

### Motivation: Random partial differential equations

Given a probability space  $(\Omega, \mathcal{F}, \mathbb{P})$  and a Hilbert space  $(H, (\cdot, \cdot)_H)$ , we consider the random partial differential equation

$$\frac{\partial}{\partial t} u(t, x) = A(t, x)u(t, x) + F(u(t, x)), \quad (t, x) \in \mathbb{T} \times \mathcal{D}$$

on a time interval  $\mathbb{T} = [0, T]$  and spatial domain  $\mathcal{D} \subset \mathbb{R}^n$  equipped with some initial and boundary conditions and

- a stochastic operator  $A$  acting on  $H$ , for example  $A = \nabla \cdot (L(t, x)\nabla)$ , where  $L = (L(t, \cdot), t \in \mathbb{T})$  is a  $H$ -valued stochastic process
- a (possibly nonlinear) operator  $F : H \rightarrow H$ .

This equation could be, for instance, a stochastic version of the porous media equation. For various applications it might be more realistic to model  $L$  discontinuously, i.e. as an infinite-dimensional Lévy process, then also called *Lévy field*.

### Approximation and simulation of Lévy fields

#### Truncated Karhunen-Loève expansions and GH fields

Let  $L = (L(t), t \in \mathbb{T})$  be a square-integrable,  $H$ -valued Lévy process on  $\mathbb{T}$ . Then there exists a non-negative, symmetric trace class covariance operator  $Q$  on  $H$  with a sequence of orthonormal eigenpairs  $((\rho_i, e_i), i \in \mathbb{N})$  and  $L$  admits the spectral decomposition

$$L(t) = \sum_{i \in \mathbb{N}} (L(t), e_i)_H e_i.$$

The Lévy field  $L$  may then be approximated by the truncated sum

$$L_N(t) = \sum_{i \in \mathbb{N}} \sqrt{\rho_i} e_i \ell_i(t),$$

where  $(\ell_i, i \in \mathbb{N})$  is a sequence of one-dimensional Lévy processes on  $\mathbb{T}$ . An important subclass of Lévy fields with various applications in finance and physics are *generalized hyperbolic (GH) fields*, which are based on the generalized hyperbolic distribution. In this case, the marginal processes  $\ell_i$  in  $L$  resp.  $L_N$  are given by one-dimensional GH processes. We use the representation of GH processes as subordinated Brownian Motions to obtain

$$L_N(t) = \sum_{i=1}^N \sqrt{\rho_i} e_i \left( \mu t + \Gamma \beta Y(t) + \sqrt{\Gamma} W_N(Y(t)) \right)_i, \quad (1)$$

where  $\mu, \beta \in \mathbb{R}^N$ ,  $\Gamma \in \mathbb{R}^{N \times N}$  are parameters of the GH field  $L$ ,  $W_N$  is a  $N$ -dimensional Brownian motion and  $Y = (Y(t), t \in \mathbb{T})$  a *generalized inverse Gaussian process* independent of  $W_N$ . Formula (1) yields an efficient sampling algorithm for  $L_N$  whose performance is independent of the truncation index  $N$ . Further, we have constructed  $L_N$  in such a way, that the approximation itself is a Lévy field with known marginal distributions. In other words, we know the law of the process  $(L_N(t, x), t \in \mathbb{T})$  for an arbitrary fixed  $x \in \mathcal{D}$ .

Often, as in the GH case, the processes  $\ell_i$  have *infinite activity*, i.e. infinitely many jumps in every compact time interval for  $\mathbb{P}$ -almost all trajectories, and, therefore, need to be approximated.

#### Fourier inversion method for one-dimensional processes

For a broad class of one-dimensional Lévy processes  $\ell_i$ , the characteristic function  $\phi_{\ell_i} : \mathbb{R} \rightarrow \mathbb{C}$  is explicitly available. We exploit the knowledge of  $\phi_{\ell_i}$  and a Fourier inversion property of characteristic functions to draw samples of the increment  $\ell_i(t + \Delta t) - \ell_i(t)$ , where  $\Delta t > 0$  is a small step in time. This procedure results in a piecewise constant approximation  $\tilde{\ell}_i$  of  $\ell_i$  on  $\mathbb{T}$  and involves numerical integration and therefore a certain error. With some relatively weak assumptions on  $\phi_{\ell_i}$ , we have shown that  $\tilde{\ell}_i$  converges to  $\ell_i$  in distribution. Further, if  $\ell_i$  has a finite  $p$ -th moment for  $p \geq 1$ , then we have the error bound

$$\sup_{t \in \mathbb{T}} \mathbb{E}[|\ell(t) - \tilde{\ell}(t)|^p] < C_{\ell, p} \Delta t$$

with some constant  $C_{\ell, p} > 0$ , hence  $\tilde{\ell}_i$  converges in  $L^p(\Omega; \mathbb{R})$  towards  $\ell_i$  uniformly on  $\mathbb{T}$ . Since  $C_{\ell, p}$  can be determined or at least bounded from above, this new result allows us to quantify the approximation error of infinite-dimensional Lévy processes.

