

# Nonlinear stochastic ordinary and partial differential equations: regularity properties and numerical approximations

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Joint works with

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Larisa Yaroslavtseva (University of Passau, Germany)

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Campus les cordeliers, Paris, France

Thursday, July 7th, 2016

## Consider

- $d, m \in \mathbb{N}$ , measurable  $D \subseteq \mathbb{R}^d$ ,  $\mu: D \rightarrow \mathbb{R}^d$ ,  $\sigma: D \rightarrow \mathbb{R}^{d \times m}$ ,
- $\xi \in D$ ,  $T > 0$ , stochastic basis  $(\Omega, \mathcal{F}, \mathbb{P}, (\mathcal{F}_t)_{t \geq 0})$ ,
- $(\mathcal{F}_t)_{t \geq 0}$ -Wiener process  $W: [0, \infty) \times \Omega \rightarrow \mathbb{R}^m$ ,
- a solution process  $X: [0, T] \times \Omega \rightarrow D$  of

$$dX_t = \mu(X_t) dt + \sigma(X_t) dW_t, \quad t \in [0, T], \quad X_0 = \xi.$$

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**Heston model** Consider  $d = m = 2$ ,  $\alpha, \gamma \in \mathbb{R}$ ,  $\beta, \delta, X_0^{(1)}, X_0^{(2)} > 0$ ,  $\rho \in [-1, 1]$ :

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Remarks:

- $X^{(2)}$  is called **Cox-Ingersoll-Ross (CIR) process**.
- It holds  $\frac{2\delta}{\beta^2} \geq 1$  if and only if it holds  $\mathbb{P}$ -a.s. that  $\forall t \in [0, T]: X_t^{(2)} > 0$ .



