} = \{\mathcal{F}_t\}_{0 \leq t \leq T}$  is not a filtration.

Let  $g : \Omega \times [0, T] \times \mathbb{R} \times \mathbb{R}^d \rightarrow \mathbb{R}$  and  $h : \Omega \times [0, T] \times \mathbb{R} \times \mathbb{R}^d \rightarrow \mathbb{R}^d$  be two progressively measurable functions with the property that there exists constants  $\kappa > 0$  and  $0 < \alpha < 1$  such that the following hypothesis are satisfied

(A.0)  $\xi \in \mathbb{L}^2(\mathcal{F}_T)$ .

(A.1) The processes  $(g(t, 0, 0))_{t \in [0, T]}$  and  $(h(t, 0, 0))_{t \in [0, T]}$  are progressively measurable such that  $\mathbb{E} \int_0^T (|g(t, 0, 0)|^2 + \|h(t, 0, 0)\|^2) dt < \infty$ .

(A.2) For any  $(y_1, z_1), (y_2, z_2) \in \mathbb{R} \times \mathbb{R}^d$ ,

$$\begin{cases} |g(t, y_1, z_1) - g(t, y_2, z_2)|^2 \leq \kappa (|y_1 - y_2|^2 + \|z_1 - z_2\|^2) \\ |h(t, y_1, z_1) - h(t, y_2, z_2)|^2 \leq \kappa |y_1 - y_2|^2 + \alpha \|z_1 - z_2\|^2. \end{cases}$$

(A.3)  $\mathbb{P}$ -a.s. for all  $(t, y)$  we have  $g(t, y, 0) = 0$  and  $h(t, y, 0) = 0$ .

(A.4)  $\mathbb{P}$ -a.s. for all  $(y, z)$  the function  $t \mapsto g(t, y, z)$  is continuous.

It was shown in [6] that, under the assumptions (A.0), (A.1) and (A.2), the backward doubly stochastic differential equation (1.1) has a unique solution.

## 2. BDSDEs and $g$ -expectation

In this section, we state a technical result, we define the notion of  $g$ -expectation for solution of BDSDEs and we prove a useful properties of  $g$ -expectation.

**2.1. A priori estimate.**

**Proposition 2.1.** *Let the assumptions (A.0), (A.1) and (A.2) hold. Then, for  $\beta \geq C(\alpha, \kappa)$  we have*

$$\begin{aligned} & \mathbb{E} \left[ \sup_{t \leq s \leq T} e^{\beta s} |Y_s|^2 + \int_t^T e^{\beta s} \|Z_s\|^2 ds \middle| \mathcal{F}_t \right] \\ & \leq C \mathbb{E} \left[ e^{\beta T} |\xi|^2 + \int_t^T e^{(\beta/2)s} |g(s, 0, 0)| ds \right]^2 + \int_t^T e^{\beta s} \|h(s, 0, 0)\|^2 ds \middle| \mathcal{F}_t \right]. \end{aligned}$$

*Proof.* Itô formula yields

$$\begin{aligned} & e^{\beta t} |Y_t|^2 + \int_t^T e^{\beta s} \|Z_s\|^2 ds + \beta \int_t^T e^{\beta s} |Y_s|^2 ds \\ & = e^{\beta T} |\xi|^2 + 2 \int_t^T e^{\beta s} Y_s g(s, Y_s, Z_s) ds + \int_t^T e^{\beta s} \|h(s, Y_s, Z_s)\|^2 ds \\ & \quad + 2 \int_t^T e^{\beta s} Y_s h(s, Y_s, Z_s) d\bar{B}_s - 2 \int_t^T e^{\beta s} Y_s Z_s dW_s. \end{aligned} \tag{2.1}$$

