

# School on Stochastic Partial Differential Equations Part 3

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## Nemytskij operator

$f \in \mathcal{C}_b^2(\mathbf{R}) = \mathcal{C}_b^2(\mathbf{R}, \mathbf{R})$  means that  $f$  has two continuous derivatives, the first and second derivatives are bounded but the zeroth derivative need not be bounded, but it follows that it grows at most linearly: (seminorm)  $|f|_{\mathcal{C}_b^1} := \|f'\|_{L^\infty(\mathbf{R})} \leq C$  implies

$$|f(s)| = |f(0) + f'(\xi)s| \leq |f(0)| + |f|_{\mathcal{C}_b^1}|s| \leq C(1 + |s|).$$

$$\begin{aligned} f \in \mathcal{C}_b^2(\mathbf{R}), \quad F: H \rightarrow H, \quad F(u)(x) &= f(u(x)) \\ \|F(u)\|_H = \|f(u)\|_{L_2} &\leq C\|1 + |u|\|_{L_2} \leq C(1 + \|u\|_H) \end{aligned}$$

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For the first Fréchet derivative we have

$$F'(u) \in \mathcal{L}(H), \quad (F'(u)v)(x) = f'(u(x))v(x), \\ \|F'(u)v\|_H = \|f'(u)v\|_{L_2} \leq \|f'(u)\|_{L_\infty} \|v\|_{L_2} \leq |f|_{\mathcal{C}_b^1} \|v\|_H \\ \|F'(u)\|_{\mathcal{L}(H)} \leq |f|_{\mathcal{C}_b^1}, \quad |F|_{\mathcal{C}_b^1(H)} := \sup_{u \in H} \|F'(u)\|_{\mathcal{L}(H)} \leq |f|_{\mathcal{C}_b^1}$$

So  $F \in \mathcal{C}_b^1(H)$  and hence globally Lipschitz  $H \rightarrow H$ . Further:  $(\mathcal{L}^{(2)}(H))$  means bounded bilinear operators  $H \rightarrow H$

$$F''(u) \in \mathcal{L}^{(2)}(H), \quad (F''(u)vw)(x) = f''(u(x))v(x)w(x) \\ \|F''(u)vw\|_H = \|f''(u)vw\|_{L_2} \leq \|f''(u)\|_{L_\infty} \|v\|_{L_4} \|w\|_{L_4} \leq |f|_{\mathcal{C}_b^2} \|v\|_{L_4} \|w\|_{L_4}$$

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No good. So  $F''(u)$  is actually not in  $\mathcal{L}^{(2)}(H)$  but in  $\mathcal{L}^{(2)}(L^4, H)$  and hence  $F \notin \mathcal{C}_b^2(H)$ !

## Nemytskij operator

But, by Sobolev's inequality,  $\|\phi\|_{L^\infty} \leq C\|\phi\|_{H^s}$  if  $s > d/2$ , so that

$$\|h\|_{H^{-s}} = \sup_{\phi} \frac{\langle h, \phi \rangle}{\|\phi\|_{H^s}} \leq \sup_{\phi} \frac{\|h\|_{L^1} \|\phi\|_{L^\infty}}{\|\phi\|_{H^s}} \leq C\|h\|_{L^1}, \quad s > d/2$$

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Then

$$\|F''(u)vw\|_{H^{-s}} = \|f''(u)vw\|_{L^1} \leq \|f''(u)\|_{L^\infty} \|v\|_{L^2} \|w\|_{L^2} \leq |f|_{C_b^2} \|v\|_H \|w\|_H$$

$$|F|_{C_b^2(H, H^{-s})} := \sup_{u \in H} \|F''(u)\|_{\mathcal{L}^{(2)}(H, H^{-s})} \leq |f|_{C_b^2}$$

So  $F''(u) \in \mathcal{L}^{(2)}(H, H^{-s})$  and  $F \in C_b^2(H, H^{-s})$ .

## Nemytskij operator

What about the diffusion coefficient  $G(X)$ ?

$$g \in \mathcal{C}_b^2(\mathbf{R}), \quad G: H \rightarrow \mathcal{L}_2^0(H), \quad (G(u)v)(x) = g(u(x))v(x)$$

$$\|G(u)\|_{\mathcal{L}_2^0(H)}^2 = \|G(u)Q^{1/2}\|_{\mathcal{L}_2(H)}^2 = \sum_{j=1}^{\infty} \|G(u)Q^{1/2}e_j\|_H^2 = \sum_{j=1}^{\infty} \gamma_j \|G(u)e_j\|_H^2$$

$$= \sum_{j=1}^{\infty} \gamma_j \|g(u)e_j\|_{L_2}^2 \leq \sum_{j=1}^{\infty} \gamma_j \|g(u)\|_{L_2}^2 \|e_j\|_{L_\infty}^2 \leq \sum_{j=1}^{\infty} \gamma_j C \|1 + \|u\|_{L_2}^2$$

$$\leq C \operatorname{Tr}(Q)(1 + \|u\|_{L_2})^2$$

Here  $\|e_j\|_{L_2} = 1$  but we had to assume  $\|e_j\|_{L_\infty} \leq C$ ! Which is true sometimes, for example,  $\sin(n\pi x)$ .

Exercise. Compute  $\|G'(u)\|_{\mathcal{L}(H, \mathcal{L}_2^0(H))}$ . What about  $\|G''(u)\|_{\mathcal{L}^{(2)}(H, \mathcal{L}_2^0(H))}$ ?

# Implementation

Euler's method for the stochastic heat equation

$$\begin{cases} X^n \in S_h, & X^0 = P_h u_0 \\ X^n - X^{n-1} + \Delta t A_h X^n = P_h \Delta W^n \end{cases}$$

$$(X^n - X^{n-1}, \chi) + \Delta t (\nabla X^n, \nabla \chi) = (\underbrace{\Delta W^n}_{\in L_2(\Omega, \dot{H}^{-1})}, \chi), \quad \forall \chi \in S_h$$

$$X^n(x) = \sum_{k=1}^{N_h} X_k^n \phi_k(x), \quad \chi = \phi_j, \quad \{\phi_j\}_1^{N_h} \text{ finite element basis functions}$$

$$\sum_{k=1}^{N_h} X_k^n (\phi_k, \phi_j) + \Delta t \sum_{k=1}^{N_h} X_k^n (\nabla \phi_k, \nabla \phi_j) = \sum_{k=1}^{N_h} X_k^{n-1} (\phi_k, \phi_j) + (\Delta W^n, \phi_j)$$

$$\mathbf{M}\mathbf{X}^n + \Delta t \mathbf{K}\mathbf{X}^n = \mathbf{M}\mathbf{X}^{n-1} + \mathbf{b}^n$$



## Implementation

How to simulate  $\mathbf{b}_j^n = (\Delta W^n, \phi_j) = (W(t_n) - W(t_{n-1}), \phi_j)$  ?

