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ON JENSEN'S INEQUALITY FOR g-EXPECTATION***

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Abstract

Briand et al. gave a counterexample showing that given g, Jensen's inequality for g-expectation usually does not hold in general. This paper proves that Jensen's inequality for g-expectation holds in general if and only if the generator g(t, z) is super-homogeneous in z. In particular, g is not necessarily convex in z.

 Keywords Backward stochastic differential equation, Jensen's inequality, gexpectation, Conditional g-expectation, Comparison theorem
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§1. Introduction

It is by now well known that there exists a unique adapted and square integrable solution to a backward stochastic differential equation (BSDE in short) of type

$$y_t = \xi + \int_t^T g(s, y_s, z_s) ds - \int_t^T z_s dB_s, \qquad 0 \le t \le T,$$
 (1.1)

providing that the generator g is Lipschitz in both variables y and z, and that ξ and the process $g(\cdot, 0, 0)$ are square integrable. We denote the unique solution of the BSDE (1.1) by $(y^{\xi}(t), z^{\xi}(t))_{t \in [0,T]}$.

In [1], $y^{\xi}(0)$, denoted by $\mathcal{E}_{g}[\xi]$, is called *g*-expectation of ξ . The notion of *g*-expectation can be considered as a nonlinear extension of the well-known Girsanov transformations. The original motivation for studying *g*-expectation comes from the theory of expected utility, which is the foundation of modern mathematical economics. Z. Chen and L. Epstein [2] gave an application of *g*-expectation to recursive utility. Since the notion of *g*-expectation was introduced, many properties of *g*-expectation have been studied in [1, 3–5]. Some properties of classical expectation are preserved (monotonicity for instance), and some results on Jensen's inequality for *g*-expectation were obtained in [3, 5]. But also in [3], the authors gave a counterexample to indicate that even for a linear function φ , which is obviously convex, Jensen's inequality for *g*-expectation usually does not hold. This yields a natural question:

What kind of generator g can make Jensen's inequality for g-expectation hold in general? Roughly speaking, for convex function $\varphi : \mathbf{R} \to \mathbf{R}$, what conditions should be given

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to the generator g such that the following inequality

$$\mathcal{E}_g[\varphi(\xi)|\mathcal{F}_t] \ge \varphi[\mathcal{E}_g(\xi|\mathcal{F}_t)]$$

will hold in general?

The objective of this paper is to investigate this problem and to prove that Jensen's inequality for g-expectation holds in general if and only if g(t, z) is super-homogeneous, and if g is convex, then Jensen's inequality for g-expectation holds in general if and only if g(t, z) is a positive-homogeneous generator; For monotonic convex function φ , we also get two necessary and sufficient conditions.

§2. Preliminaries

2.1. Notations and Assumptions

Let (Ω, \mathcal{F}, P) be a probability space and $(B_t)_{t\geq 0}$ be a *d*-dimensional standard Brownian motion on this space such that $B_0 = 0$. Let $(\mathcal{F}_t)_{t\geq 0}$ be the filtration generated by this Brownian motion

$$\mathcal{F}_t = \sigma\{B_s, s \in [0, t]\} \lor \mathcal{N}, \qquad t \in [0, T],$$

where \mathcal{N} is the set of all *P*-null subsets.

Let T > 0 be a given real number. In this paper, we always work in the space $(\Omega, \mathcal{F}_T, P)$, and only consider processes indexed by $t \in [0, T]$. For any positive integer n and $z \in \mathbb{R}^n$, |z| denotes its Euclidean norm.

We define the following usual spaces of processes:

$$\mathcal{S}_{\mathcal{F}}^{2}(0,T;\mathbf{R}) := \left\{ \psi \text{ continuous and progressively measurable; } \mathbf{E} \left[\sup_{0 \le t \le T} |\psi_{t}|^{2} \right] < \infty \right\};$$
$$\mathcal{H}_{\mathcal{F}}^{2}(0,T;\mathbf{R}^{n}) := \left\{ \psi \text{ progressively measurable; } \|\psi\|_{2}^{2} = \mathbf{E} \left[\int_{0}^{T} |\psi_{t}|^{2} dt \right] < \infty \right\}.$$

We recall the notion of g-expectation, defined in [1]. We are given a function

$$g: \Omega \times [0,T] \times \mathbf{R} \times \mathbf{R}^d \longrightarrow \mathbf{R}$$

such that the process $(g(t, y, z))_{t \in [0,T]}$ is progressively measurable for each pair (y, z) in $\mathbf{R} \times \mathbf{R}^d$, and furthermore, q satisfies some of the following assumptions:

(A1) There exists a constant $K \geq 0$, such that P -a.s., we have

$$\begin{aligned} \forall t \in [0,T], \, \forall y_1, y_2 \in \mathbf{R}, \, z_1, z_2 \in \mathbf{R}^d, \\ |g(t,y_1,z_1) - g(t,y_2,z_2)| &\leq K(|y_1 - y_2| + |z_1 - z_2|). \end{aligned}$$

(A2) The process $(g(t,0,0))_{t\in[0,T]} \in \mathcal{H}^2_{\mathcal{F}}(0,T;\mathbf{R}).$

(A3) *P*-a.s.,
$$\forall (t, y) \in [0, T] \times \mathbf{R}, g(t, y, 0) \equiv 0.$$

(A4) *P*-a.s., $\forall (y, z) \in \mathbf{R} \times \mathbf{R}^d, t \to g(t, y, z)$ is continuous.

Remark 2.1. The assumption (A3) implies the assumption (A2).

Let g satisfy the assumptions (A1) and (A2). Then for each $\xi \in L^2(\Omega, \mathcal{F}_T, P)$, there exists a unique pair $(y^{\xi}(t), z^{\xi}(t))_{t \in [0,T]}$ of adapted processes in $\mathcal{S}^2_{\mathcal{F}}(0,T;\mathbf{R}) \times \mathcal{H}^2_{\mathcal{F}}(0,T;\mathbf{R}^d)$ solving the BSDE (1.1) (see [6]). We often denote $(y^{\xi}(t), z^{\xi}(t))_{t \in [0,T]}$ by $(y_t, z_t)_{t \in [0,T]}$ in short.

