

# Numerical methods for BSDEs and nonlinear PDEs

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# FBSDE

Let  $W_t$  be a Wiener process, defined on a probability space  $(\Omega, \mathcal{F}, P)$ ,  $\mathcal{F}_t$  be its natural filtration. Consider a FBSDE

$$\begin{cases} X_t = x + \int_0^t b(s, X_s, Y_s) ds + \int_0^t \sigma(s, X_s, Y_s) dW_s, \\ Y_t = g(X_T) + \int_t^T f(s, X_s, Y_s, Z_s) ds - \int_t^T Z_s dW_s. \end{cases} \quad (1)$$

$$Y_T = g(X_T) \in \mathbf{L}^2(\mathcal{F}_t).$$

If  $b, \sigma, f, g$  are deterministic and Lipschitz continuous functions, there exist  $\mathcal{F}_t$ -adapted processes  $(X_t, Y_t, Z_t)$  satisfying (1), such that

$$E \left[ \sup_{0 \leq t \leq T} |Y_t|^2 + \int_0^T |Z_s|^2 ds \right] < \infty.$$

# Nonlinear PDE

If  $u(t, x)$  is a classical solution to

$$\begin{cases} u_t + \frac{1}{2} \text{trace}(\sigma\sigma^*(t, x, u)u_{xx}) + \\ \quad + u_x b(t, x, u) + f(t, x, u, u_x\sigma(t, x, u)) = 0, \\ u(T, x) = g(x). \end{cases} \quad (2)$$

then  $Y_t = u(t, X_t)$ ,  $Z_t = v(t, X_t) \stackrel{\triangle}{=} u_x(t, X_t)\sigma(t, X_t, u(t, X_t))$ ,  
solve (1). At the other hand one can use (1) to construct less  
regular solution to (2).

# Discretization

A natural time discretization of equation (1) is

$$\begin{cases} X_{i+1}^n \stackrel{\Delta}{=} X_i^n + b(t_i, X_i^n, Y_i^n)h + \sigma(t_i, X_i^n, Y_i^n)\Delta W_{i+1}, \\ Z_i^n \stackrel{\Delta}{=} \frac{1}{h}E_{t_i}\{Y_{i+1}^n\Delta W_{i+1}\}, \\ Y_i^n \stackrel{\Delta}{=} E_{t_i}\{Y_{i+1}^n + f(t_i, X_i^n, Y_i^n, \hat{Z}_i^n)h\}. \end{cases} \quad (3)$$

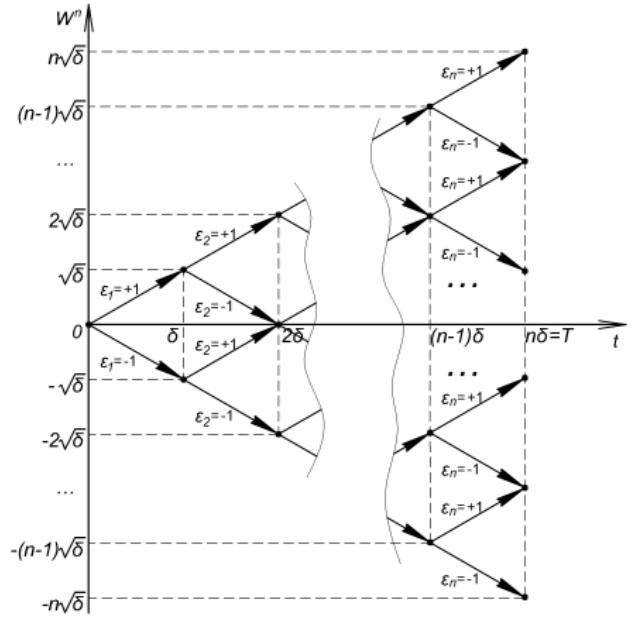
Here  $X_0^n \stackrel{\Delta}{=} x$ ,  $Y_n^n \stackrel{\Delta}{=} g(X_n^n)$ ,  $h \stackrel{\Delta}{=} \frac{T}{n}$  and  $t_i \stackrel{\Delta}{=} ih$ ,  $i = 0, 1, \dots, n$ , and  $\Delta W_{i+1} \stackrel{\Delta}{=} W_{t_{i+1}} - W_{t_i}$ .  $E_t$  denotes the conditional expectation  $E\{\cdot | \mathcal{F}_t\}$ .

