

ALGORITHMS FOR MASSIVELY PARALLEL GENERIC HP-ADAPTIVE FEM

June 05, 2020 | Marc Fehling | m.fehling@fz-juelich.de |

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- Fire safety science
- Computation fluid dynamics

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- Example: Laplace equation

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- Example: Laplace equation

Summary & Outlook

Fire safety science

- Civilian safety in metro stations
 - Smoke spread in case of fire
 - Egress routes for pedestrians
- Individual examinations necessary on complex geometries
- Experiments expensive → Alternative: Computer simulations



Figure: Experiments in metro

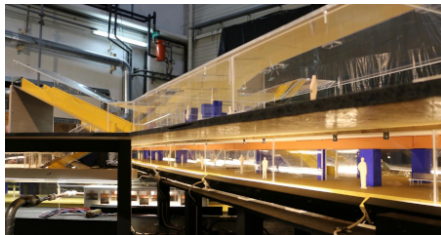


Figure: Physical model (scale 1:15)

Fire simulation

- Lots of software tools available: FDS, FireFOAM, Ansys Fluent, ...
- Flexible: Scenarios may be varied easily
- **But:** Large and complex geometries yield lots of workload
- Simulations require a lot of time

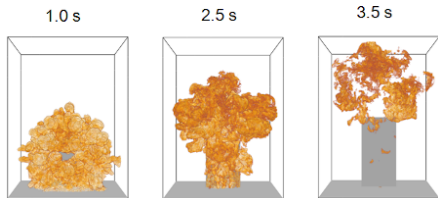


Figure: Deflagration of Heptane

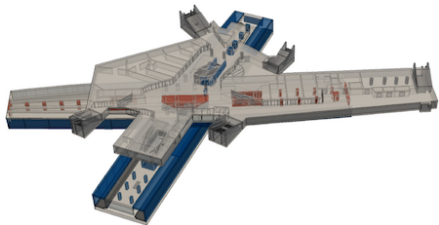


Figure: CAD model

Computational fluid dynamics

- Smoke spread modeled with incompressible Navier–Stokes equations

$$\nabla \cdot \mathbf{u} = 0$$

$$\rho_0 [\partial_t \mathbf{u} + (\mathbf{u} \cdot \nabla) \mathbf{u}] = -\nabla p + \nabla \cdot (2\mu \boldsymbol{\epsilon}) + \mathbf{f}$$

$$\rho_0 c_p [\partial_t T + (\mathbf{u} \cdot \nabla) T] = 2\mu \boldsymbol{\epsilon} : \nabla \mathbf{u} + \nabla \cdot (\kappa \nabla T) + q - \mathbf{u} \cdot \mathbf{f}$$

- Solution via numerical methods → Computational fluid dynamics
- High resolution necessary for ...
 - Large gradients in temperature and velocity
 - Turbulence
 - Flow separation

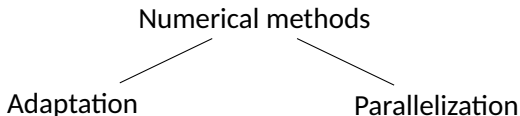


Motivation for PhD project

- Fires stay localized in general, not only during ignition phase
- Unnecessarily fine grids bind resources that could be used near the fire
- Demand for effective use of computing power

Goal

- Balance accuracy and workload by adapting resolution
- Accelerate simulations by exploiting hardware



Example: Adaptive mesh refinement

- Demonstration of adaptive mesh refinement (h -adaptive methods) via moving vortex test case as a shape-preserving potential stream.

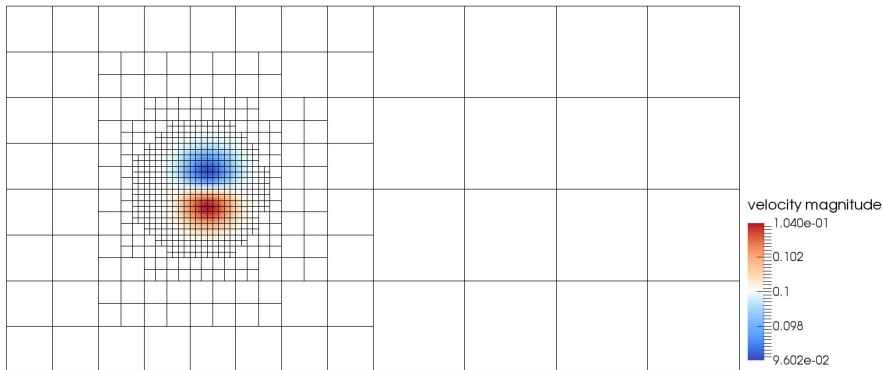
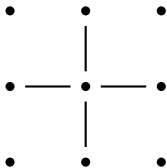