Covariance of  $\mathbf{b}^n$ :

$$\mathbf{E}(\mathbf{b}_i^n \mathbf{b}_j^n) = \mathbf{E}((\Delta W^n, \phi_i)(\Delta W^n, \phi_j)) = \Delta t (\mathbf{Q} \phi_i, \phi_j)$$

In other words:

$$\mathbf{E}(\mathbf{b}^n \otimes \mathbf{b}^n) = \Delta t \mathbf{Q}, \quad \mathbf{Q}_{ij} = (\mathbf{Q} \phi_i, \phi_j).$$

This assumes that the action of the covariance operator is known (computable). For example, integral operator with known kernel:  $(\mathbf{Q}f)(x) = \int_{\mathcal{D}} q(x, y) f(y) dy$ .

Cholesky factorization:  $\mathbf{Q} = \mathbf{L}\mathbf{L}^T$ , expensive, but done only once.

Take  $\mathbf{b}^n = \sqrt{\Delta t} \mathbf{L} \beta^n$ , where  $\beta^n \in \mathbf{R}^{N_h}$ ,  $n = 1, 2, \dots$ , are  $\mathcal{N}(0, \mathbf{I})$ , that is, generate one random vector in each time step, the components are independent normally distributed random numbers.

Then

$$\begin{aligned} \mathbf{E}(\mathbf{b}^n \otimes \mathbf{b}^n) &= \mathbf{E}(\mathbf{b}^n (\mathbf{b}^n)^T) = \Delta t \mathbf{E}(\mathbf{L} \beta^n (\mathbf{L} \beta^n)^T) \\ &= \Delta t \mathbf{L} \mathbf{E}(\beta^n (\beta^n)^T) \mathbf{L}^T = \Delta t \mathbf{L} \mathbf{L}^T = \Delta t \mathbf{Q} \end{aligned}$$

## Implementation

One situation where the action of  $Q$  is known is  $Q = I$ . Then  $\mathbf{Q}_{ij} = (Q\phi_i, \phi_j) = (\phi_i, \phi_j)$ , that is,  $\mathbf{Q} = \mathbf{M}$ , the mass matrix. It is sparse so the Cholesky factorization is not too expensive. It can also be approximated by the lumped mass matrix  $\mathbf{M}_L$ , which is diagonal and  $\mathbf{M}_L^{1/2}$  is easily computed. Then  $\mathbf{b}^n = \sqrt{\Delta t} \mathbf{M}_L^{1/2} \beta^n$  can be used.

But  $Q = I$  is of no interest unless  $d = 1$ , as we have seen.

However, it can be used (also for  $d \geq 1$ ) to generate noise increments  $\Delta W$  with prescribed covariance from the Matérn class of covariance kernels. Let  $\Delta W_I$  be a noise increment with  $Q = I$  and solve the equation

$$(\kappa I - \Delta)^{(\nu+1)/2} \Delta W = \Delta W_I \quad \text{in } \mathcal{D}.$$

Then  $\Delta W$  will have a covariance from the Matérn class with parameters  $\kappa, \nu$ . Its finite element approximation will serve as the vector  $\mathbf{b}$  above. But this equation is, in general, of fractional order  $\nu + 1$  and it is therefore not straightforward to solve.

F. Lindgren and H. Rue, *An explicit link between Gaussian fields and Gaussian Markov random fields: the stochastic partial differential equation approach*, J. R. Statist. Soc. B (2011) **73**, Part 4, pp. 423-498.

## Implementation

Another approach: truncate the orthogonal expansion (Karhunen–Loève expansion)

$$W(t) = \sum_{k=1}^{\infty} \gamma_k^{1/2} \beta_k(t) \mathbf{e}_k \approx \sum_{k=1}^M \gamma_k^{1/2} \beta_k(t) \mathbf{e}_k, \quad Q \mathbf{e}_k = \gamma_k \mathbf{e}_k.$$

The truncated expansion can be inserted in the finite element equation. This assumes that the eigenvectors of  $Q$  are known. The eigenvalues can be chosen with the desired rate of convergence  $\gamma_k \rightarrow 0$ .

## Stochastic Cahn–Hilliard equation

The Cahn–Hilliard–Cook equation:

$$\begin{cases} du - \Delta v dt = dW & \text{in } \mathcal{D} \times (0, T]; \\ v + \Delta u - f(u) = 0 & \text{in } \mathcal{D} \times (0, T]; \\ \frac{\partial u}{\partial n} = \frac{\partial v}{\partial n} = 0 & \text{on } \partial\mathcal{D} \times (0, T]; \\ u(0) = u_0 & \text{in } \mathcal{D}. \end{cases} \quad (1)$$

Here  $\mathcal{D} \subset \mathbf{R}^d$ ,  $d \leq 3$ , is a convex polygonal domain and

$$f(s) = F'(s), \quad F \text{ is a polynomial of degree 4}$$

$$F(s) \geq c_0 s^4 - c_1, \quad c_0 > 0; \quad F''(s) \geq -\beta^2,$$

Typically:  $F(s) = \frac{1}{4}(s^2 - \beta^2)^2$ ,  $f(s) = s^3 - \beta^2 s$ .

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Further assumptions:

- ▶ In order to preserve mass:  $|\mathcal{D}|^{-1} \int_{\mathcal{D}} W(t)(x) dx = 0$ .
- ▶ The initial value  $u_0$  is deterministic with  $|\mathcal{D}|^{-1} \int_{\mathcal{D}} u_0 dx = 0$ .

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**Note:** As  $\tilde{F}(s) = F(s + s_0)$  has the same structural properties as  $F$  a change of variables  $u \rightarrow u - |\mathcal{D}|^{-1} \int_{\mathcal{D}} u_0 dx$  shows that we may assume that  $|\mathcal{D}|^{-1} \int_{\mathcal{D}} u_0 dx = 0$ .

