

Multilevel Preconditioning for Isogeometric Analysis

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Setup

- Elliptic PDE on physical domain $\Omega \subset \mathbb{R}^d$, $d \geq 2$
- Isogeometric analysis $\mathbf{F} : \hat{\Omega} \rightarrow \Omega$ for parametric domain $\hat{\Omega} \subset [0, 1]^d$ with regular mapping \mathbf{F}

Goal: High order approximate solution with fast iterative method

Ingredients

- isogeometric discretization by tensor product B-splines of degree p
- iterative solution with **multilevel preconditioner** and **nested iteration**

\leadsto solution in **optimal linear complexity**

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Problem Setup

Elliptic PDE of order $2r$ on physical domain Ω

$$\begin{aligned} r = 1 : \quad & -\Delta u = f & \text{in } \Omega, & \quad u|_{\partial\Omega} = 0 \\ r = 2 : \quad & \Delta^2 u = f & \text{in } \Omega, & \quad u|_{\partial\Omega} = \mathbf{n} \cdot \nabla u|_{\partial\Omega} = 0 \end{aligned}$$

Weak operator form: for given $f \in H^{-r}(\Omega)$, find $u \in H_0^r(\Omega)$ such that

$$Au = f \quad \text{in } H^{-r}(\Omega)$$

Elliptic operator A defined by $\langle Av, w \rangle := a(v, w)$ symmetric, continuous

$$\text{and coercive on } H_0^r(\Omega): \quad \|Av\|_{H^{-r}(\Omega)} \sim \|v\|_{H^r(\Omega)}$$

Mapping $\mathbf{F} : \hat{\Omega} \rightarrow \Omega$ from parametric domain $\hat{\Omega} \subset [0, 1]^d$ to physical domain Ω

$$\mathbf{F} \text{ regular: } \|D\mathbf{F}(\mathbf{x})\| \sim 1 \text{ for all } \mathbf{x} \in \hat{\Omega}$$

Numerical Solution on Physical Space

Discretization on uniform grid: $V_h \subset H_0^r(\Omega)$ $\dim V_h < \infty$ \rightsquigarrow

$$A_h u_h = f_h \quad (*)$$

$0 < h < 1$ grid size

Goal: Realize discretization error accuracy ε

with minimal amount of work $\mathcal{O}(N(\varepsilon))$ in amount of unknowns $N(\varepsilon)$

Obstructions for fast numerical solution:

- o Large sparse linear system of equations (*) \rightsquigarrow iterative solver
- o **Convergence speed** of iterative solver depends on $\text{cond}_2(A_h)$
- o Standard discretizations with finite differences or finite elements $\rightsquigarrow \text{cond}_2(A_h) \sim h^{-2r}$
 $0 < h < 1$ grid size
- o High desired accuracy, resolution of singularities in data and/or geometry \rightsquigarrow **small h**
 \rightsquigarrow larger problem \rightsquigarrow worse condition number

Ingredients for reaching goal:

- Multilevel preconditioner C_h**
multigrid methods, BPX preconditioner, wavelet discretizations $\rightsquigarrow \text{cond}_2(C_h A_h) \sim 1$
- Nested iteration**

A-priori Estimates for Finite Elements

Quality measure: **Approximation in norm** $\|u - u_h\|_{L_2(\Omega)} \leq \varepsilon$

A-priori error estimates: $\Omega \subset \mathbb{R}^d$ $\dim V_h = N \sim h^{-d}$ uniform grid

$$\begin{aligned} & \|u - u_h\|_{L_2(\Omega)} \lesssim h^s \|u\|_{H^s(\Omega)} \quad u_h \in V_h \quad 0 \leq s \leq p+1 \\ \iff & \|u - u_N\|_{L_2(\Omega)} \lesssim N^{-s/d} \|u\|_{H^s(\Omega)} \\ & \hspace{10em} N \text{ degrees of freedom} \quad \longleftrightarrow \quad \text{accuracy } \mathcal{O}(N^{-(p+1)/d}) \end{aligned}$$