2.2. Definitions and Propositions

For the convenience of readers, we recall the notion of g-expectation and conditional g-expectation defined in [1]. We also list some basic properties of BSDEs and g-expectation. In the following Definitions 2.1 and 2.2, we always assume that g satisfies (A1) and (A3).

Definition 2.1. The g-expectation $\mathcal{E}_{q}[\cdot]: L^{2}(\Omega, \mathcal{F}_{T}, P) \longrightarrow \mathbf{R}$ is defined by

$$\mathcal{E}_q[\xi] = y^{\xi}(0).$$

Definition 2.2. The conditional g-expectation of ξ with respect to \mathcal{F}_t is defined by

$$\mathcal{E}_g[\xi|\mathcal{F}_t] = y^{\xi}(t).$$

The following Comparison Theorem is one of the great achievements of theory of BS-DEs, readers can see the proof in [7] or [8].

Proposition 2.1. (cf. [7,8]) Let g, \overline{g} satisfy (A1) and (A2), let Y_T , $\overline{Y}_T \in L^2(\Omega, \mathcal{F}_T, P)$. Let $(y(t), z(t))_{t \in [0,T]}$, $(\overline{y}(t), \overline{z}(t))_{t \in [0,T]}$ be the solutions of the following two BSDEs:

$$y_t = Y_T + \int_t^T g(s, y_s, z_s) ds - \int_t^T z_s dB_s, \qquad 0 \le t \le T;$$

$$\bar{y}_t = \overline{Y}_T + \int_t^T \bar{g}(s, \bar{y}_s, \bar{z}_s) ds - \int_t^T \bar{z}_s dB_s, \qquad 0 \le t \le T.$$

(1) If $Y_T \geq \overline{Y}_T$, $g(t, \bar{y}_t, \bar{z}_t) \geq \bar{g}(t, \bar{y}_t, \bar{z}_t)$, a.s., a.e., then we have

$$y_t \ge \bar{y}_t, \qquad a.e., a.s$$

(2) In addition, if we also assume that $P(Y_T - \overline{Y}_T > 0) > 0$, then

$$P(y_t - \bar{y}_t > 0) > 0$$
, in particular, $y_0 > \bar{y}_0$.

Propositions 2.2–2.5 come from [1], where g is assumed to satisfy (A1) and (A3).

Proposition 2.2. (1) (Preserving of constants) For each constant c, $\mathcal{E}_g[c] = c$; (2) (Monotonicity) If $X_1 \ge X_2$, a.s., then $\mathcal{E}_g[X_1] \ge \mathcal{E}_g[X_2]$;

(3) (Strict Monotonicity) If $X_1 \ge X_2$, a.s., and $P(X_1 > X_2) > 0$, then $\mathcal{E}_g[X_1] > \mathcal{E}_g[X_2]$.

Proposition 2.3. (1) If X is \mathcal{F}_t -measurable, then $\mathcal{E}_g[X|\mathcal{F}_t] = X$; (2) For all $t, s \in [0, T]$, $\mathcal{E}_g[\mathcal{E}_g[X|\mathcal{F}_t]|\mathcal{F}_s] = \mathcal{E}_g[X|\mathcal{F}_{t\wedge s}]$.

Proposition 2.4. $\mathcal{E}_g[X|\mathcal{F}_t]$ is the unique random variable η in $L^2(\Omega, \mathcal{F}_t, P)$, such that

$$\mathcal{E}_q[X1_A] = \mathcal{E}_q[\eta 1_A] \quad for \ all \ A \in \mathcal{F}_t.$$

Proposition 2.5. Let $g(\omega, t, y, z) : \Omega \times [0, T] \times \mathbf{R} \times \mathbf{R}^d \longrightarrow \mathbf{R}$ be a given function satisfying (A1) and (A3). If g does not depend on y, then we have

$$\mathcal{E}_g[X+\eta|\mathcal{F}_t] = \mathcal{E}_g[X|\mathcal{F}_t] + \eta, \qquad \forall \eta \in L^2(\Omega, \mathcal{F}_t, P), \quad \forall X \in L^2(\Omega, \mathcal{F}_T, P).$$

Proposition 2.6. (cf. [3, 8]) Let $\xi \in L^2(\Omega, \mathcal{F}_T, P)$, and let the assumptions (A1) and (A2) hold. If the process $(y_t, z_t)_{t \in [0,T]}$ is the solution of BSDE (1.1), then we have

$$\begin{split} & \mathbf{E}\Big[\sup_{t\leq s\leq T}(e^{\beta s}|y_s|^2) + \int_t^T e^{\beta s}|z_s|^2 ds|\mathcal{F}_t\Big] \\ & \leq C\mathbf{E}\Big[e^{\beta T}|\xi|^2 + \Big(\int_t^T e^{(\beta/2)s}|g(s,0,0)|ds\Big)^2\Big|\mathcal{F}_t\Big], \end{split}$$

where $\beta = 2(K + K^2)$ and C is a universal constant.

Proposition 2.7. (cf. [3]) Suppose g does not depend on y and g satisfies (A1) and (A3). Suppose moreover that for each $t \in [0,T]$, P-a.s., $z \longrightarrow g(t,z)$ is convex. Given $\xi \in L^2(\Omega, \mathcal{F}_T, P)$, let $\varphi : \mathbf{R} \longrightarrow \mathbf{R}$ be a convex function such that $\varphi(\xi) \in L^2(\Omega, \mathcal{F}_T, P)$. If P-a.s., $\partial \varphi[\mathcal{E}_g(\xi|\mathcal{F}_t)] \cap [0,1[^c \neq \emptyset, then we have$

$$P\text{-}a.s., \quad \varphi[\mathcal{E}_g(\xi|\mathcal{F}_t)] \le \mathcal{E}_g[\varphi(\xi|\mathcal{F}_t)].$$

Proposition 2.7 can be regarded as an important result on Jensen's inequality for gexpectation, but if $\partial \varphi[\mathcal{E}_g(\xi|\mathcal{F}_t)] \cap]0, 1[^c = \emptyset$, for example $\varphi(x) = x/2, \forall x \in \mathbf{R}$, Proposition 2.7 can not solve this kind of problems. It also can not tell us what kind of generator g can make Jensen's inequality hold in general.