## Abstract framework

The Cahn–Hilliard–Cook equation:

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- ▶  $W(t)$  – a  $Q$ -Wiener process in  $H$  with respect to  $(\Omega, \mathcal{F}, \mathbf{P}, \{\mathcal{F}_t\})$ :

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$$\begin{cases} dX + (A^2 X + Af(X)) \, dt = dW, & t > 0; \\ X(0) = X_0. \end{cases}$$

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- ▶ Fractional powers  $A^\alpha v = \sum_{j=1}^{\infty} \lambda_j^\alpha \langle v, \varphi_j \rangle \varphi_j$  with corresponding spaces  $\dot{H}^\alpha = D(A^{\frac{\alpha}{2}})$  and norms  $|v|_\alpha = \|A^{\frac{\alpha}{2}} v\| = (\sum_{j=1}^{\infty} \lambda_j^{2\alpha} \langle v, \varphi_j \rangle^2)^{1/2}$ .

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$$\begin{aligned} E(t)v &= e^{-tA^2} v = \sum_{j=0}^{\infty} e^{-t\lambda_j^2} \langle v, \varphi_j \rangle \varphi_j \\ &= \sum_{j=1}^{\infty} e^{-t\lambda_j^2} \langle v, \varphi_j \rangle \varphi_j + \langle v, \varphi_0 \rangle \varphi_0 \\ &= e^{-tA^2} P v + (I - P)v. \end{aligned}$$

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$$\begin{aligned} E(t)v &= e^{-tA^2} v = \sum_{j=0}^{\infty} e^{-t\lambda_j^2} \langle v, \varphi_j \rangle \varphi_j \\ &= \sum_{j=1}^{\infty} e^{-t\lambda_j^2} \langle v, \varphi_j \rangle \varphi_j + \langle v, \varphi_0 \rangle \varphi_0 \\ &= e^{-tA^2} P v + (I - P)v. \end{aligned}$$

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- ▶ Moreover:

$$\left( \int_0^t s^{2\alpha} \|A^{2\alpha+1} e^{-sA^2} v\|^2 ds \right)^{1/2} \leq C \|v\|, \quad v \in H, \alpha \geq 0.$$

- ▶  $\begin{cases} \dot{u} + A^2 u = f, & t > 0; \\ u(0) = v \end{cases} \quad \Rightarrow u(t) = e^{-tA^2} v + \int_0^t e^{-(t-s)A^2} f(s) ds.$

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A *mild solution* satisfies the equation:

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Stochastic convolution:

$$W_A(t) = \int_0^t e^{-(t-s)A^2} dW(s).$$

$$X = Y + W_A, \quad Y(t) = e^{-tA^2} X_0 - \int_0^t e^{-(t-s)A^2} Af(Y(s) + W_A(s)) ds$$

# Existence, uniqueness and regularity

## Theorem

If  $\|A^{\frac{1}{2}}Q^{\frac{1}{2}}\|_{HS} < \infty$  and  $|X_0|_1 < \infty$ , then there is a unique weak solution  $X$  of (1). Furthermore, there is  $C_T > 0$  such that

$$\mathbf{E} \sup_{t \in [0, T]} |X(t)|_1^2 + \mathbf{E} \sup_{t \in [0, T]} \|X(t)\|_{L_4}^4 \leq C_T.$$

In addition,  $X$  is also a mild solution. Furthermore, for all  $\gamma \in [0, \frac{1}{2})$ , there is a finite nonnegative random variable  $K$  such that, almost surely,

$$\sup_{t \neq s \in [0, T]} \frac{\|X(t) - X(s)\|}{|t - s|^\gamma} \leq K.$$

Da Prato and Debussche, 1996

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- ▶ Conservation of mass:

- $$W(t) \in \dot{H}, X_0 \in \dot{H} \Rightarrow P_h X_0 \in \dot{H}, X(t), X_h(t) \in \dot{H}.$$

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- ▶ note:

$$|v_h|_{1,h} = \|A_h^{1/2} v_h\| = \|\nabla v_h\| = \|A_h^{1/2} v_h\| = |v_h|_1, \quad v_h \in S_h \subset H^1(\mathcal{D}).$$



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$$\begin{cases} X_h^j - X_h^{j-1} + kA_h^2 X_h^j + kA_h P_h f(X_h^j) = P_h \Delta W^j, & t_j = jk, j = 1, 2, \dots, N, \\ X_h^0 = P_h X_0. \end{cases} \quad (2)$$

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- ▶ Mass is conserved.

## Linear CHC: regularity

$$\begin{cases} dX + A^2 X dt = dW, & t > 0 \\ X(0) = 0 \end{cases}$$

Stochastic convolution:  $X(t) = W_A(t) = \int_0^t E(t-s) dW(s)$

Seminorms:  $|v|_\beta = \|A^{\beta/2} v\| = \left( \sum_{j=1}^{\infty} \lambda_j^{\beta/2} \langle v, \varphi_j \rangle^2 \right)^{\frac{1}{2}}, \quad \dot{H}^\beta = D(A^{\beta/2}), \quad \beta \in \mathbf{R}$

Mean square:  $\|v\|_{L_2(\Omega, \dot{H}^\beta)}^2 = \mathbf{E}(|v|_\beta^2), \quad \beta \in \mathbf{R}$

### Theorem

If  $\|A^{\frac{\beta-2}{2}} Q^{\frac{1}{2}}\|_{HS} < \infty$  for some  $\beta \geq 0$ , then

$$\|W_A(t)\|_{L_2(\Omega, \dot{H}^\beta)} \leq C \|A^{\frac{\beta-2}{2}} Q^{\frac{1}{2}}\|_{HS}, \quad t \geq 0.$$

## Regularity: proof

$$\begin{aligned}\|W_A(t)\|_{L_2(\Omega, \dot{H}^\beta)}^2 &= \mathbf{E} \left( \left\| \int_0^t A^{\beta/2} E(t-s) dW(s) \right\|^2 \right) \\ &= \int_0^t \|A^{\beta/2} E(s) Q^{1/2}\|_{\text{HS}}^2 ds \\ &= \int_0^t \|AE(s) A^{(\beta-2)/2} Q^{1/2}\|_{\text{HS}}^2 ds \\ &= \sum_{k=1}^{\infty} \int_0^t \|AE(s) A^{(\beta-2)/2} Q^{1/2} \phi_k\|^2 ds \\ &\leq C \sum_{k=1}^{\infty} \|A^{(\beta-2)/2} Q^{1/2} \phi_k\|^2 \\ &= C \|A^{(\beta-2)/2} Q^{1/2}\|_{\text{HS}}^2 \quad \boxed{\int_0^t \|AE(s)v\|^2 ds \leq C \|v\|^2}\end{aligned}$$

## Linear CHC: strong convergence

$$H = L_2(\mathcal{D}), \quad \dot{H} = \left\{ v \in H : \int_{\mathcal{D}} v \, dx = 0 \right\}$$

