

Generalized reflected backward stochastic differential equations driven by a fractional Brownian motion

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This research is supported by the National Science Center, DEC-2012/07/D/ST6/02534

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Fractional Brownian motion

A Fractional Brownian motion (fBm) with Hurst parameter $H \in (0, 1)$ is a zero mean Gaussian process $B^H = \{B_t^H, t \geq 0\}$ with the covariance function

$$R_H(s, t) = E \left(B_s^H B_t^H \right) = \frac{1}{2} \left(t^{2H} + s^{2H} - |t - s|^{2H} \right).$$

The process B^H

- ▶ is not a semimartingale when $H \neq 1/2$
- ▶ for $H = 1/2$, B^H is a standard Wiener process
- ▶ has homogeneous increments, i.e. $B_{t+s}^H - B_s^H$ has the same law of B_t^H
- ▶ is self-similar, i.e. B_{at}^H has the same law as $a^H B_t^H$ for any $a > 0$
- ▶ has continuous trajectories

Let $H > 1/2$. Denote

$$\langle \xi, \eta \rangle_t = H(2H - 1) \int_0^t \int_0^t |u - v|^{2H-2} \xi(u) \eta(v) du dv,$$
$$\|\xi\|_t^2 = \langle \xi, \xi \rangle_t.$$

\mathcal{P}_T – the set of all polynomials of the fBm, i.e.

$$F(\omega) = f \left(\int_0^T \xi_1(t) dB_t^H, \dots, \int_0^T \xi_k(t) dB_t^H \right)$$

where $(\xi_n)_n$ – a sequence such that $\langle \xi_i, \xi_j \rangle_T = \delta_{ij}$.

The Malliavin derivative operator D_s^H of $F \in \mathcal{P}_T$ is defined as follows:

$$D_s^H F = \sum_{i=1}^k \frac{\partial f}{\partial x_i} \left(\int_0^T \xi_1(t) dB_t^H, \dots, \int_0^T \xi_k(t) dB_t^H \right) \xi_i(s), \quad s \in [0, T].$$

$\mathbb{D}_{1,2}$ – the Banach space being a completion of \mathcal{P}_T with the norm:

$$\|F\|_{1,2}^2 = E|F|^2 + E\|D_s^H F\|_T^2.$$

Denote

$$\mathbb{D}_t^H F = H(2H - 1) \int_0^T |t - s|^{2H-2} D_s^H F ds.$$

THEOREM.

$$E \left(\int_0^T F_s dB_s^H \right) = 0$$

and

$$E \left(\int_0^T F_s dB_s^H \right)^2 = E \left(\|F\|_T^2 + \int_0^T \int_0^T \mathbb{D}_s^H F_t \mathbb{D}_t^H F_s ds dt \right).$$

Fractional calculus

$$X_t = X_0 + \int_0^t g_s^1 ds + \int_0^t f_s^1 dB_s^H, \quad Y_t = Y_0 + \int_0^t g_s^2 ds + \int_0^t f_s^2 dB_s^H, \\ t \in [0, T] \text{ and } F \in C^{1,2}. \text{ Then}$$

$$F(t, X_t) = F(0, X_0) + \int_0^t \frac{\partial F}{\partial s}(s, X_s) ds + \int_0^t \frac{\partial F}{\partial x}(s, X_s) dX_s \\ + \frac{1}{2} \int_0^t \frac{\partial^2 F}{\partial x^2}(s, X_s) \frac{d}{ds} \left(\|f^1\|_s^2 \right) ds$$

$$X_t Y_t = X_0 Y_0 + \int_0^t X_s g_s^2 ds + \int_0^t X_s f_s^2 dB_s^H + \int_0^t Y_s g_s^1 ds \\ + \int_0^t Y_s f_s^1 dB_s^H + \int_0^t \mathbb{D}_s^H X_s f_s^2 ds + \int_0^t \mathbb{D}_s^H Y_s f_s^1 ds.$$

Let

$$\eta_t = \eta_0 + \int_0^t \sigma(s) dB_s^H, \quad t \in [0, T],$$

where $\sigma : [0, T] \rightarrow \mathbb{R}$ – a deterministic continuous differentiable function, $\sigma(t) \neq 0$, η_0 – a constant.