### Simulation of Lévy fields and mean-square error estimation

Including the aforementioned discretization, we are able to approximate a given  $H$ -valued Lévy process  $L$  by

$$\tilde{L}_N(t) = \sum_{i \in \mathbb{N}} \sqrt{\rho_i} e_i \tilde{\ell}_i(t),$$

where the processes  $\tilde{\ell}_1, \dots, \tilde{\ell}_N$  are obtained by Fourier inversion. The mean-square approximation error in  $H$  is then bounded by

$$\sup_{t \in \mathbb{T}} \|L(t) - \tilde{L}_N(t)\|_{L^2(\Omega; H)} < \left( C_{\ell, 2} \Delta t \sum_{i=1}^N \rho_i \right)^{1/2} + \left( T \sum_{i=N+1}^{\infty} \rho_i \right)^{1/2} \quad (2)$$

and we achieve  $\tilde{L}_N \xrightarrow{L^2(\Omega; H)} L$  on  $\mathbb{T}$  as  $N \rightarrow \infty$  and  $\Delta t \rightarrow 0$ .

### Numerical examples

As a test for the novel approximation method, we simulate *hyperbolic Lévy fields*, which is next to the *normal inverse Gaussian (NIG)* field the most popular example of a GH Lévy field. In the simulations below,  $\mathbb{T} := [0, 1]$ ,  $\Delta t := 2^{-6}$  and the spatial domain is  $\mathcal{D} := [0, 1]$ . We use two different covariance operators to show the effects of a varying roughness of  $Q$  on the generated fields. To equilibrate both contributions in (2),  $N$  is coupled to  $\Delta t$  and the decay of the eigenvalues  $\rho_i$  of  $Q$ . As Table 1 indicates, the truncation index decreases as the correlation throughout the field increases.