Orthogonal projector  $P: H \rightarrow \dot{H}$ ,  $W(t) \in \dot{H}$

$$W_A(t) = \int_0^t E(t-s) \, dW(s)$$

$$W_{A_h}(t) = \int_0^t E_h(t-s) P_h \, dW(s)$$

$$\begin{aligned} W_{A_h}(t) - W_A(t) &= \int_0^t (E_h(t-s) P_h - E(t-s)) \, dW(s) \\ &= \int_0^t F_h(t-s) \, dW(s) \end{aligned}$$



## Linear CHC: approximation of the semigroup

$$\begin{cases} \dot{u} + A^2 u = 0, & t > 0 \\ u(0) = v \end{cases} \quad \begin{cases} \dot{u}_h + A_h^2 u_h = 0, & t > 0 \\ u_h(0) = P_h v \end{cases}$$

$$u(t) = E(t)v$$

$$u_h(t) = E_h(t)P_h v$$

$$\text{Error: } F_h(t)v = E_h(t)P_h v - E(t)v, \quad \text{norm: } |v|_\beta = \|A^{\beta/2}v\|$$

### Theorem

- ▶  $\|F_h(t)v\| \leq Ch^\beta |v|_\beta, \quad t \geq 0, \quad \beta \in [0, 2]$
- ▶  $\left( \int_0^t \|F_h(s)v\|^2 ds \right)^{1/2} \leq Ch^\beta |\log(h)| |v|_{\beta-2}, \quad t \geq 0, \quad \beta \in [1, 2]$

Note: the FEM is based on  $(A_h)^2$  instead of  $(A^2)_h$ .

## Linear CHC: strong convergence

### Theorem

If  $\|A^{\frac{\beta-2}{2}} Q^{\frac{1}{2}}\|_{HS} < \infty$  for some  $\beta \in [1, 2]$ , then

$$\|W_{A_h}(t) - W_A(t)\|_{L_2(\Omega, H)} \leq Ch^\beta |\log(h)| \|A^{\frac{\beta-2}{2}} Q^{\frac{1}{2}}\|_{HS}, \quad t \geq 0.$$

Proof:

$$\begin{aligned} & \|W_{A_h}(t) - W_A(t)\|_{L_2(\Omega, H)}^2 \\ &= \mathbf{E} \left\| \int_0^t F_h(t-s) dW(s) \right\|^2 = \int_0^t \|F_h(t-s) Q^{1/2}\|_{HS}^2 ds \\ &= \sum_{j=1}^{\infty} \int_0^t \|F_h(t-s) Q^{1/2} \phi_j\|^2 ds \leq C \sum_{j=1}^{\infty} h^{2\beta} |\log(h)|^2 |Q^{1/2} \phi_j|_{\beta-2}^2 \\ &= Ch^{2\beta} |\log(h)|^2 \sum_{j=1}^{\infty} \|A^{(\beta-2)/2} Q^{1/2} \phi_j\|^2 \\ &= Ch^{2\beta} |\log(h)|^2 \|A^{(\beta-2)/2} Q^{1/2}\|_{HS}^2 \end{aligned}$$

## Linear CHC: strong convergence

The assumption is:  $\|A^{\frac{\beta-2}{2}} Q^{\frac{1}{2}}\|_{\text{HS}} < \infty$ ,  $\beta \in [1, 2]$ .

- ▶  $\text{Tr}(Q) = \|Q^{1/2}\|_{\text{HS}}^2 < \infty$ :  $\beta = 2$ .
- ▶  $Q = I$ , “white noise”:

$$\|A^{\frac{\beta-2}{2}}\|_{\text{HS}}^2 = \sum_{j=1}^{\infty} \lambda_j^{\beta-2} \sim \sum_{j=1}^{\infty} j^{(\beta-2)2/d} < \infty$$

if  $\beta < 2 - d/2$ . Thus:  $d = 1$ ,  $\beta \in [1, 3/2]$ .

## Linear CHC: strong convergence

Larsson and Mesforush, IMAJNA (2011).  
Euler timestepping is also studied here.

Kossioris and Zouraris, M2AN (2010) (1-D)

## Fully discrete, nonlinear: main result

### Theorem

Suppose that  $\|A^{1/2}Q^{1/2}\|_{HS} < \infty$  ( $\beta = 3$ ) and that there is  $L > 0$  such that

$$|X_0|_1 + |X_h^0|_1 + \mathcal{F}(X_h^0) + |A_h X_h^0 + P_h f(X_h^0)|_1 \leq L, \quad 0 < h < h_0,$$

where  $\mathcal{F}(u) = \int_{\mathcal{D}} F(u(x)) dx$ . Then

$$\lim_{h,k \rightarrow 0} \mathbf{E} \sup_{t_n \in [0, T]} \|X(t_n) - X_h^n\|^2 = 0.$$

Thus, we show strong convergence, but there is no error estimate, so the rate of convergence is not obtained.

## Local Lipschitz condition

The semilinear term  $Af(u)$  is not globally Lipschitz and does not have a linear growth. Instead we have local Lipschitz conditions, for example,

$$\begin{aligned} |Af(u) - Af(v)|_{-3} &= \|f(u) - f(v)\|_{-1} \leq C\|f(u) - f(v)\|_{L_{6/5}} \\ &\leq C\|(1 + u^2 + v^2)(u - v)\|_{L_{6/5}} \leq C(1 + \|u\|_{L_6}^2 + \|v\|_{L_6}^2)\|u - v\|_{L_2} \\ &\leq C(1 + |u|_1^2 + |v|_1^2)\|u - v\|_{L_2}. \end{aligned}$$

This can be used in the mild formulation, for example,

$$\begin{aligned} &\mathbf{E} \left\| \int_0^t e^{-(t-s)A^2} A(f(X(s)) - f(X_h(s))) ds \right\| \\ &= \mathbf{E} \left\| \int_0^t A^{3/2} e^{-(t-s)A^2} A^{-1/2} (f(X(s)) - f(X_h(s))) ds \right\| \\ &\leq \int_0^t \mathbf{E} \|A^{3/2} e^{-(t-s)A^2} A^{-1/2} (f(X(s)) - f(X_h(s)))\| ds \\ &\leq \int_0^t \|A^{3/2} e^{-(t-s)A^2}\|_{\mathcal{L}(H)} \mathbf{E} \|f(X(s)) - f(X_h(s))\|_{-1} ds \\ &\leq C \int_0^t (t-s)^{-3/4} \mathbf{E} [(1 + |X(s)|_1^2 + |X_h(s)|_1^2) \|X(s) - X_h(s)\|] ds \end{aligned}$$