For functions $\varphi, \psi : \mathbb{R} \rightarrow (-\infty, \infty]$ we use the following notations:

$$\partial\varphi(y) = \{\hat{y} \in \mathbb{R}; \hat{y} \cdot (v - y) + \varphi(y) \leq \varphi(v), \forall v \in \mathbb{R}\}$$

$$\text{Dom}\varphi = \{y \in \mathbb{R}; \varphi(y) < \infty\}, \quad \text{Dom}(\partial\varphi) = \{y \in \mathbb{R}; \partial\varphi(y) \neq \emptyset\}$$

$$(y, \hat{y}) \in \partial\varphi \Leftrightarrow y \in \text{Dom}(\partial\varphi), \hat{y} \in \partial\varphi(y).$$

(analogously for ψ).

GRBSDE with respect to a fBm

We consider the following generalized backward stochastic variational inequality (GBSVI) driven by a fBm: $Y_T = \xi$ and

$$dY_t + f(t, \eta_t, Y_t, Z_t)dt + g(t, \eta_t, Y_t)d\Lambda_t - Z_t dB_t^H \in \partial\varphi(Y_t)dt + \partial\psi(Y_t)d\Lambda_t.$$

DEFINITION. A solution of a GBSVI driven by a fBm B^H associated with data (ξ, f, g, Λ) is a quadruple (Y, Z, U, V) satisfying GRBSDE

$$Y_t = \xi + \int_t^T f(s, \eta_s, Y_s, Z_s)ds + \int_t^T g(s, \eta_s, Y_s)d\Lambda_s - \int_t^T Z_s dB_s^H - \int_t^T U_s ds - \int_t^T V_s d\Lambda_s, \quad t \in [0, T] \quad (1)$$

and such that $(Y_t, U_t) \in \partial\varphi$, $(Y_t, V_t) \in \partial\psi$ and

$$Y \in \tilde{\mathcal{V}}_{[0, T]}^{1/2} \cap \tilde{\mathcal{V}}_{[0, T]}^{1/2, \Lambda}, \quad Z \in \tilde{\mathcal{V}}_{[0, T]}^H, \quad U, V \in \tilde{\mathcal{V}}_{[0, T]}^H \cap \tilde{\mathcal{V}}_{[0, T]}^{H, \Lambda}.$$

THEOREM. There exists a solution of (1) such that

$$E \left(e^{\nu\Lambda_t} |Y_t|^2 + \int_t^T e^{\nu\Lambda_s} s^{2H-1} |Z_s|^2 ds + \int_t^T e^{\nu\Lambda_s} |Y_s|^2 d\Lambda_s \right) < \infty.$$

Sketch of proof. We use the Itô formula for $e^{\nu\Lambda_t} |Y_t|^2$ and facts that

$$\mathbb{D}_s^H Y_s = \int_0^s \phi(s-r) D_r^H Y_s dr = \frac{\hat{\sigma}(s)}{\sigma(s)} Z_s$$

and there exists $M > 0$ such that for all $t \in [0, T]$,

$$\frac{t^{2H-1}}{M} \leq \frac{\hat{\sigma}(t)}{\sigma(t)} \leq M t^{2H-1}.$$

Penalization scheme

We approximate functions φ and ψ by sequences of convex, C^1 class functions φ_ε , ψ_ε respectively, $\varepsilon > 0$, defined by

$$\varphi_\varepsilon(y) = \inf \left\{ \frac{1}{2\varepsilon} |y - v|^2 + \varphi(v); v \in \mathbb{R} \right\} = \frac{1}{2\varepsilon} |y - J_\varepsilon(y)|^2 + \varphi(J_\varepsilon(y)),$$
$$\psi_\varepsilon(y) = \inf \left\{ \frac{1}{2\varepsilon} |y - v|^2 + \psi(v); v \in \mathbb{R} \right\} = \frac{1}{2\varepsilon} |y - \tilde{J}_\varepsilon(y)|^2 + \psi(\tilde{J}_\varepsilon(y)),$$

where $J_\varepsilon(y) = y - \varepsilon \nabla \varphi_\varepsilon(y)$ and $\tilde{J}_\varepsilon(y) = y - \varepsilon \nabla \psi_\varepsilon(y)$.