Now, using conditions (A.2) and the algebraic inequality  $2ab \leq \frac{a^2}{\varepsilon} + \varepsilon b^2$ ,  $\varepsilon > 0$  and  $(a + b)^2 \leq (1 + \frac{1}{\varepsilon})a^2 + (1 + \varepsilon)b^2$ ,  $\varepsilon > 0$  we obtain

$$\begin{aligned} 2Y_s g(s, Y_s, Z_s) & \leq 2|Y_s| |g(s, 0, 0)| + (\varepsilon + \frac{\kappa}{\varepsilon}) |Y_s|^2 + \frac{\kappa}{\varepsilon} \|Z_s\|^2 \\ \|h(s, Y_s, Z_s)\|^2 & \leq (1 + \varepsilon) \|h(s, 0, 0)\|^2 + (1 + \frac{1}{\varepsilon}) \kappa |Y_s|^2 + (1 + \frac{1}{\varepsilon}) \alpha \|Z_s\|^2. \end{aligned}$$

By choosing  $\varepsilon = \frac{2\kappa}{1-\alpha}$ ,  $\varepsilon = \frac{3\alpha}{1-\alpha}$  and  $C(\alpha, \kappa) = (\varepsilon + \frac{\kappa}{\varepsilon}) + (1 + \frac{1}{\varepsilon})\kappa$  and plugging the last two inequalities in (2.1), we infer

$$\begin{aligned} & e^{\beta t} |Y_t|^2 + \frac{1-\alpha}{6} \int_t^T e^{\beta s} \|Z_s\|^2 ds + (\beta - C(\alpha, \kappa)) \int_t^T e^{\beta s} |Y_s|^2 ds \\ & \leq e^{\beta T} |\xi|^2 + 2 \int_t^T e^{\beta s} |Y_s| |g(s, 0, 0)| ds + \frac{1+2\alpha}{1-\alpha} \int_t^T e^{\beta s} \|h(s, 0, 0)\|^2 ds \\ & \quad + 2 \int_t^T e^{\beta s} Y_s h(s, Y_s, Z_s) d\bar{B}_s - 2 \int_t^T e^{\beta s} Y_s Z_s dW_s. \end{aligned}$$

By taking  $\beta \geq C(\alpha, \kappa)$  we obtain

$$\begin{aligned} & e^{\beta t} |Y_t|^2 + \int_t^T e^{\beta s} \|Z_s\|^2 ds \\ & \leq C \left\{ e^{\beta T} |\xi|^2 + \int_t^T e^{\beta s} |Y_s| |g(s, 0, 0)| ds + \int_t^T e^{\beta s} \|h(s, 0, 0)\|^2 ds \right. \\ & \quad \left. + \int_t^T e^{\beta s} Y_s h(s, Y_s, Z_s) d\bar{B}_s - \int_t^T e^{\beta s} Y_s Z_s dW_s \right\}. \end{aligned} \tag{2.2}$$

For fixed  $w_1 \in \Omega_1$  (see the notation in Subsection 1.1), we take the conditional expectation  $\mathbb{E}^{\mathcal{F}_{t,T}^B}[\cdot] = \mathbb{E}[\cdot | \mathcal{F}_{t,T}^B]$  to obtain

$$\begin{aligned} & \mathbb{E}^{\mathcal{F}_{t,T}^B} \left[ e^{\beta t} |Y_t|^2 + \int_t^T e^{\beta s} \|Z_s\|^2 ds \right] \\ & \leq C \mathbb{E}^{\mathcal{F}_{t,T}^B} \left[ e^{\beta T} |\xi|^2 + \int_t^T e^{\beta s} |Y_s| |g(s, 0, 0)| ds + \int_t^T e^{\beta s} \|h(s, 0, 0)\|^2 ds \right. \\ & \quad \left. - \int_t^T e^{\beta s} Y_s Z_s dW_s \right]. \end{aligned}$$

