

# Backward stochastic partial differential equations driven by infinite dimensional martingales and applications

AbdulRahman Al-Hussein

Department of Mathematics, College of Science, Qassim University,  
P. O. Box 6644, Buraydah, Saudi Arabia

E-mail: *alhusseinqu@hotmail.com*

**Workshop on Stochastic Control and Finance**  
**Roscoff, France**  
**March 18-23, 2010**

# Outline

- 1 **Aims**
- 2 **Notations**
  - Example 1
- 3 **Spaces of solutions**
- 4 **Definitions**
- 5 **Assumptions**
- 6 **Existence & Uniqueness Theorem**
- 7 **Proof**
- 8 **Some advantages**
- 9 **An application**
  - Maximum principle for controlled stochastic evolution equations
  - An example

# Aims

- To derive [the existence and uniqueness](#) of the solutions to:

$$(BSPDE) \begin{cases} -dY(t) = (A(t) Y(t) + F(t, Y(t), Z(t)Q^{1/2}(t))) dt \\ \quad -Z(t) dM(t) - dN(t), \\ Y(T) = \xi. \end{cases}$$

- to provide some applications to the [maximum principle](#) for a controlled stochastic evolution system.

# Outline

- 1 Aims
- 2 Notations**
  - Example 1
- 3 Spaces of solutions
- 4 Definitions
- 5 Assumptions
- 6 Existence & Uniqueness Theorem
- 7 Proof
- 8 Some advantages
- 9 An application
  - Maximum principle for controlled stochastic evolution equations
  - An example

- $H$  is a separable Hilbert space.
- $(\Omega, \mathcal{F}, \mathbb{P})$  is a complete probability space equipped with a right continuous filtration  $\{\mathcal{F}_t\}_{t \geq 0}$ .
- $M \in \mathcal{M}_{[0, T]}^{2, c}(H)$ , i.e  $M$  is a continuous square integrable martingale in  $H$ .
- $\langle M \rangle$  is the predictable quadratic variation of  $M$ .
- $\tilde{Q}_M$  is the predictable process taking values in the space  $L_1(H)$ , which is associated with the Doléans measure of  $M \otimes M$ .
- $\langle\langle M \rangle\rangle_t = \int_0^t \tilde{Q}_M(s) d \langle M \rangle_s$ .
- Assume:  $\exists$  a predictable process  $Q$  satisfying  $Q(t, \omega)$  is symmetric, positive definite nuclear operator on  $H$  and

$$\langle\langle M \rangle\rangle_t = \int_0^t Q(s) ds.$$

- $\xi (\xi(\omega) \in H)$  is the terminal value.

- $F : [0, T] \times \Omega \times H \times L_2(H) \rightarrow H$  is  $\mathcal{P} \otimes \mathcal{B}(H) \otimes \mathcal{B}(L_2(H)) / \mathcal{B}(H)$ -measurable.

$L_2(H)$  is the space of Hilbert-Schmidt operators on  $H$ , inner product  $\langle \cdot, \cdot \rangle_2$ , norm  $\| \cdot \|_2$ .

- $A(t, \omega)$  is a predictable unbounded linear operator on  $H$ .
- The stochastic integral  $\int_0^\cdot \Phi(s) dM(s)$  is defined for  $\Phi$  s.t. for  $(\Phi \circ \tilde{Q}_M^{1/2})(t, \omega)(h) \in L_2(H)$ , for every  $h \in H : \Phi \circ \tilde{Q}_M^{1/2}(h)$  is predictable and

$$\mathbb{E} \left[ \int_0^T \|\Phi \circ \tilde{Q}_M^{1/2}\|_2^2 d \langle M \rangle_t \right] < \infty.$$

The space of integrands  $\dashrightarrow \Lambda^2(H; \mathcal{P}, M)$ .

## Example 1

Let  $m$  be a 1-dimensional, continuous, square integrable martingale with respect to  $\{\mathcal{F}_t\}_t$  s.t.  $\langle m \rangle_t = \int_0^t h(s) ds \forall 0 \leq t \leq T$ , some cts  $h : [0, T] \rightarrow (0, \infty)$ .

