

Convergence Rates of Sparse Tensor FEM for elliptic sPDEs

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3. Infinite - Dimensional Parametric Deterministic BVP
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Problem Formulation

$$D \subset \mathbb{R}^d \quad \text{bounded, Lipschitz}, \quad d = 2, 3.$$

Consider (model) deterministic Boundary Value Problem (dBVP):

given

$$a \in L^\infty(D), \quad \operatorname{ess\,inf}_{x \in D} a(x) \geq a_0 > 0, \quad f \in H^{-1}(D) = (H_0^1(D))',$$

find $u \in H_0^1(D)$ such that

$$b(u, v) := \int_D a(x) \nabla_x u \cdot \nabla_x v dx = \int_D f(x) v(x) dx \quad \forall v \in H_0^1(D). \quad (1)$$

Existence, Uniqueness, Regularity, AFEM,:

What to do if $a(x)$ is “uncertain” ?

- Accurate numerical solutions for *one* $a(x)$ are of limited use.
- **Assume** complete statistics (joint pdf's) on data $a(x)$ available.
- Reformulate (1) as sPDE.
- Reconsider numerical solution methods for (1):
 - **Given** statistics (law) of random input data $a(\omega, x)$ (KL-expansion)
 - **compute** statistics (law) of random solution $u(\omega, x)$ ('gPC'-expansion)
 - **sampling** vs. **parsimonious param. representation of joint pdf's of** $u(\omega, x)$
 - trade **randomness** for **high-dimensionality**

Elliptic BVP with stochastic data

Given:

- probability space (Ω, Σ, P) on data space $X(D) \subseteq L^\infty(D)$, $V \subseteq H^1(D)$,
- random diffusion coefficient $a(x, \omega) \in L^\infty(\Omega, dP; X(D))$,
- deterministic source term $f \in H^{-1}(D) = (H_0^1(D))'$,

(sBVP) Find $u(x, \omega) \in L^2(\Omega, dP; H_0^1(D))$ such that

$$\mathbb{E} \left[\int_D a(x, \cdot) \nabla_x u(x, \cdot) \cdot \nabla_x v(x, \cdot) dx \right] = \mathbb{E} \left[\int_D f(x) v(x, \cdot) dx \right]$$

for all $v \in L^2(\Omega, dP; H_0^1(D))$

$a \in L^\infty(\Omega, dP; X(D))$ and $\text{essinf } a(\cdot, \cdot) \geq a_0 > 0 \Rightarrow \exists u \in L^2(\Omega, dP; H_0^1(D))$.

Monte Carlo

Sampling (sBVP): Each ‘sample’ = 1 deterministic BVP

1. Generate (in parallel) N_Ω data “samples” $\{a_j(x)\}_{j=1}^{N_\Omega}$,
2. Solve (exactly and in parallel) the N_Ω dBVPs

$$-\nabla_x \cdot (a_j(x) \nabla_x u_j) = f \quad \text{in } D, \quad u = 0 \quad \text{on } \partial D$$

3. Estimate **k -point correlations** $\mathcal{M}^k u$ from N_Ω solution samples $\{u_j(x)\}_{j=1}^{N_\Omega}$ ($k = 1$: estimate mean field $\mathbb{E}[u]$ from sample average $\mathbb{E}^{N_\Omega}[u]$).

Theorem 0: Assume $u \in L^\alpha(\Omega, V)$ for some $\alpha \in (1, 2]$ with $V = H_0^1(D)$.

Then ex. $C(\alpha)$ such that for every $N_\Omega \geq 1$ and every $0 < \varepsilon < 1$

$$P \left\{ \|\mathbb{E}[u] - \mathbb{E}^{N_\Omega}[u]\|_V \leq C\varepsilon^{-1/\alpha} N_\Omega^{-(\alpha-1)/2} \|u\|_{L^\alpha(\Omega, V)} \right\} \geq 1 - \varepsilon$$

Karhunen-Loève expansion

- separation of deterministic and stochastic variables -

Proposition 1 (Karhunen-Loève)

If $a \in L^2(\Omega, dP; L^\infty(D))$ then in $L^2(\Omega, dP; L^2(D))$,

$$a(x, \omega) = \mathbb{E}[a](x) + \sum_{m \geq 1} \sqrt{\lambda_m} \varphi_m(x) Y_m(\omega) = \mathbb{E}[a](x) + \sum_{m \geq 1} \psi_m(x) Y_m(\omega)$$

- $(\lambda_m, \varphi_m)_{m \geq 1}$ eigensequence of **covariance operator**

$$\mathcal{C}[a] : L^2(D) \rightarrow L^2(D) \quad (\mathcal{C}[a]v)(x) := \int_D C_a(x, x') v(x') dx' \quad \forall v \in L^2(D),$$

-

$$C_a(x, x') = \mathbb{E} [(a(x, \cdot) - \mathbb{E}[a](x))(a(x', \cdot) - \mathbb{E}[a](x'))]$$

-

$$Y_m(\omega) := \frac{1}{\sqrt{\lambda_m}} \int_D (a(x, \omega) - \mathbb{E}[a](x)) \varphi_m(x) dx : \Omega \rightarrow \Gamma_m \subseteq \mathbb{R} \quad m = 1, 2, \dots$$

Karhunen-Loève expansion

- convergence -

$$a(x, \omega) = E_a(x) + \sum_{m \geq 1} \sqrt{\lambda_m} \varphi_m(x) Y_m(\omega)$$

KL expansion converges in $L^2(D \times \Omega)$, not necessarily in $L^\infty(D \times \Omega)$

To ensure $L^\infty(D \times \Omega)$ convergence, must

- estimate decay rate of KL eigenvalues λ_m : Schwab & Todor JCP (2006),
- bound $\|\varphi_m\|_{L^\infty(D)}$: Todor Diss ETH (2005), SINUM (2006)
- **assume**: bounds for $\|Y_m\|_{L^\infty(\Omega)}$

Karhunen-Loève expansion

- eigenvalue estimates -

Regularity of C_a ensures decay of KL-eigenvalue sequence $(\lambda_m)_{m \geq 1}$

$C_a(x, x') : D \times D \rightarrow \mathbb{R}$ is

- **piecewise analytic on $D \times D$** if ex. **smoothness partition** $\mathcal{D} = \{D_j\}_{j=1}^J$ of D into a finite sequence of simplices D_j such that

$$\overline{D} = \bigcup_{j=1}^J \overline{D_j} \quad (2)$$

and $C_a(x, x')$ analytic in open neighbourhood of $\overline{D_j} \times \overline{D_{j'}}$ for any (j, j') .