Similarly, taking  $\mathbb{E}[\cdot | \mathcal{F}_t^W]$  in the last inequality to get

$$\begin{aligned} & \mathbb{E}^{\mathcal{F}_t} \left[ e^{\beta t} |Y_t|^2 + \int_t^T e^{\beta s} \|Z_s\|^2 ds \right] \tag{2.3} \\ & \leq C \mathbb{E}^{\mathcal{F}_t} \left[ e^{\beta T} |\xi|^2 + \int_t^T e^{\beta s} |Y_s| |g(s, 0, 0)| ds + \int_t^T e^{\beta s} \|h(s, 0, 0)\|^2 ds \right], \end{aligned}$$

where we have used the fact that

$$\mathbb{E}[\cdot | \mathcal{F}_t] = \mathbb{E}[\cdot | \mathcal{F}_{t,T}^B \otimes \mathcal{F}_t^W] = \mathbb{E}[\mathbb{E}[\cdot | \mathcal{F}_{t,T}^B] | \mathcal{F}_t^W].$$

Coming back to the inequality (2.2) we get from Burkholder-Davis-Gundy's inequality, that

$$\begin{aligned} & \mathbb{E}^{\mathcal{F}_t} \left[ \sup_{t \leq s \leq T} e^{\beta s} |Y_s|^2 \right] \\ & \leq C \mathbb{E}^{\mathcal{F}_t} \left[ e^{\beta T} |\xi|^2 + \int_t^T e^{\beta s} |Y_s| |g(s, 0, 0)| ds + \int_t^T e^{\beta s} \|h(s, 0, 0)\|^2 ds \right. \\ & \quad \left. + \left( \int_t^T e^{2\beta s} |Y_s|^2 \|h(s, Y_s, Z_s)\|^2 ds \right)^{1/2} + \left( \int_t^T e^{2\beta s} |Y_s|^2 \|Z_s\|^2 ds \right)^{1/2} \right]. \end{aligned}$$

By similar argument as before, we estimate the last term as follows

$$\begin{aligned} & C \mathbb{E}^{\mathcal{F}_t} \left[ \left( \int_t^T e^{2\beta s} |Y_s|^2 \|h(s, Y_s, Z_s)\|^2 ds \right)^{1/2} \right] \\ & \leq \frac{1}{8} \mathbb{E}^{\mathcal{F}_t} \left[ \sup_{t \leq s \leq T} e^{\beta t} |Y_t|^2 \right] + C \mathbb{E}^{\mathcal{F}_t} \left[ \int_t^T e^{\beta s} \|h(s, Y_s, Z_s)\|^2 ds \right] \\ & \leq \frac{1}{4} \mathbb{E}^{\mathcal{F}_t} \left[ \sup_{t \leq s \leq T} e^{\beta t} |Y_t|^2 \right] + C \mathbb{E}^{\mathcal{F}_t} \left[ \int_0^T e^{\beta s} \|h(s, 0, 0)\|^2 ds + \int_t^T e^{\beta s} \|Z_s\|^2 ds \right] \end{aligned}$$

and

$$\begin{aligned} & C\mathbb{E}^{\mathcal{F}_t} \left[ \left( \int_t^T e^{2\beta s} |Y_s|^2 \|Z_s\|^2 ds \right)^{1/2} \right] \\ & \leq \frac{1}{4} \mathbb{E}^{\mathcal{F}_t} \left[ \sup_{t \leq s \leq T} e^{\beta t} |Y_t|^2 \right] + C\mathbb{E}^{\mathcal{F}_t} \left[ \int_t^T e^{\beta s} \|Z_s\|^2 ds \right]. \end{aligned}$$

Using inequality (2.3) and the previous one, we deduce that

$$\begin{aligned} & \mathbb{E}^{\mathcal{F}_t} \left[ \sup_{t \leq s \leq T} e^{\beta t} |Y_t|^2 \right] \\ & \leq C\mathbb{E}^{\mathcal{F}_t} \left[ e^{\beta T} |\xi|^2 + \int_t^T e^{\beta s} |Y_s| |g(s, 0, 0)| ds + \int_0^T e^{\beta s} \|h(s, 0, 0)\|^2 ds \right]. \end{aligned}$$