# $W_t$ simulation

Set

$$W_{t_i}^n := \sqrt{h} \sum_{j=1}^i \varepsilon_j^n,$$

where  $\{\varepsilon_j^n\}_{j=1}^n$  are  
 $\{1, -1\}$ -valued i.i.d. with  
 $P\{\varepsilon_j^n = 1\} = P\{\varepsilon_j^n = -1\} = 0.5.$



## Numerical examples

Consider a simple case

$$dX_t = dW_t,$$

$$dY_t = (Y_t \cdot Z_t)dt + Z_t dW_t, \quad Y_T = \cos(W_T),$$

$y(t) = U(t, x)$  solves the Cauchy problem for the Burgers equation

$$\frac{\partial U}{\partial t} + U \frac{\partial U}{\partial x} = \frac{1}{2} \frac{\partial^2 U}{\partial x^2}, \quad U(T, x) = \cos(x).$$

## Explicit solution of Burgers equation

$$U(t, x) = \frac{\int\limits_{-\infty}^{\infty} \frac{(x-\theta)}{\tau} \exp(-G(\tau, x, \theta)) d\theta}{\int\limits_{-\infty}^{\infty} \exp(-G(\tau, x, \theta)) d\theta},$$

where

$$G(\tau, x, \theta) = \frac{(x - \theta)^2}{2\tau} + \int\limits_0^{\theta} \varphi(y) dy, \quad \varphi(y) = U(T, y), \quad \tau = T - t.$$

Explicit solution give us  $Y_0 = U(0, 0) = 0.5564803023$ , and (3) with different value  $n$  give

| $n$     | 100     | 500     | 2000    | 4000    |
|---------|---------|---------|---------|---------|
| $Y_0^n$ | 0.55588 | 0.55636 | 0.55645 | 0.55647 |

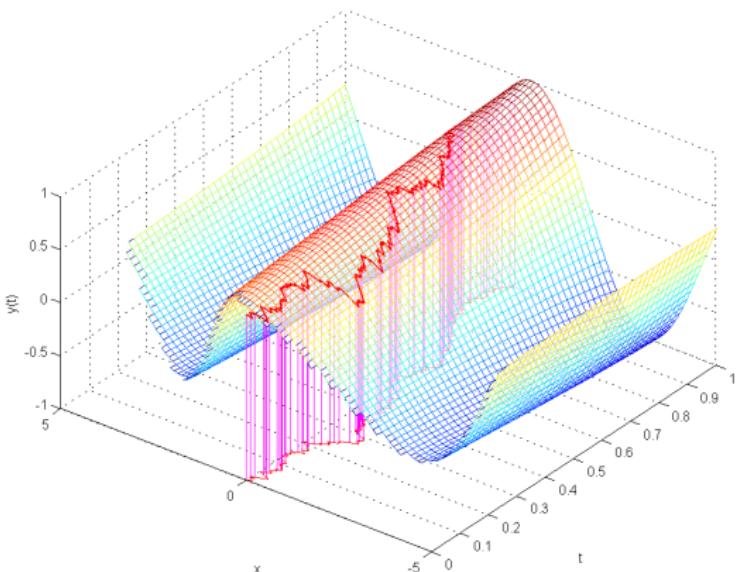


Figure: *The solution surface with one trajectory,  $n = 500$*

## Reference

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-  Bender, C. and Zhang, J. (2008). Time discretization and markovian iteration for coupled FBSDEs. *Ann. Appl. Probab.* **18** 143 – 144.