$M(t) = \beta m(t) (= \int_0^t \beta dm(s))$ , a fixed element  $\beta \neq 0$  of  $H$ .



- $M \in \mathcal{M}^{2,c}(H)$
- $\langle\langle M \rangle\rangle_t = \widetilde{\beta \otimes \beta} \int_0^t h(s) ds$ , where  $\widetilde{\beta \otimes \beta}$  is the identification of  $\beta \otimes \beta$  in  $L_1(H) : (\widetilde{\beta \otimes \beta})(k) = \langle \beta, k \rangle \beta, k \in H$ .
- $\langle M \rangle_t = |\beta|^2 \int_0^t h(s) ds$ .
- $\tilde{Q}_M = \frac{\widetilde{\beta \otimes \beta}}{|\beta|^2}$ .
- Let  $Q(t) = \widetilde{\beta \otimes \beta} h(t) \Rightarrow \langle\langle M \rangle\rangle_t = \int_0^t Q(s) ds$ .
- $Q(\cdot)$  is bounded since  $Q(t) \leq Q := \widetilde{\beta \otimes \beta} \max_{0 \leq t \leq T} h(t)$ .
- $Q^{1/2}(t)(k) = \frac{\langle \beta, k \rangle \beta}{|\beta|} h^{1/2}(t)$ . In particular  $\beta \in Q^{1/2}(t)(H)$ .

# Outline

- 1 Aims
- 2 Notations
  - Example 1
- 3 Spaces of solutions**
- 4 Definitions
- 5 Assumptions
- 6 Existence & Uniqueness Theorem
- 7 Proof
- 8 Some advantages
- 9 An application
  - Maximum principle for controlled stochastic evolution equations
  - An example



- $L^2_{\mathcal{F}}(0, T; H) := \{ \phi : [0, T] \times \Omega \rightarrow H, \text{predictable, } \mathbb{E} [\int_0^T |\phi(t)|_H^2 dt] < \infty \}$ .
- $\mathcal{B}^2(H) := L^2_{\mathcal{F}}(0, T; H) \times \Lambda^2(H; \mathcal{P}, M)$ .

This is a separable Hilbert space, the norm:

$$\begin{aligned} \|(\phi_1, \phi_2)\|_{\mathcal{B}^2(H)} &= \left( \mathbb{E} \left[ \int_0^T |\phi_1(t)|_H^2 dt \right] \right. \\ &\quad \left. + \mathbb{E} \left[ \int_0^T \|\phi_2(t) \tilde{Q}_M^{1/2}(t)\|_2^2 d \langle M \rangle_t \right] \right)^{1/2}. \end{aligned}$$

- $(V, H, V')$  is a rigged Hilbert space:
  - ▷  $V$  is a separable Hilbert space embedded continuously and densely in  $H$ .
  - ▷ By identifying  $H$  with its dual  $\Rightarrow$  get continuous and dense two inclusions:  $V \subseteq H \subseteq V'$ ,  $V'$  is the dual space of  $V$ .

# Outline

- 1 Aims
- 2 Notations
  - Example 1
- 3 Spaces of solutions
- 4 Definitions**
- 5 Assumptions
- 6 Existence & Uniqueness Theorem
- 7 Proof
- 8 Some advantages
- 9 An application
  - Maximum principle for controlled stochastic evolution equations
  - An example

## Definition 1

Two elements  $M$  and  $N$  of  $\mathcal{M}_{[0,T]}^{2,c}(H)$  are **very strongly orthogonal (VSO)** if

$$\mathbb{E} [M(u) \otimes N(u)] = \mathbb{E} [M(0) \otimes N(0)],$$

for all  $[0, T]$ -valued stopping times  $u$ .

In fact:  $M$  and  $N$  are VSO  $\Leftrightarrow \langle\langle M, N \rangle\rangle = 0$ .