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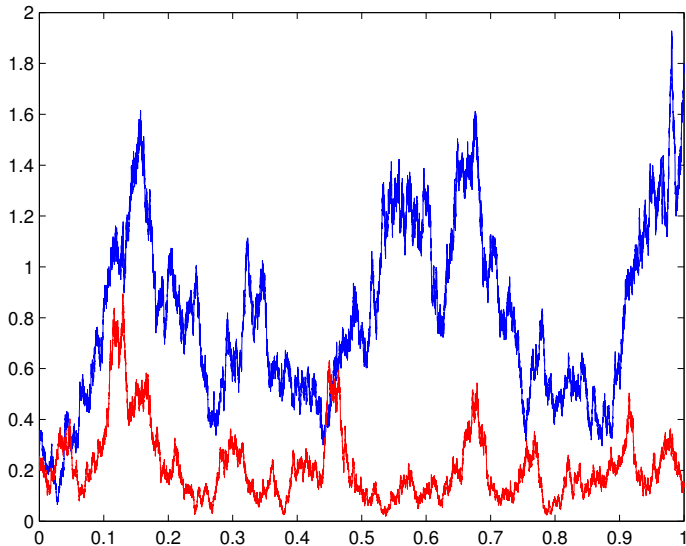
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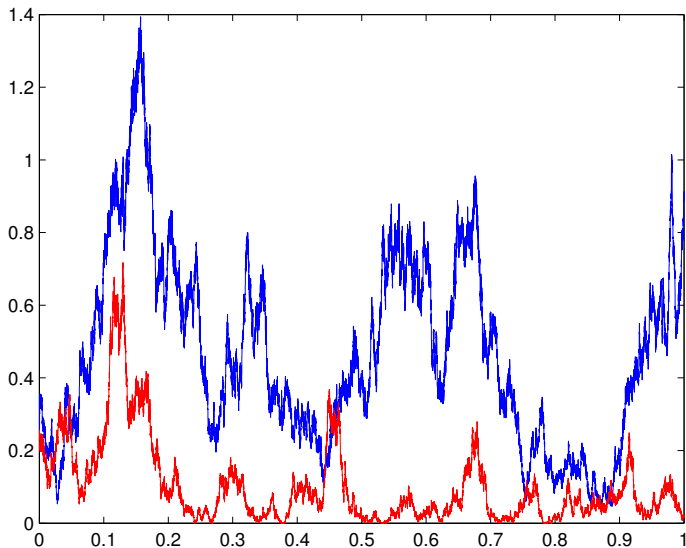
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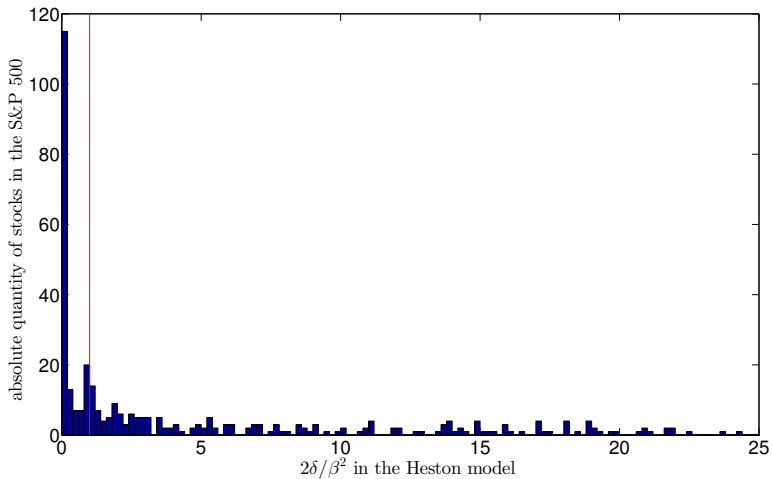
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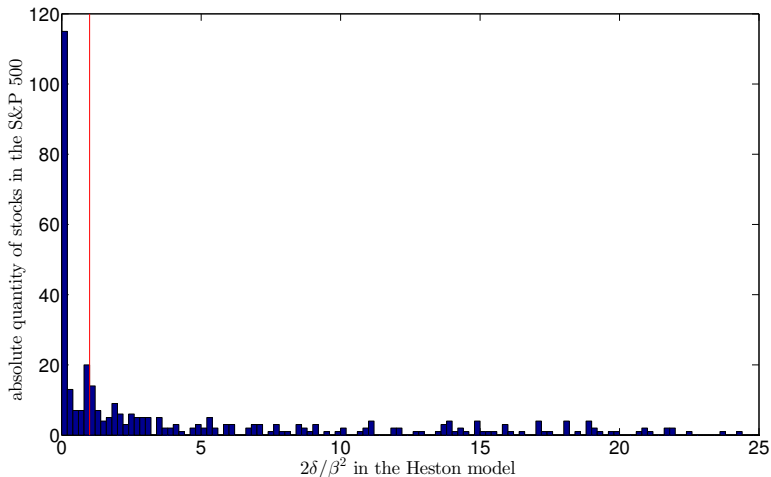
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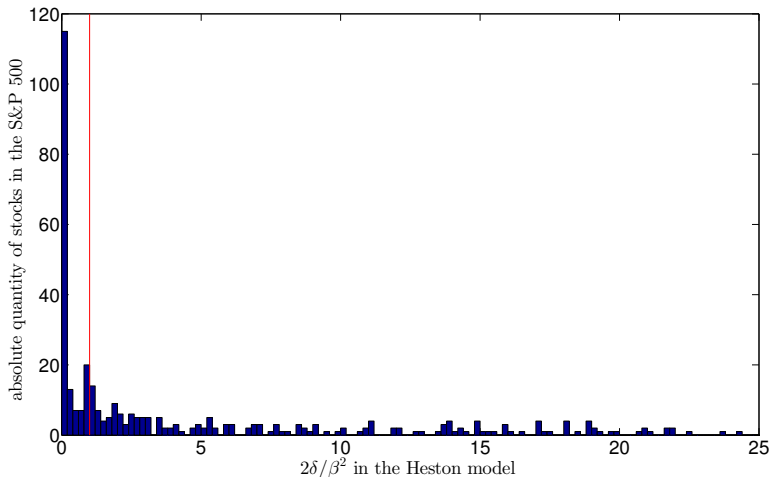