Figure: Video of velocity magnitude of moving vortex, overlaid with current mesh

Numerical methods

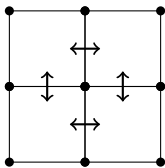
Finite differences



Difference quotients as differential operators

h-adaptive methods

Finite volumes

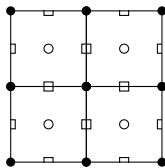


Balance fluxes on faces between volumes

Conservation laws

h-adaptive methods

Finite elements



Limit function space to piecewise polynomials

hp-adaptive methods

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Example: Laplace equation

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Finite element method

- Shape functions form nodal basis

$$\varphi_i(x_j) = \delta_{ij} = \begin{cases} 1 & \text{for } i = j \\ 0 & \text{for } i \neq j \end{cases}$$

- Q_p elements from Lagrange interpolation with degree p
- Finite element approximation is linear combination of shape functions

$$u_{hp}(x) = \sum_i u_i \varphi_i(x)$$

- Coefficients u_i are degrees of freedom

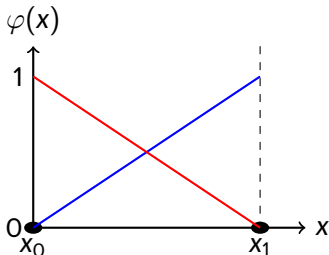


Figure: Q_1 element

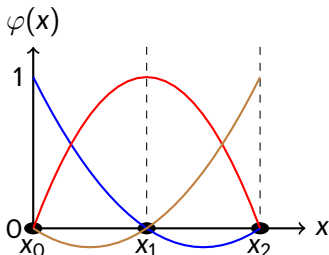


Figure: Q_2 element

Adaptive methods

- Focus computational resources on areas of interest
- Align simulation resolution with complexity of current solution
- Finite Element Method (FEM) provides two different possibilities:
 - h -adaptation: dynamic cell sizes good for irregular solutions
 - p -adaptation: dynamic function spaces good for smooth solutions
- Combination of both possible

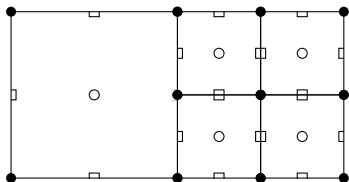


Figure: h -adaptivity

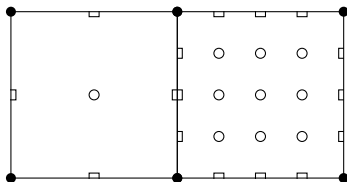


Figure: p -adaptivity

Adaptation criteria

- Which cells to adapt?
- How to adapt? h/p ?

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- Manual adaptation
- Automatic adaptation
 - Criterion to indicate adaptation
 - General approach -OR- tied to the problem

Adaptation criteria

- Which cells to adapt?
- How to adapt? h/p ?

- Manual adaptation
- Automatic adaptation
 - Criterion to indicate adaptation
 - General approach -OR- tied to the problem

- Automatic hp -decision strategies discussed in the dissertation
 - 1 Error prediction based on refinement history
[Melenk and Wohlmuth, 2001]
 - 2 Smoothness estimation by decay of Fourier coefficients
[Bangerth and Kayser-Herold, 2009]
 - 3 Smoothness estimation by decay of Legendre coefficients
[Mavriplis, 1994]

Example: Reentrant corner

- Singularity at reentrant corners for elliptic problems
- L-shaped domain:

$$\Omega = [-1, 1]^2 \setminus ([0, 1] \times [-1, 0])$$

- Manufactured Laplace problem

$$-\nabla^2 u = 0 \quad \text{on } \Omega$$

$$u = \bar{u} \quad \text{on } \partial\Omega$$

$$\bar{u} = r^{2/3} \sin(2/3 \varphi)$$

$$\|\nabla \bar{u}\| = r^{-1/3}$$

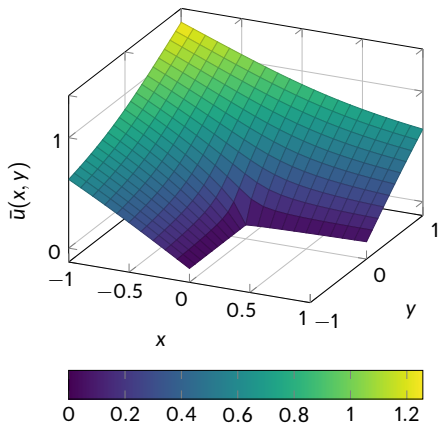


Figure: L-shaped domain

Example: Successive refinement

- Initialize coarse mesh
- Solve and refine in multiple cycles for tailored discretization

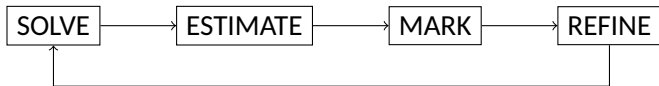


Figure: Successive refinement

- 1 Calculate refinement criteria (here: error estimates)
- 2 Flag 30%/3% of cells with highest/lowest criterion for refinement/coarsening
- 3 Calculate decision criteria (here: smoothness estimates)
- 4 Flag 90%/10% for p -/ h -adaptation

