

Quadratic BSDEs Revisited: A Forward Point of View

Tamerza, Octobre 2010

Nicole El Karoui and Pauline Barrieu

(UPMC/Ecole Polytechnique, Paris) and (LSE, London)

with the financial support of the chair "Risques financiers" de la
"Fondation du Risque".

Plan

- 1 Motivation
 - Quadratic BSDEs
 - Quadratic semimartingale
 - Algebraic characterisation
- 2 Exponential transformation and Entropic process
- 3 Quadratic variation estimates
- 4 Quadratic BSDEs

Motivation

- ▶ Quadratic BSDE's appear naturally in a lot of **optimization** problems
 - Mean variance problems
 - Utility maximization
 - Risk sensitive problem
 - Large deviations....
- ▶ From the **seminal** paper of **Kobylanski** (2000) on bounded solutions, different extensions are provided, in particular by **many people in the audience**
 - We use many ideas from **Briand and Hu** (2006), where bounded assumption are **relaxed**.
 - See also the very interesting paper of **Tevzadze** (2008).
- ▶ Need of **unified point of view** on estimates and convergence results.

Definition of Quadratic BSDE



- ▶ Probability space $(\Omega, \mathcal{F}, \mathbb{P}, (\mathcal{F}_t))$ with **continuous martingales**
- ▶ A coefficient $g(t, y, z)$ with good measurability and a terminal condition ξ_T

The equation

- ▶ An equation $-dY_t = g(t, Y_t, Z_t)dt - Z_t dW_t, \quad Y_T = \xi_T.$, where W is a d -dimensional BM
- ▶ A solution is a pair $(Y, Z) \in \mathbb{R} \times \mathbb{R}^d$ of adapted processes such that the paths of Y are continuous, and $\int_0^T |Z_t|^2 dt < \infty, \int_0^T |g(t, Y_t, Z_t)| dt < \infty, \mathbb{P}$ -a.s

The quadratic case

- ▶ **Quadratic BSDE** = quadratic coefficient = $d\mathbb{P} \otimes dt$ -a.s.
 $|g(\cdot, t, y, z)| \leq Q(t, y, z) \equiv 1/\delta|t| + c_t|y| + \frac{\delta}{2}|z|^2,$

Quadratic semimartingale : Definition

BSDEs : a forward point of view

- ▶ More flexible point of view, localization technique may be used

Definition A **quadratic semimartingale** is a continuous process $Y_t = Y_0 - V + M$ satisfying the constraint :

- ▶ **Structure condition** $\mathcal{Q}(\Lambda, C, \delta)$: There exist two adapted increasing processes (Λ, C) and $\delta > 0$ s.t.

$$d|V|_t \ll 1/\delta d\Lambda_t + |Y_t| dC_t + \frac{\delta}{2} d\langle M \rangle_t, \quad \mathbb{P}\text{-a.s.},$$

- V is a predictable process with finite total variation $|V|$
- M is a local martingale with quadratic variation $\langle M \rangle$
- \ll stands for the absolute continuity of increasing processes.

Typical examples

Notation : $\mathcal{Q}(0, 0, \delta) = q_\delta$, and $q(M) = -\frac{1}{2}\langle M \rangle = -\underline{q}(M)$, and $r_t(r_0, M) = r_0 + M_t - \frac{1}{2}\langle M \rangle_t = r_0 + r_t(M)$

Example of q -semimartingale or BSDEs

- ▶ **Log of exponential martingale** is a q -semimartingale

$$\mathcal{E}(M) = \exp(M - \frac{1}{2}\langle M \rangle) = \exp(r_t(M))$$
- ▶ **Entropic process** : If $\xi_T \in \mathbb{L}_{\text{exp}}^1$,

$$\rho_t(\xi_T) = \ln \mathbb{E}[\exp(\xi_T) | \mathcal{F}_t] = \rho_0(\xi_T) + r_t(M)$$
 and $\mathcal{E}(M)$ is a u.i. martingale