Consider a sequence of generalized BSDEs

$$Y_t^\varepsilon = \xi + \int_t^T f(s, \eta, Y_s^\varepsilon, Z_s^\varepsilon) ds + \int_t^T g(s, \eta, Y_s^\varepsilon) d\Lambda_s - \int_t^T Z_s^\varepsilon dB_s^H$$
$$- \int_t^T \nabla \varphi_\varepsilon(Y_s^\varepsilon) ds - \int_t^T \nabla \psi_\varepsilon(Y_s^\varepsilon) d\Lambda_s, \quad t \in [0, T].$$

$$(a) E \int_t^T e^{\nu\Lambda_s} s^{2H-1} (|\nabla\varphi_\varepsilon(Y_s^\varepsilon)|^2 ds + |\nabla\psi_\varepsilon(Y_s^\varepsilon)|^2 d\Lambda_s) \leq C\Theta_2(t, T)$$

$$(b) E \int_t^T e^{\nu\Lambda_s} s^{2H-1} \left(\varphi(J_\varepsilon(Y_s^\varepsilon)) + \psi(\tilde{J}_\varepsilon(Y_s^\varepsilon)) \right) d\Lambda_s \leq C\Theta_2(t, T)$$

$$(c) E e^{\nu\Lambda_t} t^{2H-1} \left(|Y_t^\varepsilon - J_\varepsilon(Y_t^\varepsilon)|^2 + |Y_t^\varepsilon - \tilde{J}_\varepsilon(Y_t^\varepsilon)|^2 \right) \leq \varepsilon \cdot C\Theta_2(t, T)$$

$$(d) E e^{\nu\Lambda_t} t^{2H-1} \left(\varphi(J_\varepsilon(Y_t^\varepsilon)) + \psi(\tilde{J}_\varepsilon(Y_t^\varepsilon)) \right) \leq C\Theta_2(t, T)$$

$$(e) E \int_t^T e^{\nu\Lambda_s} s^{2H-1} \left(|Y_s^\varepsilon - J_\varepsilon(Y_s^\varepsilon)|^2 ds + |Y_s^\varepsilon - \tilde{J}_\varepsilon(Y_s^\varepsilon)|^2 d\Lambda_s \right) \leq \varepsilon^2 \cdot C\Theta_2(t, T).$$

where

$$\begin{aligned}\Theta_2(t, T) = & E \left(T^{2H-1} e^{\nu \Lambda_T} (\varphi(\xi) + \psi(\xi)) \right. \\ & + \int_t^T e^{\nu \Lambda_s} s^{2H-1} (|\eta_s|^2 + |Y_s^\varepsilon|^2 + |Z_s^\varepsilon|^2) ds \\ & \left. + \int_t^T e^{\nu \Lambda_s} (|\eta_s|^2 + |Y_s^\varepsilon|^2) d\Lambda_s \right).\end{aligned}$$

THEOREM. $(Y^\varepsilon, Z^\varepsilon)$ is a Cauchy sequence, i.e. for $\varepsilon, \delta > 0$

$$\begin{aligned} E \left(e^{\nu \Lambda_t} t^{2H-1} |Y_t^\varepsilon - Y_t^\delta|^2 + \int_t^T e^{\nu \Lambda_s} s^{2H-2} |Y_s^\varepsilon - Y_s^\delta|^2 (ds + d\Lambda_s) \right. \\ \left. + \int_t^T e^{\nu \Lambda_s} s^{2(2H-1)} |Z_s^\varepsilon - Z_s^\delta|^2 ds \right) \\ \leq C \cdot (\varepsilon + \delta) \cdot \Theta_2(t, T). \end{aligned}$$

$$U_t \cdot (u_t - Y_t) + \varphi(Y_t) \leq \varphi(u_t) \quad \text{and} \quad V_t \cdot (v_t - Y_t) + \psi(Y_t) \leq \psi(v_t),$$

mean that $(Y_t, U_t) \in \partial\varphi$ and $(Y_t, V_t) \in \partial\psi$, $t \in [0, T]$.

$(Y^\varepsilon, Z^\varepsilon, \nabla\varphi_\varepsilon(Y^\varepsilon), \nabla\psi_\varepsilon(Y^\varepsilon))$ converges to a solution of (1).

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