

Discretization of a stable SPDE in space and time

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Starting point: Consider the SPDE

$$dU_t = \Delta U_t dt + f(U_t) dt + dW_t, \quad U_0 = u_0$$

on $H = L^2(\mathcal{O})$ where \mathcal{O} is a bounded region in \mathbb{R}^n with sufficiently smooth boundary and

- $f : H \rightarrow H$ a nonlinear operator which is Lipschitz continuous:

$$|f(u) - f(v)| \leq L|u - v|$$

- $\Delta \in L(V, V')$ with DBC.

Thereby: $V = H_0^1(\mathcal{O})$. This leads to a Gelfand triplet:

$$V \subseteq H \equiv H' \subseteq V'.$$

Random dynamical systems: Let $(\Omega, \mathcal{F}, \mathbb{P})$ be a probability space, $(X, \|\cdot\|_X)$ a normed space and $\theta = (\theta_t)_{t \in \mathbb{R}}$ a metric dynamical system.

A measurable mapping

$$\varphi : (\mathbb{R}^+ \times \Omega \times X, \mathcal{B}(\mathbb{R}^+) \otimes \mathcal{F} \otimes \mathcal{B}(X)) \rightarrow (X, \mathcal{B}(X))$$

is called a *random dynamical system* if it satisfies the cocycle property, i.e.

$$\begin{aligned}\varphi(0, \omega, x_0) &= x_0 \\ \varphi(t + \tau, \omega, x_0) &= \varphi(t, \theta_\tau \omega, \varphi(\tau, \omega, x_0))\end{aligned}$$

for all $t, \tau \geq 0, \omega \in \Omega$ and $x_0 \in X$.

Random fixed points: A measurable random variable

$$X^* : (\Omega, \mathcal{F}) \rightarrow (X, \mathcal{B}(X))$$

is called a *random fixed point* if

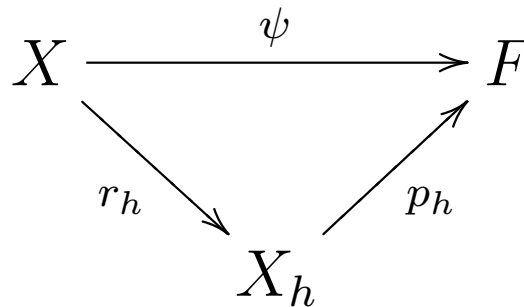
$$\varphi(t, \omega, X^*(\omega)) = X^*(\theta_t \omega)$$

holds for all $t > 0$.

Finite differences:

Definition: An *external approximation* of a space $(X, \|\cdot\|_X)$ consists of

- (i) a normed space $(F, \|\cdot\|_F)$ and an isomorphism ψ of X into F .
- (ii) a family $\{X_h, p_h, r_h\}_{h \in \mathcal{H}}$, in which for each $h \in \mathcal{H}$:
 - $(X_h, \|\cdot\|_h)$ is a normed space
 - p_h is a linear continuous mapping of X_h into F
 - r_h is a mapping of X into X_h .



For \mathcal{H} we use

$$\mathcal{H} = \prod_{i=1}^n]0, h_{i,0}] \quad \text{where} \quad 0 < h_i \leq h_{i,0}.$$

h_i is the mesh in x_i direction. For the characterization of an external approximation we need two more definitions:

Definition: The prolongations operators are said to be stable if their norms

$$\|p_h\| = \sup_{\|u_h\|_h=1} \|p_h u_h\|_F$$

can be majorized independently of h .

Definition: An external approximation is said to be stable if the following two conditions are fulfilled:

1. compatibility condition:

$$\lim_{h \rightarrow 0} \|p_h r_h u - \psi u\|_F = 0 \quad \text{for all } u \in X.$$

2. synchronization condition:

For each sequence $(u_h) \subseteq X_h$ with

$$p_h u_h \xrightarrow{h \rightarrow 0} \phi \quad \text{in } F$$

we have

$$\phi \in \psi X.$$

Now we introduce the approximation of V .