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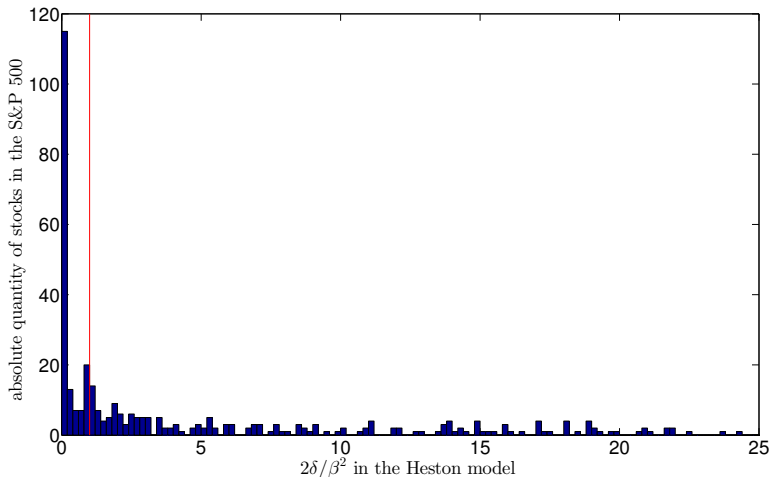
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Let  $T \in (0, \infty)$ . Then there exist *infinitely often differentiable* and *globally bounded* functions  $\mu, \sigma: \mathbb{R}^4 \rightarrow \mathbb{R}^4$  such that  $Y^N: \{0, 1, \dots, N\} \times \Omega \rightarrow \mathbb{R}^4$ ,  $N \in \mathbb{N}$ , with  $\forall N \in \mathbb{N}, n \in \{0, 1, \dots, N-1\}: Y_0^N = X_0$  and