§3. Jensen's Inequality for Super-homogeneous Generator g

In the following, we always consider the situation where the generator g does not depend on y, that is, $g: \Omega \times [0,T] \times \mathbf{R}^d \to \mathbf{R}$. We denote this kind of generator g by g(t,z). We always assume that g(t,z) satisfies (A1) and (A3).

Definition 3.1. Let g satisfy (A1) and (A3). We say that g is a super-homogeneous generator in z if g also satisfies

P-a.s.,
$$\forall (t,z) \in [0,T] \times \mathbf{R}^d$$
, $\lambda \in \mathbf{R}$: $g(t,\lambda z) \ge \lambda g(t,z)$.

Now we introduce our main results on Jensen's inequality for g-expectation.

Theorem 3.1. Let g satisfy (A1), (A3) and (A4). Then the following two conditions are equivalent:

(i) q is a super-homogeneous generator;

(ii) Jensen's inequality for g-expectation holds in general, i.e., for each $\xi \in L^2(\Omega, \mathcal{F}_T, P)$ and convex function $\varphi : \mathbf{R} \to \mathbf{R}$, if $\varphi(\xi) \in L^2(\Omega, \mathcal{F}_T, P)$, then for each $t \in [0, T]$, P-a.s.,

$$\mathcal{E}_g[\varphi(\xi)|\mathcal{F}_t] \ge \varphi[\mathcal{E}_g(\xi|\mathcal{F}_t)].$$

Proof. (i) \Rightarrow (ii). Given $\xi \in L^2(\Omega, \mathcal{F}_T, P)$ and convex function φ such that $\varphi(\xi) \in L^2(\Omega, \mathcal{F}_T, P)$, for each $t \in [0, T]$, we set $\eta_t = \varphi'_-[\mathcal{E}_g(\xi|\mathcal{F}_t)]$. Then η_t is \mathcal{F}_t -measurable. Since φ is convex, we have

$$\varphi(x) - \varphi(y) \ge \varphi'_{-}(y)(x - y), \quad \forall x, y \in R.$$

Take $x = \xi$, $y = \mathcal{E}_g(\xi | \mathcal{F}_t)$. Then we have

$$\varphi(\xi) - \varphi[\mathcal{E}_g(\xi|\mathcal{F}_t)] \ge \eta_t[\xi - \mathcal{E}_g(\xi|\mathcal{F}_t)].$$

For each positive integer n, we define

$$\Omega_{t,n} := \{ |\mathcal{E}_g(\xi|\mathcal{F}_t)| + |\eta_t| + |\varphi[\mathcal{E}_g(\xi|\mathcal{F}_t)]| \le n \}$$

Because $\mathcal{E}_{g}[\xi|\mathcal{F}_{t}]$, η_{t} , $\varphi[\mathcal{E}_{g}(\xi|\mathcal{F}_{t})]$ are all \mathcal{F}_{t} -measurable, we see that $\Omega_{t,n} \in \mathcal{F}_{t}$. We denote the indicator function of $\Omega_{t,n}$ by $\mathbf{1}_{\Omega_{t,n}}$. Set $\eta_{t,n} = \mathbf{1}_{\Omega_{t,n}}\eta_{t}$. Then we have

$$\mathbf{1}_{\Omega_{t,n}}[\varphi(\xi) - \varphi[\mathcal{E}_g(\xi|\mathcal{F}_t)]] \ge \eta_{t,n}[\xi - \mathcal{E}_g(\xi|\mathcal{F}_t)].$$
(3.1)

Since $\eta_{t,n}$, $\mathbf{1}_{\Omega_{t,n}}\varphi[\mathcal{E}_g(\xi|\mathcal{F}_t)]$ are bounded by n and $\xi, \varphi(\xi) \in L^2(\Omega, \mathcal{F}_T, P)$, we deduce that

$$\begin{aligned} \mathbf{1}_{\Omega_{t,n}}\varphi(\xi), \ \eta_{t,n}\xi \in L^2(\Omega,\mathcal{F}_T,P), \\ \mathbf{1}_{\Omega_{t,n}}\varphi[\mathcal{E}_g(\xi|\mathcal{F}_t)] \in L^2(\Omega,\mathcal{F}_t,P), \\ (\eta_{t,n}\mathcal{E}_g(\xi|\mathcal{F}_s))_{t\leq s\leq T} \in S^2_{\mathcal{F}}(t,T;R). \end{aligned}$$

From the well-known Comparison Theorem we know that conditional g-expectation $\mathcal{E}_{g}[\cdot | \mathcal{F}_{t}]$ is nondecreasing. Thus from the inequality (3.1), and by taking conditional g-expectation, we can get

$$\mathcal{E}_{g}[\mathbf{1}_{\Omega_{t,n}}[\varphi(\xi) - \varphi(\mathcal{E}_{g}(\xi|\mathcal{F}_{t}))]|\mathcal{F}_{t}] \geq \mathcal{E}_{g}[\eta_{t,n}[\xi - \mathcal{E}_{g}(\xi|\mathcal{F}_{t})]|\mathcal{F}_{t}].$$

Since $\mathbf{1}_{\Omega_{t,n}}\varphi[\mathcal{E}_g(\xi|\mathcal{F}_t)], \ \eta_{t,n}\mathcal{E}_g[\xi|\mathcal{F}_t] \in L^2(\Omega,\mathcal{F}_t,P)$, it follows from Proposition 2.5 that

$$\mathcal{E}_{g}[\mathbf{1}_{\Omega_{t,n}}\varphi(\xi)|\mathcal{F}_{t}] - \mathbf{1}_{\Omega_{t,n}}\varphi[\mathcal{E}_{g}(\xi|\mathcal{F}_{t})] \ge \mathcal{E}_{g}[\eta_{t,n}\xi|\mathcal{F}_{t}] - \eta_{t,n}\mathcal{E}_{g}[\xi|\mathcal{F}_{t}].$$
(3.2)

Let $(y_u, z_u)_{u \in [0,T]}$ be the solution of the following BSDE (3.3)

$$y_u = \xi + \int_u^T g(s, z_s) ds - \int_u^T z_s dB_s, \qquad 0 \le u \le T.$$
 (3.3)