Example of $-q$ -BSDEs

- ▶ $\underline{r}_t(M) = -r_t(-M)$ is a $-q$ -semimartingale, and if $-\xi_T \in \mathbb{L}_{\text{exp}}^1$, $\underline{r}_t(\xi_T) = -\rho_t(-\xi_T) = \underline{\rho}_0(\xi_T) - r_t(-M)$ is a solution.
 - $\mathcal{E}(-M)$ is u.i. integrable, but in general $\mathcal{E}(M)$ is not
 - True if $|\xi_T| \in \mathbb{L}_{\text{exp}}^1$

Basic properties of quadratic quasimartingale

I

Definition A **quadratic submartingale** is a continuous semimartingale $X = X_0 + M - V$ such that $-V + \frac{1}{2}\langle M \rangle = A$ is a predictable increasing process. Equivalently, $X = X_0 + r_t(M) + A$

Properties Let Y be a $\mathcal{Q}(\Lambda, C, \delta)$ -semimartingale.

- ▶ **The role of δ** : $\forall \lambda \neq 0, \lambda Y$ is a $\mathcal{Q}(\Lambda, C, \frac{\delta}{|\lambda|})$ -semimartingale, and $M^{\lambda Y} = \lambda M^Y$.
- ▶ $\lambda Y - \frac{1}{2}\lambda(\lambda - 1)\langle M \rangle$ is a $\mathcal{Q}(\lambda\Lambda, C, \delta)$ -semimartingale
- ▶ **Property of $|Y|$** :
 - Let ϵ be a predictable process such that $|\epsilon| = 1$, a.s.. Then the process $Y^\epsilon = \epsilon \cdot Y = Y_0 + \int_0^\cdot \epsilon_s dY_s$ is a $\mathcal{Q}(\Lambda, C, \delta)$ -semimartingale.

Basic properties of quadratic quasimartingale

II

- In particular, taking $\epsilon^s = \text{sign}(Y_.)$, and denoting by $L.(Y)$ the local time of $Y_.$ at 0, the process $|Y_.| - L.(Y) - 2(Y_0)^- = \epsilon^s \cdot Y_.$ is a $\mathcal{Q}(\Lambda, C, \delta)$ -semimartingale.
- If $Y_.$ is a \mathcal{Q} -quasimartingale, then $|Y_.|$ is a \mathcal{Q} -submartingale.

▶ **Structure simplification** Put

$$X_.\^{\Lambda, C}(Y) = Y_.\ + \Lambda_.\ + |Y_.| * C_., \quad \bar{X}_.\^{\Lambda, C}(Y) = e^{C_.\} |Y_.| + e^{C_.\} * \Lambda_.$$

- The processes $X_.\^{\Lambda, C}(\delta Y)$ and $X_.\^{\Lambda, C}(-\delta Y)$ are \mathcal{Q} -submartingales.
- The process $\bar{X}_.\^{\Lambda, C}(|\delta Y_.|)$ is a \mathcal{Q} -submartingale.

Characterisation of quadratic semimartingales via exponential transformation

Motivation For any $Y \in \mathcal{Q}(\Lambda, C)$, $Y = X^{\Lambda, C} + X^{\Lambda, C}(-Y)$ where $X^{\Lambda, C}$ and $X^{\Lambda, C}(-Y)$ are \mathcal{Q} -submartingales.

Main result : Converse Property

Let K be a continuous increasing process

- ▶ If there exist two **ladlag** processes with $\underline{X} + \overline{X} = K$ such that $\exp(\underline{X})$ and $\exp(\overline{X})$ are submartingales

the both processes $(\underline{X}, \overline{X})$ are **continuous** processes, and the process $Y = \overline{X} - K$ is a $\mathcal{Q}(K)$ -semimartingale.

- ▶ Not sufficient to show that $K = \Lambda + Y + |Y| * C$.
- ▶ In place, use the processes

$$U_t^{\Lambda, C}(e^Y) = e^{Y_t} + \int_0^t e^{Y_s} d\Lambda_s + \int_0^t e^{Y_s} |Y_s| dC_s$$

Proof

- ▶ From the quadratic submartingale decomposition

$$\bar{X}_t = \bar{X}_0 + \bar{M}_t - \frac{1}{2} \langle \bar{M} \rangle_t + \bar{A}_t \quad \text{and} \quad \underline{X}_t = \underline{X}_0 + \underline{M}_t - \frac{1}{2} \langle \underline{M} \rangle_t + \underline{A}_t.$$

- ▶ By uniqueness of the predictable decomposition of X and $-X$,

$$\underline{M}_t = -\bar{M}_t \quad \text{and so} \quad \langle \underline{M} \rangle_t = \langle \bar{M} \rangle_t, \quad \text{and} \quad \bar{A}_t + \underline{A}_t = \langle \bar{M} \rangle_t + 2K_t.$$

Since $\langle \bar{M} \rangle_t$ and K_t are continuous, both increasing processes \bar{A} and \underline{A} are also continuous and then \bar{X} and \underline{X} .

