

Set-Valued Stochastic Equation with Set-Valued Square Integrable Martingale

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Abstract: In this paper, we shall introduce the stochastic integral of a stochastic process with respect to set-valued square integrable martingale. Then we shall give the Aumann integral measurable theorem, and give the set-valued stochastic Lebesgue integral and set-valued square integrable martingale integral equation. The existence and uniqueness of solution to set-valued stochastic integral equation are proved. The discussion will be useful in optimal control and mathematical finance in psychological factors.

1. Introduction

Set-valued theory is used in optimal control(cf.[1]), mathematical finance(cf. [2]), fixed point theory(cf. [3]). Set-valued and fuzzy set-valued theory can be used to account for psychological factors(cf.[4,5]).In [6], stochastic control problems are discussed by stochastic integral with respect to set-valued square integral martingales. M. Malinowski *et al.* discussed the set-valued stochastic integral driven by semi-martingale and the set-valued stochastic differential equations driven by semi-martingale in [7]. J. Li *et al.* discussed set-valued stochastic Lebesgue integral and set-valued stochastic differential equation in [8,9,10]. Puri and Ralescu defined fuzzy set-valued martingales and proved convergence theorems of fuzzy set-valued martingales in [11]. J. Li *et al.* discussed the space of fuzzy set-valued square integrable martingales in [12] and fuzzy set-valued stochastic Lebesgue integral in [13].

In this paper, we shall give the set-valued stochastic integral equation :

$$F(t) = F(0) + (L) \int_0^t b(s, F(s)) ds + (M) \int_0^t \sigma(s, F(s)) dG(s)$$

the first integral is set-valued stochastic Lebesgue integral (see [9,10]), the second integral is set-valued square integrable martingale integral (see [6]). Aumann integral measurable theorem of the second integral shall be given. The existence and uniqueness of the solution to the equation shall be proved.

We organize our paper as follows: in Section 2, we shall introduce some necessary notations, definitions, and results about set-valued stochastic variables and martingales.

Furthermore, we shall give the Aumann integral measurable theorem. In Section 3, we shall give the set-valued stochastic integral equation with respect to set-valued square integrable martingale, and prove the existence and uniqueness of the solution to the set-valued stochastic integral equation.

2. Preliminary on Set-Valued Random Variables and Martingales

Throughout this paper, assume that $(\Omega, \mathcal{A}, \mu)$ is a complete probability space, the σ -field filtration $\{\mathcal{A}_t : t \in I\}$ satisfies the usual conditions (i.e. Containing all nullsets, non-decreasing and right continuous), $I = [0, T]$ with $T > 0$, R is the set of all real numbers, N is the set of all natural numbers, R^d is the d -dimensional Euclidean space with usual norm $\|\cdot\|$, $\mathfrak{B}(E)$ is the Borel field of the space E . Let $K(R^d)$ be the family of all nonempty, closed subsets of R^d . Let $K_c(R^d)$ (resp. $K_k(R^d)$, $K_{kc}(R^d)$) be the family of all nonempty closed convex (resp. compact, compact convex) subsets of R^d . For any $x \in R^d$, A, B are nonempty subsets of R^d , define the distance between x and A ,

$$d(x, A) = \inf_{y \in A} \|x - y\|,$$

$$A + B = \{a + b : a \in A, b \in B\},$$

$$A \ominus B = \{x \in R^d : x + B \subset A\}.$$

The Hausdorff metric on $K(R^d)$ is defined as

$$d_H(A, B) = \max \{ \sup_{a \in A} d(a, B), \sup_{b \in B} d(b, A) \}$$

for $A, B \in K(R^d)$. For $B \in K(R^d)$, define

$$\|B\|_K = d_H(0, B) = \sup_{a \in B} \|a\|.$$

If $F : (\mathbb{W}, \mathcal{A}) \otimes K(R^d)$ satisfies that for any open set

$O \subset R^d, F^{-1}(O) = \{\omega \in \Omega : F(\omega) \cap O \neq \Phi\} \in \mathcal{A}$, then F is called \mathcal{A} -measurable (or a set-valued random variable, random set, multivalued function.).

