

# An approach to fully coupled FBSDEs via functional differential equations

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# Outline

- 1 Brownian FBSDEs as functional differential equations
- 2 Fully coupled forward–backward stochastic dynamics
- 3 Existence and uniqueness of solutions
- 4 Related discretization algorithms for Brownian FBSDEs

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# Introduction

**Aim:** Introduce a forward approach for a general class of fully coupled FBSDEs

**Result:** System of forward equations where the coefficients depend also on the terminal values of the solution

- Conflict between forward and backward components partly avoided
- Purely probabilistic (random coefficients)
- Allows to treat other types of non–classical forward–backward equations

## Motivating observation

- $(Y_t)_{0 \leq t \leq T}$  a semimartingale on  $(\Omega, \mathcal{F}, (\mathcal{F}_t)_{0 \leq t \leq T}, P)$  with known terminal value  $Y_T = \xi \in L^1(\mathcal{F}_T)$ .
- Doob-Meyer decomposition:

$$Y_t = M_t - V_t,$$

$M$  martingale,  $V$  cont. adapted process of finite variation.

- If  $V_T$  is integrable, then:

$$\begin{aligned} M_t &= M(V, \xi)_t = E[\xi + V_T | \mathcal{F}_t] \quad \forall t \in [0, T], \\ Y_t &= Y(V, \xi)_t = E[\xi + V_T | \mathcal{F}_t] - V_t \quad \forall t \in [0, T]. \end{aligned} \quad (1.1)$$

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## Formally: alternative formulation of Brownian FBSDEs

- Probability space  $(\Omega, \mathcal{F}, P)$  with a  $m$ -dim. BM  $W$   
 $(\mathcal{F}_t)_{0 \leq t \leq T}$  corresponding augmented filtration
- Classical fully coupled FBSDE of the form

$$\begin{cases} dY_t &= -f(t, X_t, Y_t, Z_t)dt + Z_t dW_t, & Y_T = \Phi(X_T), \\ dX_t &= \mu(t, X_t, Y_t, Z_t)dt + \sigma(t, X_t, Y_t)dW_t, & X_0 = x, \end{cases} \quad (1.2)$$

where  $f : \Omega \times [0, T] \times \mathbb{R}^n \times \mathbb{R}^d \times \mathbb{R}^{d \times m} \rightarrow \mathbb{R}^d$ ,

$\mu : \Omega \times [0, T] \times \mathbb{R}^n \times \mathbb{R}^d \times \mathbb{R}^{d \times m} \rightarrow \mathbb{R}^n$ ,

$\sigma : \Omega \times [0, T] \times \mathbb{R}^n \times \mathbb{R}^d \rightarrow \mathbb{R}^{n \times m}$ ,  $\Phi : \Omega \times \mathbb{R}^n \rightarrow \mathbb{R}^d$ .



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## Formally: alternative formulation of Brownian FBSDEs

Define an associated system of functional differential equations:

$$\begin{cases} dV_t = f(t, X_t, Y(V, X)_t, Z(V, X)_t)dt, \\ dX_t = \mu(t, X_t, Y(V, X)_t, Z(V, X)_t)dt + \sigma(t, X_t, Y(V, X)_t)dW_t \end{cases} \quad (1.3)$$

with **initial conditions**  $V_0 = 0$ ,  $X_0 = x$ , where

$$\begin{aligned} M(V, X)_t &:= E[\Phi(X_T) + V_T | \mathcal{F}_t], \\ Y(V, X)_t &:= E[\Phi(X_T) + V_T | \mathcal{F}_t] - V_t, \\ Z(V, X)_t &:= D_t M(V, X)_T = D_t(\Phi(X_T) + V_T) \quad \forall t \in [0, T]. \end{aligned} \quad (1.4)$$