- **piecewise $H^{t,t}$ on $D \times D$** if

$$V_a \in H_{pw}^{t,t}(D \times D) := \bigcap_{i,j \leq J} L^2(D_i, H^t(D_j))$$

Karhunen-Loève expansion

- eigenvalue estimates -

- $(H, \langle \cdot, \cdot \rangle)$ Hilbert space,
- $\mathcal{C} \in \mathcal{K}(H)$ compact, s.a.,
- eigenpair sequence $(\lambda_m, \phi_m)_{m \geq 1}$.

If $\mathcal{C}_m \in \mathcal{B}(H)$ is any operator of rank at most m ,

$$\lambda_{m+1} \leq \|\mathcal{C} - \mathcal{C}_m\|_{\mathcal{B}(H)}. \quad (3)$$

(e.g. Pinkus 1985: n -widths in Approx. Theory).

Karhunen-Loève expansion

- eigenvalue estimates -

Proposition 2 (KL-eigenvalue decay)

- ($>$ exponential KL decay: *Gaussian* $C_a(x, x')$)

$$C_a(x, x') := \sigma^2 \exp(-\gamma|x - x'|^2) \implies 0 \leq \lambda_m \leq c(\gamma, \sigma)/m! \quad \forall m \geq 1$$

- (exponential KL decay: *Piecewise analytic* $C_a(x, x')$)

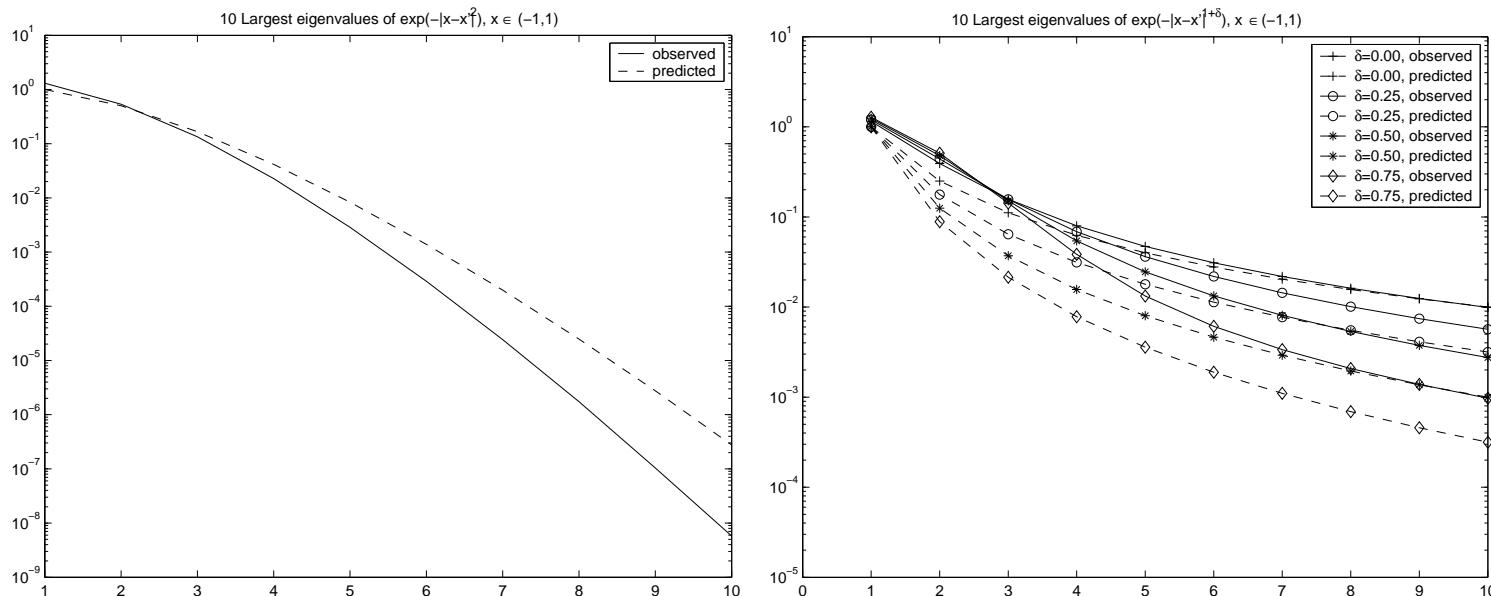
$$C_a \text{ pw analytic on } D \times D \implies \exists c > 0 \quad 0 \leq \lambda_m \leq c \exp(-bm^{1/d}) \quad \forall m \geq 1$$

- (Algebraic KL-eigenvalue decay for p.w. $H^t(D)$ -kernels)

$$C_a \in H_{pw}^{t,t}(D \times D) \ (t \geq d/2) \implies 0 \leq \lambda_m \leq cm^{-t/d} \quad \forall m \geq 1$$

Karhunen-Loève expansion

- eigenvalue estimates -



Karhunen-Loève expansion

- eigenfunction estimates -

Regularity of C_a ensures L^∞ bounds for L^2 -scaled eigenfunctions $(\varphi_m)_{m \geq 1}$

Theorem 3 (Schwab & Todor JCP 2006)

Assume

$$C_a \in H_{pw}^{t,t}(D \times D) \quad \text{for} \quad t > d.$$

Then

$$\forall \delta > 0 \quad \text{ex. } C(\delta) > 0 \quad \text{s.t.} \quad \forall m \geq 1 : \quad \|\varphi_m\|_{L^\infty(D)} \leq C(\delta) \lambda_m^{-\delta}.$$