Approximation rate determined by

- (i) (piecewise polynomials of degree $p \rightsquigarrow$) approximation order $p+1$ of V_h
- (ii) space dimension d
- (iii) amount of smoothness of u in L_2

Target:

Realize discretization error accuracy $\varepsilon \sim h^{p+1} \sim 2^{-(p+1)J}$ for grid with spacing $h \sim 2^{-J}$

Problem complexity: For $h \sim 2^{-J}$ a total of $N \sim 2^{Jd}$ unknowns

Optimal complexity for iterative solver: Minimal amount of work is $\mathcal{O}(N)$

Isogeometric Elements of Degree p

Mesh on $[0, 1]$: $\Xi := \{\xi_1, \dots, \xi_{n+p+1}\}$ a p -open knot vector such that

$$0 = \xi_1 = \dots = \xi_{p+1} < \xi_{p+2} \leq \dots \leq \xi_n < \xi_{n+1} = \dots = \xi_{n+p+1} = 1$$

(allowing internal repeated knots having multiplicity $m_i \leq p - r + 1$)

B-splines of degree p on Ξ defined recursively by

$$p = 0: \quad N_{i,0}(\zeta) = \begin{cases} 1 & \text{if } \xi_i \leq \zeta < \xi_{i+1}, \\ 0 & \text{otherwise,} \end{cases}$$

$$p \geq 1: \quad N_{i,p}(\zeta) = \frac{\zeta - \xi_i}{\xi_{i+p} - \xi_i} N_{i,p-1}(\zeta) + \frac{\xi_{i+p+1} - \zeta}{\xi_{i+p+1} - \xi_{i+1}} N_{i+1,p-1}(\zeta)$$

$\leadsto n$ B-splines spanning spline space of piecewise polynomials of degree p

with $p - m_i$ continuous derivatives at inner nodes

$$S_h(\hat{\Omega}) := \text{span} \left\{ B_i(\mathbf{x}) := \prod_{\ell=1}^d N_{i_\ell,p}(x_\ell), \quad i = 1, \dots, N := nd \right\}$$

Mapping $\mathbf{F} = (F_1, \dots, F_d)^T \quad F_i \in S_{\tilde{h}}(\hat{\Omega})$ for some $\tilde{h} \gg h$

$$\leadsto V_h := \{v_h \in H_0^r(\Omega) : v_h \circ \mathbf{F} \in S_h(\hat{\Omega})\}$$

$$\|v\|_{L_2(\Omega)}^2 := \int_{\hat{\Omega}} |v(\mathbf{F}(\mathbf{x}))|^2 \|D\mathbf{F}(\mathbf{x})\| dx$$

Theorem

(S) Uniform stability of basis with respect to $L_2(\Omega)$

$$\left\| \sum_{i=1}^N c_i B_i \circ \mathbf{F}^{-1} \right\|_{L_2(\Omega)}^2 \sim \sum_{i=1}^N |c_i|^2 =: \|\mathbf{c}\|_{\ell_2}^2 \quad \text{for any } \mathbf{c} \in \ell_2 \quad \text{constants} = c(p, d) \neq c(h)$$

(J) Direct or Jackson estimates

$$\inf_{v_h \in V_h} \|v - v_h\|_{L_2(\Omega)} \lesssim h^s |v|_{H^s(\Omega)} \quad \text{for any } v \in H^s(\Omega) \quad 0 \leq s \leq p+1$$

(B) Inverse or Bernstein estimates

$$\|v_h\|_{H^s(\Omega)} \lesssim h^{-s} \|v_h\|_{L_2(\Omega)} \quad \text{for any } v_h \in V_h \text{ and } 0 \leq s \leq p$$

Multilevel Preconditioner

Asymptotically **optimal preconditioner**: C_h such that

$$\text{cond}_2(C_h A_h) \sim 1$$

and **setup** and **application** of C_h in optimal linear complexity $\mathcal{O}(N)$

Schwarz iterative schemes based on **subspace corrections**

↪ **Multilevel schemes** yielding **optimal** preconditioners:

- ▶ **Multiplicative schemes** ↪ **multigrid methods** Brandt, Braess, Bramble, Hackbusch, Zulehner ...
IgA: Gahleitner, Kraus, Tomar ...
- ▶ **Additive schemes** ↪ **BPX preconditioner; wavelet discretization** Bramble, Pasciak, Xu, Yserentant, Oswald, Dahmen, Kunoth ...