## Definition 2

A **solution** of the:

$$(BSPDE) \begin{cases} -dY(t) = (A(t) Y(t) + F(t, Y(t), Z(t)Q^{1/2}(t))) dt \\ \quad -Z(t) dM(t) - dN(t), \quad 0 \leq t \leq T, \\ Y(T) = \xi, \end{cases}$$

is a triple  $(Y, Z, N) \in L^2_{\mathcal{F}}(0, T; V) \times \Lambda^2(H; \mathcal{P}, M) \times \mathcal{M}_{[0, T]}^{2, c}(H)$  s.t.  
 $\forall t \in [0, T]$ :

$$Y(t) = \xi + \int_t^T (A(s) Y(s) + F(s, Y(s), Z(s)Q^{1/2}(s))) ds \\ - \int_t^T Z(s) dM(s) - \int_t^T dN(s),$$

$N(0) = 0$  and  $N$  is VSO to  $M$ .

# Outline

- 1 Aims
- 2 Notations
  - Example 1
- 3 Spaces of solutions
- 4 Definitions
- 5 Assumptions**
- 6 Existence & Uniqueness Theorem
- 7 Proof
- 8 Some advantages
- 9 An application
  - Maximum principle for controlled stochastic evolution equations
  - An example

- (A1)  $F : [0, T] \times \Omega \times H \times L_2(H) \rightarrow H$  is a mapping such that the following properties are verified.
  - (i)  $F$  is  $\mathcal{P} \otimes \mathcal{B}(H) \otimes \mathcal{B}(L_2(H)) / \mathcal{B}(H)$ -measurable.
  - (ii)  $\mathbb{E} [\int_0^T |F(t, 0, 0)|^2 dt] < \infty$ , where  $F(t, 0, 0) = F(t, \omega, 0, 0)$ .
  - (iii)  $\exists k_1 > 0$  such that  $\forall y, y' \in H, \forall z, z' \in L_2(H)$   
 $|F(t, \omega, y, z) - F(t, \omega, y', z')|^2 \leq k_1 (|y - y'|^2 + \|z - z'\|_2^2)$ ,  
uniformly in  $(t, \omega)$ .
- (A2)  $\xi \in L^2(\Omega, \mathcal{F}_T, \mathbb{P}; H)$ .
- (A3) There exists a predictable process  $Q$  satisfying  $Q(t, \omega)$  is symmetric, positive definite nuclear operator on  $H$  and

$$\langle\langle M \rangle\rangle_t = \int_0^t Q(s) ds.$$

- (A4) Every square integrable  $H$ -valued martingale with respect to  $\{\mathcal{F}_t, 0 \leq t \leq T\}$  has a continuous version.
- (A5)  $A(t, \omega)$  is a linear operator on  $H$ ,  $\mathcal{P}$ -measurable, belongs to  $L(V; V')$  uniformly in  $(t, \omega)$  and satisfies the following conditions:

(i)  $A(t, \omega)$  satisfies the coercivity condition in the sense that

$$2 [A(t, \omega) y, y] + \alpha |y|_V^2 \leq \lambda |y|_H^2 \quad \text{a.e. } t \in [0, T], \quad \text{a.s. } \forall y \in V,$$

for some  $\alpha, \lambda > 0$ .

(ii)  $A(t, \omega)$  is uniformly continuous, i.e.  $\exists k_3 \geq 0$  such that for all  $(t, \omega)$

$$|A(t, \omega) y|_{V'} \leq k_3 |y|_V,$$

for every  $y \in V$ .

# Outline

- 1 Aims
- 2 Notations
  - Example 1
- 3 Spaces of solutions
- 4 Definitions
- 5 Assumptions
- 6 Existence & Uniqueness Theorem**
- 7 Proof
- 8 Some advantages
- 9 An application
  - Maximum principle for controlled stochastic evolution equations
  - An example



## Theorem 1

Assume (A1)–(A5). Then there exists a unique solution  $(Y, Z, N)$  of the

$$(BSPDE) \begin{cases} -dY(t) = (A(t) Y(t) + F(t, Y(t), Z(t)Q^{1/2}(t))) dt \\ \quad -Z(t) dM(t) - dN(t), \quad 0 \leq t \leq T, \\ Y(T) = \xi \end{cases}$$

in  $L^2_{\mathcal{F}}(0, T; V) \times \Lambda^2(H; \mathcal{P}, M) \times \mathcal{M}^{2,c}_{[0,T]}(H)$ .

