

Efficient p -multigrid solvers for Isogeometric Analysis

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Isogeometric Analysis (IgA) is an extension to the finite element method (FEM).

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Many common solvers for linear systems do not perform well for IgA discretizations.

Isogeometric Analysis

Multigrid

Smoothers

Numerical results

Where next?

IgA: Variational form

Start with a PDE

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$$\begin{cases} -\Delta u & = f \text{ in } \Omega, \\ u & = 0 \text{ on } \partial\Omega. \end{cases}$$

Multiply with test function $v \in V$, integrate over Ω , integration by parts, Gauss divergence theorem

$$-\int_{\Omega} \Delta u v \, d\Omega = \int_{\Omega} f v \, d\Omega,$$

$$-\int_{\Omega} \operatorname{div}(\nabla u v) \, d\Omega + \int_{\Omega} \nabla u \nabla v \, d\Omega = \int_{\Omega} f v \, d\Omega,$$

$$-\int_{\partial\Omega} \nabla u v \cdot n \, d\Gamma + \int_{\Omega} \nabla u \nabla v \, d\Omega = \int_{\Omega} f v \, d\Omega,$$

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$$a(u, v) = \langle f, v \rangle.$$

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Inserting this into the variational form we see

$$a(u_h, \phi_j) = \sum_{i=1}^n u_i a(\phi_i, \phi_j) = \langle f, \phi_j \rangle \quad \text{for } j = 1, \dots, n.$$

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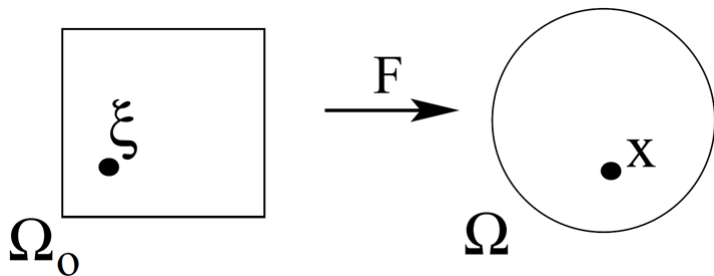
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Leading to the matrix equation

$A\mathbf{u} = \mathbf{f}$, where

$$A_{i,j} = a(\phi_i, \phi_j), \quad f_i = \langle f, \phi_i \rangle \quad i, j = 1, \dots, n.$$



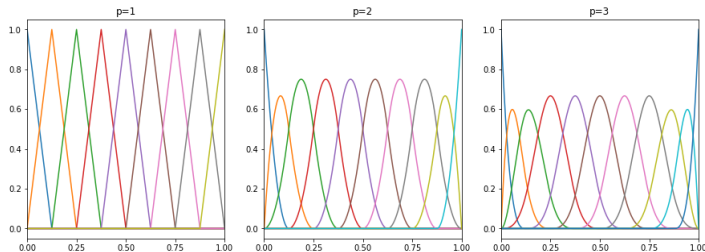


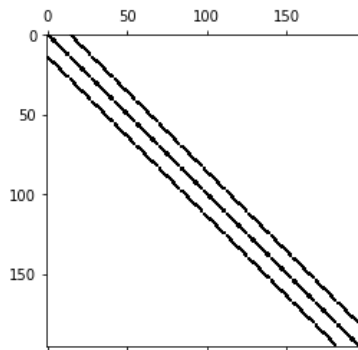
IgA: B-Spline basis functions

Cox-de Boor formula

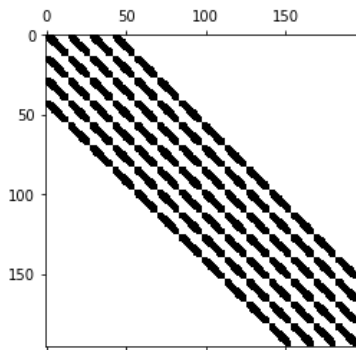
$$\phi_{i,0}(\xi) = \begin{cases} 1 & \text{if } \xi_i \leq \xi < \xi_{i+1}, \\ 0 & \text{else,} \end{cases}$$

$$\phi_{i,p}(\xi) = \begin{cases} \frac{\xi - \xi_i}{\xi_{i+p} - \xi_i} \phi_{i,p-1}(\xi) + \frac{\xi_{i+p+1} - \xi}{\xi_{i+p+1} - \xi_{i+1}} \phi_{i+1,p-1}(\xi) & \text{if well defined,} \\ 0 & \text{else.} \end{cases}$$





(a) $p = 1, h = 1/16$



(b) $p = 3, h = 1/16$

Sparisty pattern of various system matrices.