Relevant idea from Approximation Theory: **Multilevel characterization** of function spaces
and **norm equivalences**

Not optimal are preconditioners based on domain decomposition, overlapping Schwarz, hierarchical basis preconditioners. . . Beirao da Veiga, Cho, Pavarino. Scacci, Kleiss, Pechstein, Jüttler, Langer ...

Multilevel Characterization of Function Spaces

$V_h \longleftrightarrow V_j \quad h \sim 2^{-j} \quad j$ resolution level

Multiresolution $V_{j_0} \subset V_{j_0+1} \subset \dots \subset V_j \subset V_{j+1} \subset \dots \subset H_0^r(\Omega)$

$$\text{clos}_{H^r(\Omega)} \left(\bigcup_{j=j_0}^{\infty} V_j \right) = H_0^r(\Omega)$$

Linear orthogonal projectors $Q_j : H_0^r(\Omega) \rightarrow V_j$ s.th. $Q_j Q_\ell = Q_j$ for $j \leq \ell \rightsquigarrow Q_j - Q_{j-1}$ projector

Corollary

(S) Φ_j uniformly stable basis for V_j : $\|\mathbf{c}\|_{\ell_2} \sim \|\mathbf{c}^T \Phi_j\|_{L_2(\Omega)}$

(J) Jackson estimate

$$\inf_{v_j \in V_j} \|v - v_j\|_{L_2(\Omega)} \lesssim 2^{-sj} \|v\|_{H^s(\Omega)} \quad v \in H^s(\Omega) \quad 0 < s \leq \delta$$

(B) Bernstein inequality $\|v_j\|_{H^s(\Omega)} \lesssim 2^{sj} \|v_j\|_{L_2(\Omega)} \quad v_j \in V_j \quad s < \tau$

\implies Norm equivalence

$$\|v\|_{H^r(\Omega)}^2 \sim \sum_{j=j_0}^J 2^{2rj} \|(Q_j - Q_{j-1})v\|_{L_2(\Omega)}^2 \quad v \in V_J$$

Proof: Norm equivalence on $\hat{\Omega}$ [Dahmen, Kunoth '92] [Oswald '92]

(J) and discrete Hardy inequality \rightsquigarrow upper estimate for $\|\cdot\|_{H^s(\Omega)}$

(B), $\|Q_j\|_{L_2(\Omega)} \lesssim 1$ and Whitney estimate \rightsquigarrow lower estimate

F regular mapping

Norm Equivalence for Optimal Preconditioning

Corollary: For $H_0^r(\Omega)$ $C_J^{-1} := A_{j_0} Q_{j_0} \circ \mathbf{F}^{-1} + \sum_{j=j_0}^J 2^{2rj} (Q_j - Q_{j-1}) \circ \mathbf{F}^{-1}$
 is optimal preconditioner for $A_J : V_J \rightarrow V_J$: $\text{cond}_2(C_J^{1/2} A_J C_J^{1/2}) \sim 1$ as $J \rightarrow \infty$

Proof: [Jaffard '92], [Dahmen, Kunoth '92], [Oswald '92]
 Isomorphism $\|Av\|_{H^{-r}(\Omega)} \sim \|v\|_{H^r(\Omega)}$ combined with norm equivalence for $H_0^r(\Omega)$ and \mathbf{F} regular mapping

BPX realization of C_J^{-1} : replace $C_J = A_{j_0} Q_{j_0} \circ \mathbf{F}^{-1} + \sum_{j=j_0}^J 2^{-2rj} (Q_j - Q_{j-1}) \circ \mathbf{F}^{-1}$

by spectrally equivalent preconditioner

$$\hat{C}_J := A_{j_0} Q_{j_0} \circ \mathbf{F}^{-1} + \sum_{j=j_0}^J 2^{-2rj} Q_j \circ \mathbf{F}^{-1} \quad \text{is optimal}$$

developed by [Bramble, Pasciak, Xu '90], optimality proved by [Dahmen, Kunoth '92], [Oswald '92]
 Hierarchical basis preconditioner by [Yserentant '89] **not optimal**