Hence:

$$b_m := \frac{\lambda_m^{1/2} \|\varphi_m\|_{L^\infty(D)}}{\inf_{x \in D} \mathbb{E}[a]} \leq C(\delta) \lambda_m^{1/2-\delta} \leq C(\delta) m^{-t/2d-\delta}$$

Karhunen-Loève expansion

- convergence rate -

Conclusion:

KL expansion of

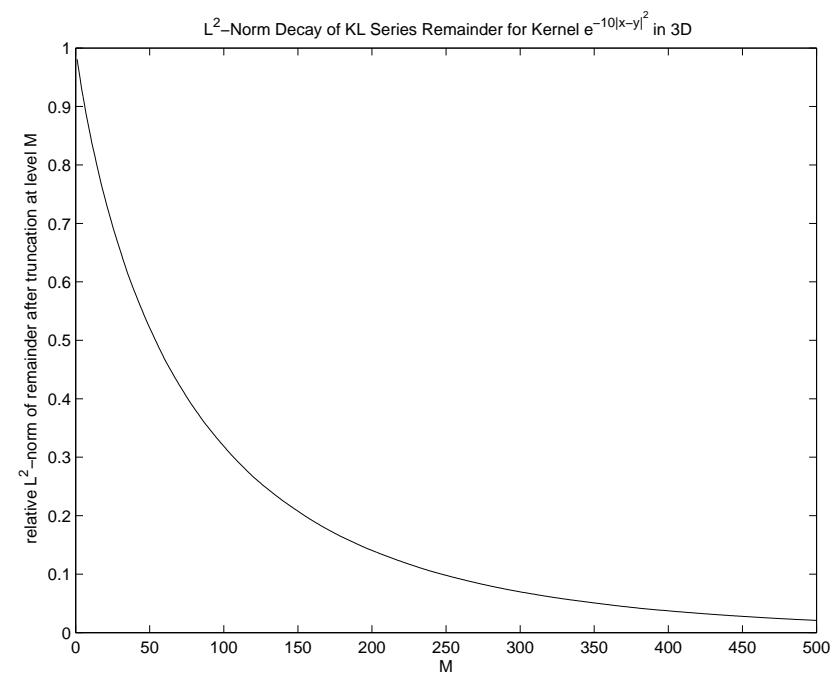
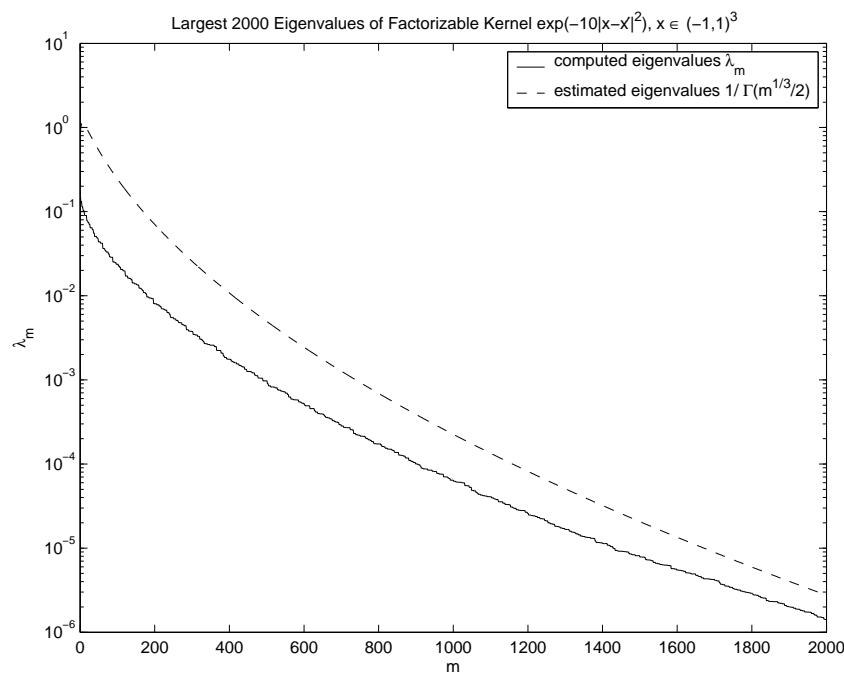
$$a(x, \omega) \in L^2(\Omega, dP; L^\infty(D))$$

converges **uniformly and exponentially** on $D \times \Omega$ if

- $C_a(x, x')$ piecewise analytic
- $(Y_m(\omega))_{m \geq 1}$ uniformly bounded on Ω (e.g. $Y_m(\omega) \sim \mathcal{U}(-1, 1)$)

Karhunen-Loève expansion

- convergence rate -



Karhunen-Loève expansion

- truncation from infinite to finite dimension $\textcolor{red}{M}$ -

$\infty > \textcolor{red}{M} \in \mathbb{N}$ KL-truncation order

$$a_M(x, \omega) := \mathbb{E}[a](x) + \sum_{m \geq 1}^{\textcolor{red}{M}} \sqrt{\lambda_m} \varphi_m(x) Y_m(\omega)$$

(SBVP) with stochastic coefficient $a(x, \omega)$

$$-\operatorname{div}(a(x, \omega) \nabla_x u(x, \omega)) = f(x) \quad \text{in } L^2(\Omega, dP; H^{-1}(D))$$

(SBVP) $_M$ with truncated stochastic coefficient $a_M(x, \omega)$

$$-\operatorname{div}(a_M(x, \omega) \nabla_x u_M(x, \omega)) = f(x) \quad \text{in } L^2(\Omega, dP; H^{-1}(D))$$

Theorem 4 If C_a pw analytic and $(Y_m)_{m \geq 1}$ uniformly bounded, then $\forall \delta > 0$ ex. $b, C(\delta), M_0 > 0$ such that (SBVP) $_M$ well-posed for $M \geq M_0$ and

$$\|u - u_{\textcolor{red}{M}}\|_{L^2(\Omega; H_0^1(D))} \leq \begin{cases} C \exp(-b \textcolor{red}{M}^{1/d}) & \forall \textcolor{red}{M} \geq M_0 \quad \text{if } C_a \text{ pw analytic} \\ C(\delta) \textcolor{red}{M}^{-t/2d+1-\delta} & \forall \textcolor{red}{M} \geq M_0 \quad \text{if } C_a \in H_{pw}^{t,t}(D \times D) \end{cases}$$

