

# **Nonlinear Approximation and Besov Regularity of a Class of Random Fields**

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# Outline

- I. The Stochastic Model
- II. Linear and Best  $N$ -Term Approximation
- III. Besov Regularity

Joint work with

*P. Cioica, S. Dahlke, S. Kinzel (all from Marburg),  
F. Lindner, R. Schilling (both from Dresden),  
and T. Raasch (Mainz).*

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# I. The Stochastic Model

## Given

- a bounded Lipschitz domain  $D \subset \mathbb{R}^d$ ,
- a Riesz basis  $(\psi_{j,k})_{j \in \mathbb{N}_0, k \in \nabla_j}$  for  $L_2(D)$  with  $\#\nabla_j \asymp 2^{jd}$ ,  
i.e., for

$$f = \sum_{j=0}^{\infty} \sum_{k \in \nabla_j} c_{j,k} \cdot \psi_{j,k} \in L_2(D)$$

it holds

$$\|f\|^2 \asymp \sum_{j=0}^{\infty} \sum_{k \in \nabla_j} c_{j,k}^2$$

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Let  $Y_{j,k}, Z_{j,k}$  be independent with

$$Y_{j,k} \sim B(1, 2^{-\beta jd}), \quad Z_{j,k} \sim N(0, 2^{-\alpha jd}).$$

## Define

$$X = \sum_{j=0}^{\infty} \sum_{k \in \nabla_j} Y_{j,k} Z_{j,k} \cdot \psi_{j,k}.$$

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Note that  $X$  is Gaussian iff  $\beta = 0$ .

## II. Linear and Best $N$ -Term Approximation

Put

$$e(\widehat{X}) = \left( E \|X - \widehat{X}\|_{L_2(D)}^2 \right)^{1/2}.$$

**Linear approximation errors**

$$e_N^{\text{lin}} = \inf e(\widehat{X}_N)$$

with infimum over all  $\widehat{X}_N$  such that

$$\dim \text{span}(\widehat{X}_N(\Omega)) \leq N,$$

see *Mathé (1990)*.

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**Theorem 1 C-D-K-L-R-R-S (2010)**

$$e_N^{\text{lin}} \asymp N^{-\frac{\alpha+\beta-1}{2}}.$$



For  $f = \sum_{j=0}^{\infty} \sum_{k \in \nabla_j} c_{j,k} \cdot \psi_{j,k} \in L_2(D)$

$$\eta(f) = \#\{(j, k) : c_{j,k} \neq 0\}.$$

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Two variants of **best  $N$ -term approximation**

$$e_N = \inf e(\hat{X}_N)$$

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and

$$e_N^{\text{avg}} = \inf e(\hat{X}_N)$$

with infimum over all  $\hat{X}_N$  such that

$$E(\eta(\hat{X}_N)) \leq N.$$

**Theorem 2 C-D-K-L-R-R-S (2010)**

$$e_N^{\text{avg}} \asymp \begin{cases} N^{-\frac{\alpha+\beta-1}{2(1-\beta)}}, & \text{if } \beta < 1 \\ 2^{-\alpha d N/2}, & \text{if } \beta = 1, \end{cases}$$

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and for every  $\varepsilon > 0$

$$e_N \preceq \begin{cases} N^{-\frac{\alpha+\beta-1}{2(1-\beta)}} + \varepsilon, & \text{if } \beta < 1 \\ N^{-1/\varepsilon}, & \text{if } \beta = 1. \end{cases}$$

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## Remark

- Simulation of  $\widehat{X}_N$  that yields upper bound for  $e_N^{\text{avg}}$  at cost  $N(\ln N)^2$ .
- For  $\beta < 1$

$$e_{N^{1-\beta}}^{\text{avg}} \preceq e_N^{\text{lin}}.$$

Nonlinear approximation of stochastic processes,  $D = [0, 1]$ ,

- wavelet methods for piecewise stationary processes, *Cohen, d'Ales (1997), Cohen, Daubechies, Guleryuz, Orchard (2002)*,
- free Knot splines for Brownian motion, SDEs, *Kon, Plaskota (2005), Creutzig, Müller-Gronbach, R (2007), Slassi (2010)*,
- free Knot splines for Lévy driven SDEs, *Dereich (2010), Dereich, Heidenreich (2010)*.

# III. Besov Regularity

In the sequel  $p, q, s > 0$ .

**Assumption** For all  $f = \sum_{j=0}^{\infty} \sum_{k \in \nabla_j} c_{j,k} \cdot \psi_{k,k} \in L_2(D)$

$$\|f\|_{B_q^s(L_p(D))}^q \asymp \sum_{j=0}^{\infty} 2^{j(s+d(1/2-1/p))q} \left( \sum_{k \in \nabla_j} c_{j,k}^p \right)^{q/p}$$



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**Theorem 3 C-D-K-L-R-R-S (2010)**

$X \in B_q^s(L_p(D))$   $P$ -a.s. iff

$$\frac{\alpha - 1}{2} + \frac{\beta}{p} > \frac{s}{d},$$

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**Remark** See *Abramovich, Sapatinas, Silverman (1998)* and *Bochkina (2006)* for  $d = 1$  and  $p, q \geq 1$ .

Application: Bayesian non-parametric regression.

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Proof: A.s convergence or weak convergence of

$$\sum_{k \in \nabla_j} Y_{j,k} \left( 2^{\alpha j d / 2} |Z_{j,k}| \right)^p .$$

## Work in progress

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### Parabolic SPDE

$$dU(t) = (\Delta U(t) + A(t, U(t))) dt + B(t, U(t)) dX(t).$$