

# Quasilinear parabolic systems and FBSDEs with quadratic growth

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Thanks to Gordan Žitković for many helpful discussions!

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- 1 Introduction
- 2 Main results
- 3 Ideas of the proof

# The FBSDE

We consider the FBSDE

$$\begin{cases} dX_t = b(t, X_t, Y_t, Z_t)dt + \sigma(t, X_t, Y_t)dB_t, \\ dY_t = -f(t, X_t, Y_t, Z_t)dt + Z_t dB_t, \\ Y_T = g(X_T), \quad X_0 = x_0. \end{cases}$$

**Probabilistic setup:**  $(\Omega, \mathcal{F}, \mathbb{P})$ ,  $d$ -dimensional Brownian motion with filtration  $\mathbb{F}$

**Data:** 4 deterministic functions

- drift  $b = b(t, x, y, z) : [0, T] \times \mathbb{R}^d \times \mathbb{R}^n \times (\mathbb{R}^d)^n \rightarrow \mathbb{R}^d$
- volatility  $\sigma = \sigma(t, x, y) : [0, T] \times \mathbb{R}^d \times \mathbb{R}^n \rightarrow \mathbb{R}^{d \times d}$
- driver  $f = f(t, x, y, z) : [0, T] \times \mathbb{R}^d \times \mathbb{R}^n \times (\mathbb{R}^d)^n \rightarrow \mathbb{R}^n$
- terminal condition  $g = g(x) : \mathbb{R}^d \rightarrow \mathbb{R}^n$

**Unknowns:** adapted processes  $X$ ,  $Y$ ,  $Z$  taking values in  $\mathbb{R}^d$ ,  $\mathbb{R}^n$ , and  $(\mathbb{R}^d)^n$ .

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# Features of interest

We are interested in the situation where

- 1  $n > 1$ , i.e.  $Y$  is multi-dimensional
- 2  $f$  has quadratic growth:  $|f(t, x, y, z)| \leq C(1 + |z|^2)$
- 3 the equations are *strongly coupled*, in the sense that  $\sigma = \sigma(t, x, y)$  depends on  $y$  (but  $\sigma$  non-degenerate)

These three issues have received considerable attention... we highlight

- Kobylanski (2000) handles feature 2, but not 1 or 3
- Delarue (2002) handles features 1 and 3 simultaneously, but not 2
- Xing and Žitković (2018) handles 1 and 2 simultaneously, but not 3

**Goal:** Generalize the works above by establishing well-posedness when 1,2,3 are all present.

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# The PDE

By the “4-step scheme” (Ma, Protter, and Yong 1994), solving the FBSDE boils down to solving

$$\begin{cases} \partial_t u^i + \text{tr}(a(t, x, u) D^2 u^i) + f^i(t, x, u, Du) = 0, & (t, x) \in [0, T] \times \mathbb{R}^d, \\ u^i(T, x) = g^i(x) & x \in \mathbb{R}^d. \end{cases}$$

**Data:** 3 functions

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# What can go wrong?

Assume  $\sigma$  and  $g$  are nice. What conditions on the quadratic driver  $f$  guarantee a smooth solution?

**Good news:** can expect a smooth solution on  $[T - \epsilon, T]$  (small-time well-posedness)

**Bad news:** global existence/uniqueness may fail, because

- $u$  may blow up in finite time (blow-up)
- even if  $u$  stays bounded,  $Du$  may blow up (gradient blow-up)

**More good news:** if we manage to prove a gradient estimate (an a-priori estimate on  $\|Du\|_{L^\infty}$ ), then

- can be bootstrapped to higher regularity, in particular estimates on  $\|u\|_{C^{2,\alpha}}$
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# The key question

Based on the previous slide, we know that understanding the PDE system boils down to understanding the following.