Let $f = \{f(t), \mathcal{A}_t, t \in I\}$ be a R^d -valued adapted stochastic process. Let $L^p[\mathbb{W}, \mathcal{A}, m, R^d]$ be the set of R^d -valued \mathcal{A}_t -measurable random variable x with $E[\|x\|^p] < \infty$ ($1 \leq p < \infty$), write $\|x\|_p = [E[\|x\|^p]]^{1/p}$.

Let $S_F^p(\mathcal{A}) = \{f \in L^p[\Omega, \mathcal{A}, \mu; R^d] : f(\omega) \in F(\omega), a.e. \omega \in \Omega\}$

For short, denoted by S_F^p . A set-valued random variable

$F : \Omega \rightarrow K(R^d)$ is called *integrable* if S_F^1 is non-empty. F is L^p -bounded if and only if the real-valued random variable $\|F\|_K \in L^p[\Omega; R]$

If F is L^1 -bounded, then F is also called *integrably bounded*. Let

$L^p[\mathbb{W}, \mathcal{F}, m, K(R^d)]$ be the family of all $K(R^d)$ -valued L^p -bounded \mathcal{F} -measurable random variables.

Similarly, we have notations $L^p[\mathbb{W}, \mathcal{F}, m, K_c(R^d)]$, $L^p[\mathbb{W}, \mathcal{F}, m, K_{lc}(R^d)]$, $L^p[\mathbb{W}, K(R^d)]$ and so on. A

set-valued stochastic process $F = \{F(t), \mathcal{A}_t, t \in I\}$ is called a set-valued martingale if

(i) $F = \{F(t), \mathcal{A}_t, t \in I\}$ is adapted and for any $t \in I, F(t)$ is L^1 -bounded;

(ii) for any $t \leq s, t \in I, E[F(s) | \mathcal{A}_t] = F(t)$, a.e. (μ).

A set-valued martingale $F = \{F(t), \mathcal{A}_t, t \in I\}$ is called square integrable, if $\sup_{t \in I} E[\|F(t)\|_K^2] < \infty$.

Denote $CMS(F)$ as the set of R^d -valued continuous martingale selections of set-valued square integrable martingale F .

Concerning more notations, definitions and more results of set-valued random variable, set-valued martingales, readers could refer to [6,14].

Definition (cf.[6]) Assume $F = \{F(t), \mathcal{A}_t, t \in I\}$ is a separable square integrable set-valued martingale and for any fixed $\omega \in \Omega, F(\omega, t)$ is lower

semicontinuous with $F(0) = 0$ a.e., g is a predictable bounded stochastic process. For any $\omega \in \Omega, t \in I$, define

$$(A) \int_0^t g(s, \omega) dF(s, \omega) = \{ \int_0^t g(s, \omega) df(s, \omega) : f \in CMS(F) \},$$

then $(A) \int_0^t g(s, \omega) dF(s, \omega)$ is said to be the *Aumann type stochastic integral* of g with respect to the set-valued square integrable martingale F .

By Theorem 4.4 and Theorem 4.7 in [6], we have the following Aumann integral measurability theorem:

Theorem 1 Assume that $F = \{F(t), \mathcal{A}_t, t \in I\}$ is a separable square integrable set-valued martingale and for any fixed $\omega \in \Omega, F(\omega, t)$ is lower semicontinuous with $F(0) = 0$ a.e., g is a predictable bounded stochastic process. For any $\omega \in \Omega, t \in I$, we have that

$$cl \{ (A) \int_0^t g(s, \omega) dF(s, \omega) \} = cl \{ \int_0^t g(s, \omega) df(s, \omega) : f \in CMS(F) \}$$

is a set-valued random variable.

The above theorem is useful to discuss the theory of Aumann integral, etc.