Vectors \mathbf{u} and \mathbf{f} consist of coefficients to basis functions.

Non-zero structure of the system matrix A .

Goal: determine \mathbf{u} in $A\mathbf{u} = \mathbf{f}$.

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Two-grid cycle:

1. Relax ν_1 times on $A\mathbf{u} = \mathbf{f}$ with initial guess \mathbf{v} .
2. Compute the residual $\mathbf{r} = \mathbf{f} - A\mathbf{v}$.
3. Restrict the residual $\tilde{\mathbf{r}} = \mathcal{I}_R\mathbf{r}$.
4. Solve $\tilde{A}\tilde{\mathbf{e}} = \tilde{\mathbf{r}}$.
5. Prolongate the error $\mathbf{e} = \mathcal{I}_P\tilde{\mathbf{e}}$.
6. Update the guess $\mathbf{v} \leftarrow \mathbf{v} + \mathbf{e}$.
7. Relax ν_2 times on $A\mathbf{u} = \mathbf{f}$ with initial guess \mathbf{v} .

Motivation: We know most solving methods perform badly for higher values of p .

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What does this mean for \tilde{A} , $\tilde{\mathbf{v}}$?

Intergrid operators are based on an L_2 -projection.

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The prolongation- and restriction operator are given by

$$\mathcal{I}_1^p = (M_p)^{-1} P_1^p,$$

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Here the mass matrices M and transfer matrices P are defined as

$$(M_p)_{i,j} = \int_{\Omega} \phi_{i,p} \phi_{j,p} \, d\Omega, \quad (P_1^p)_{i,j} = \int_{\Omega} \phi_{i,p} \phi_{j,1} \, d\Omega,$$

$$(M_1)_{i,j} = \int_{\Omega} \phi_{i,1} \phi_{j,1} \, d\Omega, \quad (P_p^1)_{i,j} = \int_{\Omega} \phi_{i,1} \phi_{j,p} \, d\Omega.$$

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The method chosen is an ILUT-smoother, based on an incomplete LU factorization.

MG: ILUT smoother

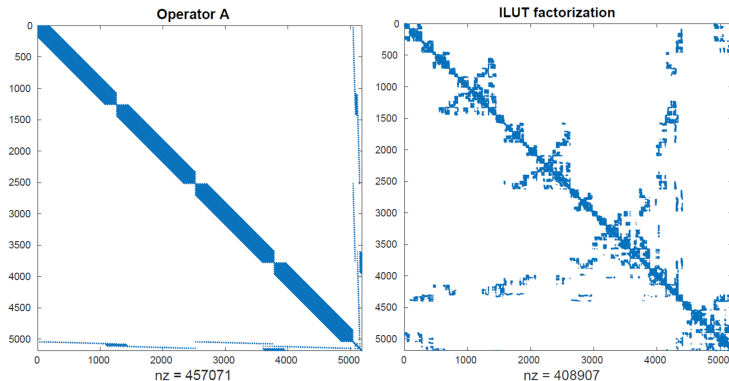


Figure: Sparsity pattern of A and its ILUT factorization.
Image from "A block ILUT smoother for multipatch geometries in Isogeometric Analysis" by R. Tienen, M. Möller and K. Vuik

We want a method where convergence is independent of p .
This is achieved by a p -multigrid method, with coarsening based on a lower order discretization.
As a smoother on the fine level we can use an ILUT smoother.

NR: Problem description

2-dimensional homogeneous Poisson equation.

$$\begin{cases} -\Delta u = f, & \text{on } [0, 1]^2 \\ u = 0, & \text{on the boundary.} \end{cases}$$

With right hand side $f(x) = 2\pi^2 \sin(\pi x) \sin(\pi y)$ and initial guess the zero vector $u^0 \equiv 0$.