Example: Successive refinement

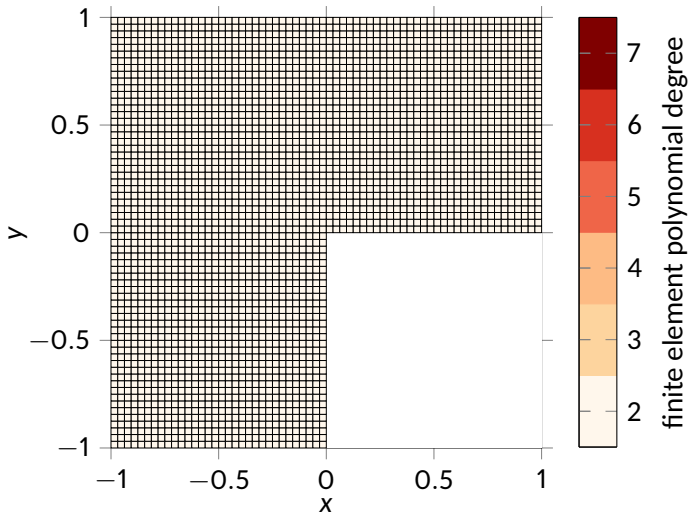


Figure: Polynomial degrees in cycle 0. Zoom 100%.

Example: Successive refinement

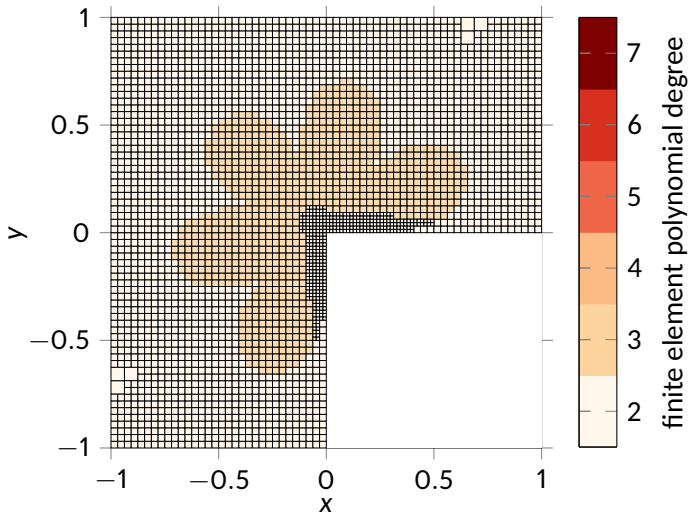


Figure: Polynomial degrees in cycle 1. Zoom 100%.

Example: Successive refinement

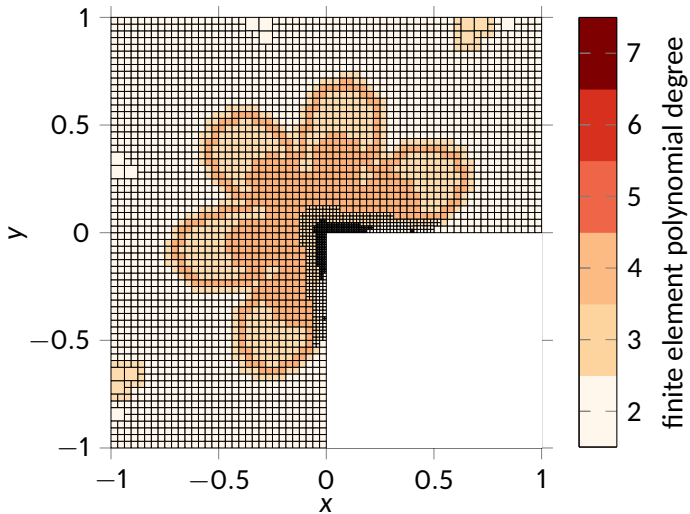


Figure: Polynomial degrees in cycle 2. Zoom 100%.

Example: Successive refinement

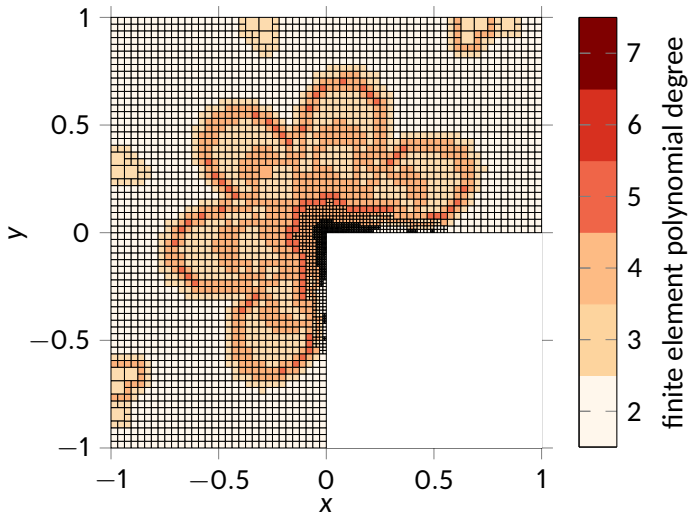


Figure: Polynomial degrees in cycle 3. Zoom 100%.