# Outline

- 1 Aims
- 2 Notations
  - Example 1
- 3 Spaces of solutions
- 4 Definitions
- 5 Assumptions
- 6 Existence & Uniqueness Theorem
- 7 Proof**
- 8 Some advantages
- 9 An application
  - Maximum principle for controlled stochastic evolution equations
  - An example

We shall divide the proof of Theorem 1 into different cases:

### Lemma 1

Suppose that  $F \in L^2_{\mathcal{F}}(0, T; H)$  and (A2)–(A5) hold. Then

$$Y(t) = \xi + \int_t^T (A(s) Y(s) + F(s)) ds \\ - \int_t^T Z(s) dM(s) - \int_t^T dN(s)$$

attains a unique solution:

$$(Y, Z, N) \in L^2_{\mathcal{F}}(0, T; V) \times \Lambda^2(H; \mathcal{P}, M) \times \mathcal{M}^{2,c}_{[0,T]}(H).$$

## Proof of Lemma 1

▷ The proof of Lemma 1 is achieved through the method of Galerkin's finite dimensional approximation, e.g. by following Pardoux and Rozovskii.

▷ It can be found in

Al-Hussein, A., Backward stochastic partial differential equations driven by infinite dimensional martingales and applications, Stochastics, Vol. 81, No. 6, 2009, 601-626.

## Lemma 2

Assume that (A2)–(A5) hold and  $F$  satisfies:

- (A1)'  $F : [0, T] \times \Omega \times L_2(H) \rightarrow H$  is a mapping s.t.

(i)  $F$  is  $\mathcal{P} \otimes \mathcal{B}(L_2(H))\mathcal{B}(H)$ -measurable.

(ii)  $\mathbb{E} [\int_0^T |F(t, 0, 0)|^2 dt] < \infty$ .

(iii)  $\exists k_2 > 0$  such that  $\forall z, z' \in L_2(H)$

$$|F(t, \omega, z) - F(t, \omega, z')|^2 \leq k_2 (|y - y'|^2 + \|z - z'\|_2^2),$$

uniformly in  $(t, \omega)$ .

Then there exists a unique solution  $(Y, Z, N)$  of the BSPDE:

## continued Lemma 2

$$\begin{aligned}
 Y(t) = & \xi + \int_t^T (A(s) Y(s) + F(s, Z(s) Q^{1/2}(s))) ds \\
 & - \int_t^T Z(s) dM(s) - \int_t^T dN(s)
 \end{aligned}$$

in  $L^2_{\mathcal{F}}(0, T; V) \times \Lambda^2(H; \mathcal{P}, M) \times \mathcal{M}^{2,c}(H)$ .

**Final step:**

- ▷ We establish the existence of solutions to our original (BSPDE).
- ▷ Let  $Y_0 \equiv 0$ , define recursively using Lemma 2 the BSPDE:

$$\begin{aligned}
 Y_n(t) = & \xi + \int_t^T (A(s) Y_n(s) + F(s, Y_{n-1}(s), Z_n(s) \mathcal{Q}^{1/2}(s))) ds \\
 & - \int_t^T Z_n(s) dM(s) - \int_t^T dN_n(s), \quad 0 \leq t \leq T,
 \end{aligned}$$

for  $n \geq 1$ .

- ▷ The solutions  $(Y_n, Z_n, N_n)$  lie in

$$L^2_{\mathcal{F}}(0, T; V) \times \Lambda^2(H; \mathcal{P}, M) \times \mathcal{M}^{2,c}(H) \text{ for each } n \geq 1.$$