Then for the given $t \in [0, T]$, we have

$$\eta_{t,n} y_u = \eta_{t,n} \xi + \int_u^T \eta_{t,n} g(s, z_s) ds - \int_u^T \eta_{t,n} z_s dB_s, \qquad t \le u \le T.$$
(3.4)

We define function $g_1(s, z)$ in this way: for each $(s, z) \in [t, T] \times \mathbf{R}^d$,

$$g_1(s,z) := \begin{cases} \eta_{t,n}g(s,z/\eta_{t,n}), & \text{ if } \eta_{t,n} \neq 0; \\ 0, & \text{ if } \eta_{t,n} = 0. \end{cases}$$

Since $\eta_{t,n}$ is bounded, the following BSDE

$$\bar{y}_u = \eta_{t,n}\xi + \int_u^T g_1(s,\bar{z}_s)ds - \int_u^T \bar{z}_s dB_s, \qquad t \le u \le T$$
 (3.5)

has a unique solution in $S^2_{\mathcal{F}}(t, T; \mathbf{R}) \times \mathcal{H}^2_{\mathcal{F}}(t, T; \mathbf{R}^d)$. We denote it by $(\bar{y}_s, \bar{z}_s)_{s \in [t,T]}$. Also from that $\eta_{t,n}$ is bounded we know that $(\eta_{t,n}y_s, \eta_{t,n}z_s)_{s \in [t,T]}$ is in $S^2_{\mathcal{F}}(t, T; \mathbf{R}) \times \mathcal{H}^2_{\mathcal{F}}(t, T; \mathbf{R}^d)$. From (3.4) and the definition of g_1 , we conclude that the solution of BSDE (3.5) is just $(\eta_{t,n}y_s, \eta_{t,n}z_s)_{s \in [t,T]}$.

Consider the solutions of BSDE (3.5) and the following BSDE (3.6):

$$\tilde{y}_u = \eta_{t,n}\xi + \int_u^T g(s, \tilde{z}_s)ds - \int_u^T \tilde{z}_s dB_s, \qquad t \le u \le T.$$
(3.6)

Due to the super-homogeneity of g(t, z) in z, we can get that for each $s \in [t, T]$, P-a.s.,

 $g(s,\eta_{t,n}z_s) \ge \eta_{t,n}g(s,z_s).$

Combining this with the definition of g_1 , we have, *P*-a.s.,

$$\forall s \in [t,T], \quad g(s,\bar{z}_s) = g(s,\eta_{t,n}z_s) \ge \eta_{t,n}g(s,z_s) = g_1(s,\eta_{t,n}z_s) = g_1(s,\bar{z}_s).$$

Thus from Comparison Theorem, we have, P-a.s.,

$$\mathcal{E}_{g}[\eta_{t,n}\xi|\mathcal{F}_{t}] = \tilde{y}_{t} \ge \bar{y}_{t} = \eta_{t,n}y_{t} = \eta_{t,n}\mathcal{E}_{g}[\xi|\mathcal{F}_{t}].$$
(3.7)

Coming back to (3.2), we can get

$$\mathcal{E}_{g}[\mathbf{1}_{\Omega_{t,n}}\varphi(\xi)|\mathcal{F}_{t}] - \mathbf{1}_{\Omega_{t,n}}\varphi[\mathcal{E}_{g}(\xi|\mathcal{F}_{t})] \geq \mathcal{E}_{g}[\eta_{t,n}\xi|\mathcal{F}_{t}] - \eta_{t,n}\mathcal{E}_{g}[\xi|\mathcal{F}_{t}] \geq 0.$$

Applying Lebesgue's dominated convergence theorem to $(\mathbf{1}_{\Omega_{t,n}}\varphi(\xi))_{n=1}^{\infty}$, we can get easily that

$$L^2 - \lim_{n \to \infty} \mathbf{1}_{\Omega_{t,n}} \varphi(\xi) = \varphi(\xi).$$

Since that $\xi \to \mathcal{E}_g(\xi|\mathcal{F}_t)$ is a continuous map from $L^2(\mathcal{F}_T)$ into $L^2(\mathcal{F}_t)$ (see [1, Lemma 36.9]), it follows that

$$L^{2} - \lim_{n \to \infty} \mathcal{E}_{g}[\mathbf{1}_{\Omega_{t,n}}\varphi(\xi)|\mathcal{F}_{t}] = \mathcal{E}_{g}[\varphi(\xi)|\mathcal{F}_{t}].$$

Thus for the given $t \in [0,T]$, there exists a subsequence $(\mathcal{E}_g[\varphi(\xi)\mathbf{1}_{\Omega_{t,n_i}}|\mathcal{F}_t])_{i=1}^{\infty}$ such that, *P*-a.s.,

$$\lim_{i \to \infty} \mathcal{E}_g[\varphi(\xi) \mathbf{1}_{\Omega_{t,n_i}} | \mathcal{F}_t] = \mathcal{E}_g[\varphi(\xi) | \mathcal{F}_t].$$

On the other hand, by the definition of $\Omega_{t,n}$, we can get, *P*-a.s.,

$$\lim_{n \to \infty} \mathbf{1}_{\Omega_{t,n}} \varphi[\mathcal{E}_g(\xi | \mathcal{F}_t)] = \varphi[\mathcal{E}_g(\xi | \mathcal{F}_t)].$$

Hence we can assert that (i) implies (ii). Indeed, P-a.s.,

$$\mathcal{E}_{g}[\varphi(\xi)|\mathcal{F}_{t}] = \lim_{i \to \infty} \mathcal{E}_{g}[\mathbf{1}_{\Omega_{t,n_{i}}}\varphi(\xi)|\mathcal{F}_{t}] \geq \lim_{i \to \infty} \mathbf{1}_{\Omega_{t,n_{i}}}\varphi[\mathcal{E}_{g}(\xi|\mathcal{F}_{t})] = \varphi[\mathcal{E}_{g}(\xi|\mathcal{F}_{t})].$$