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# Setting

- $(\Omega, \mathcal{F}, P)$  probability space with a  $m$ -dim. BM  $W$ ,  $(\mathcal{F}_t)_{0 \leq t \leq T}$  with usual assumptions
- $\mathcal{C}([0, T], \mathbb{R}^d) := \{V : \Omega \times [0, T] \rightarrow \mathbb{R}^d \mid V \text{ continuous and adapted, } E[\max_j \sup_t |V_t^j|^2] < \infty\}$
- $\mathcal{C}_0([0, T], \mathbb{R}^d) := \mathcal{C}([0, T], \mathbb{R}^d) \cap \{V \mid V_0 = 0\}$
- $\mathcal{M}^2([0, T], \mathbb{R}^d) := \{M : \Omega \times [0, T] \rightarrow \mathbb{R}^d \mid M \text{ square integrable martingale on } [0, T]\}$

# Setting

- $\|V\|_{C[0,T]} := \sqrt{E[\sup_{0 \leq t \leq T} |V_t|^2]}$
- $\mathcal{S}([0, T], \mathbb{R}^d) := \mathcal{C}([0, T], \mathbb{R}^d) \oplus \mathcal{M}^2([0, T], \mathbb{R}^d)$
- $\mathcal{H}^2([0, T], \mathbb{R}^p) := \{Z : \Omega \times [0, T] \rightarrow \mathbb{R}^p \mid Z \text{ predictable, } \|Z\|_{\mathcal{H}^2[0,T]}^2 := E[\int_0^T |Z_t|^2 dt] < \infty\}$

# Fully coupled forward–backward stochastic dynamics

- General filtration  $\Rightarrow$  No martingale representation!  
 $\Rightarrow$  Substitute  $Z$  by  $L(M)$ , where  $L$  nonlinear functional mapping  $\mathcal{M}^2([0, T], \mathbb{R}^d)$  into  $p$ -dim. adapted processes
- This leads us to the following generalization of (1.2):

$$\begin{cases} dY_t = -f(t, X_t, Y_t, L(M)_t)dt + dM_t, & Y_T = \Phi(X_T), \\ dX_t = \mu(t, X_t, Y_t, L(M)_t)dt + \sigma(t, X_t, Y_t)dW_t, & X_0 = x. \end{cases} \quad (2.1)$$

A **solution** to (2.1) is then a triplet of adapted processes  $(X, Y, M)$  satisfying the integral formulation of (2.1) and such that  $M$  is a square-integrable martingale.

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# Fully coupled forward–backward stochastic dynamics

Reduce the problem (2.1) to a system of functional differential equations:

$$\begin{cases} dV_t = f(t, X_t, Y(V, X)_t, L(M(V, X))_t)dt, \\ dX_t = \mu(t, X_t, Y(V, X)_t, L(M(V, X))_t)dt + \sigma(t, X_t, Y(V, X)_t)dW_t, \end{cases} \quad (2.2)$$

with initial conditions  $V_0 = 0$ ,  $X_0 = x$ .

**Then:** if  $(V, X)$  solves (2.2),  $(X, Y(V, X), M(V, X))$  solves (2.1).

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# Local existence and uniqueness

- Derive sufficient conditions on the coefficients and on  $L$  to guarantee existence and uniqueness of solutions
- For short time intervals: existence and uniqueness under weak assumptions on  $L$   
⇒ Possibility to treat other types of functionals  $L$  not fitting in the classical framework

## Local existence and uniqueness

- Derive sufficient conditions on the coefficients and on  $L$  to guarantee existence and uniqueness of solutions
- For short time intervals: existence and uniqueness under weak assumptions on  $L$   
⇒ Possibility to treat other types of functionals  $L$  not fitting in the classical framework

# Assumptions

## Assumption (A1)

The coefficients  $f$ ,  $\mu$ ,  $\sigma$  and  $\Phi$  satisfy Assumption **(A1)** if there exists a constant  $K > 0$  such that:

(A1.1) For any  $(x, y, z) \in \mathbb{R}^n \times \mathbb{R}^d \times \mathbb{R}^p$ ,  $f(\cdot, x, y, z)$ ,  $\mu(\cdot, x, y, z)$  and  $\sigma(\cdot, x, y)$  are  $\mathbb{F}$ -adapted and  $\Phi(\cdot, x)$  is  $\mathbb{F}_T$ -measurable.