Example: Successive refinement

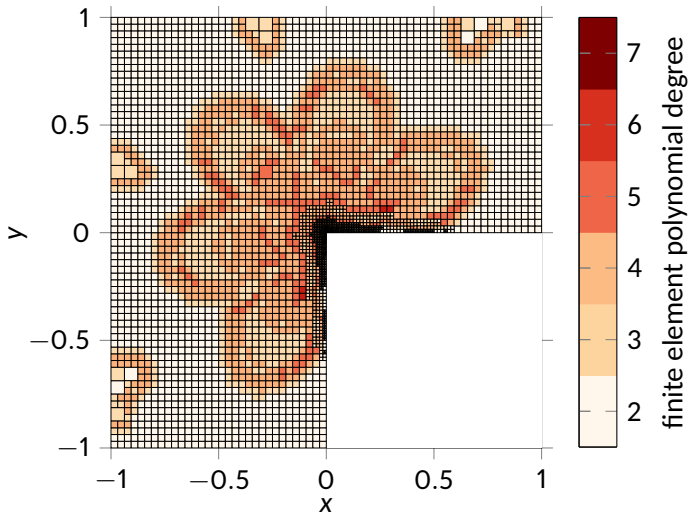


Figure: Polynomial degrees in cycle 4. Zoom 100%.

Example: Successive refinement

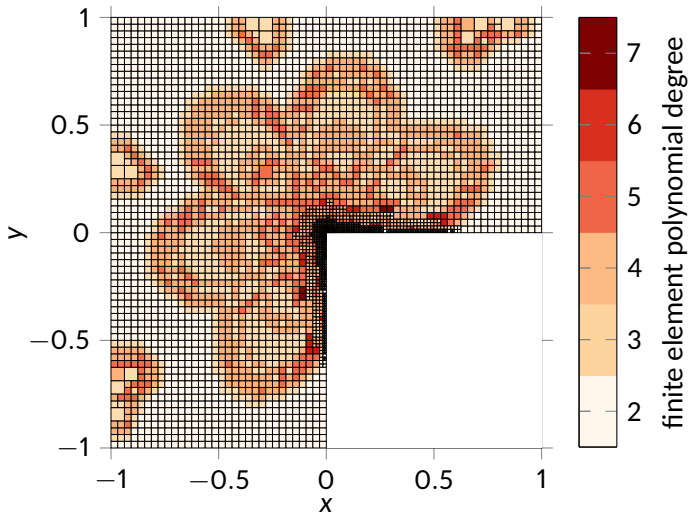


Figure: Polynomial degrees in cycle 5. Zoom 100%.

Example: Successive refinement

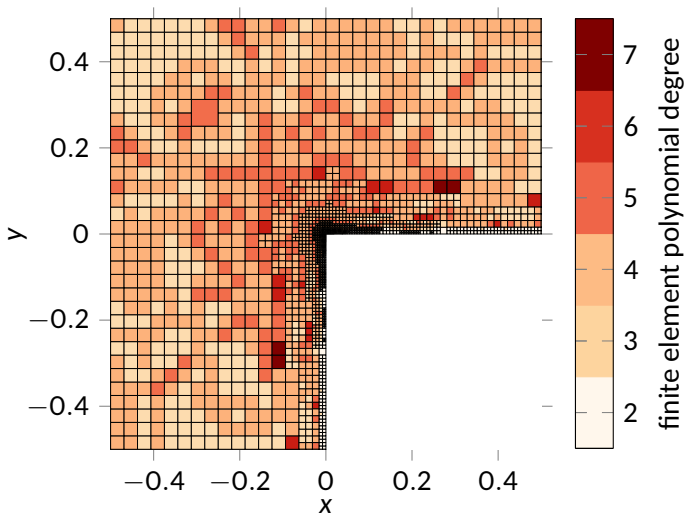


Figure: Polynomial degrees in cycle 5. Zoom 200%.

Example: Successive refinement

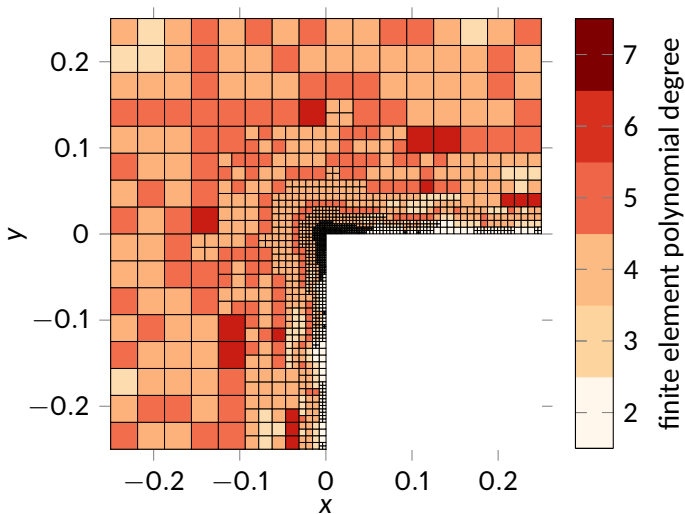


Figure: Polynomial degrees in cycle 5. Zoom 400%.