**The key question:** what conditions on  $f$  will guarantee an a-priori estimate of  $\|Du\|_{L^\infty}$ ?

Break this up into three questions:

- 1 when can we get an estimate on  $\|u\|_{L^\infty}$ ?
- 2 when does bound on  $\|u\|_{L^\infty}$  imply bound on  $\|u\|_{C^\alpha}$ ?
- 3 when does bound on  $\|u\|_{C^\alpha}$  imply bound on  $\|Du\|_{L^\infty}$ ?

We focus in this talk on the answers to 2 and 3.

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# Some answers

Here are the two main structural conditions.

$$|f^i(t, x, u, p)| \leq C(1 + |p^i||p| + \sum_{j < i} |p^j|^2 + |p|^{2-\epsilon}) \quad (H_{BF})$$

$$\begin{cases} |f(t, x, u, p) - f(t, x', u', p)| \leq C(1 + |p|^2)(|x - x'| + |u - u'|), \text{ and} \\ |f(t, x, u, p) - f(t, x, u, p')| \leq C(1 + |p| + |p'|)|p - p'| \end{cases} \quad (H_{Reg})$$

Theorem (J. 2022)

*Under  $H_{BF}$ , an estimate on  $\|u\|_{L^\infty}$  implies an estimate on  $\|u\|_{C^\alpha}$ .*

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## Theorem (J. 2022)

*Assume that that all data is jointly continuous, and*

- ①  $\sigma = \sigma(t, x, y)$  is non-degenerate and Lipschitz in  $(x, y)$
- ②  $g = g(x)$  bounded and Lipschitz
- ③  $b = b(t, x, y, z)$  Lipschitz in  $(x, y, z)$ ,  $|b(t, x, y, z)| \leq C(1 + |y| + |z|)$
- ④  $f$  satisfies  $H_{BF}$  and  $H_{Reg}$ , and  $H_{AB}$  (a technical condition to get  $\|u\|_{L^\infty} < \infty$ )

*Then there is a solution to the FBSDE*

$$\begin{cases} dX_t = b(t, X_t, Y_t, Z_t)dt + \sigma(t, X_t, Y_t)dB_t, \\ dY_t = -f(t, X_t, Y_t, Z_t)dt + Z_tdB_t, \\ Y_T = g(X_T), \quad X_0 = x_0. \end{cases}$$

Our results can be viewed as...

- a generalization of the results of Delarue (2002 and 2003) to the quadratic case
  - Delarue (2002) gives existence for FBSDEs with Lipschitz data
  - Delarue (2003) gives probabilistic approach to Hölder and gradient estimates in the Lipschitz case
- a generalization of the results of Bensoussan and Frehse (2002), Xing and Žitković (2018), and Harter and Richou (2019) to the case  $\sigma = \sigma(t, x, y)$  (versus  $\sigma = \sigma(t, x)$ )
  - Bensoussan and Frehse obtain Hölder and Sobolev estimates in a bounded domain via PDE arguments
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# The big picture

## The Hölder estimate

- The basic idea is the same as the one used by Delarue in Lipschitz setting - combine Krylov Safonov estimates with BMO-martingale theory
- Execution is different - concept of *sliceability* is used to deal with quadratic growth

## The gradient estimate

- Again, sliceability is key
- Hölder estimate implies a-priori sliceability of  $Z$
- Probabilistic representation of  $Du$  via linear BSDE with sliceable coefficients (thanks to sliceability of  $Z$ !)
- Conclude using results from J. and Žitković 2021 (see also Delbaen and Tang 2008 and Harter and Richou 2019)

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From here on, we suppose we have a nice solution  $u$  to the PDE system, and we define for  $t_0, x_0 \in [0, T] \times \mathbb{R}^d$  the triple  $(X^{t_0, x_0}, Y^{t_0, x_0}, Z^{t_0, x_0})$  by

$$\begin{aligned} X_t^{t_0, x_0} &= x_0 + \int_{t_0}^t \sigma(s, X_s^{t_0, x_0}, u(s, X_s^{t_0, x_0})) dB_s, \quad t_0 \leq t \leq T \\ Y^{t_0, x_0} &= u(\cdot, X^{t_0, x_0}), \quad Z^{t_0, x_0} = \sigma(\cdot, X^{t, x}, Y^{t, x}) Du(\cdot, X^{t, x}). \end{aligned}$$