(A1.2) For every  $t \in [0, T]$ ,  $(x, y, z), (x', y', z') \in \mathbb{R}^n \times \mathbb{R}^d \times \mathbb{R}^p$ ,

$$|f(t, x, y, z) - f(t, x', y', z')| \leq K(|x - x'| + |y - y'| + |z - z'|),$$

$$|\Phi(x) - \Phi(x')| \leq K|x - x'|,$$

$$|\mu(t, x, y, z) - \mu(t, x, y', z')| \leq K(|y - y'| + |z - z'|),$$

$$|\sigma(t, x, y) - \sigma(t, x', y')|^2 \leq K(|x - x'|^2 + |y - y'|^2).$$

# Assumptions

## Assumption (A1)

(A1.3) For every  $t \in [0, T]$ ,  $(y, z) \in \mathbb{R}^d \times \mathbb{R}^p$ ,  $x, x' \in \mathbb{R}^n$ ,

$$(x - x')^T (\mu(t, x, y, z) - \mu(t, x', y, z)) \leq K|x - x'|^2.$$

(A1.4) For every  $t \in [0, T]$ ,  $(x, y, z) \in \mathbb{R}^n \times \mathbb{R}^d \times \mathbb{R}^p$ ,

$$|f(t, x, y, z)| \leq K(1 + |x| + |y| + |z|),$$

$$|\Phi(x)| \leq K(1 + |x|),$$

$$|\mu(t, x, y, z)| \leq K(1 + |x| + |y| + |z|),$$

$$|\sigma(t, x, y)| \leq K(1 + |x| + |y|).$$

(A1.5) The functions  $u \mapsto \mu(t, u, y, z)$  is continuous for all  $t \in [0, T]$ ,  $(y, z) \in \mathbb{R}^d \times \mathbb{R}^p$ .

# Assumptions

## Assumption (A2)

The functional  $L$  satisfies Assumption **(A2)** if there exists a constant  $K > 0$  such that:

(A2.1)  $L$  maps  $\mathcal{M}^2([0, T], \mathbb{R}^d)$  into  $\mathcal{O}([0, T], \mathbb{R}^p)$ , where  
 $\mathcal{O}([0, T], \mathbb{R}^p) \in \{\mathcal{H}^2([0, T], \mathbb{R}^p), \mathcal{C}([0, T], \mathbb{R}^p)\}$ .

(A2.2)  $L$  is bounded and Lipschitz continuous, i.e.

$$\begin{aligned}\|L(M)\|_{\mathcal{O}([0, T])} &\leq K\|M\|_{\mathcal{C}([0, T])}, \\ \|L(M) - L(M')\|_{\mathcal{O}([0, T])} &\leq K\|M - M'\|_{\mathcal{C}([0, T])} \quad \forall M, M' \in \mathcal{M}^2.\end{aligned}$$

## Examples for $L$

### Example (1)

- $(\mathcal{F}_t)_{0 \leq t \leq T}$  augmented filtration generated by  $W$
- Choose  $\mathcal{O}([0, T], \mathbb{R}^p) = \mathcal{H}^2([0, T], \mathbb{R}^{d \times m})$
- $L : \mathcal{M}^2([0, T], \mathbb{R}^d) \rightarrow \mathcal{H}^2([0, T], \mathbb{R}^{d \times m})$  defined via the Itô representation theorem, i.e.

$$M_t^i = E[M_t^i] + \sum_{j=1}^m \int_0^t L(M)_s^{i,j} dW_s^j, \quad i = 1, \dots, d.$$

- Classical fully coupled FBSDEs ( $L(M(X, V)) = Z(X, V)$ )



## Examples for $L$

### Example (2)

- $(\mathcal{F}_t)_{0 \leq t \leq T}$  with usual assumptions
- Choose  $\mathcal{O}([0, T], \mathbb{R}^p) = \mathcal{H}^2([0, T], \mathbb{R}^{d \times m})$
- $L : \mathcal{M}^2([0, T], \mathbb{R}^d) \rightarrow \mathcal{H}^2([0, T], \mathbb{R}^{d \times m})$  given by the integrand process in the orthogonal decomposition w.r.t.  $W$ , i.e.