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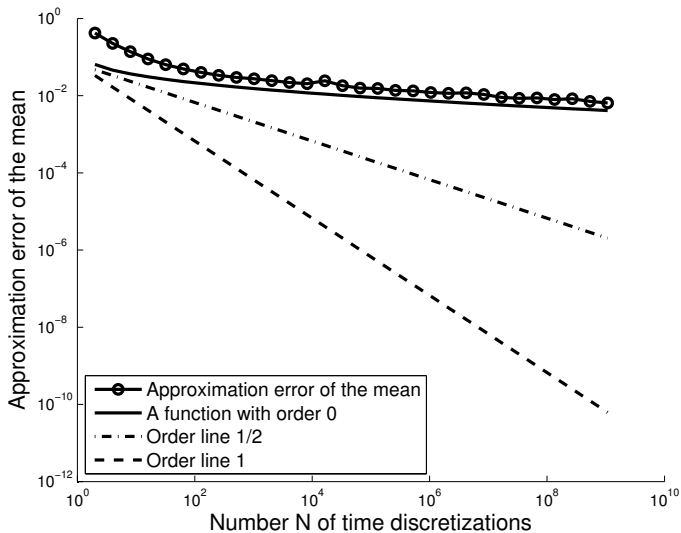
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Plot of  $\|\mathbb{E}[X_T] - \mathbb{E}[Y_N^N]\|$  for  $T = 2$  and  $N \in \{2^1, 2^2, \dots, 2^{30}\}$ .



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Let  $(a_n)_{n \in \mathbb{N}} \subseteq (0, \infty)$  satisfy  $\lim_{n \rightarrow \infty} a_n = 0$ . Then there exist *globally bounded* functions  $\mu, \sigma \in C^\infty(\mathbb{R}^4, \mathbb{R}^4)$  such that  $\forall n \in \mathbb{N}$ :

