

# Discretisations of stochastic Allen-Cahn equations

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*Singular SPDEs, invariant measures and discrete models*

Joint works with Ana Djurdjevac (FU Berlin), Helena Kremp (TU Wien), Harprit Singh (University of Edinburgh)

Main example: [stochastic Allen-Cahn equation](#)

$$(\partial_t - \Delta)u = u - u^3 + \xi$$

on  $[0, 1] \times \mathbb{T}$ ,  $\mathbb{T} = \mathbb{R}/\mathbb{Z}$ , where  $\xi$  is space-time white noise.

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Some variations:

- Replace  $u - u^3$  by  $f(u)$ , where  $f$  is nice (e.g. globally Lipschitz)
- Replace  $u - u^3$  by  $P(u)$ , where  $P$  is a polynomial of odd degree with negative leading order coefficient
- Replace  $\mathbb{T}$  by  $\mathbb{T}^d$

The noise:  $\xi$  is the space-time white noise: a centered, Gaussian random field with covariance

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Reference object, 3: The solution of the **linear** equation

$$(\partial_t - \Delta)\Psi = \xi$$

is  $1/4 - \varepsilon$ -Hölder continuous in time,  $1/2 - \varepsilon$ -Hölder continuous in space for any  $\varepsilon > 0$ .

In dimensions 2 and higher  $\Psi$  itself is a distribution.

Our focus: Discretisation, error estimates.

- Spatial scheme on scale  $N^{-1}$
- Temporal scheme on scale  $M^{-1}$
- Random variables sampled

Disclaimer:

- Will not comment on initial condition
- Will drop any  $\varepsilon$ -s in rates of convergence



[Gyöngy '99]: Approximation  $v^{M,N}$  by

- Finite differences in space
- Finite differences in time
- Random variables from rectangular increments of  $W$

$$\begin{aligned}\Xi_{i,j}^{N,M} &:= \xi\left(\left[\frac{i}{M}, \frac{i+1}{M}\right] \times \left[\frac{j}{N}, \frac{j+1}{N}\right]\right) \\ &= W_{\frac{i+1}{M}, \frac{j+1}{N}} - W_{\frac{i+1}{M}, \frac{j}{N}} - W_{\frac{i}{M}, \frac{j+1}{N}} + W_{\frac{i}{M}, \frac{j}{N}}.\end{aligned}$$

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The scheme:

$$\begin{aligned}M\left(v^{N,M}\left(\frac{i+1}{M}, \frac{j}{N}\right) - v^{N,M}\left(\frac{i}{M}, \frac{j}{N}\right)\right) &= f\left(v^{N,M}\left(\frac{i}{M}, \frac{j}{N}\right)\right) + MN\Xi_{i,j}^{N,M} \\ + N^2\left(v^{N,M}\left(\frac{i}{M}, \frac{j+1}{N}\right) + v^{N,M}\left(\frac{i}{M}, \frac{j-1}{N}\right) - 2v^{N,M}\left(\frac{i}{M}, \frac{j}{N}\right)\right)\end{aligned}$$

## Theorem (Gyöngy '99)

Let  $p \geq 2$ . If  $M^{-1} < (1/2)N^{-2}$  and  $f$  is globally Lipschitz continuous, then there exists a constant  $C$  such that for all  $M, N$

$$\sup_{i,j} \left( \mathbb{E} \left( u \left( \frac{i}{M}, \frac{j}{N} \right) - v^{N,M} \left( \frac{i}{M}, \frac{j}{N} \right) \right)^p \right)^{1/p} \leq C(M^{-1/4} + N^{-1/2}).$$

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### Theorem (Davie-Gaines '01)

Let  $R_1, \dots, R_{NM}$  be the rectangles of the grid with meshsize  $M^{-1}, N^{-1}$ . There exists a constant  $c > 0$  such that for all  $M, N$

$$\inf_{\varphi \text{ meas.}} \left( \mathbf{E} (\Psi(1,0) - \varphi(\xi(R_{11}), \dots, \xi(R_{NM})))^2 \right)^{1/2} \geq c(M^{-1/4} + N^{-1/2}).$$

[Becker-Gess-Jentzen-Kloeden '23]

- Spectral Galerkin truncation in space
- Tamed exponential Euler scheme in time
- Random variables from the Wiener increments on each Fourier mode of the noise:

$$B_{\frac{i+1}{M}}^k - B_{\frac{i}{M}}^k$$

Same upper and lower bounds hold for  $f(u) = u - u^3$ .

[Butkovsky-Dareiotis-G. '23]: The theorem of [Gyöngy '99] holds verbatim even if  $f$  is just a **bounded measurable** function.

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Tangential analogy: For SDEs

$$dX_t = b(X_t) dt + \sigma(X_t) dB_t$$

For the Euler (=finite difference) scheme one has

- Rate 1/2
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The previous discretisations “treat additive noise as multiplicative” because the stochastic integral  $\Psi$  is *approximated*.