Combining this with the fact that

$$\begin{aligned} & C\mathbb{E}^{\mathcal{F}_t} \left[ \int_t^T e^{\beta s} |Y_s| |g(s, 0, 0)| ds \right] \\ & \leq \frac{1}{2} \mathbb{E}^{\mathcal{F}_t} \left[ \sup_{0 \leq t \leq T} e^{\beta t} |Y_t|^2 \right] + \frac{C^2}{2} \mathbb{E}^{\mathcal{F}_t} \left[ \int_t^T e^{(\beta/2)s} |g(s, 0, 0)| ds \right]^2, \end{aligned}$$

we obtain the desired result.  $\square$

**2.2.  $g$ -expectation for BDSDEs.** Assuming that the data  $(\xi, g, h)$  satisfies the assumptions (A.0) – (A.3). Let us introduce the operator  $\mathcal{E}_{g,h}$  as: for any  $\xi \in \mathbb{L}^2(\Omega, \mathcal{F}_T, \mathbb{P})$ , denote by  $\mathcal{E}_{g,h}(\xi)$  and  $\mathcal{E}_{g,h}[\xi|\mathcal{F}_t]$  the initial value  $Y_0$  and the value  $Y_t$  at time  $t$  of the solution to BDSDE (1.1) respectively. For a stopping time  $\tau$ , the operator  $\mathcal{E}_{g,h}[\xi|\mathcal{F}_\tau]$  can be defined in an identical way.

Note that, since  $Y_0$  is  $\mathcal{F}_{0,T}^B$ -measurable, then  $\mathcal{E}_{g,h}(\xi)$  is not deterministic.

**N.B.** For simplicity, we remove the dependence of  $h$  in the sequel.

The following properties are a direct consequence to the comparison, strict comparison theorem and the uniqueness of the solution for BDSDEs

**Proposition 2.2.** *Let us suppose that  $\xi_1, \xi_2 \in \mathbb{L}^2(\mathcal{F}_T)$  and  $\tau, \nu$  are two stopping times with value in  $[0, T]$ .*

- (1) *For any  $X \in \mathbb{L}^2(\mathcal{F}_0)$ ,  $\mathcal{E}_g(X) = X$ . In particular  $\mathcal{E}_g(c) = c$  for any constant  $c \in \mathbb{R}$ .*
- (2) *If  $\xi_1 \leq \xi_2$  a.s. then  $\mathcal{E}_g(\xi_1) \leq \mathcal{E}_g(\xi_2)$  a.s.*
- (3) *If  $\xi_1 \leq \xi_2$  a.s. and  $\mathbb{P}(\xi_1 < \xi_2) > 0$  then  $\mathcal{E}_g(\xi_1) < \mathcal{E}_g(\xi_2)$  a.s.*
- (4) *If  $\tau \leq \nu$ ,  $\mathcal{E}_g[\mathcal{E}_g[\xi_1|\mathcal{F}_\nu]|\mathcal{F}_\tau] = \mathcal{E}_g[\xi_1|\mathcal{F}_\tau]$ .  
In particular,  $\mathcal{E}_g(\xi_1) = \mathcal{E}_g(\mathcal{E}_g[\xi_1|\mathcal{F}_\nu])$ .*

### 3. Converse Comparison Theorem for BDSDEs

Let  $\sigma : \mathbb{R}^n \rightarrow \mathbb{R}^n \times \mathbb{R}^d$  be a globally  $\kappa$ -Lipschitz function. It is well known that, for  $(t, x) \in [0, T) \times \mathbb{R}^n$ , there exists a unique process  $X^{t,x}$  solution of the stochastic

differential equation

$$X_s^{t,x} = x + \int_t^{t \vee s} \sigma(X_r^{t,x}) dW_r, \quad 0 \leq s \leq T$$

which satisfies

$$\mathbb{E} \sup_{u \leq s \leq u'} \|X_s^{t,x}\|^2 \leq C(1 + \|x\|^2)|u - u'|. \quad (3.1)$$