High dimensional deterministic bvp

$$a_M : D \times \Omega \rightarrow \mathbb{R}, \quad a_M(x, \omega) = \mathbb{E}[a](x) + \sum_{m \geq 1}^M \sqrt{\lambda_m} \varphi_m(x) Y_m(\omega)$$

Assumption

$(Y_m)_{m \geq 1}$ independent, uniformly bounded family of rv's
 (e.g. Y_m uniformly distributed in $\Gamma_m = I = (-1/2, 1/2)$, $m = 1, 2, 3, \dots$)

$$\begin{array}{ccc} \text{Random variable } Y_m & \longrightarrow & \text{Parameter } y_m \in I \\ (Y_1, Y_2, \dots, Y_M) & \longrightarrow & y = (y_1, y_2, \dots, y_M) \in I^M \\ dP = \rho(y) dy & = & \bigotimes_{m \geq 1} \rho_m(y_m) dy_m \end{array}$$

$$\tilde{a}_M : D \times I^M \rightarrow \mathbb{R}, \quad \tilde{a}_M(x, y) = \mathbb{E}[a](x) + \sum_{m \geq 1}^M \sqrt{\lambda_m} \varphi_m(x) y_m$$

High dimensional deterministic bvp

stochastic bvp

$$-\operatorname{div}(a_M(x, \omega) \nabla_x u_M(x, \omega)) = f(x) \quad \text{in } H^{-1}(D), \quad P\text{-a.e. } \omega \in \Omega$$

M -dimensional, parametric deterministic bvp

$$-\operatorname{div}(\tilde{a}_M(x; y_1, y_2, \dots, y_M) \nabla_x \tilde{u}_M(x, y)) = f(x) \quad \text{in } H^{-1}(D), \quad \forall y \in I^M$$

Proposition 5(Doob & Dynkin)

Under **Assumption**, the parametric deterministic bvp is well-posed and

$$u_M(x, \omega) = \tilde{u}_M(x, Y_1(\omega), Y_2(\omega), \dots, Y_M(\omega))$$

High dimensional deterministic bvp

- stochastic semi-discretization -

$$\tilde{a}_M(x, y) := E_a(x) + \sum_{m \geq 1}^M \sqrt{\lambda_m} \varphi_m(x) y_m$$

M -dimensional, parametric deterministic bvp: find

$\tilde{u}_M \in L_\rho^2(I^M; H_0^1(D)) \simeq L_\rho^2(I^M) \otimes H_0^1(D) = S \otimes V$: $\forall v \in L_\rho^2(I^M; H_0^1(D))$ holds

$$\int_{I^M} \left(\int_D \tilde{a}_M(x, y) \nabla_x \tilde{u}_M(x, y) \cdot \nabla_x v(x, y) dx \right) \rho(y) dy = \int_{I^M} \int_D f(x) v(x) dx \rho(y) dy$$

Galerkin semi-discretization in y (sGFEM):

$$S^M \subset L_\rho^2(I^M), \quad \hat{N}_\Omega := \dim S^M < \infty \quad dBVPs$$

find $\tilde{U}_M \in S^M \otimes H_0^1(D)$ such that $\forall v \in S^M \otimes H_0^1(D)$:

$$\int_{I^M} \left(\int_D \tilde{a}_M(x, y) \nabla_x \tilde{U}_M(x, y) \cdot \nabla_x v(x, y) dx \right) \rho(y) dy = \int_{I^M} \int_D f(x) v(x) dx \rho(y) dy$$

sGFEM for high dimensional deterministic bvp

- stochastic semi-discretization -

Quasi-Optimality:

$$\|u - \tilde{U}_M\|_{L^2_\rho(H_0^1)} \leq C \inf_{v \in S^M \otimes H_0^1(D)} \|u - v\|_{L^2_\rho(H_0^1)}$$

$\tilde{a}_M(x, y)$ affine in $y \Rightarrow \tilde{u}_M(x, y)$ analytic in $y \Rightarrow S^M$ polyn. space w.r.to y

task: solve dbvp with KL-accuracy* $O(\exp(-cM^{1/d}))$ in “low complexity” **

*how to choose the polynomial space in $y = (y_1, y_2, \dots, y_M)$?

**how to choose a basis \mathcal{B} of \mathcal{P} ?

sGFEM for high dimensional deterministic bvp

- stochastic semi-discretization: p.w. analytic C_a -

'ANOVA' type Tensor Product Spaces in I^M :

For $M, \mu \geq 0$, $1 \leq \nu << M \in \mathbb{N}_0$ define index sets

$$\Lambda_{\mu,\nu}^M := \{\alpha \in \mathbb{N}_0^M \mid |\alpha|_1 \leq \mu, \quad |\alpha|_0 \leq \nu\} \subset \mathbb{N}_0^M, \quad (4)$$

polynomial subspaces (N. Wiener (1938))

$$\mathbf{y}^\alpha = y_1^{\alpha_1} y_2^{\alpha_2} \dots y_m^{\alpha_m} \dots, \quad \mathcal{L}_\alpha(y) = L_{\alpha_1}(y_1) L_{\alpha_2}(y_2) \dots \quad \alpha \in \Lambda$$

$$S^M = \mathcal{P}(\Lambda_{\mu,\nu}^M) := \text{span}\{\mathbf{y}^\alpha \mid \alpha \in \Lambda_{\mu,\nu}^M\} \subset L^2(I^M), \quad (5)$$

$$S^M = \mathcal{L}(\Lambda_{\mu,\nu}^M) := \text{span}\{\mathcal{L}_\alpha(y) \mid \alpha \in \Lambda_{\mu,\nu}^M\} \subset L^2(I^M), \quad (6)$$

sGFEM for high dimensional deterministic bvp

- stochastic semi-discretization: p.w. analytic C_a -

Theorem 6 (Todor + Sc IMA Journ Numer. Anal. (2007))
If ex. $b, C, \kappa > 0$ s.t.