Optimal BPX-type Preconditioners

$$\hat{C}_J := A_{j_0} Q_{j_0} \circ \mathbf{F}^{-1} + \sum_{j=j_0}^J 2^{-2rj} Q_j \circ \mathbf{F}^{-1} \text{ using } Q_j = \sum_{i \in I_j} (\cdot, B_{i,j} \circ \mathbf{F}^{-1})_{L_2(\Omega)} B_{i,j} \circ \mathbf{F}^{-1}$$

$$\begin{aligned} \rightsquigarrow G_J &= A_{j_0}^{-1} \sum_{i \in I_0} (\cdot, B_{i,j_0} \circ \mathbf{F}^{-1})_{L_2(\Omega)} B_{i,j_0} \circ \mathbf{F}^{-1} \\ &\quad + \sum_{j=j_0+1}^J 2^{-2jr} \sum_{i \in I_j} (\cdot, B_{i,j} \circ \mathbf{F}^{-1})_{L_2(\Omega)} B_{i,j} \circ \mathbf{F}^{-1} \end{aligned}$$

Implementation: refinement relation for B-splines

\rightsquigarrow prolongation $\mathbf{I}_j^{j+1} : V_j \rightarrow V_{j+1}$

restriction $\mathbf{I}_{j+1}^j = (\mathbf{I}_j^{j+1})^T$

For $p = 2$, restriction is $\mathbf{I}_{j+1}^j = 2^{-1/2}$

$$\left[\begin{array}{cccccccc} \frac{1}{2} & & & & & & & \\ & \frac{9}{8} & \frac{3}{8} & & & & & \\ & \frac{1}{4} & \frac{3}{4} & & & & & \\ & & \frac{3}{4} & \frac{1}{4} & & & & \\ & & \frac{1}{4} & \frac{3}{4} & \frac{3}{4} & \frac{1}{4} & & \\ & & & & \ddots & \ddots & & \\ & & & & & \frac{1}{4} & \frac{3}{4} & \\ & & & & & & \frac{3}{8} & \frac{1}{4} \\ & & & & & & \frac{3}{8} & \frac{9}{8} \\ & & & & & & & \frac{1}{2} \end{array} \right] \in \mathbb{R}^{2^j \times 2^{j+1}}$$

$$\rightsquigarrow \mathbf{I}_j^J := \mathbf{I}_{J-1}^J \mathbf{I}_{J-2}^{J-1} \dots \mathbf{I}_j^{j+1} \quad \text{and} \quad \mathbf{I}_j^J := \mathbf{I}_{j+1}^j \mathbf{I}_{j+2}^{j+1} \dots \mathbf{I}_J^{J-1}$$

$$\rightsquigarrow \tilde{\mathbf{G}}_J = \sum_{j=j_0}^J \mathbf{I}_j^J (\text{diag } \mathbf{A}_j)^{-1} \mathbf{I}_j^J \circ \mathbf{F}^{-1}$$

First Numerical Results

Condition numbers $\text{cond}_2(\tilde{\mathbf{G}}_J \mathbf{A}_J)$ for Laplacian on $\Omega = (0, 1)^d$ for $d = 1, 2, 3$
with above BPX preconditioning

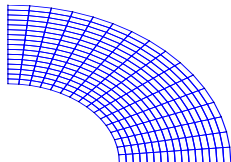
level j	interval				square				cube			
	$p = 1$	$p = 2$	$p = 3$	$p = 4$	$p = 1$	$p = 2$	$p = 3$	$p = 4$	$p = 1$	$p = 2$	$p = 3$	$p = 4$
3	7.43	3.81	7.03	5.93	5.93	7.31	22.8	133	3.49	39.5	356	5957
4	8.87	4.40	9.47	7.81	5.00	9.03	40.2	225	4.85	50.8	624	9478
5	10.2	4.67	11.0	9.36	5.70	9.72	51.8	293	5.75	56.6	795	11887
6	11.3	4.87	12.1	10.7	6.27	10.1	58.7	340	6.40	59.7	895	13185
7	12.2	5.00	12.7	11.5	6.74	10.4	63.1	371	6.91	61.3	961	13211
8	13.0	5.10	13.0	11.9	7.14	10.5	66.0	391	7.34	62.2	990	13234
9	13.7	5.17	13.2	12.1	7.48	10.6	68.0	403	7.70	62.6	1016	13255
10	14.2	5.22	13.4	12.2	7.77	10.6	69.3	411	7.99	62.9	1040	