For any fixed  $(t, x, y, p) \in [0, T] \times \mathbb{R}^n \times \mathbb{R} \times \mathbb{R}^n$  and for any  $\varepsilon > 0$ , let us denote  $({}^\varepsilon Y^{t,x,y,p}, {}^\varepsilon Z^{t,x,y,p}) = (Y^\varepsilon, Z^\varepsilon)$  the solution on  $[0, t + \varepsilon]$  of the following BDSDE

$$\begin{aligned} Y_s^\varepsilon &= y + p(X_{t+\varepsilon}^{t,x} - x) + \int_s^{t+\varepsilon} g(r, Y_r^\varepsilon, Z_r^\varepsilon) dr \\ &\quad + \int_s^{t+\varepsilon} h(r, Y_r^\varepsilon, Z_r^\varepsilon) d\bar{B}_r - \int_s^{t+\varepsilon} Z_r^\varepsilon dW_r. \end{aligned}$$

To introduce our converse comparison theorem, we need the following lemma.

**Lemma 3.1.** *Let the assumptions (A.0)–(A.2) and (A.4) hold and supposing that*

$$\mathbb{E} \left( \sup_{0 \leq t \leq T} |g(t, 0, 0)|^2 \right) < \infty \quad \text{and} \quad h(t, 0, 0) = 0 \quad \text{for all } t.$$

Then,

$$\frac{1}{\varepsilon} ({}^\varepsilon Y_t^{t,x,y,p} - y) \xrightarrow[\varepsilon \rightarrow 0^+]{\mathbb{L}^2} g(t, y, \sigma(x)p).$$

*Proof.* For  $s \in [t, t + \varepsilon]$ , put  $(\tilde{Y}_s^\varepsilon, \tilde{Z}_s^\varepsilon) = (Y_s^\varepsilon - (y + p^*(X_s^{t,x} - x)), Z_s^\varepsilon - \sigma(X_s^{t,x})p)$ . Then  $d\tilde{Y}_s^\varepsilon = dY_s^\varepsilon - p dX_s^{t,x}$ ,  $\tilde{Y}_{t+\varepsilon}^\varepsilon = 0$ . That is

$$\begin{aligned} \tilde{Y}_s^\varepsilon &= \int_s^{t+\varepsilon} g(r, y + p^*(X_r^{t,x} - x) + \tilde{Y}_r^\varepsilon, \sigma(X_r^{t,x})p + \tilde{Z}_r^\varepsilon) dr \\ &\quad + \int_s^{t+\varepsilon} h(r, y + p^*(X_r^{t,x} - x) + \tilde{Y}_r^\varepsilon, \sigma(X_r^{t,x})p + \tilde{Z}_r^\varepsilon) d\bar{B}_r - \int_s^{t+\varepsilon} \tilde{Z}_r^\varepsilon dW_r. \end{aligned}$$

Combining Proposition 2.1, Assumption (A.2) and that  $\sigma$  is Lipschitz, we obtain

$$\begin{aligned} &\mathbb{E} \left[ \sup_{t \leq s \leq t+\varepsilon} |\tilde{Y}_s^\varepsilon|^2 + \int_t^{t+\varepsilon} \|\tilde{Z}_s^\varepsilon\|^2 ds \middle| \mathcal{F}_t \right] \\ &\leq C \mathbb{E} \left[ \left| \int_t^{t+\varepsilon} g(r, y + p(X_r^{t,x} - x), \sigma(X_r^{t,x})p) dr \right|^2 \middle| \mathcal{F}_t \right] \\ &\leq C(\kappa, x, y, p) \varepsilon^2 \mathbb{E} \left[ \sup_{t \leq s \leq t+\varepsilon} (\|X_s^{t,x}\|^2 + |g(s, 0, 0)|^2) \middle| \mathcal{F}_t \right], \end{aligned}$$

which implies by taking expectation and using (3.1) that

$$\mathbb{E} \left( \sup_{t \leq s \leq t+\varepsilon} |\tilde{Y}_s^\varepsilon|^2 + \int_t^{t+\varepsilon} \|\tilde{Z}_s^\varepsilon\|^2 ds \right) \leq C\varepsilon^2. \quad (3.2)$$