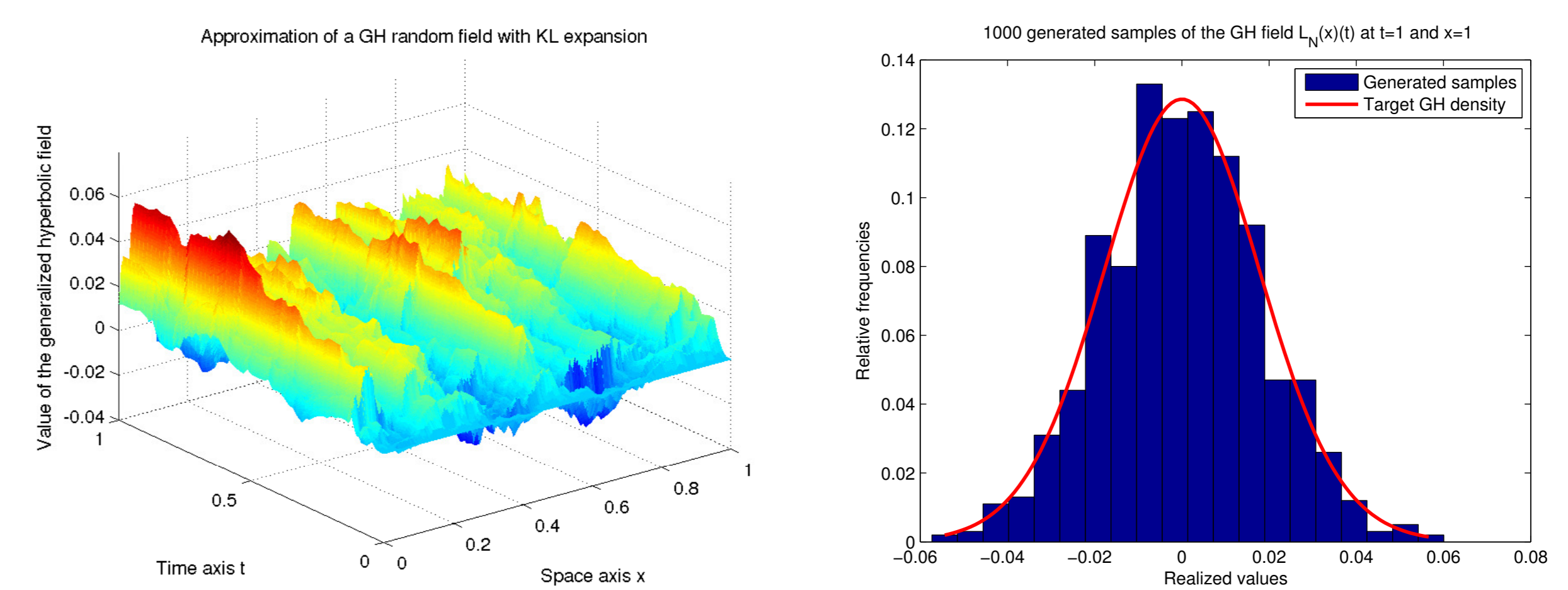


Figure 1 Sample and empirical distribution of a hyperbolic field with exponential covariance operator.

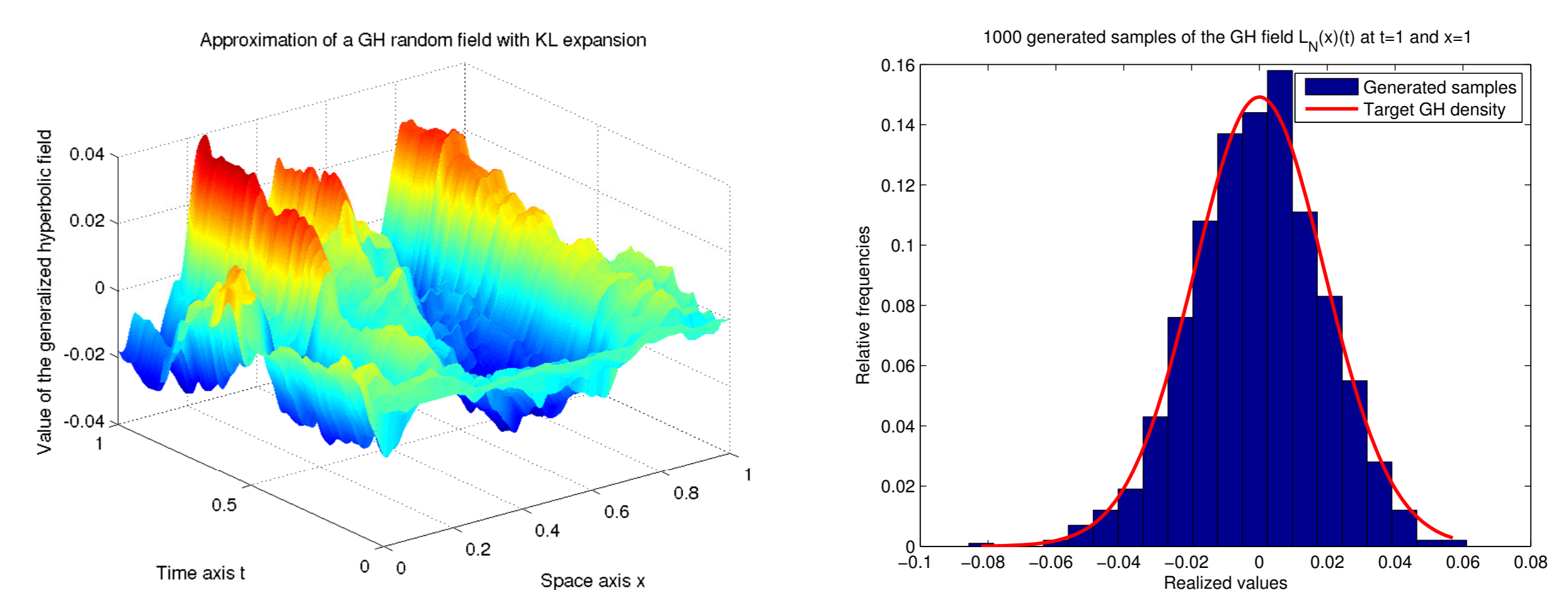


Figure 2 Sample and empirical distribution of a hyperbolic field with Matérn-1.5 covariance operator.

	Truncation index $N$	$L^2(\Omega; H)$ -error in % of the overall variance	Simulation time
Exponential covariance	132	6.30%	0.0759 sec.
Matérn-1.5 covariance	18	6.07%	0.0752 sec.

Table 1 Approximation errors and simulation times for hyperbolic Lévy fields.

### References

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- [3] S. Peszat and J. Zabczyk. *Stochastic Partial Differential Equations with Lévy Noise*, volume 113 of *Encyclopedia of Mathematics and its Applications*. Cambridge University Press, Cambridge, 2007.