$$M_t^i = E[M_t^i] + \sum_{j=1}^m \int_0^t L(M)_s^{ij} dW_s^j + (M')_t^i, \quad i = 1, \dots, d.$$

## Examples for $L$

### Example (3)

- $(\mathcal{F}_t)_{0 \leq t \leq T}$  quasi-left continuous
- For  $M \in \mathcal{M}^2([0, T], \mathbb{R})$  consider the decomposition

$$M = M^c + M^d$$

$M^c$  continuous martingale null at 0,  $M^d$  purely discontinuous martingale

- Choose  $\mathcal{O}([0, T], \mathbb{R}^p) = \mathcal{C}([0, T], \mathbb{R}^d)$ .  
 $L : \mathcal{M}^2([0, T], \mathbb{R}^d) \rightarrow \mathcal{C}([0, T], \mathbb{R}^d)$  defined by

$$L(M)_t^i := \sqrt{\langle (M^c)^i, (M^c)^i \rangle_t}, \quad i = 1, \dots, d.$$

## Existence of local solutions

### Theorem

*Under the assumptions **(A1)** and **(A2)** there is a constant  $\tau_K$  so that, for  $T \leq \tau_K$ , (2.2) admits a unique solution  $(X, V)$  satisfying*

$$\|X\|_{C[0, T]} + \|V\|_{C[0, T]} < \infty.$$

*Moreover, the solution processes  $V$  and  $X$  are continuous.*

## Sketch of proof

### Sketch of proof

Define the mapping  $\mathbb{L} : \mathcal{C}([0, T], \mathbb{R}^n) \times \mathcal{C}([0, T], \mathbb{R}^d) \rightarrow \mathcal{C}([0, T], \mathbb{R}^n) \times \mathcal{C}([0, T], \mathbb{R}^d)$  by  $\mathbb{L}(X, V) := (\tilde{X}, \tilde{V})$ , where  $\tilde{X}$  solution of the forward SDE

$$\begin{cases} \tilde{X}_0 = x, \\ d\tilde{X}_t = \mu(t, \tilde{X}_t, Y(V, X)_t, L(M(V, X))_t)dt + \sigma(t, \tilde{X}_t, Y(V, X)_t)dW_t, \end{cases}$$

whereas  $\tilde{V}$  is explicitly given by

$$\tilde{V}_t = \int_0^t f(s, \tilde{X}_s, Y(V, X)_s, L(M(V, X))_s)ds.$$

$(X, V)$  solves (2.2) if and only if it is a fixed point of  $\mathbb{L}$ .

## Global solution

- Extension of the local solutions to global ones: still work in progress
- The study of the simple decoupled case suggests that additional assumptions on  $L$  are needed!
- For  $[T_2, T_1] \subset [0, T]$ , define the restriction  $L_{[T_2, T_1]}$  from  $\mathcal{M}^2([T_2, T_1], \mathbb{R}^d)$  to  $\mathcal{O}([T_2, T_1], \mathbb{R}^m)$  by

$$L_{[T_2, T_1]}(N)_t := L(\tilde{N})_t, \quad N \in \mathcal{M}^2([T_2, T_1], \mathbb{R}^d),$$

where  $\tilde{N}_t := E[N_{T_1} | \mathcal{F}_t]$ ,  $t \in [0, T]$ , is the extension of  $N$  to  $\mathcal{M}^2([0, T], \mathbb{R}^d)$ .

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## Global solution

### Assumption (A2')

We say that  $L$  satisfies **(A2')** if it satisfies **(A2)** as well as  
**(A2.3)** (Local-in-time property) For  $0 \leq T_2 < T_1 \leq T$  and  
 $M \in \mathcal{M}^2([0, T], \mathbb{R}^d)$ ,

$$L(M) = L_{[T_2, T_1]}(\widehat{M}) \text{ on } (T_2, T_1), \quad \text{where } \widehat{M} = M|_{[T_2, T_1]}.$$