- ▶ From Radon-Nikodym's Theorem, $d\bar{A}_t = \alpha_t d(\frac{1}{2} \langle \bar{M} \rangle_t + K_t)$ with $0 \leq \alpha_t \leq 2$.

Substituting \bar{A} into the decomposition of Y , we get

$$dY_t = -(1 - \alpha_t) d(\frac{1}{2} \langle \bar{M} \rangle_t + K_t) + dM_t \quad \text{with} \quad |1 - \alpha_t| \leq 1$$

Therefore, Y is a $\mathcal{Q}(K, 0, 1)$ -semimartingale.



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 - Uniform integrability and Inequalities
- 3 Quadratic variation estimates
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The class $(\mathcal{D}_{\text{exp}})$ and Inequalities I

Definition of the class $(\mathcal{D}_{\text{exp}})$

- ▶ The "class (\mathcal{D}) ", ((\mathcal{D}) for Doob)=Optional process X , s.t $|X|$ is **dominated** by a u.i martingale.
- ▶ X such that $\exp(|X|)$ is in class (\mathcal{D}) is in the class $(\mathcal{D}_{\text{exp}})$.
- ▶ **(\mathcal{D}) -submartingales** S are characterized by "submartingale inequalities"

$$\text{for } \sigma \leq \tau \leq T, \quad S_\sigma \leq \mathbb{E}[S_\tau | \mathcal{F}_\sigma], \text{ a.s..}$$

- ▶ If S is positive, $X_t = \ln S_t$ verifies the so-called *entropic inequalities* : $X_\sigma \leq \rho_\sigma(X_\tau)$ *a.s.*

Submartingale Inequalities and Characterization I

\mathcal{Q} semimartingales characterization.

An optional process X with $\exp(|X_T|) \in \mathbb{L}^1$ is a \mathcal{Q} -semimartingale in $(\mathcal{D}_{\text{exp}})$ if and only if $\underline{\rho}_\sigma(X_T) \leq X_\sigma \leq \rho_\sigma(X_T)$ a.s.

$\mathcal{Q}(\Lambda, C)$ semimartingales characterization. : (Briand and Hu)

Assume $\bar{X}_T^{\Lambda, C}(|Y_T|) = e^{C_T} |Y_T| + \int_0^T e^{C_s} d\Lambda_s$ in $\mathbb{L}_{\text{exp}}^1$ and

- ▶ $|Y_t| \leq \rho_t(e^{C_{t,T}} |Y_T| + \int_t^T e^{C_{t,s}} d\Lambda_s) = \ln \Phi_t(|Y_T|)$
- ▶ or equivalently $\exp(e^{C_t} |Y_t| + \int_0^t e^{C_s} d\Lambda_s)$ is a (\mathcal{D}) submartingale

Submartingale Inequalities and Characterization II

- ▶ $\Phi(|Y_T|)$ and $U^{\Lambda, C}(\Phi(|Y_T|))$ are (\mathcal{D}) -supermartingales.
- ▶ any Y s.t. $e^{|Y|} \leq \Phi_t(|Y_T|)$ is a $\mathcal{Q}(\Lambda, C)$ semimartingale if and only if $U^{\Lambda, C}(e^Y)$ and $U^{\Lambda, C}(e^{-Y})$ satisfies submartingale inequalities.
- ▶ Y is said to be in $\mathcal{S}_Q(|\mathcal{Y}_T|, \Lambda, C)$

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Quadratic variation estimates I

Theorem

Let $Y. \in \mathcal{S}_Q(|\eta_T|, \Lambda, C)$, and $\bar{X}_T^{\Lambda, C}(|Y_T|) = e^{C_T} |\eta_T| + \int_0^T e^{C_s} d\Lambda_s$.