Example: Successive refinement

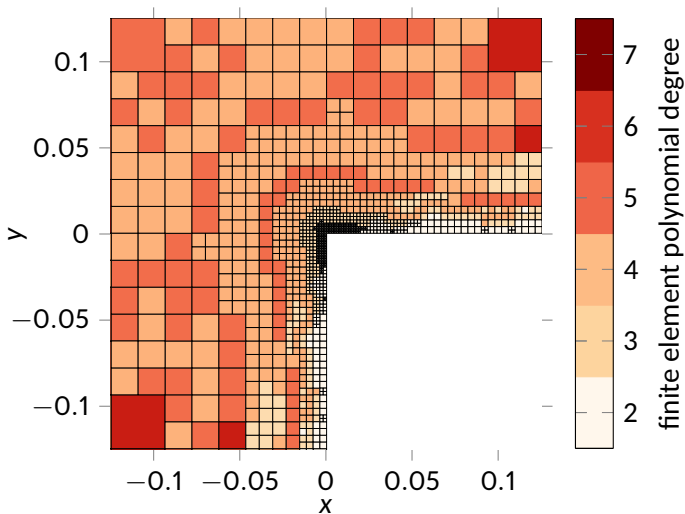
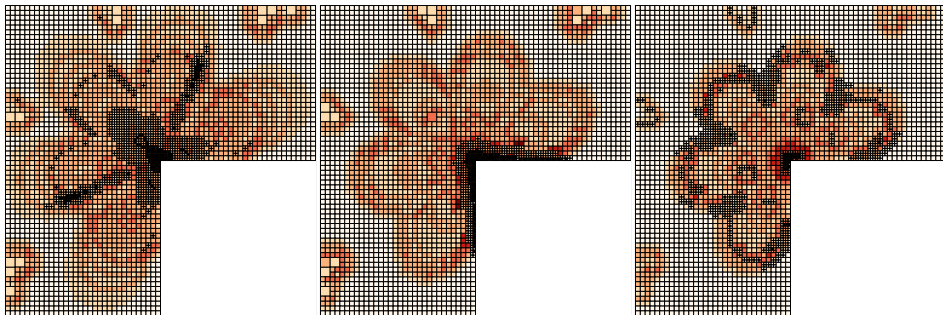


Figure: Polynomial degrees in cycle 5. Zoom 800%.

Example: Successive refinement



(a) Fourier coefficient decay

(b) Legendre coefficient decay

(c) Refinement history

Figure: Mesh and polynomial degrees of finite elements after 5 consecutive *hp*-adaptations.

Example: Successive refinement

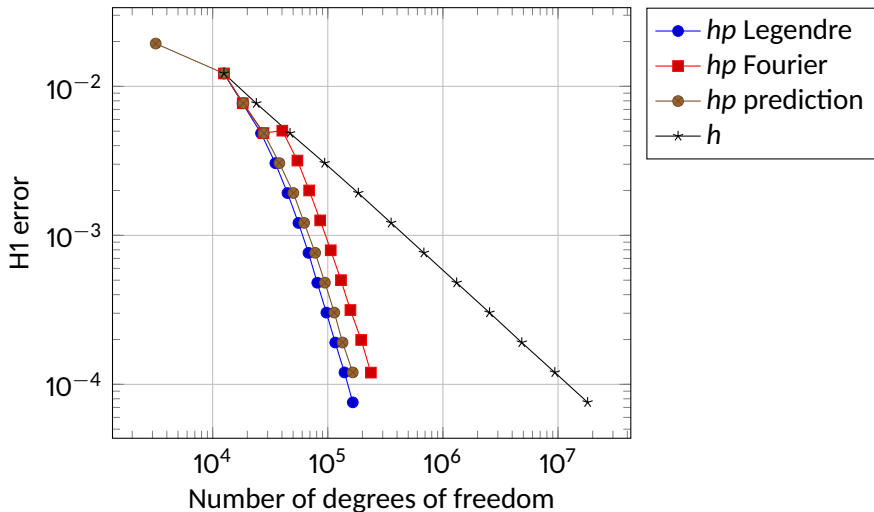


Figure: Error convergence for different strategies

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Summary & Outlook

Parallelization

- Current computer architectures provide multi-core processors
 - Supercomputers arrange those on distributed nodes
- Using all resources efficiently requires parallelization
 - Distribution of workload and memory demand
- Our approach: Distribution of domain on several processes
 - Each subdomain needs relevant part of the global solution
 - Requires a layer of so called ghost cells
 - Involves communication between processors

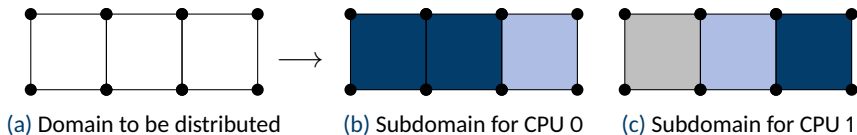


Figure: Illustration of **locally owned**, **ghost**, and **artificial** cells

Parallel hp-adaptive FEM

- Combination of hp-adaptive methods with parallelisation
- The non-trivial parts are:
 - 1 Enumeration of degrees of freedom, independent of number of subdomains
 - 2 Consignment of contiguous memory chunks for data transfer
 - 3 Weighted repartitioning for load balancing