## Local Lipschitz condition

$$\begin{aligned} & \mathbf{E} \left\| \int_0^t e^{-(t-s)A^2} A(f(X(s)) - f(X_h(s))) ds \right\| \\ & \leq C \int_0^t (t-s)^{-3/4} \mathbf{E}[(1 + |X(s)|_1^2 + |X_h(s)|_1^2) \|X(s) - X_h(s)\|] ds \end{aligned}$$

Here, by Hölder's inequality,

$$\begin{aligned} & \mathbf{E}[(1 + |X(s)|_1^2 + |X_h(s)|_1^2) \|X(s) - X_h(s)\|] \\ & \leq (1 + \mathbf{E}|X(s)|_1^4 + \mathbf{E}|X_h(s)|_1^4)^{1/2} (\mathbf{E}\|X(s) - X_h(s)\|^2)^{1/2}. \end{aligned}$$

Thus,

$$\begin{aligned} & \mathbf{E} \left\| \int_0^t e^{-(t-s)A^2} A(f(X(s)) - f(X_h(s))) ds \right\| \\ & \leq C_t \sup_{s \in [0, t]} (1 + \mathbf{E}|X(s)|_1^4 + \mathbf{E}|X_h(s)|_1^4)^{1/2} \sup_{s \in [0, t]} (\mathbf{E}\|X(s) - X_h(s)\|^2)^{1/2}. \end{aligned}$$

This indicates the need for moment bounds for  $X(t)$ ,  $X_h(t)$  and  $X_h^n$ .

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Thus,

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This indicates the need for moment bounds for  $X(t)$ ,  $X_h(t)$  and  $X_h^n$ .

However, the norm on the left is  $L_1(\Omega, H)$  while on the right we have

$\|X(s) - X_h(s)\|_{L_2(\Omega, H)}$ . We need to have the same  $L_2(\Omega, H)$  on both sides. But then we must have  $\|X\|_{L_\infty(\Omega \times [0, t], \dot{H}^1)}$  and  $\|X_h\|_{L_\infty(\Omega \times [0, t], \dot{H}^1)}$  in the Lipschitz constant.

This is not possible, but we will find a work-around, which yields convergence but no error estimate.



## Main obstacles

The proof of moment bounds must take advantage of structural properties of  $f(u)$ , and must therefore be based on the weak formulation instead of the mild formulation. Some difficulties:

- ▶ It follows from the conditions on  $F$  that

$$\langle f(x), x \rangle \geq -C_0 - C_1 \|x\|^2,$$

and even

$$\langle A^{\frac{1}{2}} f(x), A^{\frac{1}{2}} x \rangle = \langle \nabla f(x), \nabla x \rangle \geq -c |x|_1^2.$$

## Main obstacles

- Matters are worse for the finite element approximation. While

$$\langle P_h f(v_h), v_h \rangle = \langle f(v_h), v_h \rangle \geq -C_0 - C_1 \|v_h\|^2, \quad v_h \in S_h, \quad (3)$$

unfortunately

$$\langle A_h^{\frac{1}{2}} P_h f(v_h), A_h^{\frac{1}{2}} v_h \rangle = \langle A_h^{\frac{1}{2}} P_h f(v_h), A_h^{\frac{1}{2}} v_h \rangle \not\geq -c |v_h|_1^2, \quad v_h \in S_h, \quad (4)$$

The operators  $P_h$ ,  $e^{-tA^2}$ ,  $e^{-tA_h^2}$  act globally in  $\mathcal{D}$  and do not preserve the pointwise structural properties of  $f(u)$ . Therefore, they cannot be exploited in connection with the mild formulation, and the  $P_h$  must be removed as in (3). This cannot be done in (4).

## Deterministic error estimates

- ▶ The following error estimates are readily available in the literature:

$$\|(E(t_n) - R_{k,h}^n)P_h v\| \leq C(h^\beta + k^{\beta/4})|v|_\beta, \quad t_n \geq 0, \quad \beta \in [0, 2]$$

$$\|(E(t_n) - R_{k,h}^n)P_h v\| \leq C(h^\beta + k^{\beta/4})t_n^{-(\beta-\gamma)/4}|v|_\gamma, \quad t > 0, \\ \gamma \in [-1, 1], \quad \max(0, \gamma) \leq \beta \leq 2.$$

## Deterministic error estimates

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$$\|(E(t_n) - R_{k,h}^n)P_h v\| \leq C(h^\beta + k^{\beta/4})|v|_\beta, \quad t_n \geq 0, \quad \beta \in [0, 2]$$

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- ▶ These lead to an error bound for the stochastic convolution:

### Lemma

Let  $\gamma \in (0, \frac{1}{2}]$ ,  $\beta \in [1, 2]$ , and  $p > \frac{2}{\gamma}$ . Then there is  $C = C(p, \gamma, T)$  such that

$$\left( \mathbf{E} \left( \sup_{t_n \in [0, T]} \|W_A(t_n) - W_{A_h}^n\|^p \right) \right)^{1/p} \leq C(h^\beta + k^{\beta/4}) \|A^{(\beta-2)/2+\gamma} Q^{1/2}\|_{HS}.$$

The proof is based on a "factorization argument" from Da Prato–Zabczyk. We will use  $\beta = 2$ ,  $\gamma = \frac{1}{2}$ ,  $\|A^{1/2}Q^{1/2}\|_{HS}$ .

## Deterministic Cahn–Hilliard equation

Gradient flow in  $\dot{H}^{-1}$  for the energy functional:

$$\begin{aligned} J(u) &= \frac{1}{2} \|\nabla u\|^2 + \int_{\mathcal{D}} F(u) \, dx \\ &= \frac{1}{2} \|u\|_1^2 + \mathcal{F}(u), \quad u \in \dot{H}^1, \quad F(s) = \frac{1}{4} s^4 - \frac{1}{2} s^2 \\ J'(u) &= Au + f(u) \end{aligned}$$

Deterministic case:  $\dot{u} + A(Au + f(u)) = 0$ , that is

$$\dot{u} + AJ'(u) = 0, \quad \text{or } \dot{u} = -AJ'(u).$$

Multiply by  $A^{-1}\dot{u}$ :

$$\begin{aligned} \langle \dot{u}, A^{-1}\dot{u} \rangle + \langle AJ'(u), A^{-1}\dot{u} \rangle &= 0 \\ |\dot{u}|_{-1}^2 + D_t J'(u) &= 0 \\ D_t J(u) = -|\dot{u}|_{-1}^2 &\leq 0 \\ J(u(t)) \leq J(u_0), \quad t \geq 0 &\quad (\text{Lyapunov functional}) \end{aligned}$$

Note:  $J(u(t))$  is equivalent to  $\|u(t)\|_1^2 + \|u(t)\|_{L^4(\mathcal{D})}^4$ , so we get a bound for these norms.