## continued final step:

- ▷ We show  $\{Y_n\}$ ,  $\{Z_n\}$  and  $\{N_n\}$  are Cauchy sequences in  $L^2_{\mathcal{F}}(0, T; V)$ ,  $\Lambda^2(H; \mathcal{P}, M)$  and  $\mathcal{M}^{2,c}(H)$ , respectively.
- ▷ Let  $Y$ ,  $Z$  and  $N$  denote the limits of these sequences.
- ▷ Then we show the very strong orthogonality between  $N$  and  $M$ .
- ▷ Now this convergence together with (A1) and (A5)(ii) allows us to let  $n \rightarrow \infty$  in the previous sequence of BSPDEs to obtain:

$$\begin{aligned}
 Y(t) &= \xi + \int_t^T (A(s) Y(s) + F(s, Y(s), Z(s))) ds \\
 &\quad - \int_t^T Z(s) dM(s) - \int_t^T dN(s), \quad 0 \leq t \leq T.
 \end{aligned}$$

- ▷ Hence  $(Y, Z, N)$  is a solution to (BSPDE).



# Outline

- 1 Aims
- 2 Notations
  - Example 1
- 3 Spaces of solutions
- 4 Definitions
- 5 Assumptions
- 6 Existence & Uniqueness Theorem
- 7 Proof
- 8 Some advantages**
- 9 An application
  - Maximum principle for controlled stochastic evolution equations
  - An example

- These types of BSPDEs seem to be fresh!
- Yong, J. and Zhou, X. Y. [Springer 1999] insist on the condition:  
“  $\{\mathcal{F}_t\}_{t \geq 0}$  is the natural filtration generated by  $W(t)$ , augmented by all the  $\mathbb{P}$ -null sets in  $\mathcal{F}$  ”  
to study the adjoint equation of SDEs.
- Useful for studying the stochastic maximum principle for infinite dimensional controlled stochastic evolution systems.

- These types of BSPDEs seem to be fresh!
- Yong, J. and Zhou, X. Y. [Springer 1999] insist on the condition:  
“  $\{\mathcal{F}_t\}_{t \geq 0}$  is the natural filtration generated by  $W(t)$ , augmented by all the  $\mathbb{P}$ -null sets in  $\mathcal{F}$  ”  
to study the adjoint equation of SDEs.
- Useful for studying the stochastic maximum principle for infinite dimensional controlled stochastic evolution systems.

- These types of BSPDEs seem to be fresh!
- Yong, J. and Zhou, X. Y. [Springer 1999] insist on the condition:  
“  $\{\mathcal{F}_t\}_{t \geq 0}$  is the natural filtration generated by  $W(t)$ , augmented by all the  $\mathbb{P}$ -null sets in  $\mathcal{F}$  ”  
to study the adjoint equation of SDEs.
- Useful for studying the stochastic maximum principle for infinite dimensional controlled stochastic evolution systems.

# Outline

- 1 Aims
- 2 Notations
  - Example 1
- 3 Spaces of solutions
- 4 Definitions
- 5 Assumptions
- 6 Existence & Uniqueness Theorem
- 7 Proof
- 8 Some advantages
- 9 **An application**
  - Maximum principle for controlled stochastic evolution equations
  - An example

Consider the following stochastic evolution equation (SEE):

$$\begin{cases} dX^{\nu(\cdot)}(t) = (A(t)X^{\nu(\cdot)}(t) + \Psi(X^{\nu(\cdot)}(t), \nu(t))) dt + G(X^{\nu(\cdot)}(t)) dM(t), \\ X^{\nu(\cdot)}(0) = x \in H. \end{cases} \quad (1)$$

- $\nu : [0, T] \times \Omega \rightarrow U$  ( $U$  is a sep. Hilbert space) is **admissible** if  $\nu \in L^2_{\mathcal{F}}(0, T; U)$ .

The set of admissible controls  $\mathcal{U}_{ad}$ .

- The **cost functional**:

$$J(x, \nu(\cdot)) := \mathbb{E} \left[ \int_0^T g(X^{\nu(\cdot)}(t), \nu(t)) dt + \phi(X^{\nu(\cdot)}(T)) \right]. \quad (2)$$

- Define

$$J^*(x) := \inf \{ J(x, \nu(\cdot)) : \nu(\cdot) \in \mathcal{U}_{ad} \}. \quad (3)$$

The control problem for this (SEE) is to find a control  $\nu^*(\cdot)$  and the corresponding solution  $X^{\nu^*(\cdot)}$  s.t.

$$J^*(x) = J(x, \nu^*(\cdot)). \quad (4)$$

- $\nu^*(\cdot)$  is an optimal control,
- $X^{\nu^*(\cdot)}$  is an optimal solution,
- $J^*$  is the value function,
- the pair  $(X^{\nu^*(\cdot)}, \nu^*(\cdot))$  is called an optimal pair of the stochastic control problem (1)-(4).