**(A2.4)** (Differential property) For  $0 \leq T_2 < T_1 \leq T$  and  
 $N \in \mathcal{M}^2([T_2, T_1], \mathbb{R}^d)$ ,

$$L_{[T_2, T_1]}(N - N_{T_2}) = L_{[T_2, T_1]}(N) \text{ on } (T_2, T_1).$$



## Global solution

- **Main idea:** Derive some uniform estimates for the solution over short time intervals, extend the solution to any time interval while still keeping that estimate.
- Well known from classical theory ([Delarue], [Zhang]): additional assumptions on the coefficients are needed.
- It can be proven that, under the assumption (A2') on  $L$  and under the same assumptions on the coefficients as in [Delarue] or [Zhang], the system (2.2) has a unique solution on any time interval

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## Brownian FBSDEs

The functional differential equation approach and the related contraction mapping opens the door to a new class of discretization algorithms.

Assume we have a classical FBSDE in a Brownian filtration:

$$\begin{cases} dY_t &= -f(t, X_t, Y_t, Z_t)dt + Z_t dW_t, & Y_T = \Phi(X_T), \\ dX_t &= \mu(t, X_t, Y_t, Z_t)dt + \sigma(t, X_t, Y_t) dW_t, & X_0 = x. \end{cases}$$

$$\Leftrightarrow \begin{cases} dV_t &= f(t, X_t, Y(V, X)_t, Z(V, X)_t)dt, \\ dX_t &= \mu(t, X_t, Y(V, X)_t, Z(V, X)_t)dt + \sigma(t, X_t, Y(V, X)_t) dW_t. \end{cases}$$

## Numerical approximation

$\pi = (t_0, \dots, t_N)$  partition of  $[0, T]$ . For  $p \in \mathbb{N}$ , define  $V^{\pi,p}$  and  $X^{\pi,p}$  recursively on  $\pi$  by  $V^{\pi,0} \equiv 0$ ,  $X^{\pi,0} \equiv x$  and

$$X_0^{\pi,p+1} = x, \quad V_0^{\pi,p+1} = 0,$$

$$X_{t_{i+1}}^{\pi,p+1} = X_{t_i}^{\pi,p+1} + \mu(t_i, X_{t_i}^{\pi,p+1}, Y(V^{\pi,p}, X^{\pi,p})_{t_i}, Z(V^{\pi,p}, X^{\pi,p})_{t_i})\Delta t_i \\ + \sigma(t_i, X_{t_i}^{\pi,p+1}, Y^{\pi,p}(V, X)_{t_i})(\Delta W_{t_i})^T,$$

$$V_{t_{i+1}}^{\pi,p+1} = V_{t_i}^{\pi,p+1} + f(t_i, X_{t_i}^{\pi,p+1}, Y(V^{\pi,p}, X^{\pi,p})_{t_i}, Z(V^{\pi,p}, X^{\pi,p})_{t_i})\Delta t_i$$

for  $i = 0, \dots, N-1$  and  $p \geq 1$ , where

$$Y(V^{\pi,p}, X^{\pi,p})_{t_i} = E[\Phi(X_T^{\pi,p}) + V_T^{\pi,p} | \mathcal{F}_{t_i}] - V_{t_i}^{\pi,p},$$

$$Z(V^{\pi,p}, X^{\pi,p})_{t_i} = \frac{1}{\Delta t_i} E \left[ Y(V^{\pi,p}, X^{\pi,p})_{t_{i+1}} (\Delta W_{t_i})^T | \mathcal{F}_{t_i} \right].$$

# Numerical approximation

- Motivated by the continuous time results
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- Conjecture: the algorithm converges to the true solution of the FBSDE

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## The decoupled case

The convergence can easily be proved in the decoupled case:

### Theorem

*Assume that  $f$ ,  $\mu$ ,  $\sigma$  and  $\Phi$  are Lipschitz in the space variables and  $1/2$ -Hölder in the time variable. Then there is a constant  $C$ , depending only on the Lipschitz constants involved and the dimension of the problem, such that*

$$\sup_{0 \leq t \leq T} E[|V_t - V_t^{p,\pi}|^2] + \sup_{0 \leq t \leq T} E[|X_t - X_t^{p,\pi}|^2] \leq C \left( |\pi| + \left( \frac{1}{2} + C|\pi| \right)^p \right).$$

## Some references



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Thank you for your attention!