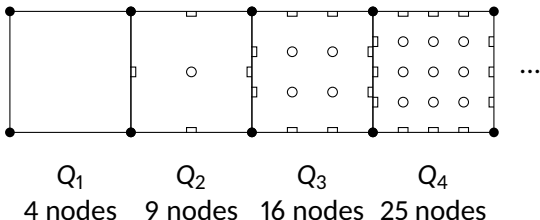


Figure: Different finite elements and their number of nodes in 2D

Example: Load balancing

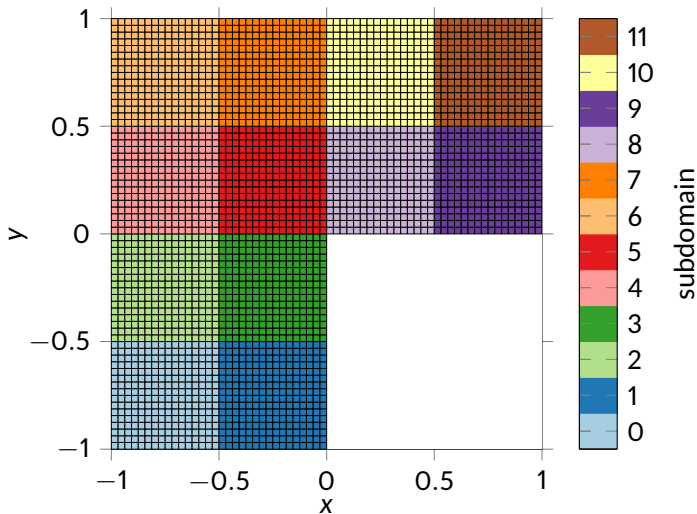


Figure: Mesh decomposition in cycle 0. Weights assigned to cells are $\propto n_{\text{dofs}}^{1.9}$.

Example: Load balancing

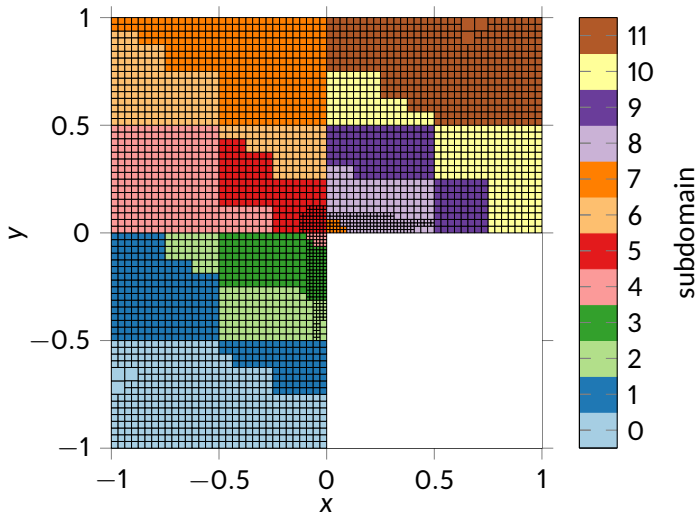


Figure: Mesh decomposition in cycle 1. Weights assigned to cells are $\propto n_{\text{dofs}}^{1.9}$.

Example: Load balancing

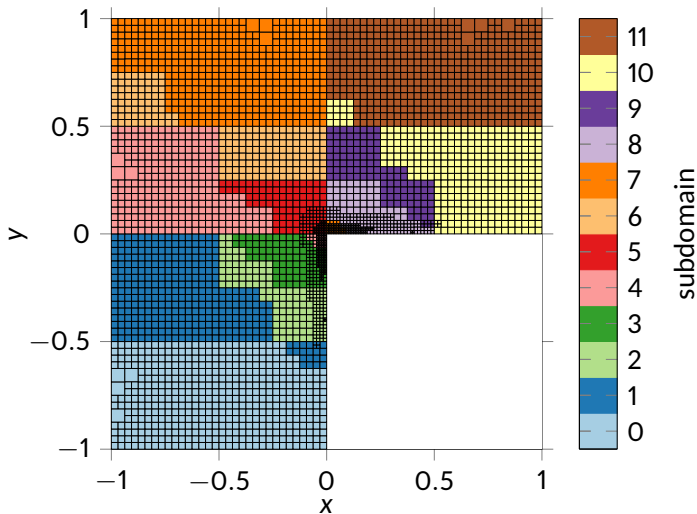


Figure: Mesh decomposition in cycle 2. Weights assigned to cells are $\propto n_{\text{dofs}}^{1.9}$.