- ▶ $\frac{1}{2} \mathbb{E}[\langle M \rangle_{S, T} | \mathcal{F}_S] \leq \Phi_{S, T}(|\eta_T|) \mathbf{1}_{\{S < T\}}$
- ▶ Let $p^\eta = \sup\{p; \mathbb{E}[\exp(p\bar{X}_T^C)] < +\infty\}$. Then $p^\eta \geq 1$ and $\forall p \in [1, p^\eta[$, $\mathbb{E}[\langle M \rangle_T^p] \leq (2p)^p \mathbb{E}[\exp(p\bar{X}_T^C)]$.
- ▶ If for any $S \leq T$, $\Phi_{S, T}(|\eta_T|)$ is bounded, then the martingale M is in BMO.

Sketch proof

- ▶ From Kobylanski, if $v(x) = e^x - 1 - x$, and $V_t^{\Lambda, C}(e^{|Y_t|}) = v(|Y_t|) + \int_0^t v'(|Y_s|)(d\Lambda_s + |Y_s|dC_s)$, then $V_t^{\Lambda, C}(e^{|Y_t|}) - \frac{1}{2} \langle M \rangle_t$ is a (\mathcal{D}) -submartingale.
- ▶ Neveu-Garsia Lemma

Neveu-Garsia Lemma I

The Lemma Let A , a predictable increasing process and U a random variable, positive and integrable. Assume that for $S \leq T$,

$$\mathbb{E}[A_T - A_S \mathbf{1}_{\{0 < S \leq T\}} | \mathcal{F}_S] \leq \mathbb{E}[U \mathbf{1}_{\{S < T\}} | \mathcal{F}_S],$$

- ▶ $\forall r \geq 1, \quad \mathbb{E}[A_T^r] \leq r^r \mathbb{E}[U^r]$
- ▶ More generally, for any **convex** function F s.t.
 $p = \sup_{x>0} (x(\ln F)'(x)) < +\infty,$

$$\mathbb{E}[F(A_T)] \leq \mathbb{E}[F(pU)].$$

Total Variation Estimates The total variation of V , s.t.

$Y_t = Y_0 + M_t - V_t$ satisfies for $1 \leq p < p^\eta$

$$\mathbb{E}[|V|_T^p] \leq (2p)^p \mathbb{E}[\exp(p\bar{X}_T^C)],$$

Strong convergence of martingale parts I

Theorem

Assume the sequence (Y^n) of $\mathcal{S}_Q(|\eta_T|, \Lambda, C)$ -quasimartingales to be a Cauchy sequence for the a.s. uniform convergence, i.e. $\sup_{t \leq T} |Y_t^n - Y_t^{n+p}|$ tends to 0 almost surely when $n \rightarrow \infty$. Different types of convergence hold true for the processes (M^n, V^n) of the decomposition $Y^n = Y_0^n + M^n - V^n$.

Martingales convergence

- ▶ The sequence of martingales (M^n) converges to a continuous martingale M_\cdot in \mathbb{H}^1 .
- ▶ If, for some $p > 1$, $\bar{X}_T^{\Lambda, C}(|\eta_T|) \in \mathbb{L}_{\text{exp}}^p$, the sequence of martingales (M^n) converges to a continuous martingale M_\cdot in \mathbb{H}^{2p} .
- ▶ If $\Phi(|\eta_T|)$ is bounded, the sequence of martingales (M^n) converges to a continuous martingale M_\cdot in the BMO-space.

Theorem

Semimartingale convergence

- ▶ The sequence (V^n) converges uniformly to a finite variation process V satisfying the structure condition $\mathcal{Q}(\Lambda, C)$ at least in \mathbb{L}^1 .
- ▶ The sequence of $\mathcal{S}_Q(|\eta_T|, \Lambda, C)$ -quasimartingales (Y^n) converges to the continuous quadratic $\mathcal{S}_Q(|\eta_T|, \Lambda, C)$ -quasimartingale $Y = Y_0 + M - V$.