$d = 1$: no dependence on p

Numerical Results: Dependence on Parametric Mapping F

Condition numbers of the BPX-preconditioned Laplacian on an analytic arc

level	$p = 1$	$p = 2$	$p = 3$	$p = 4$
3	5.04 (21.8)	12.4 (8.64)	31.8 (31.8)	184 (184)
4	11.1 (90.2)	16.3 (34.3)	54.7 (32.9)	291 (173)
5	25.3 (368)	19.0 (139)	70.1 (98.9)	376 (171)
6	31.9 (1492)	21.4 (560)	79.2 (401)	436 (322)
7	37.4 (6015)	23.1 (2255)	84.4 (1620)	471 (1297)
8	42.1 (241721)	24.3 (9062)	87.3 (6506)	490 (5217)
9	45.7 (969301)	25.2 (36353)	89.0 (26121)	500 (20945)
10	48.8 (388690)	25.9 (145774)	90.1 (104745)	505 (83975)

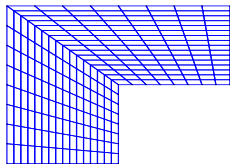


(numbers in brackets: no preconditioning)

Numerical Results: Dependence on Parametric Mapping F

... relative to a C^0 -parametrization of an L-shaped domain

level	$p = 1$	$p = 2$	$p = 3$	$p = 4$
3	14.0 (25.8)	13.4 (10.2)	33.5 (33.5)	194 (194)
4	25.2 (108)	20.6 (41.1)	56.7 (34.7)	301 (182)
5	36.9 (452)	26.8 (168)	72.1 (123)	383 (180)
6	47.9 (1845)	31.8 (689)	80.5 (500)	442 (400)
7	57.4 (7465)	35.4 (2790)	85.5 (2025)	477 (1620)
8	65.3 (30047)	38.0 (11244)	88.3 (8157)	496 (6533)
9	71.8 (120603)	40.0 (45172)	90.0 (32773)	505 (26264)
10	77.0 (483618)	41.2 (181140)	91.0 (131418)	511 (105381)

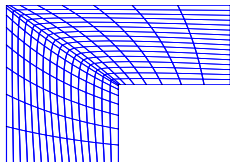


(numbers in brackets: no preconditioning)

Numerical Results: Dependence on Parametric Mapping F

Condition numbers of the BPX-preconditioned Laplacian relative to a **singular** C^1 -parametrization of an L-shaped domain

level	$p = 1$	$p = 2$	$p = 3$	$p = 4$
3	15.0 (28.8)	14.7 (13.2)	32.8 (32.8)	185 (185)
4	44.1 (133)	36.2 (53.7)	56.7 (38.5)	303 (189)
5	91.1 (568)	70.9 (225)	95.7 (158)	388 (196)
6	167 (2341)	147 (931)	155 (639)	463 (557)
7	443 (9502)	385 (3804)	385 (2587)	887 (2350)
8	1136 (38544)	960 (15353)	1021 (10491)	2417 (9604)
9	2797 (155276)	2301 (61619)	2588 (42355)	6251 (38695)
10	6664 (622565)	5362 (247091)	6318 (169844)	15505 (155143)



Further Improvement: BPX with SSOR

Split $\mathbf{A}_j = \mathbf{L}_j + \mathbf{D}_j + \mathbf{L}_j^T$ and replace $(\text{diag } \mathbf{A}_j)^{-1}$ by SSOR preconditioner

$$\tilde{\mathbf{G}}_J = \sum_{j=j_0}^J \mathbf{I}_j^J (\mathbf{D}_j + \mathbf{L}_j)^{-T} \mathbf{D}_j (\mathbf{D}_j + \mathbf{L}_j)^{-1} \mathbf{I}_j^j \circ \mathbf{F}^{-1} \quad (\text{application in optimal } \mathcal{O}(N) \text{ complexity})$$