Example: Load balancing

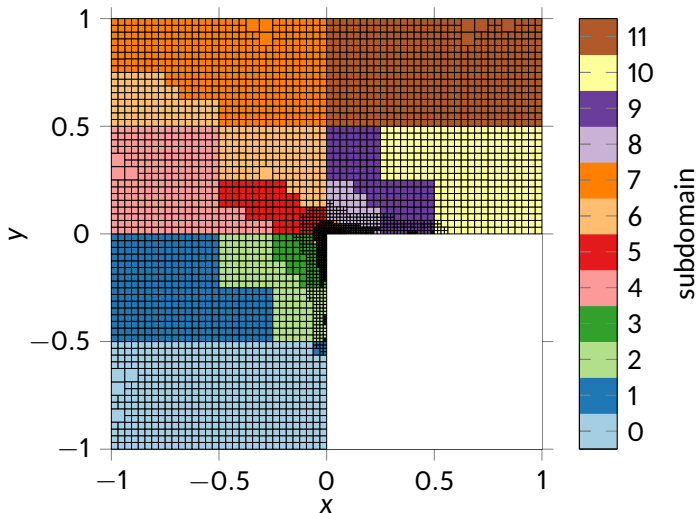


Figure: Mesh decomposition in cycle 3. Weights assigned to cells are $\propto n_{\text{dofs}}^{1.9}$.

Example: Load balancing

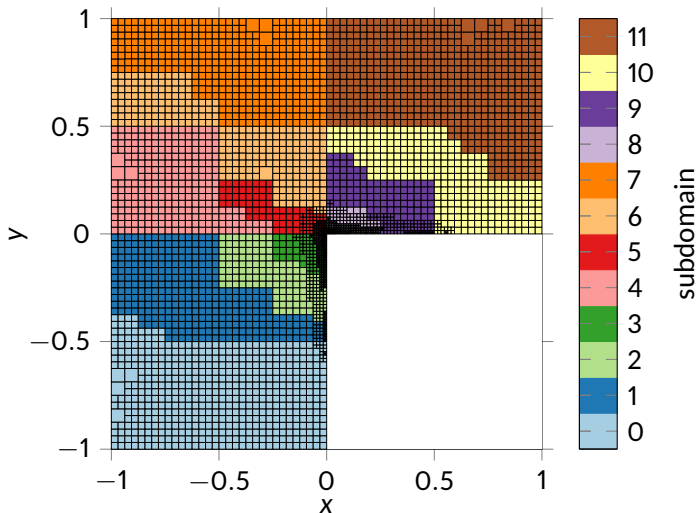


Figure: Mesh decomposition in cycle 4. Weights assigned to cells are $\propto n_{\text{dofs}}^{1.9}$.

Example: Load balancing

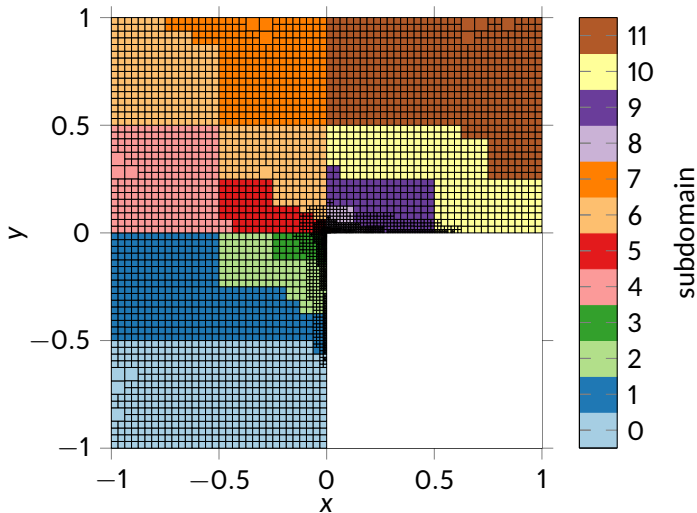


Figure: Mesh decomposition in cycle 5. Weights assigned to cells are $\propto n_{\text{dofs}}^{1.9}$.

Example: Strong scaling

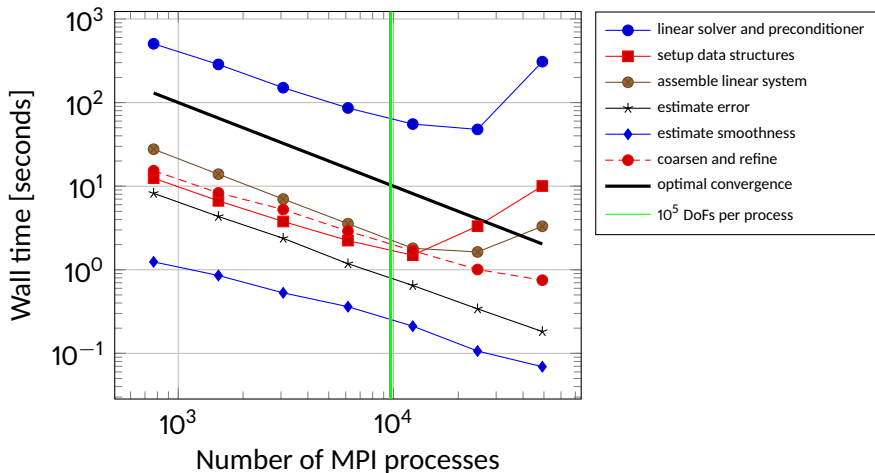


Figure: Strong scaling for fixed problem size of ~ 970 million degrees of freedom