Let $F = L^2(\mathcal{O})^{n+1}$ and

$$\psi : V \rightarrow F, \quad u \mapsto (u, D_1 u, D_2 u, \dots, D_n u).$$

The approximated space V_h is given by

$$V_h = \left\{ u_h \mid u_h(x) = \sum_{P \in \mathring{\mathcal{O}}_h^1} u_h(P) 1_{\sigma_h(P)}(x), \quad u_h(P) \in \mathbb{R}^n \right\}$$

and the prolongation resp. restriction operators by

$$\begin{aligned} p_h u_h &= (u_h, \delta_1 u_h, \delta_2 u_h, \dots, \delta_n u_h) \\ (r_h u)(P) &= \frac{1}{m(\sigma_h(P))} \int_{\sigma_h(P)} u(x) \, dx \quad \forall P \in \mathring{\mathcal{O}}_h^1. \end{aligned}$$

Theorem: The preceding external approximation of V is stable and convergent. Moreover, for all h , V_h is finite dimensional, a subspace of $L^2(\mathcal{O})$ and a Hilbert space with the scalar product

$$((u_h, v_h))_h = \sum_{i=0}^n (\delta_i u_h, \delta_i v_h).$$

The next theorem is a discrete analogon of the Poincaré inequality and enables us to equip V_h with another scalar product:

Theorem: Let $u_h \in V_h$. Then

$$|u_h| \leq d_{\mathcal{O}} |\delta_i u_h|.$$

Therefore $((\cdot, \cdot))_h$ may start at $i = 1$.

From SPDE to RPDE: It's very convenient for our approach to rewrite the SPDE as a RPDE. For that we need to introduce the Ornstein-Uhlenbeck-process:

Definition: The unique stationary solution of the SPDE

$$dU_t = \Delta U_t dt + dW_t$$

is called Ornstein-Uhlenbeck-process. It's given by

$$\widehat{U}(\theta_t \omega) \equiv \widehat{U}_t(\omega) = \int_{-\infty}^t S_{\Delta}(t-s) dW_s(\omega)$$

where $S_{\Delta}(t-s)$ denotes the strongly continuous semigroup generated by Δ .

The OU-process is tempered, i.e. the mapping

$$t \mapsto |\widehat{U}(\theta_t \omega)|$$

has subexponential growth.

Subtracting the solution of the OU-process from any solution U_t of the SPDE we see that the difference $X_t = U_t - \widehat{U}_t$ is pathwise differential in time and satisfies pathwise the RPDE

$$\frac{\partial}{\partial t} X_t = \Delta X_t + f(X_t + \widehat{U}_t).$$

Spatial approximation: We are now able to approximate the given RPDE by finite differences. For that we define

$$X_t^h(x) = \sum_{P \in \mathring{O}_h^1} X_t^h(P) 1_{\sigma_h(P)}(x), \quad X_t^h(P) \in \mathbb{R}^n$$

$$\Delta_h = \sum_{i=1}^n \delta_i^2 \Rightarrow \langle -\Delta_h u_h, u_h \rangle \geq d_{\mathcal{O}}^{-2} |u_h|^2$$

$$\widehat{U}_t^h(P) = (r_h \widehat{U}_t)(P) \Rightarrow |\widehat{U}_t^h(\theta_t \omega)| \leq |\widehat{U}_t(\theta_t \omega)|$$

and get

$$\frac{\partial}{\partial t} X_t^h = \Delta_h X_t^h + f(X_t^h + \widehat{U}_t^h), \quad X_0^h = x_0^h. \quad (1)$$

Theorem: Let

$$C := \frac{1}{d_{\mathcal{O}}^2} - L > 0.$$

Then the solution of (1) generates a RDS φ_h which has a unique random fixed point.

Discretization in time: Implicit Euler scheme:

$$X_{m+1}^h = X_m^h + \left[\Delta_h X_{m+1}^h + f_h(X_{m+1}^h + \widehat{U}_{(m+1)k}^h) \right] k, \quad X_0^h = x_0^h \quad (2)$$

with

$$f_h(u_h)(x) := \sum_{P \in \mathring{\mathcal{O}}_h^1} 1_{\sigma_h(P)}(x) \frac{1}{m(\sigma_h(P))} \int_{\sigma_h(P)} f(u_h)(y) dy.$$

Then f_h is Lipschitz continuous with constant L .

Theorem: Let

$$C := \frac{1}{d_{\mathcal{O}}^2} - L > 0.$$

Then the solution of (2) generates a RDS $\varphi_{h,k}$ which has a unique random fixed point.

Convergence of the spatial discretization:

Theorem: As $h \rightarrow 0$ it holds

$$X_h^*(\omega) \rightarrow X^*(\omega) \quad \text{in } H.$$

Therefore we get the following diagram:

$$\begin{array}{ccc} \varphi(t, \theta_{-t}\omega, x_0) & \longrightarrow & X^*(\omega) \\ & & \uparrow h \rightarrow 0 \\ \varphi_h(t, \theta_{-t}\omega, x_0^h) & \longrightarrow & X_h^*(\omega) \\ & & \uparrow k \rightarrow 0 \\ \varphi_{h,k}(nk, \theta_{-nk}\omega, x_0^h) & \longrightarrow & X_{h,k}^*(\omega) \end{array}$$

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