$$\lambda_m \leq C \exp(-bm^\kappa) \quad m \rightarrow \infty,$$

ex. $c_3, c_4, c_r > 0$ such that for

$$\mu = \lceil c_3 M^\kappa \rceil, \quad \nu = \lceil c_4 M^{\kappa/(\kappa+1)} \rceil \quad (7)$$

holds, as $M \rightarrow \infty$ for **polynomial subspace** $\mathcal{P}(\Lambda_{\mu,\nu}^M) \otimes H_0^1(D)$

i. (\mathcal{P}): ex. $b, \hat{c} > 0$ s.t.

$$\begin{aligned} \inf_{v \in \mathcal{P}(\Lambda_{\mu,\nu}^M) \otimes H_0^1(D)} \|\tilde{u}_M - v\|_{L^\infty(I^M; H_0^1(D))} &\lesssim \exp(-bM^{1/d}) \\ N_\Omega := \dim \mathcal{P}(\Lambda_{\mu,\nu}^M) &\lesssim \exp(\hat{c}M^{1/(d+1)} \log(M)) \end{aligned} \quad (8)$$

ii. sGFEM converges w. **spectral rate**:

$$\forall s > 0 : \text{ex. } C(s) \text{ s.t. } \inf_{v \in \mathcal{P}(\Lambda_{\mu,\nu}^M) \otimes H_0^1(D)} \|\tilde{u}_M - v\|_{L^\infty(I^M; H_0^1(D))} \leq C(s) N_\Omega^{-s}$$

iii. (\mathcal{B}): In L_ρ^2 -ONbasis of $\mathcal{P}(\Lambda_{\mu,\nu}^M)$ the stiffness matrix of $(sBVP)_M$ in I^M is well-conditioned and sparse (at most $O(M)$ nontrivial “entries” / row)

High dimensional dBVP - nonlinear approximation

$$C_a \in H_{pw}^{t,t}(D \times D), \quad t < \infty$$

Recall: $C_a \in H_{pw}^{t,t}(D \times D) \Rightarrow a(x, \omega) = \mathbb{E}[a](x) + \sum_{m \geq 1} \psi_m(x) Y_m(\omega)$

$$\lambda_m \lesssim m^{-t/d}, \quad \|\psi_m\|_{L^\infty(D)} = \lambda_m^{1/2} \|\varphi_m\|_{L^\infty(D)} \lesssim m^{-t/2d-\delta} \quad m = 1, 2, \dots$$

KL - convergence rate: if $t > 2d$ then ex. $M_0 > 0$ such that

$$\|u - u_M\|_{L^2(\Omega; H_0^1(D))} \lesssim \|a - a_M\|_{L^\infty(\Omega; L^\infty(D))} \leq CM^{-s} \quad \forall M \geq M_0, \quad 0 < s < t/2d - 1.$$

- Convergence rate of sGFEM of $O(N_\Omega^{-s'})$ possible?
- Which Λ ?
- Which $s'(t) > 1/2$?

High dimensional dBVP - nonlinear approximation

$$C_a \in H_{pw}^{t,t}(D \times D), \quad t < \infty$$

Benchmark:

find $M \leq \infty$ and $\Lambda \subset \mathbb{N}_0^{\mathbb{N}}$ by *nonlinear, best N_Ω -term approximation*.

Notation:

- $\Lambda = \{ \text{all sequences } \alpha = (\alpha_m)_{m=1}^{\infty} \subset \mathbb{N}_0 \text{ w. finite support } \} \subset \mathbb{N}_0^{\mathbb{N}}$
- Λ countable: define

$$\Lambda(M) := \{ \alpha : \text{supp } \alpha \subset \{1, \dots, M\} \} \subset \Lambda, \quad M = 1, 2, \dots$$

Then $\Lambda(M)$ countable and $\Lambda = \bigcup_{M=1}^{\infty} \Lambda(M)$, hence countable.

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$$\alpha \in \Lambda \quad \Rightarrow \quad |\alpha|_0 := \# \text{supp } \alpha < \infty, \quad |\alpha|_1 := \sum_{m \geq 1} \alpha_m < \infty.$$

High dimensional dBVP - nonlinear approximation

$$C_a \in H_{pw}^{t,t}(D \times D), \quad t < \infty$$

Let

$$y \in U := B_1(\ell_\infty) = (-1, 1)^\infty \quad V = H_0^1(D).$$

Then

$$u = \sum_{\alpha \in \Lambda} u_\alpha y^\alpha = \sum_{\alpha \in \Lambda} c_\alpha \mathcal{L}_\alpha(y) \in L^\infty(U, V)$$

where, for any $\alpha \in \Lambda$,

$$u_\alpha := \frac{1}{\alpha!} (D_y^\alpha u)(0) \in V, \quad c_\alpha := \int_U u(x, y) \mathcal{L}_\alpha(y) \rho(y) dy \in V.$$

High dimensional dBVP - nonlinear approximation

$$C_a \in H_{pw}^{t,t}(D \times D), \quad t < \infty$$

For any $\Lambda_0 \subset \Lambda$,

$$\inf_{v \in \mathcal{L}(\Lambda_0) \otimes V} \|u - v\|_{L_\rho^2(U,V)}^2 \leq \|u - \sum_{\alpha \in \Lambda_0} c_\alpha \mathcal{L}_\alpha\|_{L_\rho^2(U,V)}^2 \leq \sum_{\alpha \notin \Lambda_0^c} \|c_\alpha\|_V^2,$$

$$\inf_{v \in \mathcal{P}(\Lambda_0) \otimes V} \|u - v\|_{L^\infty(U,V)} \leq \|u - \sum_{\alpha \in \Lambda_0} u_\alpha y^\alpha\|_{L^\infty(U,V)} \leq \sum_{\alpha \notin \Lambda_0^c} \|u_\alpha\|_V$$

ℓ^τ -summability of $\{\|u_\alpha\|_V : \alpha \in \Lambda\}$, $\{\|c_\alpha\|_V : \alpha \in \Lambda\}$?

A-priori estimates of $\|u_\alpha\|_V$, $\|c_\alpha\|_V$ for $\alpha \in \Lambda$.