On the other hand, we have

$$\begin{aligned} & \frac{1}{\varepsilon} (\varepsilon Y_t^{t,x,y,p} - y) - g(t, y, \sigma(x)p) \\ &= \frac{1}{\varepsilon} \mathbb{E} \left[ \int_t^{t+\varepsilon} \left\{ g \left( r, y + p(X_r^{t,x} - x) + \tilde{Y}_r^\varepsilon, \sigma(X_r^{t,x})p + \tilde{Z}_r^\varepsilon \right) \right. \right. \\ & \quad \left. \left. - g \left( r, y + p(X_r^{t,x} - x), \sigma(X_r^{t,x})p \right) \right\} dr \middle| \mathcal{F}_t \right] \\ & \quad + \frac{1}{\varepsilon} \mathbb{E} \left[ \int_t^{t+\varepsilon} \left\{ g \left( r, y + p(X_r^{t,x} - x), \sigma(X_r^{t,x})p \right) - g(r, y, \sigma(x)p) \right\} dr \middle| \mathcal{F}_t \right] \\ & \quad + \frac{1}{\varepsilon} \mathbb{E} \left[ \int_t^{t+\varepsilon} \left\{ g(r, y, \sigma(x)p) - g(t, y, \sigma(x)p) \right\} dr \middle| \mathcal{F}_t \right] \\ &= T_1 + T_2 + T_3. \end{aligned}$$

First, combining Jensen and Hölder inequality we obtain

$$\begin{aligned} \mathbb{E}|T_1|^2 &\leq \frac{1}{\varepsilon} \mathbb{E} \int_t^{t+\varepsilon} \left| g \left( r, y + p(X_r^{t,x} - x) + \tilde{Y}_r^\varepsilon, \sigma(X_r^{t,x})p + \tilde{Z}_r^\varepsilon \right) \right. \\ & \quad \left. - g \left( r, y + p(X_r^{t,x} - x), \sigma(X_r^{t,x})p \right) \right|^2 dr. \end{aligned}$$

From (A.2) and (3.2) we have

$$\mathbb{E}|T_1|^2 \leq \frac{\kappa}{\varepsilon} \mathbb{E} \int_t^{t+\varepsilon} (|\tilde{Y}_r^\varepsilon|^2 + \|\tilde{Z}_r^\varepsilon\|^2) dr \leq C(\varepsilon^2 + \varepsilon)$$

which show the convergence of  $T_1$  to 0 as  $\varepsilon \rightarrow 0$ . On the same way, we get

$$\mathbb{E}|T_2|^2 \leq \frac{\kappa}{\varepsilon} \|p\|^2 \mathbb{E} \int_t^{t+\varepsilon} \|X_r^{t,x} - x\|^2 dr \leq C \mathbb{E} \int_t^{t+\varepsilon} \|\sigma(X_r^{t,x})\|^2 dr \leq C\varepsilon.$$

Consequently,  $T_2 \rightarrow 0$  as  $\varepsilon \rightarrow 0$  in  $L^2$ .