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- *Conjectured\**: [Davie-Gaines '01], [Jentzen-Kloeden '08]
- *Realistic*:  $\mathcal{F}\Psi(k)$  on the (temporal) grid are still Gaussian with known covariance
- *Proved* to provide *some* improvement: from temporal rate  $1/4$  to  $1/2$  [Jentzen '11], [Wang '20]

Combining [Bréhier-Cui-Hong '18] and [Jentzen-Kloeden '08], consider:

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The scheme:

$$v^{M,N}\left(\frac{i+1}{M}, \cdot\right) = \Pi_N P_{M-1} \Phi\left(v^{M,N}\left(\frac{i}{M}, \cdot\right)\right) + \Pi_N\left(\Psi\left(\frac{i+1}{M}, \cdot\right) - P_{M-1}\Psi\left(\frac{i}{M}, \cdot\right)\right),$$

where  $P$  is the heat semigroup,  $\Pi_N$  is the projection on the first  $N$  Fourier modes, and  $\Phi$  is the solution flow of the ODE  $\dot{x} = f(x)$ .

## Theorem (Djurdjevac-G-Kremp '24)

Let  $p \geq 2$ , let  $f$  have polynomially growing derivatives up to order 3 and bounded from above first derivative. Then there exists a constant  $C$  such that for all  $M, N$

$$\left( \mathbb{E} \sup_{i,j} \left( u\left(\frac{i}{M}, \frac{j}{N}\right) - v^{N,M}\left(\frac{i}{M}, \frac{j}{N}\right) \right)^p \right)^{1/p} \leq C(M^{-1} + N^{-1/2}).$$

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### Proposition (Djurđjevac-G-Kremp '24)

Let  $u$  be the solution with  $f(x) = x$ . There exists a constant  $c > 0$  such that for all  $M, N$

$$\inf_{\varphi \text{ meas.}} \left( \mathbb{E} \left( u(1,0) - \varphi\left(\hat{\Psi}\left(\frac{1}{M}, 0\right), \dots, \hat{\Psi}(1, N)\right) \right)^2 \right)^{1/2} \geq c(M^{-1} + N^{-1/2}).$$

The bottleneck is the quadrature error

$$E_M := \left| \int_0^T P_{T-s}(f(\Psi_s) - f(\Psi_{k_M(s)})) ds \right|.$$

Here  $P$ : heat kernel,  $k_M(s)$ : the last gridpoint before  $s$ . Even for  $f \in C_c^\infty$  nontrivial!



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- Triangle inequality, regularity of  $\Psi$ :  $E_M \lesssim M^{-1/4}$
- Triangle inequality, regularity of  $\mathcal{F}\Psi(k)$ :  $E_M \lesssim M^{-1/2}$

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- Stochastic sewing [Lê '20]: if  $X$  is fBM with  $H = 1/4$ , then

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- second+third point:  $E_M \lesssim M^{-1}$

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...by **distributional topology**. This is:

- *Well-motivated*:  $C^\alpha$  for  $\alpha < 0$  are the natural solution spaces in higher dimensions

And the rate improves with lowering  $\alpha$ : [Hairer '14] for mollifier approximations, [Ma-Zhu '21], [Ma-Wang-Yang '24] for discrete approximations in  $d = 2$

- *Promising*: although  $\Psi$  is  $1/4$  Hölder in time one has for all  $\alpha \in (-1/2, 1/2)$ ,  $\varepsilon > 0$

$$\|\Psi_t - \Psi_s\|_{C^\alpha(\mathbb{T})} \lesssim |t - s|^{1/4 - \alpha/2 - \varepsilon}.$$

- *Problematic*: in distributional spaces  $u^3$  is not defined (and we have no renormalisation!)

Setup and assumptions:

- Scheme exactly as in [Gyöngy '99]
- Temporal scale  $cN^{-2}$ ,  $c < 1/8$ , spatial scale  $N^{-1}$ .
- Nonlinearity  $f(u) = -u^k + P(u)$ , where  $k$  odd and  $P$  is polynomial of order  $k - 1$ .

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### Theorem (G-Singh '22)

*Let  $\theta \in (-1/2, 0]$ . Then there exists an almost surely finite random variable  $\eta$  such that for all  $n \in \mathbb{N}$*

$$\sup_i \left\| u_{\frac{i}{N^2}} - v_{\frac{i}{N^2}}^N \right\|_{C^\theta(\mathbb{T}_N)} \leq \eta N^{-1/2+\theta}.$$

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Lower bound of order  $N^{-1}$  holds already for a fixed Fourier mode.



The rate one could expect from the general theory is

$$\text{true regularity} - \text{critical regularity}$$

E.g. for  $\Phi_3^4$ :  $-1/2 - (-1) = 1/2$

for  $k = 7$  in 1 dimensions:  $1/2 - (-1/3) = 5/6 < 1$ .

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But error and solution can be considered in different spaces:

$$\begin{aligned}\|u^3 - (v^N)^3\|_{C^{-1/2}} &= \|(u - v^N)(u^2 + (v^N)^2)\|_{C^{-1/2}} \\ &\lesssim \|u - v^N\|_{C^{-1/2}} \|u^2 + (v^N)^2\|_{C^{1/2}}\end{aligned}$$

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In higher dimensions this is done on the level of the remainders (in progress with Marco Cacace)

Thank you!