# The adjoint equation

Define the **Hamiltonian**

$$\mathcal{H} : [0, T] \times H \times U \times H \times L_2(H) \rightarrow \mathbb{R},$$

$$\mathcal{H}(t, x, \nu, y, z) := -g(x, \nu) + \langle \Psi(x, \nu), y \rangle + \langle G(x) \mathcal{Q}^{1/2}(t), z \rangle_2. \quad (5)$$

**The adjoint equation:**

$$\begin{cases} -dY^{\nu(\cdot)}(t) = [A^*(t) Y^{\nu(\cdot)}(t) + \nabla_x \mathcal{H}(X^{\nu(\cdot)}(t), \nu(t), Y^{\nu(\cdot)}(t), Z^{\nu(\cdot)}(t))] dt \\ \quad \quad \quad \quad \quad \quad \quad \quad \quad \quad \quad \quad \quad \quad \quad -Z^{\nu(\cdot)}(t) dM(t) - dN^{\nu(\cdot)}(t), \\ Y^{\nu(\cdot)}(T) = -\nabla \phi(X^{\nu(\cdot)}(T)). \end{cases} \quad (6)$$

$A^*(t)$  is the adjoint operator of  $A(t)$ .



## Theorem 2

Given  $\nu^*(\cdot) \in \mathcal{U}_{ad}$  assume  $\exists$  unique solutions  $X^{\nu^*(\cdot)}$ ,  $(Y^{\nu^*(\cdot)}, Z^{\nu^*(\cdot)}, N^{\nu^*(\cdot)})$  of the corresponding SEE (1) and its adjoint equation (6).

Suppose that

(i)  $\Psi$ ,  $G$ ,  $g$  and  $\phi$  are given  $C_b^1$  mappings and  $\phi$  is convex,

(ii)  $\mathcal{H}(t, \cdot, \cdot, Y^{\nu^*(\cdot)}(t), Z^{\nu^*(\cdot)}(t))$  is concave for all  $t \in [0, T]$  - a.s.,

(iii) 
$$\begin{aligned} \mathcal{H}(t, X^{\nu^*(\cdot)}(t), \nu^*(t), Y^{\nu^*(\cdot)}(t), Z^{\nu^*(\cdot)}(t)) \\ = \max_{\nu \in \mathcal{U}} \mathcal{H}(t, X^{\nu^*(\cdot)}(t), \nu, Y^{\nu^*(\cdot)}(t), Z^{\nu^*(\cdot)}(t)) \end{aligned}$$

for a.e.  $t \in [0, T]$  - a.s.

Then  $(X^{\nu^*(\cdot)}, \nu^*(\cdot))$  is an optimal pair for the problem (1)-(4).

## Proof.

The proof of Theorem 2 can be found in the paper:

[Al-Hussein, A., Maximum principle for controlled stochastic evolution equations, Warwick-preprint, 2009.](#)

- ▶ Let  $H = L^2(\mathbb{R}^n)$ ,  $V = \mathbb{H}^1(\mathbb{R}^n)$  and  $V' = \mathbb{H}^{-1}(\mathbb{R}^n)$ .
- ▶  $M$  is the continuous martingale given in Example 1,  $\beta \neq 0$  a fixed element of  $H$ .
- ▶  $U$  is a separable Hilbert space (space of controls).
- ▶ Assume  $\tilde{F} : U \rightarrow H$  is a bounded linear operator.
- ▶ Consider the SEE:

$$\begin{cases} dX(t) = (A X(t) + \tilde{F} \nu(t)) dt + \langle X(t), \beta \rangle dM(t), \\ X(0) = x \in H. \end{cases} \quad (7)$$