3. Set-Valued Stochastic Integral Equation

We consider the following set-valued stochastic integral equation

$$F(t) = F(0) + (L) \int_0^t b(s, F(s)) ds + (M) \int_0^t \sigma(s, F(s)) dG(s),$$

(1)

Where for any $t \in I, F \in L^2(K(R^d))$ with initial condition $F(0)$ being an L^2 -bounded set-valued

random variable, $b : I \times K(R^d) \rightarrow K(R^d)$ is measurable, $\sigma : I \times K(R^d) \rightarrow R^d \otimes R^m$ is bounded predictable measurable, $G(s)$ is a set-valued

square integral martingale. $(L) \int_0^t b(s, F(s)) ds$ is set-valued stochastic Lebesgue integral (see [9]),

$(M) \int_0^t \sigma(s, F(s)) dG(s)$ is stochastic integral with

respect to set-valued square integral martingale (see Definition 4.5 in [6]).

Theorem 2 Assume that $b(t, F), s(t, F)$ satisfy the following conditions:

(1) Linear bound condition:

$$\|b(t, F)\|^2 + \|s(t, F)\|^2 \leq a^2(1 + \|F\|_K^2);$$

(2) Lipschitz continuous condition:

$$\|b(t, F_1(t)) - b(t, F_2(t))\| + \|s(t, F_1(t)) - s(t, F_2(t))\| \leq a d_H(F_1(t), F_2(t))$$

where $a > 0$ is a constant;

(3) Set-valued integral inequality:

$$Ed_H^2\left(\int_0^t \sigma(s, F_1(s))dG_s, \int_0^t \sigma(s, F_2(s))dG_s\right) \leq E \int_0^t d_H^2(F_1(s), F_2(s))ds$$

Then for any $F(0) \in L^2(K(R^d))$, there exists a unique solution to equation (1).

Proof: First, it is obviously the existence of b, s which satisfies (1),(2),(3).

If $F(t) \in L^2(K(R^d))$, for $t \in I$, then

$$E\|b(t, F(t))\|_K^2 + E\|s(t, F(t))\|^2 \leq a^2(1 + E\|F\|_K^2).$$

Without loss of generality, assume that $b \in L^2(K(R^d))$, $s \in \mathcal{V}L^2(R^d \times R^m)$.

Step 1. $F_0(t) = F(0)$,

$$F_{n+1}(t) = F(0) + \int_0^t b(s, F_n(s))ds + \int_0^t \sigma(s, F_n(s))dG_s, n \geq 0$$

We shall prove that for any $n \geq 0$, F_n satisfies:

(a) $F_n \in L^2(K(R^d))$, for $\forall t \in I$

(b) $\lim_{s \rightarrow t} Ed_H^2(F_n(t), F_n(s)) = 0$.

For $n = 0$, it is obviously right. Suppose that F_n has properties (a),(b) for any fixed n , we shall prove so does F_{n+1} .

$$Y(t) := \int_0^t b(s_1, F_n(s_1))ds_1$$

Let (a) holds to $Y(t)$ by Theorem 3.6 in [9],

$$Ed_H^2(Y(t), Y(s)) = Ed_H^2(cI(Y(s) + \int_s^t b(s_1, F_n(s_1))ds_1), Y(s))$$

$$\leq Ed_H^2(\int_s^t b(s_1, F_n(s_1))ds_1, 0)$$

$$= E\|\int_s^t b(s_1, F_n(s_1))ds_1\|_K^2$$

By the Linear bound condition (1) and norm property

(β)

of set-valued Lebesgue integral, it satisfies

Let $Z(t) := \int_0^t \sigma(s_1, F_n(s_1))dG_{s_1}$. It is obviously that (a) holds.

$$\begin{aligned} Ed_H^2(Z(t), Z(s)) &= Ed_H^2\left(\int_0^t \sigma(s_1, F_n(s_1))dG_{s_1}, \int_0^s \sigma(s_1, F_n(s_1))dG_{s_1}\right) \\ &= Ed_H^2\left(\int_0^s \sigma(s_1, F_n(s_1))ds_1 + \left(\int_0^t \sigma(s_1, F_n(s_1))dG_{s_1} - \int_0^s \sigma(s_1, F_n(s_1))dG_{s_1}\right), \int_0^s \sigma(s_1, F_n(s_1))dG_{s_1}\right) \\ &\leq Ed_H^2\left(\int_0^t \sigma(s_1, F_n(s_1))dG_{s_1} - \int_0^s \sigma(s_1, F_n(s_1))dG_{s_1}, 0\right) \\ &\leq Ed_H^2\left(\int_s^t \sigma(s_1, F_n(s_1))dG_{s_1}, 0\right) \end{aligned}$$

By Theorem 1 and Theorem 4.7 in [6], the set-valued integral $\int_s^t \sigma(s_1, F_n(s_1))dG_{s_1}$ is represented by Aumann integral.