(ii) \Rightarrow (i). Firstly we show that for each $z \in \mathbf{R}^d$, $t \in [0, T[$,

$$L^{2} - \lim_{n \to \infty} n[\mathcal{E}_{g}(z \cdot (B_{t+1/n} - B_{t})|\mathcal{F}_{t})] = g(t, z).$$
(3.8)

(3.8) is a special case of [3, Proposition 2.3]. But for the convenience of readers and the completeness of our proof, here we give a straightforward proof. For each given $z \in \mathbf{R}^d$, $t \in [0, T[$, we choose a large enough positive integer n, such that $t + 1/n \leq T$. We denote by $(y_{s,n}, z_{s,n})_{s \in [t,t+1/n]}$ the solution of the following BSDE:

$$y_s = z \cdot (B_{t+1/n} - B_t) + \int_s^{t+1/n} g(u, z_u) du - \int_s^{t+1/n} z_u dB_u, \qquad t \le s \le t + 1/n.$$
(3.9)

We set

$$\bar{y}_{s,n} = y_{s,n} - z \cdot (B_s - B_t), \quad \bar{z}_{s,n} = z_{s,n} - z.$$

Then we have $y_{t,n} = \bar{y}_{t,n}$ and

$$\bar{y}_{s,n} = \int_{s}^{t+1/n} g(u, \bar{z}_{u,n} + z) du - \int_{s}^{t+1/n} \bar{z}_{u,n} dB_u, \qquad t \le s \le t+1/n.$$
(3.10)

Since

$$\mathcal{E}_g[z \cdot (B_{t+1/n} - B_t)|\mathcal{F}_t] = y_{t,n} = \bar{y}_{t,n} = \mathbf{E}\Big[\int_t^{t+1/n} g(s, \bar{z}_{s,n} + z)ds|\mathcal{F}_t\Big],$$

by the classical Jensen's inequality and Hölder's inequality, we have

$$\begin{split} \mathbf{E}[n\mathcal{E}_{g}[z \cdot (B_{t+\frac{1}{n}} - B_{t})|\mathcal{F}_{t}] - g(t, z)]^{2} \\ &= \mathbf{E}\Big[n\mathbf{E}\Big[\int_{t}^{t+\frac{1}{n}} (g(s, \bar{z}_{s,n} + z) - g(t, z))ds|\mathcal{F}_{t}\Big]\Big]^{2} \\ &\leq n^{2}\mathbf{E}\Big[\int_{t}^{t+1/n} (g(s, \bar{z}_{s,n} + z) - g(t, z))ds\Big]^{2} \\ &\leq n\mathbf{E}\int_{t}^{t+1/n} |g(s, \bar{z}_{s,n} + z) - g(t, z)|^{2}ds \\ &\leq 2n\mathbf{E}\int_{t}^{t+1/n} |g(s, \bar{z}_{s,n} + z) - g(s, z)|^{2}ds \\ &+ 2n\mathbf{E}\int_{t}^{t+1/n} |g(s, z) - g(t, z)|^{2}ds. \end{split}$$
(3.11)

By (A1), Proposition 2.6 and (A3), we know that there exists a universal constant ${\cal C}$ such that

$$\begin{split} &2n\mathbf{E}\int_{t}^{t+1/n}|g(s,\bar{z}_{s,n}+z)-g(s,z)|^{2}ds\\ &\leq 2nK^{2}\mathbf{E}\int_{t}^{t+1/n}|\bar{z}_{s,n}|^{2}ds\\ &\leq 2nK^{2}C\mathbf{E}\Big(\int_{t}^{t+1/n}|g(s,z)|ds\Big)^{2}\\ &\leq 2nK^{2}C\mathbf{E}\Big(\int_{t}^{t+1/n}K|z|ds\Big)^{2}\\ &= 2K^{4}C|z|^{2}/n, \end{split}$$

where K is the Lipschitz constant.

By (A4), we know that

P-a.s.,
$$\lim_{n \to \infty} 2n \int_{t}^{t+1/n} |g(s,z) - g(t,z)|^2 ds = 0.$$

In view of (A3) and (A1), we have

$$2n\int_{t}^{t+1/n}|g(s,z)-g(t,z)|^{2}ds \leq 2n\int_{t}^{t+1/n}(2K|z|)^{2}ds = 8K^{2}|z|^{2}.$$

It follows from Lebesgue's dominated convergence theorem that

$$\lim_{n \to \infty} 2n \mathbf{E} \int_{t}^{t+1/n} |g(s,z) - g(t,z)|^2 ds = 0.$$

Then coming back to (3.11), we can get

$$\lim_{n \to \infty} \mathbf{E} [n \mathcal{E}_g (z \cdot (B_{t+1/n} - B_t) | \mathcal{F}_t) - g(t, z)]^2$$

$$\leq \lim_{n \to \infty} 2K^4 C |z|^2 / n + \lim_{n \to \infty} 2n \mathbf{E} \int_t^{t+1/n} |g(s, z) - g(t, z)|^2 ds = 0.$$

Therefore we have

$$L^{2} - \lim_{n \to \infty} n[\mathcal{E}_{g}(z \cdot (B_{t+1/n} - B_{t})|\mathcal{F}_{t})] = g(t, z)$$

Secondly we prove that for each triple $(t, z, \lambda) \in [0, T] \times \mathbf{R}^d \times \mathbf{R}$, we have

$$P\text{-a.s.}, \quad g(t, \lambda z) \ge \lambda g(t, z). \tag{3.12}$$

Given $\lambda \in \mathbf{R}$, we define a corresponding convex function $\varphi_{\lambda} : \mathbf{R} \to \mathbf{R}$, such that $\varphi_{\lambda}(x) = \lambda x, \forall x \in \mathbf{R}$. Given $t \in [0, T[$, let us pick a large enough positive integer n, such that $t + 1/n \leq T$. Then for each $z \in \mathbf{R}^d$, it is obvious that $\varphi_{\lambda}(z \cdot (B_{t+1/n} - B_t)) \in L^2(\Omega, \mathcal{F}_T, P)$. By (ii), we know that, P-a.s.,

$$\mathcal{E}_g[\varphi_\lambda(z \cdot (B_{t+1/n} - B_t))|\mathcal{F}_t] \ge \varphi_\lambda[\mathcal{E}_g(z \cdot (B_{t+1/n} - B_t)|\mathcal{F}_t)]$$

that is, *P*-a.s.,

$$\mathcal{E}_g[\lambda z \cdot (B_{t+1/n} - B_t) | \mathcal{F}_t] \ge \lambda [\mathcal{E}_g(z \cdot (B_{t+1/n} - B_t) | \mathcal{F}_t)].$$
(3.13)