It remains to show that  $T_3$  converges to 0 as  $\varepsilon \rightarrow 0$ . Indeed, using Hölder inequality,

$$\mathbb{E}|T_3|^2 \leq \mathbb{E} \left( \frac{1}{\varepsilon} \int_t^{t+\varepsilon} |g(r, y, \sigma(x)p) - g(t, y, \sigma(x)p)|^2 dr \right)$$

Since the process  $(g(t, y, z))_{t \in [0, T]}$  is continuous, the right hand in the last inequality goes to 0 as  $\varepsilon \rightarrow 0$ . Moreover, we have

$$\frac{1}{\varepsilon} \int_t^{t+\varepsilon} |g(r, y, \sigma(x)p) - g(t, y, \sigma(x)p)|^2 dr \leq C \left( 1 + \sup_{0 \leq s \leq T} |g(s, 0, 0)|^2 \right) \in \mathbb{L}^1(\Omega).$$

Then it follows from the dominated convergence theorem that  $T_3$  converges in  $\mathbb{L}^2(\Omega)$  to 0 which concludes the proof of the lemma.  $\square$

Now we establish a converse comparison theorem for BDSDEs. For this, let  $(Y_t^i(\xi), Z_t^i(\xi))$ ;  $i = 1, 2$  be solution of the following BDSDE

$$Y_t^i(\xi) = \xi + \int_t^T g_i(s, Y_s^i(\xi), Z_s^i(\xi)) ds + \int_t^T h(s, Y_s^i(\xi), Z_s^i(\xi)) d\bar{B}_s - \int_t^T Z_s^i(\xi) dW_s. \tag{3.3}$$

Then we have the next theorem.



**Theorem 3.2.** *Suppose that  $h$  and  $g_i$ ,  $i = 1, 2$  verify the assumptions (A.2), (A.3) and (A.4). If  $\forall \xi \in \mathbb{L}^2(\mathcal{F}_T) \forall t \in [0, T], Y_t^1(\xi) \leq Y_t^2(\xi)$ , then*

$$\mathbb{P} - a.s. \forall t \in [0, T] \forall (y, z) \in \mathbb{R} \times \mathbb{R}^d, g_1(t, y, z) \leq g_2(t, y, z).$$

*Proof.* For any fixed  $(t, y, z)$ , we have for a subsequence,  $\mathbb{P}$ -a.s.

$$\begin{aligned} g_1(t, y, z) &= \lim_{\varepsilon \rightarrow 0^+} \frac{1}{\varepsilon} \{y + Y_t^1(y + z(W_{t+\varepsilon} - W_t))\} \\ &\leq \lim_{\varepsilon \rightarrow 0^+} \frac{1}{\varepsilon} \{y + Y_t^2(z(W_{t+\varepsilon} - W_t))\} = g_2(t, y, z). \end{aligned}$$

By continuity we get the desired result.  $\square$

#### 4. General Converse Comparison Theorem for BDSDEs

Now we give our second main result, which is a general converse theorem for BDSDEs. Roughly speaking, if we can compare the solutions of two BDSDEs with data  $(\xi, g_j, h)$ ,  $j = 1, 2$  at time  $t = 0$  with the same coefficient  $h$  and the same terminal condition  $\xi$ , for all terminal conditions, can we compare the generators  $g_j$ ,  $j = 1, 2$ ?

**Theorem 4.1.** *Suppose that  $h$  and  $g_j$ ,  $j = 1, 2$  verify the assumptions (A.2), (A.3) and (A.4). Then the following two conditions are equivalent*

- (i)  $\forall \xi \in \mathbb{L}^2(\mathcal{F}_T), \mathcal{E}_{g_1}(\xi) \geq \mathcal{E}_{g_2}(\xi)$
- (ii)  $\mathbb{P}$ -a.s.  $\forall t \in [0, T] \forall (y, z) \in \mathbb{R} \times \mathbb{R}^d, g_1(t, y, z) \geq g_2(t, y, z)$ .