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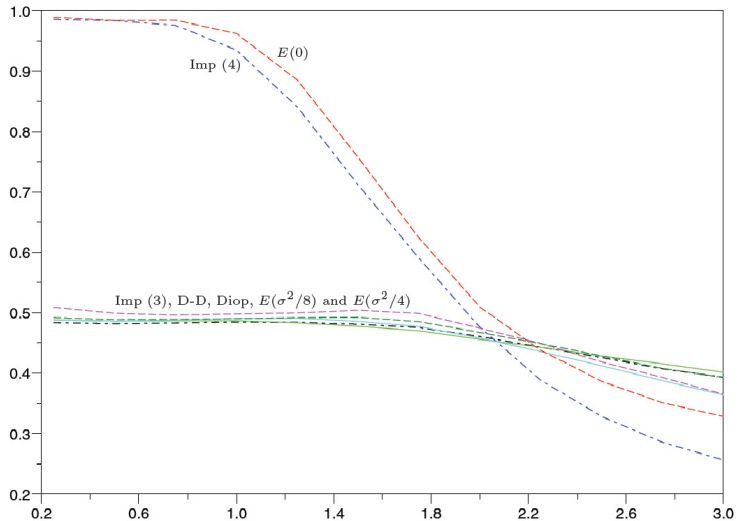
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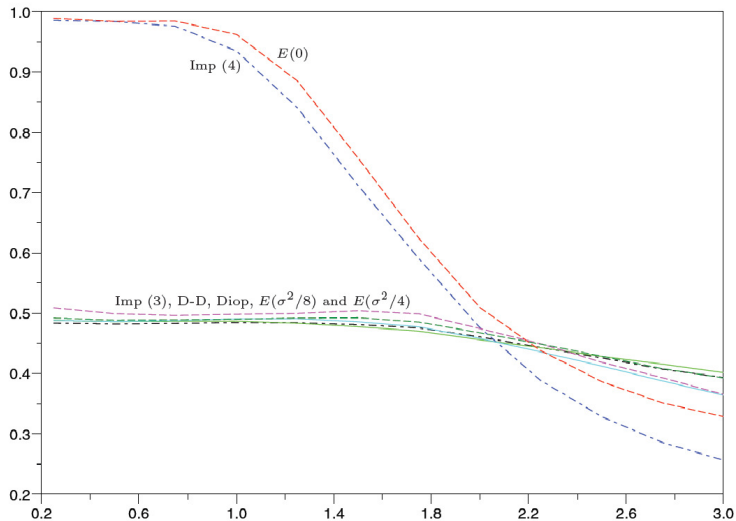
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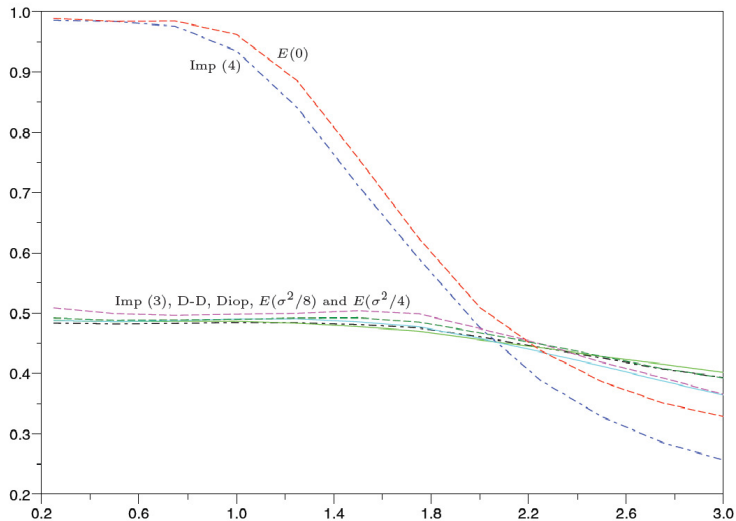
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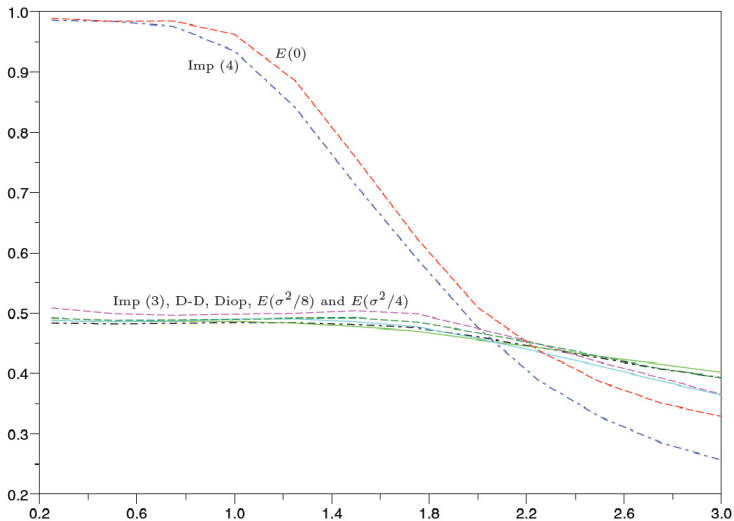
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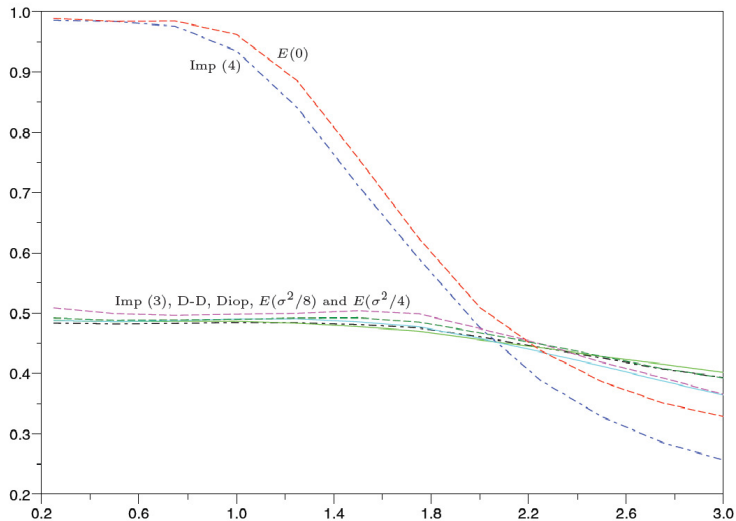
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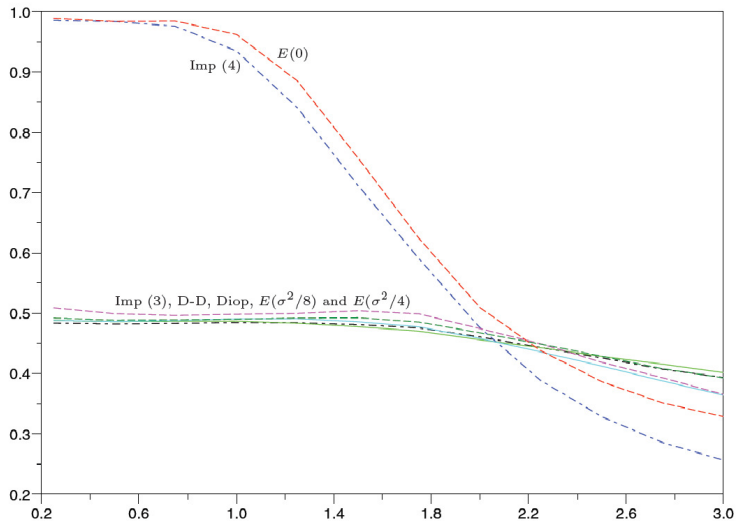
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## Theorem (Hairer, Hutzenthaler & J, AOP 2015)