Because of (3.8), we know there exists a subsequence $\{n_k\}_{k=1}^{\infty}$ such that

$$\begin{aligned} P\text{-a.s.,} \quad &\lim_{k \to \infty} n_k [\mathcal{E}_g(\lambda z \cdot (B_{t+1/n_k} - B_t) | \mathcal{F}_t)] = g(t, \lambda z), \\ P\text{-a.s.,} \quad &\lim_{k \to \infty} \lambda n_k [\mathcal{E}_g(z \cdot (B_{t+1/n_k} - B_t) | \mathcal{F}_t)] = \lambda g(t, z). \end{aligned}$$

Thus for the given $t \in [0, T[, z \in \mathbf{R}^d, \lambda \in \mathbf{R}, \text{ by (3.13)}, \text{ we have}$

$$P\text{-a.s.}, \quad g(t, \lambda z) \ge \lambda g(t, z).$$

By (A4), we know that for each z, the process $t \longrightarrow g(t, z)$ is continuous. Hence we have

$$P\text{-a.s.}, \quad g(T,\lambda z) = \lim_{\varepsilon \to 0^+} g(T-\varepsilon,\lambda z) \geq \lim_{\varepsilon \to 0^+} \lambda g(T-\varepsilon,z) = \lambda g(T,z).$$

Therefore we can get (3.12) immediately. The proof is complete.

Remark 3.1. When we prove that (i) implies (ii), we do not need (A4).

Example 3.1. Let $g : \mathbf{R} \to \mathbf{R}$ be defined as follows: $g(z) = z^4$, if $|z| \le 1$ and g(z) = 4|z| - 3, if |z| > 1. We can see clearly that though g is convex, g is not super-homogeneous. Thus for this generator g, by Theorem 3.1, we know that Jensen's inequality for g-expectation does not hold in general.

In fact, if we take T = 1, $\xi = B_T - T$ and $\varphi(x) = \frac{x}{3}$, $\forall x \in \mathbf{R}$, then we can verify that $(B_t - t, 1)_{t \in [0,T]}$ is the solution of the following BSDE:

$$y_t = \xi + \int_t^T g(z_s)ds - \int_t^T z_s dB_s, \qquad 0 \le t \le T,$$

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and $(\frac{B_t}{3} - \frac{26T+t}{81}, \frac{1}{3})_{t \in [0,T]}$ is the solution of the following BSDE:

$$\bar{y}_t = \varphi(\xi) + \int_t^T g(\bar{z}_s) ds - \int_t^T \bar{z}_s dB_s, \qquad 0 \le t \le T.$$

We can calculate that

$$\mathcal{E}_g[\varphi(\xi)|\mathcal{F}_t] - \varphi[\mathcal{E}_g(\xi|\mathcal{F}_t)] = \frac{26}{81}(t-T) < 0, \quad \text{when } t < T.$$

Example 3.1 yields a natural question: What kind of convex generator g can make Jensen's inequality for g-expectation hold in general? The following Theorem 3.2 will answer this question.

Definition 3.2. We call a generator g(t, z) is positive-homogeneous in z if

$$P\text{-}a.s., \quad \forall \lambda \ge 0, \ t \in [0,T], \ z \in \mathbf{R}^d, \quad g(t,\lambda z) = \lambda g(t,z).$$

Theorem 3.2. Suppose g satisfies (A1), (A3) and (A4). Suppose moreover that for each $t \in \mathbf{R}$, P-a.s., $z \longrightarrow g(t, z)$ is convex in z. Then the following two conditions are equivalent:

(i) g(t, z) is positive-homogeneous in z;

(ii) Jensen's inequality for g-expectation holds in general.

Proof. By Theorem 3.1, it suffices to prove that if g(t, z) is convex in z and $g(t, 0) \equiv 0$, then g(t, z) is positive-homogeneous in z if and only if g(t, z) is super-homogeneous.

Suppose g(t, z) is positive-homogeneous in z. We only need to consider the case when $\lambda \leq 0$. For each $\lambda \leq 0$, $(t, z) \in [0, T] \times \mathbf{R}^d$, since g is convex and $g(t, 0) \equiv 0$, we have, P-a.s.,

$$0 = g(t,0) = g\left(t, \frac{\lambda z}{2} + \frac{(-\lambda)z}{2}\right) \le \frac{g(t,\lambda z)}{2} + \frac{g(t,-\lambda z)}{2} = \frac{g(t,\lambda z)}{2} + \frac{-\lambda g(t,z)}{2}.$$

Thus we have

$$P\text{-a.s.}, \quad \forall \lambda \le 0, \ (t,z) \in [0,T] \times \mathbf{R}^d, \quad g(t,\lambda z) \ge \lambda g(t,z).$$

Hence g(t, z) is super-homogeneous.

Suppose g(t, z) is super-homogeneous. For each given triple $(t, z, \lambda) \in [0, T] \times \mathbf{R}^d \times \mathbf{R}_+$, if $0 \le \lambda \le 1$, then by the convexity of g and (A3) we have

P-a.s.,
$$g(t, \lambda z) \leq \lambda g(z)$$
.

Thus by the super-homogeneity of g, we have, P-a.s.,

$$\forall \lambda \in [0,1], \ t \in [0,T], \quad g(t,\lambda z) = \lambda g(t,z).$$
(3.14)

For $\lambda > 1$, it follows from (3.14) that *P*-a.s.,

$$\lambda g(t,z) = \lambda g\Big(t,\frac{1}{\lambda}\times(\lambda z)\Big) = \lambda\times\frac{1}{\lambda}\times g(t,\lambda z) = g(t,\lambda z).$$

Thus g(t, z) is positive-homogeneous. This completes the proof.