If  $(t_0, x_0)$  is not important, we just write  $(X, Y, Z)$ . Note that

$$\begin{cases} dX_t = \sigma(t, X_t, Y_t) dB_t, \\ dY_t = -f(t, X_t, Y_t, \sigma^{-1}(t, X_t, Y_t) Z_t) dt + Z_t dB_t. \end{cases}$$

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## Notation:

$$\|Z\|_{\text{bmo}}^2 = \sup_{\tau} \|\mathbb{E}[\int_{\tau}^T |Z|^2 dt | \mathcal{F}_{\tau}]\|_{L^{\infty}}, \quad \|\alpha\|_{\text{bmo}^{1/2}} = \|\sqrt{|\alpha|}\|_{\text{bmo}}^2$$

$Z$  is **sliceable** if  $\|Z1_{[t-\delta, t]}\|_{\text{bmo}}$  is small for  $\delta$  small.

## Definition

A **c-Lyapunov pair**  $(h, k)$  is a smooth function  $h = h(y) : \mathbb{R}^n \rightarrow \mathbb{R}$  and a constant  $k$  such that  $h(0) = 0$ ,  $Dh(0) = 0$ , and for  $|y| \leq c$ ,

$$\frac{1}{2} \sum_{i,j=1}^n (D^2 h(y))_{ij} z^i \cdot z^j - Dh(y) \cdot f(t, x, u, \sigma^{-1}(t, x, u)z) \geq |z|^2 - k.$$

The point is that  $(h, k)$  is a  $c$ -Lyapunov function, then

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# The connection between Hölder regularity and sliceability

## Lemma (Xing and Žitković)

*Under  $H_{BF}$ , for any  $c > 0$  there exists a  $c$ -Lyapunov pair  $(h, k)$ .*

Now suppose we have a  $c$ -Lyapunov pair  $(h, k)$  and  $|Y| \leq c$ . Then for  $t - \delta \leq \tau \leq t$ ...

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Thus under  $H_{BF}$ ,

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**Sliceability implies Hölder:** Suppose we have a bounded solution  $v$  to

$$\partial_t v + \operatorname{tr}(a D^2 u) + b \cdot Du + k = 0, \quad (t, x) \in (0, T) \times \mathbb{R}^d$$

Krylov-Safonov estimates show that

$$a = \frac{1}{2} \sigma \sigma^T \text{ bounded and elliptic \& } b, k \text{ bounded} \implies v \text{ is Hölder,}$$

at least away from  $t = T$ .

Ideas from Delarue (2003) show that the same argument works when

$$\sup_{(t,x)} \|b(\cdot, X^{t,x})\|_{\text{bmo}} < \infty$$

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Because the arguments from previous slide don't use regularity of  $\sigma$ , they apply to quasi-linear setting. Under  $H_{BF}$ , we can write

$$\partial_t u^1 + \operatorname{tr}(a D^2 u^i) + \tilde{b} \cdot Du^1 + \tilde{f} = 0,$$

where

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Then since  $Z^{(t,x)} = \sigma Du(\cdot, X^{(t,x)})$ , we find

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which in turn implies  $Z^1$  is sliceable. It turns out similar reasoning lets us show

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which lets us prove by induction that  $u$  is Hölder. The full chain of reasoning is

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## Main estimates:

- ① under  $H_{\text{BF}}$ , bound on  $\|u\|_{L^\infty} \implies$  bound on  $\|u\|_{C^\alpha}$
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## Takeaways:

- ① sliceability is a useful concept in regularity theory, in particular...
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