Let  $T \in (0, \infty)$ . Then there exist *infinitely often differentiable* and *globally bounded* functions  $\mu, \sigma: \mathbb{R}^4 \rightarrow \mathbb{R}^4$  such that  $Y^N: \{0, 1, \dots, N\} \times \Omega \rightarrow \mathbb{R}^4$ ,  $N \in \mathbb{N}$ , satisfying  $\forall N \in \mathbb{N}, n \in \{0, 1, \dots, N-1\}: Y_0^N = X_0$  and

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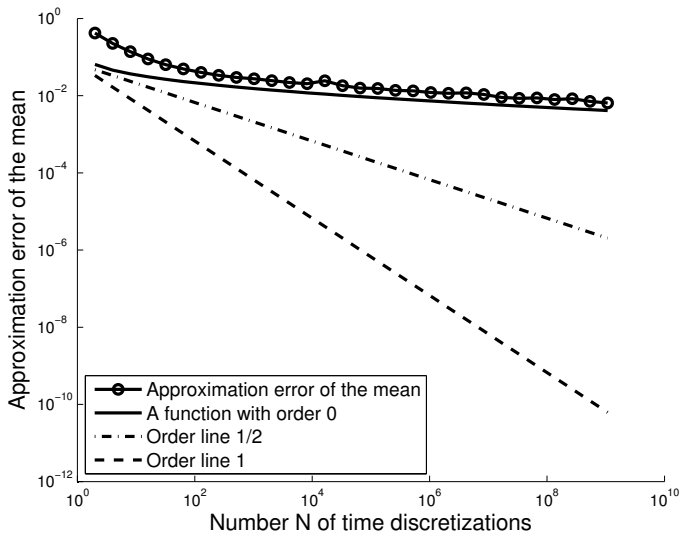
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Plot of  $\|\mathbb{E}[X_T] - \mathbb{E}[Y_N^N]\|$  for  $T = 2$  and  $N \in \{2^1, 2^2, \dots, 2^{30}\}$ .



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(iv) **Cox-Ingersoll-Ross process** ( $d = 1$ ):  $\delta, \beta > 0, \gamma \in \mathbb{R}, t \in [0, T]$ :

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**Key difficulty:**  $\mathbb{E} [ |\bar{X}_s - \bar{X}_{[s]_{T/N}}| ] \leq \dots$  or, more generally,  $\mathbb{E} [ |\bar{X}_{t_1} - \bar{X}_{t_2}| ] \leq \dots$

(iv) **Cox-Ingersoll-Ross process** ( $d = 1$ ):  $\delta, \beta > 0, \gamma \in \mathbb{R}, t \in [0, T]$ :

$$dX_t = (\delta - \gamma X_t) dt + \beta \sqrt{X_t} dW_t.$$

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$$\mathbb{E} \left[ \sup_{t \in [0, T]} |X_t - Y_t^N| \right] \leq C \cdot N^{\varepsilon - \min\{\theta, 1/2\}}$$

(Hutzenthaler, J & Noll 2014) where  $(Y_t^N)_{t \in [0, T]}$ ,  $N \in \mathbb{N}$ , are linearly-interpolated drift-implicit square-root Euler approximations (Alfonsi 2005). **A few ideas in the proof in the case  $\gamma = 0, \beta = 2$ :**  $\bar{\mu}(0) := 0, \bar{\mu}(x) := \frac{\theta}{x}$  for  $x \in (0, \infty)$  and  $\bar{X}_t := \sqrt{X_t}, \bar{Y}_t^N := \sqrt{Y_t^N}$  for  $t \in [0, T]$  satisfy  $\forall t \in [0, T]$ :  $\mathbb{P}$ -a.s.:

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This allows us to obtain  $\forall p \in (0, \infty)$ :

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for all  $N \geq 44$ .

Lower bounds for approximations of stochastic (O)DEs in the literature: E.g., Clark & Cameron 1978 (Rate  $\frac{1}{2}$ ), Davie & Gaines 2000 (Rate  $\frac{1}{6}$ ), Müller-Gronbach 2002 (Rate  $\frac{1}{2}$  up to a logarithmic term), Müller-Gronbach & Ritter 2007 & 2008 (Rate  $\alpha \in (0, \infty)$  depending on the regularity of the problem), Hutzenthaler, J & Kloeden 2011 (divergence), Kruse 2012 (Rate  $1, \frac{3}{2}, \dots$ ),  $\dots$

Here it is established that Euler's method converges **without any arbitrarily small polynomial rate of convergence.**

Consider

$$dX_1(t) = \mathbb{1}_{(1,\infty)}(X_4(t)) \cdot e^{\frac{-1}{([X_4(t)]^2-1)}} \cdot \cos\left(\left[X_3(t) - \int_0^1 e^{\frac{-1}{(1-u^2)}} du\right] e^{[X_2(t)]^3}\right) dt$$

$$dX_2(t) = dW_t$$

$$dX_3(t) = \mathbb{1}_{(-1,1)}(X_4(t)) \cdot e^{\frac{-1}{(1-[X_4(t)]^2)}} dt$$

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