Corollary 3.1. Given $\mu \ge 0$, let the generator $g(t, z) = \mu |z|, \forall (t, z) \in [0, T] \times \mathbf{R}^d$. Then Jensen's inequality for g-expectation holds in general.

This kind of g-expectation $\mathcal{E}_{g}[\cdot]$ plays a key role in [4].

§4. Jensen's Inequality for Monotonic Convex Function φ

In this section, we will consider the following problem: If g is independent of y, φ is a monotonic convex function, then what conditions should be given to the generator g, such that Jensen's inequality for g-expectation holds for φ ? We will give two necessary and sufficient conditions to solve this problem, one condition is for increasing convex function φ , the other condition is for decreasing convex function φ .

Theorem 4.1. Let g satisfy (A1), (A3) and (A4). Then the following two conditions are equivalent:

(i) *P-a.s.*, $\forall (t, z, \lambda) \in [0, T] \times \mathbf{R}^d \times \mathbf{R}_+, g(t, \lambda z) \ge \lambda g(t, z);$

(ii) Jensen's inequality for g-expectation holds for increasing convex function, i.e., for each $\xi \in L^2(\Omega, \mathcal{F}_T, P)$ and increasing convex function $\varphi : \mathbf{R} \to \mathbf{R}$, if $\varphi(\xi) \in L^2(\Omega, \mathcal{F}_T, P)$, then for each $t \in [0, T]$, P-a.s.,

$$\mathcal{E}_g[\varphi(\xi)|\mathcal{F}_t] \ge \varphi[\mathcal{E}_g(\xi|\mathcal{F}_t)].$$

Proof. (i) \Rightarrow (ii). Given $\xi \in L^2(\Omega, \mathcal{F}_T, P)$ and increasing convex function φ such that $\varphi(\xi) \in L^2(\Omega, \mathcal{F}_T, P)$. For each $t \in [0, T]$ and positive integer n, just as in the proof of Theorem 3.1, we set or define

$$\eta_t = \varphi'_-[\mathcal{E}_g(\xi|\mathcal{F}_t)], \quad \Omega_{t,n} := \{|\mathcal{E}_g[\xi|\mathcal{F}_t]| + |\eta_t| + |\varphi[\mathcal{E}_g(\xi|\mathcal{F}_t)]| \le n\}, \quad \eta_{t,n} = \mathbf{1}_{\Omega_{t,n}}\eta_t.$$

We already know that

$$\begin{split} &\Omega_{t,n} \in \mathcal{F}_t, \eta_{t,n}, \mathbf{1}_{\Omega_{t,n}} \text{ are } \mathcal{F}_t\text{-measurable;} \\ &\eta_{t,n}, \ \mathbf{1}_{\Omega_{t,n}} \varphi[\mathcal{E}_g(\xi|\mathcal{F}_t)] \text{ are bounded by } n; \\ &\mathbf{1}_{\Omega_{t,n}} \varphi(\xi), \ \eta_{t,n} \xi \in L^2(\Omega, \mathcal{F}_T, P), \ \mathbf{1}_{\Omega_{t,n}} \varphi[\mathcal{E}_g(\xi|\mathcal{F}_t)] \in L^2(\Omega, \mathcal{F}_t, P); \\ &(\eta_{t,n} \mathcal{E}_g(\xi|\mathcal{F}_s))_{s \in [t,T]} \in \mathcal{S}_{\mathcal{F}}^2(t,T;R). \end{split}$$

Moreover, we also know that

$$\mathcal{E}_{g}[\mathbf{1}_{\Omega_{t,n}}\varphi(\xi)|\mathcal{F}_{t}] - \mathbf{1}_{\Omega_{t,n}}\varphi[\mathcal{E}_{g}(\xi|\mathcal{F}_{t})] \ge \mathcal{E}_{g}[\eta_{t,n}\xi|\mathcal{F}_{t}] - \eta_{t,n}\mathcal{E}_{g}[\xi|\mathcal{F}_{t}]. \tag{4.1}$$

Let $(y_u, z_u)_{u \in [0,T]}$ be the unique square integrable solution of the following BSDE:

$$y_u = \xi + \int_u^T g(s, z_s) ds - \int_u^T z_s dB_s, \qquad 0 \le u \le T.$$
 (4.2)

Then for the given $t \in [0, T]$, we have

$$\eta_{t,n} y_u = \eta_{t,n} \xi + \int_u^T \eta_{t,n} g(s, z_s) ds - \int_u^T \eta_{t,n} z_s dB_s, \qquad t \le u \le T.$$
(4.3)

For the given t, again we define function $g_1(s, z)$ in this way: for each $(s, z) \in [t, T] \times \mathbf{R}^d$,

$$g_1(s,z) := \begin{cases} \eta_{t,n} g(s, z/\eta_{t,n}), & \text{if } \eta_{t,n} \neq 0; \\ 0, & \text{if } \eta_{t,n} = 0. \end{cases}$$

Consider the solutions of the following BSDE (4.4) and BSDE (4.5):

$$\bar{y}_u = \eta_{t,n}\xi + \int_u^T g_1(s,\bar{z}_s)ds - \int_u^T \bar{z}_s dB_s, \qquad t \le u \le T,$$
(4.4)

$$\tilde{y}_u = \eta_{t,n}\xi + \int_u^T g(s, \tilde{z}_s)ds - \int_u^T \tilde{z}_s dB_s, \qquad t \le u \le T.$$
(4.5)

Analogous to the proof of Theorem 3.1, from (4.3) we deduce that $(\eta_{t,n}y_s, \eta_{t,n}z_s)_{s \in [t,T]}$ is the unique solution of BSDE (4.4).