## Deterministic Cahn–Hilliard equation

Another proof.

$$\dot{u} + A^2 u + Af(u) = 0,$$

Do not eliminate the chemical potential  $v = Au + f(u)$ :

$$\begin{cases} \dot{u} + Av = 0, \\ v = Au + f(u) \end{cases}$$

Multiply by  $v$ :

$$\langle \dot{u}, v \rangle + \langle Av, v \rangle = 0$$

But we have

$$J(u) = \frac{1}{2} \|A^{1/2} u\|^2 + \int_{\mathcal{D}} F(u) \, dx$$

$$J'(u) = Au + f(u) = v$$

$$D_t J(u) = \langle J'(u), \dot{u} \rangle = \langle v, \dot{u} \rangle$$

Therefore:

$$D_t J(u) + |v|_1^2 = 0$$

$$J(u(t)) + \int_0^t |v|_1^2 \, ds = J(u_0), \quad t \geq 0$$

$$J(u(t)) \leq J(u_0), \quad t \geq 0 \quad (\text{Lyapunov functional})$$

## Cahn–Hilliard–Cook equation

Time continuous stochastic case:

### Theorem

If  $\|A^{1/2}Q^{1/2}\|_{HS}^2 < \infty$ , then

$$\mathbf{E}[J(X(t))] \leq C(t), \quad \mathbf{E}[J(X_h(t))] \leq C(t), \quad t \geq 0.$$

Moreover,

$$\mathbf{E} \left[ \sup_{t \in [0, T]} \left( |X(t)|_1^2 + \|X(t)\|_{L_4}^4 \right) \right] \leq K_T,$$

$$\mathbf{E} \left[ \sup_{t \in [0, T]} \left( |X_h(t)|_1^2 + \|X_h(t)\|_{L_4}^4 \right) \right] \leq K_T.$$

## Cahn–Hilliard–Cook equation

Proof for  $J(X_h(t))$ :

$$J(u_h) = \frac{1}{2} \|A_h^{1/2} u_h\|^2 + \int_{\mathcal{D}} F(u_h) dx$$

$$J'(u_h) = A_h u_h + P_h f(u_h)$$

$$J''(u_h) = A_h + P_h [f'(u_h) \cdot]$$

$$\begin{aligned} dX_h &= -A_h(A_h X_h + P_h f(X_h)) dt + P_h dW \\ &= -A_h J'(X_h) dt + P_h dW \end{aligned}$$

Itô's formula: (with  $Q_h = P_h Q P_h$ )

$$\begin{aligned} J(X_h(t)) &= J(X_h(0)) + \int_0^t \langle J'(X_h(s)), dX_h(s) \rangle + \frac{1}{2} \int_0^t \text{Tr}(J''(X_h(s)) Q_h) ds \\ &= J(P_h X_0) - \int_0^t \langle J'(X_h(s)), A_h J'(X_h(s)) \rangle ds \\ &\quad + \int_0^t \langle J'(X_h(s)), P_h dW(s) \rangle + \frac{1}{2} \int_0^t \text{Tr}(J''(X_h(s)) Q_h) ds. \end{aligned}$$



## Cahn-Hilliard-Cook equation

$$\begin{aligned} & \mathbf{E}[J(X_h(t))] + \mathbf{E}\left[\int_0^t |J'(X_h(s))|_1^2 ds\right] \\ &= \mathbf{E}[J(P_h X_0)] + \underbrace{\mathbf{E}\left[\int_0^t \langle J'(X_h(s)), P_h dW(s) \rangle\right]}_{=0} \\ & \quad + \frac{1}{2} \mathbf{E}\left[\int_0^t \text{Tr}(J''(X_h(s))Q_h) ds\right] \\ &= \mathbf{E}[J(P_h X_0)] + \frac{1}{2} \mathbf{E}\left[\int_0^t \text{Tr}(A_h Q_h) ds\right] + \frac{1}{2} \mathbf{E}\left[\int_0^t \text{Tr}([f'(X_h(s)) \cdot] Q_h) ds\right] \end{aligned}$$

Proof completed by bounding these terms and using Gronwall's lemma.

For example,  $\text{Tr}(A_h Q_h) \leq C \|A^{1/2} Q^{1/2}\|_{\text{HS}}$ .

For the second variant: take  $\sup_{t \in [0, T]}$  before  $\mathbf{E}$ .

## Cahn-Hilliard-Cook equation

$$\begin{aligned} & \mathbf{E}[J(X_h(t))] + \mathbf{E}\left[\int_0^t |J'(X_h(s))|_1^2 ds\right] \\ &= \mathbf{E}[J(P_h X_0)] + \underbrace{\mathbf{E}\left[\int_0^t \langle J'(X_h(s)), P_h dW(s) \rangle\right]}_{=0} \\ & \quad + \frac{1}{2} \mathbf{E}\left[\int_0^t \text{Tr}(J''(X_h(s))Q_h) ds\right] \\ &= \mathbf{E}[J(P_h X_0)] + \frac{1}{2} \mathbf{E}\left[\int_0^t \text{Tr}(A_h Q_h) ds\right] + \frac{1}{2} \mathbf{E}\left[\int_0^t \text{Tr}([f'(X_h(s)) \cdot] Q_h) ds\right] \end{aligned}$$

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This was used to prove convergence (without rate) for the spatially semidiscrete approximation. Kovács–Larsson–Mesforush, 2011.

Now we want to imitate this for the fully discrete approximation.

Difficulty: no Itô formula.