High dimensional dBVP - Regularity I

$$C_a \in H_{pw}^{t,t}(D \times D), \quad t < \infty$$

$$a(x, y) = a_0(x) + \sum_{m=1}^{\infty} y_m \psi_m(x), \quad y = (y_1, y_2, \dots) \in U$$

$$\text{essinf}_{x \in D} \{a_0(x)\} \geq a_{0,min} > 0, \quad \psi_m := \lambda_m^{1/2} \varphi_m(x), \quad m = 1, 2, \dots$$

Recall: if $C_a \in H_{pw}^{t,t}(D \times D)$, then for any $0 < s < t/2d$,

$$b_m := \frac{\|\psi_m\|_{L^\infty(D)}}{a_{0,min}}, \quad \tilde{b}_m := \frac{\|\psi_m\|_{L^\infty(D)}}{2a_{min}} \lesssim \lambda_m^{1/2} \|\varphi_m\|_{L^\infty(D)} \lesssim m^{-s}, \quad m = 1, 2, \dots$$

Then $\{b_m\}_{m \geq 1} \in \ell^\tau$ for any $1/\tau < t/(2d)$ and,

$$\forall \alpha \in \Lambda : \quad b^\alpha = \prod_{m \geq 1} b_m^{\alpha_m} < \infty.$$

High dimensional dBVP - Regularity II

$$C_a \in H_{pw}^{t,t}(D \times D), \quad t < \infty$$

Theorem 7

Let $\{b_m\}$, $\{\tilde{b}_m\}$ be as above and denote

$$\beta_m := b_m m^{1+\delta}, \quad \tilde{\beta}_m := \tilde{b}_m m^{1+\delta}, \quad \delta > 0.$$

Then, for every $\alpha \in \Lambda$,

$$\|u_\alpha\|_V \lesssim \|f\|_{V^*} \begin{cases} \frac{|\alpha|!}{\alpha!} b^\alpha, \\ \beta^\alpha \end{cases} \quad \text{and} \quad \|c_\alpha\|_V \lesssim \|f\|_{V^*} \begin{cases} \frac{|\alpha|!}{\alpha!} \tilde{b}^\alpha \\ \tilde{\beta}^\alpha \end{cases} \quad \alpha \in \Lambda.$$

High dimensional dBVP - Regularity III

$$C_a \in H_{pw}^{t,t}(D \times D), \quad t < \infty$$

Idea of Proof: consider

$$b = \{b_m\}_{m=1}^{\infty} \in \ell_1, \quad \|b\|_{\ell_1} \leq \gamma < 1,$$

$$y = (y_1, y_2, y_3, \dots) \in B_1(\ell_{\infty})$$

Then for

$$a(y) = 1 + b_1 y_1 + \dots = 1 + \sum_{m=1}^{\infty} b_m y_m$$

we have

$$a(y)u(y) = 1 \iff u(y) = \frac{1}{a(y)} = \frac{1}{1 + b_1 y_1 + b_2 y_2 + \dots}, \quad y \in B_1(\ell_{\infty}).$$

$a(y)$ is *linear in each y_i* and

$u(y)$ is *continuous in $y \in B_1(\ell_{\infty})$* and *analytic in each y_i* .

Moreover, for every $\alpha \in \Lambda$,

$$D_y^\alpha u(y) = D_y^\alpha [1 + b_1 y_1 + b_2 y_2 + \dots]^{-1} = (-1)^{|\alpha|} [a(y)]^{-|\alpha|} |\alpha|! \prod_{m \geq 1} b_m^{\alpha_m}$$

First bound: ‘real variable’ bootstrap argument.
 (Todor and Schwab, IMA Journ. Numer. Anal. 2007)

Second bound: ‘several complex variables’ argument.

Hartogs’ Theorem and Cauchy’s inequalities for functions $u(\cdot, z)$ which are *separately analytic w.r. to each z_i in the Bernstein ellipses \mathcal{E}_{ρ_i} , $i = 1, 2, \dots$* , imply $u(\cdot, z)$ are
jointly analytic in $z = (z_1, z_2, \dots)$ in the polycylinders

$$[-1, 1]^M \subset \mathcal{E}_{\rho_1} \times \dots \times \mathcal{E}_{\rho_M} \subset \mathbb{C}^M \quad \text{with } \rho_m \sim 1/b_m \rightarrow \infty \text{ as } m \rightarrow \infty$$

for any M .

High dimensional dBVP - convergence rate of sGFEM I

$$C_a \in H_{pw}^{t,t}(D \times D), \quad t < \infty$$

Best N_Ω -term approximation:

find finite sets $\Lambda_0 \subset \Lambda$ of monomials y^ν / Legendre polynomials $\mathcal{L}_\nu(y)$, $\nu \in \Lambda_0$, of cardinality $N_\Omega = \#\Lambda_0 \rightarrow \infty$, and “coefficients” $c_\mu \in V$ such that

$$\|u - \sum_{\nu \in \Lambda_0} c_\nu \mathcal{L}_\nu\|_{L_p^2(U,V)} \leq C(\#\Lambda_0)^{-\sigma(t)},$$

with, hopefully, $\sigma(t) > 1/2$ (MCM).

ok for $\sigma = \frac{1}{\tau} - \frac{1}{2} < \frac{t}{2d} - \frac{1}{2}$ **provided** $\{\|c_\alpha\|_V : \alpha \in \Lambda\} \in \ell^\tau$!