For the given $t \in [0, T]$ and φ , since φ is increasing, we have

$$\eta_t = \varphi'_{-}[\mathcal{E}_g(\xi|\mathcal{F}_t)] \ge 0, \quad \eta_{t,n} = \mathbf{1}_{\Omega_{t,n}} \eta_t \ge 0.$$

In view of (i), for each $s \in [t, T]$, *P*-a.s., we have

$$g(s,\eta_{t,n}z_s) \ge \eta_{t,n}g(s,z_s). \tag{4.6}$$

Therefore, for each $s \in [t, T]$, we can get, *P*-a.s.,

$$g(s, \bar{z}_s) = g(s, \eta_{t,n} z_s) \ge \eta_{t,n} g(s, z_s) = g_1(s, \eta_{t,n} z_s) = g_1(s, \bar{z}_s).$$

Thus from Comparison Theorem we have

$$P\text{-a.s.}, \quad \mathcal{E}_g[\eta_{t,n}\xi|\mathcal{F}_t] = \tilde{y}_t \ge \bar{y}_t = \eta_{t,n}y_t = \eta_{t,n}\mathcal{E}_g[\xi|\mathcal{F}_t]. \tag{4.7}$$

This with (4.1), it follows that

$$\mathcal{E}_{g}[\mathbf{1}_{\Omega_{t,n}}\varphi(\xi)|\mathcal{F}_{t}] - \mathbf{1}_{\Omega_{t,n}}\varphi[\mathcal{E}_{g}(\xi|\mathcal{F}_{t})] \ge \mathcal{E}_{g}[\eta_{t,n}\xi|\mathcal{F}_{t}] - \eta_{t,n}\mathcal{E}_{g}[\xi|\mathcal{F}_{t}] \ge 0$$

Applying Lebesgue's dominated theorem to $(\mathbf{1}_{\Omega_{t,n}}\varphi(\xi))_{n=1}^{\infty}$, we can get easily that

$$L^2 - \lim_{n \to \infty} \mathbf{1}_{\Omega_{t,n}} \varphi(\xi) = \varphi(\xi).$$

Similarly to the proof of Theorem 3.1, we can get

$$L^{2} - \lim_{n \to \infty} \mathcal{E}_{g}[\mathbf{1}_{\Omega_{t,n}}\varphi(\xi)|\mathcal{F}_{t}] = \mathcal{E}_{g}[\varphi(\xi)|\mathcal{F}_{t}].$$

Hence for each $t \in [0, T]$, *P*-a,s., we have

$$\mathcal{E}_g[\varphi(\xi)|\mathcal{F}_t] \ge \varphi[\mathcal{E}_g(\xi|\mathcal{F}_t)].$$

(ii) \Rightarrow (i). Given $\lambda \geq 0$, we define a corresponding increasing convex function φ_{λ} : $\mathbf{R} \rightarrow \mathbf{R}$, such that $\varphi_{\lambda}(x) = \lambda x$, $\forall x \in \mathbf{R}$. For each $t \in [0, T[, z \in \mathbf{R}^d]$, let us pick a large enough positive integer n, such that $t + 1/n \leq T$. It is obvious that $\varphi_{\lambda}(z \cdot (B_{t+1/n} - B_t)) \in L^2(\Omega, \mathcal{F}_T, P)$. By (ii), we know that Jensen's inequality holds for the increasing function φ_{λ} . Thus we have, P-a.s.,

$$\mathcal{E}_g[\varphi_\lambda(z \cdot (B_{t+1/n} - B_t))|\mathcal{F}_t] \ge \varphi_\lambda[\mathcal{E}_g(z \cdot (B_{t+1/n} - B_t)|\mathcal{F}_t)]$$

that is, *P*-a.s.,

$$\mathcal{E}_g[\lambda z \cdot (B_{t+1/n} - B_t) | \mathcal{F}_t] \ge \lambda [\mathcal{E}_g(z \cdot (B_{t+1/n} - B_t) | \mathcal{F}_t)].$$
(4.8)

By (3.8), we know that there exists a subsequence $\{n_k\}_{k=1}^{\infty}$ such that

$$\begin{aligned} P\text{-a.s.,} \quad &\lim_{k \to \infty} n_k [\mathcal{E}_g(\lambda z \cdot (B_{t+1/n_k} - B_t) | \mathcal{F}_t)] = g(t, \lambda z), \\ P\text{-a.s.,} \quad &\lim_{k \to \infty} \lambda n_k [\mathcal{E}_g(z \cdot (B_{t+1/n_k} - B_t) | \mathcal{F}_t] = \lambda g(t, z). \end{aligned}$$

Thus for each $t \in [0, T[, z \in \mathbf{R}^d, \lambda \ge 0, \text{ it follows from (4.8) that}$

$$P\text{-a.s.}, \quad g(t, \lambda z) \ge \lambda g(t, z). \tag{4.9}$$

(A4) and (4.9) imply that

$$P\text{-a.s.}, \quad g(T,\lambda z) = \lim_{\varepsilon \to 0^+} g(T-\varepsilon,\lambda z) \ge \lim_{\varepsilon \to 0^+} \lambda g(T-\varepsilon,z) = \lambda g(T,z).$$

Hence (ii) implies (i). The proof is complete.

Corollary 4.1. Given $\mu \ge 0$, let the generator $g(t, z) = -\mu |z|, \forall (t, z) \in [0, T] \times \mathbf{R}^d$. Then Jensen's inequality for g-expectation holds for increasing convex function φ .

Similarly we can get the following

Theorem 4.2. Let g satisfy (A1), (A3) and (A4). Then the following two conditions are equivalent:

(i) *P*-a.s., $\forall \lambda \leq 0, (t,z) \in [0,T] \times \mathbf{R}^d, g(t,\lambda z) \geq \lambda g(t,z);$

(ii) Jensen's inequality for g-expectation holds for decreasing convex function, i.e., for each $\xi \in L^2(\Omega, \mathcal{F}_T, P)$ and decreasing convex function $\varphi : \mathbf{R} \to \mathbf{R}$, if $\varphi(\xi) \in L^2(\Omega, \mathcal{F}_T, P)$, then for each $t \in [0, T]$, P-a.s.,

$$\mathcal{E}_g[\varphi(\xi)|\mathcal{F}_t] \ge \varphi[\mathcal{E}_g(\xi|\mathcal{F}_t)].$$

Proof. The proof of Theorem 4.2 is similar to that of Theorem 4.1. We omit it.

By Theorem 4.2, we can obtain the following corollary immediately.

Corollary 4.2. Let g satisfy (A1) and (A3). If P-a.s., $\forall (t, z) \in [0, T] \times \mathbb{R}^d$, $g(t, z) \ge 0$, then Jensen's inequality for g-expectation holds for decreasing convex function φ .

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