$$A = \frac{1}{2} \Delta.$$

- ▷ Given a function  $c : \mathbb{R}^n \rightarrow \mathbb{R}$  of  $H$  assume the cost functional:

$$J(x, \nu(\cdot)) := \mathbb{E} \left[ \int_0^T |\nu(t)|_U^2 dt \right] + \mathbb{E} \left[ \langle c, X(T) \rangle_H \right] \quad (8)$$

and the value function:

$$J^*(x) := \inf \{ J(x, \nu(\cdot)) : \nu \in \mathcal{U}_{ad} \}. \quad (9)$$

- ▷ Take for  $(x, \nu) \in H \times U$ :

$$\Psi(x, \nu) = \tilde{F} \nu, \quad G(x) = \langle x, \beta \rangle,$$

$$g(x, \nu) = |\nu|^2, \quad \phi(x) = \langle c, x \rangle.$$

- ▷ The Hamiltonian is:

$$\mathcal{H} : [0, T] \times H \times U \times H \times L_2(H) \rightarrow \mathbb{R},$$

$$\mathcal{H}(t, x, \nu, y, z) = -|\nu|_U^2 + \langle \tilde{F}\nu, y \rangle + \langle x, \beta \rangle \langle Q^{1/2}(t), z \rangle_2,$$

$$(t, x, \nu, y, z) \in [0, T] \times H \times U \times H \times L_2(H).$$

- ▷ Consider the adjoint BSPDE:

$$\begin{cases} -dY(t) = \left[ \frac{1}{2} \Delta Y(t) + \langle Q^{1/2}(t), Z(t) Q^{1/2}(t) \rangle_2 \beta \right] dt \\ \quad -Z(t) dM(t) - dN(t), \\ Y(T) = -c. \end{cases} \quad (10)$$

## An example

- ▶ The assumptions of Theorem 1 are satisfied for this BSPDE.
- ▶ Consequently it has a unique solution  $(Y, Z, N)$ .
- ▶ Since  $Y(T)$  is non-random we can choose  $Z(t) = 0$  and  $N(t) = 0$  for each  $t \in [0, T]$ . So:

$$\begin{cases} \frac{\partial}{\partial t} Y(t) = -\frac{1}{2} \Delta Y(t) \\ Y(T) = -c. \end{cases}$$

- ▶ Thus  $Y(t) = -S(T-t)c$ , where

$$(S(r)c)(\sigma) = \frac{1}{(2\pi r)^{n/2}} \int_{\mathbb{R}^n} c(x) e^{\left(\frac{-|\sigma-x|^2}{2r}\right)} dx, \quad \sigma \in \mathbb{R}^n, r > 0.$$

- ▶ By uniqueness of solutions of the BSPDE (10) this triple  $(Y, 0, 0)$  is actually its unique solution.

## An example

- ▷ Now for fixed  $(t, x, y, z)$ ,  
 $U \ni \nu \mapsto H(t, x, \nu, y, z) \in \mathbb{R}$  takes its maximum at  $\nu = -\frac{1}{2} \tilde{F}^* y$ .
- ▷ Then we can let

$$\nu^*(t, \omega) = -\frac{1}{2} \tilde{F}^* Y(t, \omega) (\in U), \quad (11)$$

as a candidate for an optimal control.

- ▷ With this choice all the requirements in Theorem 2 are verified.  
 $\Rightarrow$  this candidate (11) is an optimal control for the problem (7)-(9).

- ▷ An optimal solution  $\hat{X}$  is given by the solution of the equation:

$$\begin{cases} d\hat{X}(t) = (A \hat{X}(t) - \frac{1}{2} \tilde{F} \tilde{F}^* Y(t)) dt + \langle \hat{X}(t), \beta \rangle dM(t), \\ \hat{X}(0) = x \in H. \end{cases}$$

- ▷ The value function takes then the formula:

$$J^*(x) = \mathbb{E} \left[ \int_0^T \left| -\frac{1}{2} \tilde{F}^* Y(t, \omega) \right|_U^2 dt \right] + \mathbb{E} \left[ \langle c, \hat{X}(T) \rangle \right].$$