When are

$$\{\tilde{\beta}^\nu : \nu \in \Lambda\}, \quad \left\{ \frac{|\nu|!}{\nu!} \tilde{b}^\nu : \nu \in \Lambda \right\} \in \ell^\tau?$$

High dimensional dBVP - convergence rate of sGFEM II

$$C_a \in H_{pw}^{t,t}(D \times D), \quad t < \infty$$

Consider for $b = \{b_m\}$ the conditions

$$\|b\|_{\ell^1} = |b_1| + |b_2| + \dots < 1, \quad (9)$$

$$\|b\|_{\ell^\infty} = \sup_m \{|b_m|\} < 1, \quad (10)$$

and

$$b \in \ell^\tau \quad \text{for some } \tau < 1. \quad (11)$$

Proposition 8 (Cohen, DeVore & CS '08)

(10) and (11) \rightarrow

$$\{b^\alpha : \alpha \in \Lambda\} \in \ell^\tau,$$

(9) and (11) \rightarrow

$$\left\{ \frac{|\alpha|!}{\alpha!} b^\alpha : \alpha \in \Lambda \right\} \in \ell^\tau. \quad (12)$$

High dimensional dBVP - convergence rate of sGFEM III

$$C_a \in H_{pw}^{t,t}(D \times D), \quad t < \infty$$

Theorem 9

If

$$C_a \in H_{pw}^{t,t}(D \times D), \quad 2d < t < \infty$$

then the solution $u(x, y) \in L^2(U, V)$ admits the Legendre expansion

$$u(x, y) = \sum_{\alpha \in \Lambda} c_\alpha(x) \mathcal{L}_\alpha(y)$$

and the ‘coefficient sequence’ $\{c_\alpha : \alpha \in \Lambda\} \subset V$ satisfies

$$\{\|c_\alpha\|_V : \alpha \in \Lambda\} \in \ell^\tau \quad \text{for} \quad 1 < \frac{1}{\tau} < \frac{t}{2d}.$$

There exists a sequence $\{\Lambda_\ell\}_{\ell=1}^\infty$ of finitely supported index sets such that the sGFEM based on $\mathcal{P}(\Lambda_\ell) \otimes V$ converges with rate

$$\|u - u_{\Lambda_\ell}\|_{L_\rho^2(U, V)} \lesssim (\#\Lambda_\ell)^{-\sigma}, \quad 0 < \sigma < \frac{t}{2d} - \frac{1}{2}.$$

Number Theory (finding Λ_ℓ) - I

Problem: Theorem 9 does *not* give a constructive way of finding sets

$$\Lambda_1 \subset \Lambda_2 \subset \dots \subset \Lambda_\ell \subset \dots \Lambda$$

Two strategies: 1. Adaptive sGFEM, 2. A-priori selection of the Λ_ℓ .

Consider 2:

given $\mu = (\mu_1, \mu_2, \dots) > 0$ s.t. $1 > \mu_1 \geq \mu_2 \geq \dots \geq \mu_m \rightarrow 0$ and $\varepsilon > 0$, define

$$\Lambda_\varepsilon(\mu) := \{\alpha \in \mathbb{N}_0^\mathbb{N} \mid \mu^\alpha \geq \varepsilon\} = \{\alpha \in \mathbb{N}_0^\mathbb{N} \mid \mu^{-\alpha} \leq 1/\varepsilon\}$$

Based on [sharp] bounds on coefficients $\|u_\alpha\|_V$, $\|c_\alpha\|_V$,
 $\Lambda_\varepsilon(b)$ will contain essentially the $\#\Lambda_\varepsilon(b)$ ‘largest coefficients’ $\|u_\alpha\|_V$, $\|c_\alpha\|_V$.

Monotonicity: for any $\mu, \bar{\mu} \in \Lambda$ as above and any $\bar{\varepsilon}, \varepsilon > 0$

1. $M_\varepsilon(\mu) := \max_{m \in \mathbb{N}} \{\mu_m \geq \varepsilon\} < \infty$,
2. $\forall \alpha \in \Lambda_\varepsilon(\mu) : \text{supp}(\alpha) \subset \{1, 2, \dots, M_\varepsilon(\mu)\}$,
3. $\bar{\varepsilon} \geq \varepsilon$ implies $\Lambda_{\bar{\varepsilon}}(\mu) \subseteq \Lambda_\varepsilon(\mu)$,
4. $\mu \preceq \bar{\mu}$ implies $\Lambda_\varepsilon(\mu) \subseteq \Lambda_\varepsilon(\bar{\mu})$.

Number Theory (finding Λ_ℓ) - II

Consider *comparison sequences* $\nu_\sigma \in \ell^\infty$ such that

$$b \preceq \nu_\sigma := \{(m+1)^{-\sigma} : m = 1, 2, \dots\}, \quad \sigma > 0.$$

Note:

$$b \preceq \nu_\sigma \quad \Rightarrow \quad \forall \alpha \in \Lambda : \quad b^\alpha \leq \nu_\sigma^\alpha$$

Scaling: given $\sigma > 0$, find

$$\Lambda_\varepsilon(\nu_\sigma) = \Lambda_{\varepsilon^{1/\sigma}}(\nu_1) = \left\{ \alpha \in \mathbb{N}_0^\mathbb{N} : \prod_{m \in \mathbb{N}} (m+1)^{\alpha_m} \leq \varepsilon^{-1/\sigma} \right\}, \quad \varepsilon \rightarrow 0.$$

$\alpha \in \Lambda_\varepsilon(\nu_\sigma)$ iff α **multiplicative partition of some integer** $x < \varepsilon^{-1/\sigma}$.

Complexity (finding Λ_ℓ)

Proposition 10 (R. Andreev '08)

1. $\Lambda_\varepsilon(\nu_\sigma)$ can be localized in work and memory growing log-linearly in $\#\Lambda_\varepsilon(\nu_\sigma)$.
2. As $\varepsilon \rightarrow 0$,

$$\#\Lambda_\varepsilon(\nu_\sigma) \sim x \frac{e^{2\sqrt{\log x}}}{2\sqrt{\pi}(\log x)^{3/4}} \quad \text{with} \quad x = \varepsilon^{-1/\sigma},$$

- E.R. Canfield, Paul Erdős and C. Pomerance:
On a Problem of Oppenheim concerning “Factorisatio Numerorum”
J. Number Theory **17** (1983) 1 ff.,
- G. Szekeres and P. Turán:
Über das zweite Hauptproblem der “Factorisatio Numerorum”
Acta Litt. Sci. Szeged **6** (1933) 143-154.

Sparse Tensor sGFEM - I

Issues:

- Selection of polynomial basis:
gPC from 1-d polyn. orthog. w.r.to *separable* (prior) $\rho(y)$.
- So far, only semidiscrete approximation:
Fully Discrete sGFEM needs (A)FEM in D : $c_\mu \rightarrow c_\mu^L \in V_L \subset V$.