Therefore,

$$\lim_{s \rightarrow t} \int_s^t \sigma(s_1, F_n(s_1))dG_{s_1} = \{0\}, \text{ a. e.}$$

Thus, we have $\lim_{s \rightarrow t} Ed_H^2(Z(t), Z(s)) = 0$.

Step 2. Now we shall prove that F_n converges to F .

Let

$$F_0(t) = F(0),$$

$$F_{n+1}(t) = F(t) + \int_0^t b(s, F_n(s))ds + \int_0^t \sigma(s, F_n(s))dG_s, n \geq 0$$

We have

$$\begin{aligned} d_H(F_1(t), F_0(t)) &= d_H(F_0(t) + \int_0^t b(s, F_0(s))ds + \int_0^t \sigma(s, F_0(s))dG_s, F_0(t)) \\ &\leq \varphi_H\left(\int_0^t b(s, F_0(s))ds, 0\right) + d_H\left(\int_0^t \sigma(s, F_0(s))dG_s, 0\right) \\ &= \|\int_0^t b(s, F_0(s))ds\|_K + \|\int_0^t \sigma(s, F_0(s))dG_s\|_K \end{aligned}$$

For the first part, we have

$$\begin{aligned} E\|\int_0^t b(s, F_0(s))ds\|_K^2 &\leq E\left(\int_0^t \|f(s, F_0(s))\|_K ds\right)^2 \\ &\leq tE\left(\int_0^t \|b(s, F_0(s))\|_K^2 ds\right) \\ &= t\int_0^t E\|b(s, F_0(s))\|_K^2 ds \\ &\leq ta^2\int_0^t (1 + E\|F_0(s)\|_K^2)ds \\ &\leq A^2t, \end{aligned}$$

For the second part, by inequality (3), we have

$$\begin{aligned}
 & E \left\| \int_0^t \sigma(s, F_0(s)) dG_s \right\|_K^2 \\
 &= E d_H^2 \left(\int_0^t \sigma(s, F_0(s)) dG_s, 0 \right) \\
 &\leq E \int_0^t d_H^2(F_0(s), 0) ds \\
 &= tE \left\| F(0) \right\|_K^2 \\
 \text{Let } B^2 &= E \left\| F(0) \right\|_K^2, \\
 & E d_H^2(F_1(0), F_0(t)) \\
 &\leq 2E \left\| \int_0^t b(s, F_0(s)) ds \right\|_K^2 + 2E \left\| \int_0^t \sigma(s, F_0(s)) dG_s \right\|_K^2 \\
 &\leq 2(A + B)^2 t.
 \end{aligned}$$

By the same way, we have

$$\begin{aligned}
 & E d_H^2(F_{n+1}(t), F_n(t)) \\
 &= E(d_H(F(0) + \int_0^t b(s, F_n(s)) ds + \int_0^t \sigma(s, F_n(s)) dG_s, F(0) \\
 &\quad + \int_0^t b(s, F_{n-1}(s)) ds + \int_0^t \sigma(s, F_{n-1}(s)) dG_s) \\
 &\leq E d_H(F(0), F(0)) + E d_H(\int_0^t b(s, F_n(s)) ds, \int_0^t b(s, F_{n-1}(s)) ds) \\
 &\quad + E d_H(\int_0^t \sigma(s, F_n(s)) dG_s, \int_0^t \sigma(s, F_{n-1}(s)) dG_s) \\
 &\leq 2E d_H^2(\int_0^t b(s, F_n(s)) ds, \int_0^t b(s, F_{n-1}(s)) ds) \\
 &\quad + 2E d_H^2(\int_0^t \sigma(s, F_n(s)) dG_s, \int_0^t \sigma(s, F_{n-1}(s)) dG_s) \\
 &\leq 2tE \int_0^t d_H^2(b(s, F_n(s)), b(s, F_{n-1}(s))) ds \\
 &\quad + 2E \int_0^t d_H^2(F_n(s), F_{n-1}(s)) ds \\
 &\leq 2ta^2 E \int_0^t d_H^2(F_n(s), F_{n-1}(s)) ds \\
 &\quad + 2E \int_0^t d_H^2(F_n(s), F_{n-1}(s)) ds \\
 &= 2(ta^2 + 1)E \int_0^t d_H^2(F_n(s), F_{n-1}(s)) ds
 \end{aligned}$$