*Proof.* By the Comparison Theorem (see [8]), it is obvious that (ii)  $\Rightarrow$  (i). We need to prove that (i)  $\Rightarrow$  (ii). For each  $\delta > 0$  and  $(y, z) \in \mathbb{R} \times \mathbb{R}^d$ , let

$$\tau_\delta = \tau_\delta(y, z) \triangleq \inf\{t \geq 0; g_1(t, y, z) \leq g_2(t, y, z) - \delta\} \wedge T.$$

If (ii) does not hold, then there exists  $\delta > 0$ ,  $(y, z) \in \mathbb{R} \times \mathbb{R}^d$  such that  $\mathbb{P}(\tau_\delta < T) > 0$ . For such a triplet  $(\delta, y, z)$ , we consider the following SDE defined over the interval  $[\tau_\delta, T]$

$$Y_t^j = y - \int_{\tau_\delta}^t g_j(s, Y_s^j, z) ds - \int_{\tau_\delta}^t h(s, Y_s^j, z) d\bar{B}_s + \int_{\tau_\delta}^t z dW_s.$$

It is well known that the above equation (for  $j = 1, 2$ ) admits a unique solution  $Y^j$ . On the other hand, let define a new stopping time

$$\nu_\delta = \nu_\delta(y, z) \triangleq \inf\{t \geq \tau_\delta; g_1(t, Y_t^1, z) \geq g_2(t, Y_t^2, z) - \frac{\delta}{2}\} \wedge T.$$

Thanks to the assumption (A.4), it follows that  $\{\tau_\delta < \nu_\delta\} = \{\tau_\delta < T\}$  and so  $\mathbb{P}(\tau_\delta < \nu_\delta) > 0$ . Using Tanaka's formula we can write

$$\begin{aligned} (Y_{\nu_\delta}^1 - Y_{\nu_\delta}^2)^+ &= - \int_{\tau_\delta}^{\nu_\delta} \mathbb{I}_{(Y_s^1 - Y_s^2)} \{g_1(s, Y_s^1, z) - g_2(s, Y_s^2, z)\} ds \\ &\quad - \int_{\tau_\delta}^{\nu_\delta} \mathbb{I}_{(Y_s^1 - Y_s^2)} \{h(s, Y_s^1, z) - h(s, Y_s^2, z)\} d\bar{B}_s + L_{\nu_\delta} - L_{\tau_\delta}, \end{aligned}$$

where  $L$  is the local time associated with the  $Y^1 - Y^2$ . Taking expectation, we conclude that

$$\begin{aligned} \mathbb{E} (Y_{\nu_\delta}^1 - Y_{\nu_\delta}^2)^+ &\geq -\mathbb{E} \int_{\tau_\delta}^{\nu_\delta} \mathbb{I}_{(Y_s^1 - Y_s^2)} \{g_1(s, Y_s^1, z) - g_2(s, Y_s^2, z)\} ds \\ &\geq \frac{\delta}{2} \mathbb{E}(\nu_\delta - \tau_\delta) > 0 \end{aligned}$$

and consequently  $Y_{\nu_\delta}^1 > Y_{\nu_\delta}^2$  on  $\{\tau_\delta < \nu_\delta\}$ . Now, since  $(Y^j, z)$ ,  $j = 1, 2$  are solutions of BDSDEs (1.1) with data  $(Y_T^j, g_j, h)$ , then we obtain from Proposition 2.2 (4) that

$$\mathcal{E}_{g_1}[Y_{\nu_\delta}^1 | \mathcal{F}_{\tau_\delta}] = \mathcal{E}_{g_1}[Y_T^1 | \mathcal{F}_{\tau_\delta}] = y = \mathcal{E}_{g_2}[Y_T^2 | \mathcal{F}_{\tau_\delta}] = \mathcal{E}_{g_2}[Y_{\nu_\delta}^2 | \mathcal{F}_{\tau_\delta}]$$

and then  $\mathcal{E}_{g_1}(Y_{\nu_\delta}^1) = \mathcal{E}_{g_2}(Y_{\nu_\delta}^2)$ . Moreover, from Proposition 2.2 (3) and the assumption (i) it follows that  $\mathcal{E}_{g_1}(Y_{\nu_\delta}^1) > \mathcal{E}_{g_1}(Y_{\nu_\delta}^2) \geq \mathcal{E}_{g_2}(Y_{\nu_\delta}^2)$ . Which is a contradiction.  $\square$

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