Sparse Tensor sGFEM - II

Regularity of deterministic problem:

- Smoothness Scale of approximation spaces:

$$V = \mathcal{A}^0 \supset \mathcal{A}^1 \supset \mathcal{A}^2 \supset \mathcal{A}^s \dots \supset C^\infty(\overline{D})$$

Examples (Dirichlet Problem w. random coefficient):

1. (Isotropic Sobolev Scale, h -FEM in D on quasiuniform meshes)

$$\mathcal{A}^0 = H_0^1(D), \quad \mathcal{A}^s = H^s(D) \cap H_0^1(D), \quad s > 1.$$

2. (Weighted Kondrat'ev Scale, h -FEM in D on graded meshes)

$$\mathcal{A}^0 = H_0^1(D), \quad \mathcal{A}^s = V_\beta^s(D) \cap H_0^1(D), \quad s > 1 \quad \text{integer.}$$

3. (Besov Scale, h -AFEM in D)

$$\mathcal{A}^0 = H_0^1(D), \quad \mathcal{A}^s = B_{2,\infty}^s(D) \cap H_0^1(D), \quad s > 1.$$

- Spatial regularity at order $s' > 1$:

$$u(y) = \sum_{\alpha \in \Lambda} y^\alpha u_\alpha, \quad y \in U, \quad u_\alpha \in \mathcal{A}^{s'} \subset V. \quad (13)$$

Sparse Tensor sGFEM - III

Proposition 11 (sGFEM /full tensor approximation):

Assume

1. spatial regularity (13) of order $s' > 1$: $\psi_\mu \in \mathcal{A}^{s'}$,
2. Covariance regularity of order t :

$$C_a \in H_{pw}^{t,t}(D \times D), \quad t > 2d,$$

3. multilevel spatial approximation scale:

$$V_0 \subset V_1 \subset V_2 \subset \dots \subset V$$

with uniformly bounded, V -stable and quasioptimal projectors $P_\ell : V \rightarrow V_\ell$ and

$$N_{D,\ell} := \dim V_\ell = O(2^{\ell d}), \quad \ell \rightarrow \infty.$$

Then ex. $\{\Lambda_k\}_{k=1}^\infty$ with $\Lambda_k \subset \Lambda$, $\#\Lambda_k \sim N_\Omega(k) \rightarrow \infty$ and

$$\|u - \sum_{\alpha \in \Lambda_k} y^\alpha P_\ell \psi_\alpha\|_{L^\infty(U;V)} \lesssim N_\Omega^{-r(t)} + N_D^{-s'/d}$$

with *total ‘number of DOF’*

$$N_{total} = N_\Omega N_D$$

Note: for Monte Carlo we have $r = 1/2$.

Sparse Tensorization (not possible for MCM):

$$u = \sum_{\alpha \in \Lambda} c_\alpha(x) \mathcal{L}_\alpha(y), \quad u_{\Lambda_\ell} = \sum_{\alpha \in \Lambda_\ell} c_\alpha(x) \mathcal{L}_\alpha(y).$$

Sparse Tensor discretization: approximate c_α by $c_{\ell(\alpha)} \in V_{\ell(\alpha)} \subset V$, $\alpha \in \Lambda_k$.

Error:

$$\|u - \sum_{\alpha \in \Lambda_k} c_{\ell(\alpha)} \mathcal{L}_\alpha\|_{L^2(U,V)}^2 \leq \|u - \sum_{\alpha \in \Lambda_k} c_\alpha \mathcal{L}_\alpha\|_{L^2(U,V)}^2 + \sum_{\alpha \in \Lambda_k} \|c_\alpha - P_{\ell(\alpha)} c_\alpha\|_V^2$$

If $\{\|c_\alpha\|_{\mathcal{A}^s} : \alpha \in \Lambda\} \in \ell^\tau(t)$, proper selection of Λ_ℓ and $\ell(\alpha)$ gives rate

$$N_\Omega^{-r} + N_D^{-s'/d} \quad \text{with} \quad N_{total} = N_\Omega \log N_D + N_D \log N_\Omega.$$

Conclusion

- Elliptic PDE with stochastic coefficients:
Variational Formulation, Existence, Uniqueness
- Karh  nen - Lo  ve Expansion of Random Input Data:
Exponential pointwise convergence for p.w. analytic C_a ,
Algebraic pointwise convergence for $C_a \in H_{pw}^{t,t}$,
- Fast Computation of KL-expansion in general domains $D \subset \mathbb{R}^3$
by gFMM (Rokhlin and Greengard), \mathcal{H} -Matrix techniques (Hackbusch etal),...
- Transform SPDE into parametric, deterministic PDE on $U = (-1, 1)^\infty$
- Truncation to $M < \infty$ dimensions; conditional expectation; error estimates.
- Convergence Rates of h -, p - type sGFEM for sparse tensor approximations of

$$B_1(\ell_\infty) \ni y \rightarrow u(y, \cdot) \in V$$

- Stochastic Regularity of random solution = domain of analyticity of parametric, deterministic problem
- Finite Smoothness Covariance: algebraic Conv. of h -sGFEM and Spectral Conv. of p -sGFEM as

$$h \rightarrow 0, \quad p \rightarrow \infty, \quad M \rightarrow \infty.$$

- Sparse collocation on input-KL adapted ‘lattices’ of integration points in $(-1, 1)^M$ (Schwab and Todor IMAJNA (2007))
- Sparse tensorization of D and Ω Galerkin discretizations:

$$N_{total} \simeq N_\Omega \log N_D + N_D \log N_\Omega$$

- Diffusion problems in physical dimension $d = 2$, with $M = 1500$.
- requires *mixed regularity* in *both*, D and U : if

$$u = \sum_{\alpha \in \Lambda} c_\alpha(x) \mathcal{L}_\alpha(y), \quad \text{then} \quad \{\|c_\alpha\|_{\mathcal{A}^s} : \alpha \in \Lambda\} \in \ell^\tau$$

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