## Moment bound

Fully discrete scheme:

$$X_h^j - X_h^{j-1} + kA_h(A_h X_h^j + P_h f(X_h^j)) = P_h \Delta W^j$$

Define  $Y_h^j = A_h X_h^j + P_h f(X_h^j)$  (discrete chemical potential).

$$X_h^j - X_h^{j-1} + kA_h Y_h^j = P_h \Delta W^j$$

## Theorem

Let  $p \geq 1$ . If  $\|A^{1/2}Q^{1/2}\|_{HS} \leq L$  and

$$|X_h^0|_1 + \mathcal{F}(X_h^0) + |Y_h^0|_1 \leq L,$$

then there exists  $C, k_0 > 0$ , depending on  $p, T$  and  $L$ , such that

$$\mathbf{E} \sup_{1 \leq j \leq N} |X_h^j|_1^{2p} + \mathbf{E} \sup_{1 \leq n \leq N} \mathcal{F}(X_h^n)^p + \mathbf{E} \left( \sum_{j=1}^N k |Y_h^j|_1^2 \right)^p \leq C, \quad k \leq k_0.$$

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**Proof.** Multiply by  $A_h^{-1}(X_h^j - X_h^{j-1})$  does not work:

$$\begin{aligned} \langle X_h^j - X_h^{j-1}, A_h^{-1}(X_h^j - X_h^{j-1}) \rangle + k \langle A_h(A_h X_h^j + P_h f(X_h^j)), A_h^{-1}(X_h^j - X_h^{j-1}) \rangle \\ = \langle P_h \Delta W^j, A_h^{-1}(X_h^j - X_h^{j-1}) \rangle \end{aligned}$$

$$|X_h^j - X_h^{j-1}|_{-1,h}^2 + k(J(X_h^j) - J(X_h^{j-1})) + \dots = \underbrace{\frac{1}{2} |P_h \Delta W^j|_{-1,h}^2}_{\approx k} + \frac{1}{2} |X_h^j - X_h^{j-1}|_{-1,h}^2$$

## Proof, continued

$$X_h^j - X_h^{j-1} + kA_h Y_h^j = P_h \Delta W^j$$

Multiply by  $Y_h^j$ :

$$\langle X_h^j - X_h^{j-1}, Y_h^j \rangle + k|Y_h^j|_1^2 = \langle Y_h^j, P_h \Delta W^j \rangle$$

Here:

$$\begin{aligned} \langle X_h^j - X_h^{j-1}, Y_h^j \rangle &= \langle X_h^j - X_h^{j-1}, A_h X_h^j \rangle + \langle X_h^j - X_h^{j-1}, P_h f(X_h^j) \rangle \\ &\geq \frac{1}{2} (|X_h^j|_1^2 - |X_h^{j-1}|_1^2 + |X_h^j - X_h^{j-1}|_1^2) \\ &\quad + \mathcal{F}(X_h^j) - \mathcal{F}(X_h^{j-1}) - \beta^2 \|X_h^j - X_h^{j-1}\|^2 \\ &= J(X_h^j) - J(X_h^{j-1}) + \frac{1}{2} |X_h^j - X_h^{j-1}|_1^2 - \beta^2 \|X_h^j - X_h^{j-1}\|^2 \end{aligned}$$

Sum up (with  $\Delta X_h^j := X_h^j - X_h^{j-1}$ ):

$$J(X_h^n) + k \sum_{j=1}^n |Y_h^j|_1^2 + \frac{1}{2} \sum_{j=1}^n |\Delta X_h^j|_1^2 \leq J(X_h^0) + \sum_{j=1}^n \langle Y_h^j, P_h \Delta W^j \rangle + \beta^2 \sum_{j=1}^n \|\Delta X_h^j\|^2$$

Here:

$$\sum_{j=1}^n \langle Y_h^j, P_h \Delta W^j \rangle = \sum_{j=1}^n \langle Y_h^{j-1}, P_h \Delta W^j \rangle + \sum_{j=1}^n \langle \Delta Y_h^j, P_h \Delta W^j \rangle$$

Raise to power  $p$ , take  $\sup_{1 \leq n \leq N}$  and then **E**.

Most difficult term:

$$\sum_{j=1}^n \langle \Delta Y_h^j, P_h \Delta W^j \rangle = \sum_{j=1}^n \langle A_h(X_h^j - X_h^{j-1}) + P_h(f(X_h^j) - f(X_h^{j-1})), P_h \Delta W^j \rangle$$

Delicate calculation of the Lipschitz constant of  $f$  using auxiliary moment bounds...

## Auxiliary moment bound 1

### Lemma

Let  $p \geq 1$ . If  $\|Q^{1/2}\|_{HS} < \infty$  and  $|X_0|_{-1,h} < L$  then there exists a  $C > 0$  and  $k_0 > 0$  such that, for  $0 < k < k_0$ ,

$$\mathbf{E} \sup_{1 \leq j \leq N} |X_h^j|_{-1,h}^{2p} \leq C,$$

$$\mathbf{E} \left( \sum_{j=1}^n \left( |X_h^j - X_h^{j-1}|_{-1,h}^2 + k |X_h^j|_1^2 \right) \right)^p \leq C.$$

## Auxiliary moment bound 1

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**Proof.** Take inner products with  $A_h^{-1} X_h^j$  and use

$$\langle P_h f(v_h), v_h \rangle = \langle f(v_h), v_h \rangle \geq -C_0 - C_1 \|v_h\|^2.$$

Apply Gronwall's Lemma to finish.



## Auxiliary moment bound 2

### Lemma

Suppose that  $\|Q^{1/2}\|_{HS} < \infty$  and  $\|X^0\| + |X_h^0|_{-1,h} < L$ . Then, for every  $\epsilon, \delta > 0$  and  $p \geq 1$ , there is  $C = C(T, \epsilon, \delta, p, X^0, \|Q^{1/2}\|_{HS}) > 0$  and  $K = K(T, p) > 0$  and  $k_0 = k_0(\epsilon, p)$  such that for  $0 < k < k_0$ ,

$$\mathbf{E} \sup_{1 \leq j \leq N} \|X_h^j\|^{2p} + \mathbf{E} \left( \sum_{j=1}^N \|X_h^j - X_h^{j-1}\|^2 \right)^p \leq C + K\delta \mathbf{E} \left( \sum_{j=1}^N k |Y_h^j|_1^2 \right)^{\frac{1+\epsilon}{2} p}.$$

## Auxiliary moment bound 2

### Lemma

Suppose that  $\|Q^{1/2}\|_{HS} < \infty$  and  $\|X^0\| + |X_h^0|_{-1,h} < L$ . Then, for every  $\epsilon, \delta > 0$  and  $p \geq 1$ , there is  $C = C(T, \epsilon, \delta, p, X^0, \|Q^{1/2}\|_{HS}) > 0$  and  $K = K(T, p) > 0$  and  $k_0 = k_0(\epsilon, p)$  such that for  $0 < k < k_0$ ,

$$\mathbf{E} \sup_{1 \leq j \leq N} \|X_h^j\|^{2p} + \mathbf{E} \left( \sum_{j=1}^N \|X_h^j - X_h^{j-1}\|^2 \right)^p \leq C + K\delta \mathbf{E} \left( \sum_{j=1}^N k |Y_h^j|_1^2 \right)^{\frac{1+\epsilon}{2} p}.$$

**Proof.** Taking inner products with  $X_h^j$  but keep in mind that

$$\langle A_h^{\frac{1}{2}} P_h f(v_h), A_h^{\frac{1}{2}} v_h \rangle \not\geq -c |v_h|_1^2, \quad v_h \in S_h.$$