Iterating the above process, we get

$$E d_H^2(F_{n+1}(t), F_n(t)) \leq 2^{n+1} (ta^2 + 1)^{2n} (A + B)^2 \frac{t^{n+1}}{(n+1)!}$$

i.e.

$$\begin{aligned}
 \Delta_n(t) &= E \sup_{s \in [0, t]} d_H^2(F_{n+1}(s), F_n(s)) \\
 &\leq 2^{n+1} (ta^2 + 1)^{2n} (A + B)^2 \frac{t^{n+1}}{(n+1)!}
 \end{aligned}$$

Thus,

$$\Delta_n(T) \leq 2^{n+1} (Ta^2 + 1)^{2n} (A + B)^2 \frac{T^{n+1}}{(n+1)!}$$

Therefore, we have

$$\sum_{n=1}^{\infty} \Delta_n(T) < \infty.$$

This ensures the existence of the strong solution.

Step 3. Let F, G be two solutions to equation (1). The uniqueness of the solution is the same to the proof of the existence.

Remark 3 Furthermore, fuzzy set-valued martingale integral, fuzzy set-valued stochastic Lebesgue integral [13], and fuzzy set-valued stochastic integral would be useful in the area.

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References

- [1] M. Kisielewicz, Differential Inclusions and Optimal Control, Kluwer Academic Publishers, 1991.
- [2] I. Karatzas, Lectures on the Mathematics of Finance, American Mathematical Society, Providence, RI, 1997.
- [3] A. Fryszkowski, Fixed Point Theory for Decomposable Sets, Kluwer Academic Publishers, 2004.
- [4] S. Li and L. Guan, “Fuzzy set-valued Gaussian process and Brownian motion,” Information Sciences, vol.177, pp. 3251–3259, August 2007.
- [5] J. Li and J. Wang, “On a fluid queue with a fuzzy set-valued Gaussian process input,” unpublished.
- [6] S. Li, J. Li and X. Li, “Stochastic integral w.r.t. set-valued square integrable martingale,” Journal of Mathematical Analysis and Applications vol.370, pp. 659-671, October 2010.
- [7] M. Malinowski, “On a new set-valued stochastic integral with respect to semimartingales and its applications,” Journal of Mathematical Analysis and Applications vol.408, pp.669-680, Decembe 2013.
- [8] J. Li S. Li and Y. Ogura, “Strong solution of Ito type set-valued stochastic differential equation,” Acta Mathematica Sinica, English Series, vol.26, pp. 1739-1748, September 2010.
- [9] J. Li and S. Li, “Set-valued stochastic Lebesgue integral and representation theorems, International Journal of Computational Intelligence Systems,” vol.1, pp. 177-187, May 2008.
- [10] J. Li and S. Li, “Ito type set-valued stochastic differential equation,” Journal of Uncertain Systems, vol.3, pp.52-63, February 2009.
- [11] M. L. Puri and D. A. Ralescu, “Convergence theorem for fuzzy martingales,” Journal of Mathematical Analysis and Applications vol.160, pp.107-122, September 1991.
- [12] J. Li, S. Li and Y. Xue, “The Space of Fuzzy Set-Valued Square Integrable Martingales,” IEEE

International Conference on Fuzzy Systems, pp. 872-876, August 2009.

- [13] J. Li and J. Wang, "Fuzzy set-valued stochastic Lebesgue integral," *Fuzzy Sets and Systems*, vol.200, pp.48-64, August 2012.
- [14] S. Li, Y. Ogura and V. Kreinovich, *Limit Theorems and Applications of Set-Valued and Fuzzy Set-Valued Random Variables*, Kluwer Academic Publishers, 2002.