Spectral condition numbers of BPX-preconditioned Laplacian for cubic B-splines ($p = 3$) on different geometries and using a SSOR preconditioning on each level

level	square, simple BPX	square	analytic arc	\mathcal{C}^0 -map of L-shaped domain	singular \mathcal{C}^1 -map of L-shaped domain
3	22.8	3.61	3.65	3.67	3.80
4	40.2	6.58	6.97	7.01	7.05
5	51.8	8.47	10.2	10.2	14.8
6	58.7	9.73	13.1	13.2	32.2
7	63.1	10.5	14.9	15.2	77.7
8	66.0	11.0	15.9	16.3	180
9	68.0	11.2	16.5	17.0	411
10	69.3	11.4	16.9	17.7	933

Ingredients for Efficient Numerical Solution: Nested Iteration

Recall **goal**: realize discretization error accuracy $\varepsilon_J \sim h^2 \sim 2^{-2J}$ for grid with spacing $h \sim 2^{-J}$ with minimal amount of work $\mathcal{O}(N)$ $N \sim 2^{Jd}$ unknowns

Naive strategy:

- ▶ Iterate only on highest level J and iterate until discretization error accuracy needs $\mathcal{O}(J) = \mathcal{O}(-\log \varepsilon_J)$ iterations to achieve prescribed discretization error accuracy $\varepsilon_J \sim 2^{-2J}$
- ▶ Each application of optimally conditioned \mathbf{A}_J requires $\mathcal{O}(N_J)$ arithmetic operations
 \leadsto a total of $\mathcal{O}(J N_J)$ arithmetic operations iterating on finest level only

Theorem:

Starting with coarsest level j_0 , solve $\mathbf{A}_j \mathbf{y}_j = \mathbf{f}_j$ on each level j up to discretization error accuracy ε_j and prolongate result from level j to next level $j+1$ as initial guess

\leadsto Optimal preconditioner + nested iteration yields method of optimal complexity $\mathcal{O}(N_J)$

to reach discretization error accuracy on finest level J

Proof: For multiplicative Schwarz schemes: known as full multigrid

For additive preconditioners: optimal condition of $\mathbf{A}_j \leadsto$ fixed amount of iterations on each level to reach discretization error accuracy on that level;

spaces nested and $N_j \sim 2^{dj}$ and geometric series argument

[Dahmen, Kunoth, Schneider '99]

Summary

- ▶ Proof for **optimal multilevel preconditioner** such that $\text{cond}_2(\tilde{\mathbf{G}}_J \mathbf{A}_J) = \mathcal{O}(1)$ based on generating system from hierarchical B-splines
(\neq (suboptimal) hierarchical basis preconditioner)
- ▶ **Optimal complexity** of the solver: optimal preconditioner + nested iteration
- ▶ **Use of SSOR within BPX** strongly reduces effect of isomorphism constants, parametric mappings...

Remarks and Outlook

- ▶ Multigrid preconditioners for high order FE based on low order functions [Sundar, Stadler, Biros '14]
- ▶ Extensions for nurbs, generalized splines
- ▶ Dependence of condition numbers on p ? [Manni et al '14], [Gahalaut, Tomar, Douglas '14], [Zulehner et al '14]
- ▶ A-posteriori error estimates \leadsto local refinement
 \leadsto BPX preconditioner possible Gianelli, Jüttler, Simeon, Vazquez ...

(Further) Literature

BPX for C^0 , C^1 elements

BPX for C^0 , C^1 elements on sphere

(Monotone) multigrid for higher order B-splines (for pricing American options)

[Dahmen, Kunoth '92] [Kunoth '94] [Oswald '92, '94]

[Maes, Kunoth, Bultheel '07]

[Holtz, Kunoth '07]

Iteration history for basis functions of degree $p = 2$ and $p = 3$