However we may use from the previous result that with any  $q \geq 1$ ,

$$\mathbf{E} \left( \sum_{j=1}^n k |X_h^j|_1^2 \right)^q \leq C.$$

## Uniform bounds

We now have

$$\mathbf{E} \sup_{1 \leq j \leq N} (|X_h^j|_1^2 + \|X_h^j\|_{L^4}^4) \leq K_T$$

$$\mathbf{E} \sup_{t \in [0, T]} |X(t)|_1^2 + \mathbf{E} \sup_{t \in [0, T]} \|X(t)\|_{L^4}^4 \leq K_T$$

$$\mathbf{E} \sup_{t_n \in [0, T]} \frac{\|W_A(t_n) - W_{A_h}^n\|^2}{(h^2 + k^{1/2})^2} \leq K_T$$

Therefore, by Chebyshev's inequality, for any  $\epsilon$  there is  $\Omega_{h,k}^\epsilon \subset \Omega$  with  $\mathbf{P}(\Omega_{h,k}^\epsilon) > 1 - \epsilon$ , such that

$$\sup_{1 \leq j \leq N} (|X_h^j|_1^2 + \|X_h^j\|_{L^4}^4) \leq K_{\epsilon, T}$$

$$\sup_{t \in [0, T]} |X(t)|_1^2 + \mathbf{E} \sup_{t \in [0, T]} \|X(t)\|_{L^4}^4 \leq K_{\epsilon, T}$$

$$\sup_{t_n \in [0, T]} \frac{\|W_A(t_n) - W_{A_h}^n\|^2}{(h^2 + k^{1/2})^2} \leq K_{\epsilon, T}$$

on  $\Omega_{h,k}^\epsilon$ . We can now estimate the error using deterministic methods on  $\Omega_{h,k}^\epsilon$ .

## Chebyshev's inequality

Let  $F$  be a random variable with  $\mathbf{E}[F] \leq K$ .

Chebyshev's inequality gives, for every  $\alpha > 0$ ,

$$\mathbf{P}\left(\{\omega \in \Omega : F > \alpha\}\right) \leq \frac{1}{\alpha} \mathbf{E}[F] \leq \frac{K}{\alpha}.$$

We choose  $\alpha = \epsilon^{-1}K$  and set  $\Omega_\epsilon = \{\omega \in \Omega : F \leq \epsilon^{-1}K\}$ . Then

$$\mathbf{P}(\Omega_\epsilon) = 1 - \mathbf{P}\left(\{\omega \in \Omega : F > \alpha\}\right) \geq 1 - \epsilon.$$

Recall that we have

$$\mathbf{E} \sup_{1 \leq j \leq N} (|X_h^j|_1^2 + \|X_h^j\|_{L^4}^4) \leq K_T$$

$$\mathbf{E} \sup_{t \in [0, T]} |X(t)|_1^2 + \mathbf{E} \sup_{t \in [0, T]} \|X(t)\|_{L^4}^4 \leq K_T$$

$$\mathbf{E} \sup_{t_n \in [0, T]} \frac{\|W_A(t_n) - W_{A_h}^n\|^2}{(h^2 + k^{1/2})^2} \leq K_T$$

We apply Chebyshev with  $F = \sup_{1 \leq j \leq N} (|X_h^j|_1^2 + \|X_h^j\|_{L^4}^4)$ , and so on.

## Theorem

Suppose that  $\|A^{1/2}Q^{1/2}\|_{HS} < \infty$  and that

$$|X_h^0|_1 + \mathcal{F}(X_h^0) + |Y_h^0|_1 + |X_0|_2 \leq L.$$

Let  $h, k > 0$  small and  $0 < \epsilon, \delta < 1$ . Then, there is  $\Omega_{h,k}^\epsilon \subset \Omega$  with  $\mathbf{P}(\Omega_{h,k}^\epsilon) > 1 - \epsilon$ , and  $C = C(T, L, \epsilon, \delta)$  such that for all  $\omega \in \Omega_{h,k}^\epsilon$ ,

$$\|X(t_n) - X_h^n\| \leq C(h^{2(1-\delta)} + k^{1/2(1-\delta)}), \quad t_n \in [0, T].$$

**Proof.** The proof is based on the mild formulation, using the moment bounds, the error estimate for the stochastic convolution, and Gronwall's inequality on  $\Omega_{h,k}^\epsilon$ .

**Remark.** If only

$$|X_h^0|_1 + \mathcal{F}(X_h^0) + |Y_h^0|_1 + |X_0|_1 \leq L,$$

then

$$\|X(t_n) - X_h^n\| \leq C(h + k^{1/4}), \quad t_n \in [0, T].$$

## Proof of the main result

The main result:

$$\lim_{h,k \rightarrow 0} \mathbf{E} \sup_{t_n \in [0, T]} \|X(t_n) - X_h^n\|^2 = 0.$$

Proof:

$$\begin{aligned} \mathbf{E} \sup_{0 \leq t_n \leq T} \|X(t_n) - X_h^n\|^2 &\leq \int_{\Omega_{h,k}^\epsilon} \sup_{0 \leq t_n \leq T} \|X(t_n) - X_h^n\|^2 d\mathbf{P} \\ &\quad + 2 \int_{(\Omega_{h,k}^\epsilon)^c} \sup_{0 \leq t_n \leq T} (\|X(t_n)\|^2 + \|X_h^n\|^2) d\mathbf{P} \\ &\leq C_\epsilon (h^2 + k^{1/2}) + 4\epsilon^{1/2} \left( \int_{(\Omega_{h,k}^\epsilon)^c} \sup_{0 \leq t_n \leq T} (\|X(t_n)\|^4 + \|X_h^n\|^4) d\mathbf{P} \right)^{1/2} \\ &\leq C_\epsilon (h^2 + k^{1/2}) + 4\epsilon^{1/2} \left( \mathbf{E} \sup_{0 \leq t_n \leq T} (\|X(t_n)\|^4 + \|X_h^n\|^4) \right)^{1/2} \\ &\leq C_\epsilon (h^2 + k^{1/2}) + 4\epsilon^{1/2} |\mathcal{D}|^{1/2} \left( \mathbf{E} \sup_{0 \leq t_n \leq T} (\|X(t_n)\|_{L_4}^4 + \|X_h^n\|_{L_4}^4) \right)^{1/2} \\ &\leq C_\epsilon (h^2 + k^{1/2}) + 4\epsilon^{1/2} |\mathcal{D}|^{1/2} K_T. \end{aligned}$$