Sketch of the proof ($B^{i,j} = |V^i| + |V^j|$)

$$\begin{aligned} \mathbb{E}[\langle M_{\sigma,T}^{i,j} \rangle | \mathcal{F}_\sigma] &\leq \mathbb{E}[\max |Y_{\sigma,T}^{i,j}|^2 \mathbf{1}_{\{\sigma < T\}} | \mathcal{F}_\sigma] + \mathbb{E}\left[\int_\sigma^T \max |Y_{\sigma,s}^{i,j}| dE\right] \\ &\leq \mathbb{E}[(\max |Y_T^{i,j}|^2 + \max |Y_T^{i,j}| B_T^{i,j}) \mathbf{1}_{\{\sigma < T\}} | \mathcal{F}_\sigma] \end{aligned}$$

+ Neveu-Garsia Lemma

Monotone convergence I

Theorem

Let assume the sequence (Y^n) to be a monotone sequence of $\mathcal{S}_Q(|\eta_T|, \Lambda, C)$ -quasimartingales converging almost surely to a process Y . Then, Y is continuous, the convergence is uniform and all properties given in previous Theorem hold true.

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Quadratic BSDEs I

- ▶ An equation $-dY_t = g(t, Y_t, Z_t)dt - Z_t dW_t$, $Y_T = \xi_T$, where W is a d -dimensional BM
- ▶ Quadratic BSDE = quadratic coefficient = $d\mathbb{P} \otimes dt$ -a.s.
 $|g(\cdot, t, y, z)| \leq Q(t, y, z) \equiv 1/\delta|l_t| + c_t|y| + \frac{\delta}{2}|z|^2$,

The "linear" case $|g(\cdot, t, y, z)| \leq |l_t| + c_t|y| + k_t|z|$

- ▶ Main observation

Linear \Rightarrow quadratic

- ▶ $k_t|z| \leq \frac{1}{2}(\frac{1}{\varepsilon}|k_t|^2 + \varepsilon(|z|^2))$

Strongly Quadratic BSDEs I

- ▶ Let g be a coefficient satisfying $g(t, y, z) = f(t, y, z) + \frac{\delta}{2}|z|^2$, where $f(t, y, z) \leq l_t + c_t|y| + k|y|$.
- ▶ Assume that $\bar{X}_T = \exp((\delta + \epsilon)(e_T^C|\xi_T| + \int_0^T e^{C_s}(l_s + \frac{1}{2}k_\epsilon^2)ds)$ is finite,
and let $\phi_t^\epsilon = \rho_{\delta+\epsilon,t}(e_{t,T}^C|\xi_T| + \int_t^T e^{C_{t,s}}(l_s + \frac{1}{2}k_\epsilon^2)ds)$

Then, there exists a maximal solution of the BSDEs satisfying $Y \leq \phi^\epsilon$ obtained from the linear growth BSDEs with generator

$$g_n = f(t, y, z) + \frac{\delta}{2}|z| \inf(|z|, n)$$

The General Existence result I

Theorem

Assume that $\mathbb{E}[\bar{X}_T = \exp(\rho(\delta(e_T^C|\xi_T| + \int_0^T e^{C_s}(I_s ds))ds)]$ is finite, and let $\phi_t = \rho_{\delta,t}(e_{t,T}^C|\xi_T| + \int_t^T e^{C_{t,s}}(I_s ds))$

- ▶ The strongly quadratic growth coefficient g^n defined as : $g^n(t, y, z) = g(t, y, z) \vee (-c_l + ac - \delta n|z| + \frac{\delta}{2}|z|^2)$ are decreasing to g
- ▶ There exists a minimal solution (Y^n, Z^n) dominated by ϕ_t $-dY_t^n = g^n(t, Y_t^n, Z_t^n)dt - Z_t^n dW_t$ and the sequence Y^n is decreasing.
- ▶ There exists a minimal solution (Y, Z) dominated by ϕ to the BSDE, $-dY_t = g(t, Y_t, Z_t)dt - Z_t dW_t$.



Perspectives

- ▶ With Anis and his PhD-student, Quadratic BSDEs with jumps
- ▶ Different extensions and BMO case
- ▶ Optimisation problems

Thanks to Mingyou XU for very stimulating discussions

Thank you for your attention