The primary thing we are interested in for our multigrid program is to see how many steps it takes to converge. For this we use the stopping criterium

$$\frac{\|\mathbf{r}^k\|}{\|\mathbf{r}^0\|} < 10^{-8}.$$

where \mathbf{r}^k and \mathbf{r}^0 are the residuals after k and 0 steps respectively.

NR: h -multigrid

	$p=1$	$p=2$	$p=3$	$p=4$	$p=5$	$p=6$	
Ndof=5	4	8	33	XX	XX	XX	2-dim Poisson problem, IgA discretization
Ndof=9	6	8	28	116	730	XX	h -multigrid, smoother = GS
Ndof=17	8	9	25	74	311	840	steps till convergence
Ndof=33	8	9	24	72	228	720	
Ndof=65							

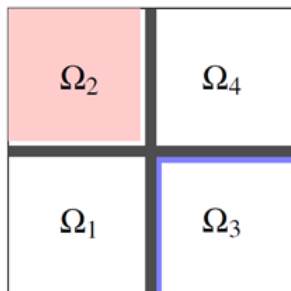
NR: p -multigrid + Gauss Seidel

	$p=1$	$p=2$	$p=3$	$p=4$	$p=5$	$p=6$	
Nel=4	-	9	30	113	388	708	2-dim Poisson problem, IgA discretization
Nel=8	-	8	22	84	347	1042	p -multigrid, smoother = GS
Nel=16	-	7	23	68	254	793	steps till convergence
Nel=32	-	5	21	63	204	774	
Nel=64	-						

NR: p -multigrid + ILUT smoother

	$p=1$	$p=2$	$p=3$	$p=4$	$p=5$	$p=6$	
Nel=4	-	2	2	2	2	3	2-dim Poisson problem, IgA discretization p -multigrid, smoother = ILUT steps till convergence
Nel=8	-	3	2	2	2	2	
Nel=16	-	3	3	3	3	3	
Nel=32	-	3	3	3	3	3	
Nel=64	-						

Next: Multipatch



$$\begin{bmatrix} A_{11} & 0 & 0 & 0 & A_{\Gamma 1} \\ 0 & A_{22} & 0 & 0 & A_{\Gamma 2} \\ 0 & 0 & A_{33} & 0 & A_{\Gamma 3} \\ 0 & 0 & 0 & A_{44} & A_{\Gamma 4} \\ A_{1\Gamma} & A_{2\Gamma} & A_{3\Gamma} & A_{4\Gamma} & A_{\Gamma\Gamma} \end{bmatrix}$$

Next: Block Structure

$$A = \begin{bmatrix} A_{11} & & 0 & A_{\Gamma 1} \\ & \ddots & & \vdots \\ 0 & & A_{KK} & A_{\Gamma K} \\ A_{1\Gamma} & \dots & A_{K\Gamma} & A_{\Gamma\Gamma} \end{bmatrix}.$$

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$$A = LU = \begin{bmatrix} L_1 & & & \\ & \ddots & & \\ & & L_K & \\ B_1 & \dots & B_K & I \end{bmatrix} \begin{bmatrix} U_1 & & & C_1 \\ & \ddots & & \vdots \\ & & U_K & C_k \\ & & & S \end{bmatrix}.$$

Next: Block ILUT

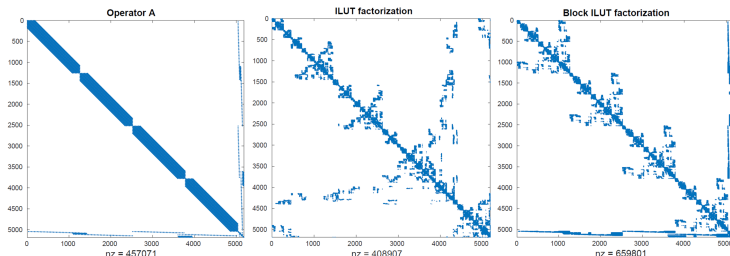


Figure: Sparsity pattern of A , its global ILUT factorization and its block ILUT factorization.

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