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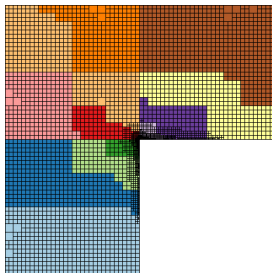
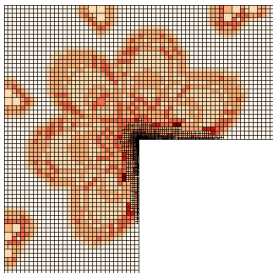
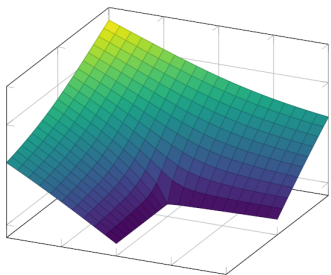
Example: Laplace equation

Summary & Outlook

Summary & Outlook

- New algorithm for massively parallel hp-adaptive methods, generally applicable for any FEM software
- Reference implementation in `deal.II` involves:
 - Enumeration of degrees of freedom, independent of number of subdomains
 - Consignment of contiguous memory chunks for data transfer
 - Weighted repartitioning for load balancing
 - Selection of adaptation strategies for *hp*-FEM
- Future steps:
 - p-Multigrid methods
 - MatrixFree methods
 - Provide tutorial in `deal.II` as a manual for a broader audience
 - More applications





MASSIVELY PARALLEL HP-ADAPTIVE FEM

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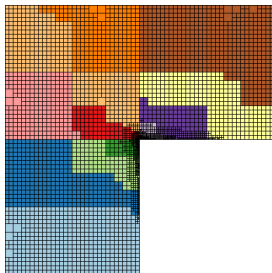
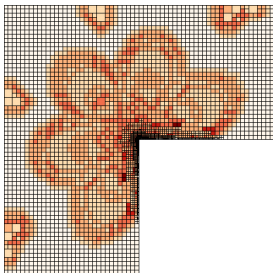
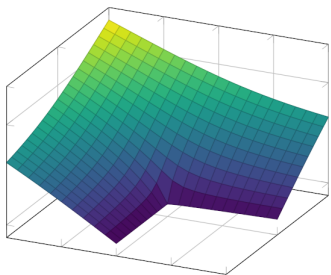
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MASSIVELY PARALLEL HP-ADAPTIVE FEM

Expected questions

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INS assumptions

- Incompressibility assumption: Mach number < 0.3
- Buoyancy force with Boussinesq approximation:
 $f \approx \rho_0 g - \rho_0 g \beta (T - T_0)$

Karniadakis scheme

- Explicit convection step

$$(u^* - u^n)/k = -\nabla \cdot (u^n u^n) + f^{n+1}$$

- Pressure Poisson equation and velocity reconstruction

$$\nabla^2 p^{n+1} = -1/k \nabla \cdot u^*$$

$$u^{**} = u^* - k \nabla p^{n+1}$$

- Implicit diffusion step

$$(u^{n+1} - u^{**})/k = \nabla \cdot (2\nu \epsilon)$$

Dimensionless quantities

- Reynolds number: inertial/viscous forces (turbulence behavior)
- Rayleigh number: time scales for thermal transport diffusion/convection
- Grashof number: buoyant/viscous forces
- Prandtl number: momentum/thermal diffusivity

Types of PDEs

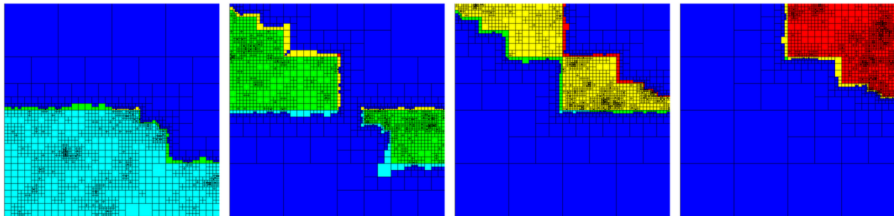
General form of second order PDE with two variables

$$Au_{xx} + 2Bu_{xy} + Cu_{yy} + Du_x + Eu_y + Fu + G = 0$$

- Elliptic equations: $B^2 - AC < 0$
Laplace equation: $-u_{xx} - u_{yy} = 0$
- Hyperbolic equations: $B^2 - AC > 0$
Wave equation: $u_{tt} - c^2u_{xx} = 0$
- Parabolic equations: $B^2 - AC = 0$
Heat equation: $u_t - \alpha u_{xx} = 0$

Artificial cells

- Every process knows about global coarse mesh
- Correct level representation only on locally owned cells
- 2:1 mesh balance
- Artificial cells as coarse as possible



Space-filling curve

- Z-order space filling curve
- Partitioning via leaves of tree structure

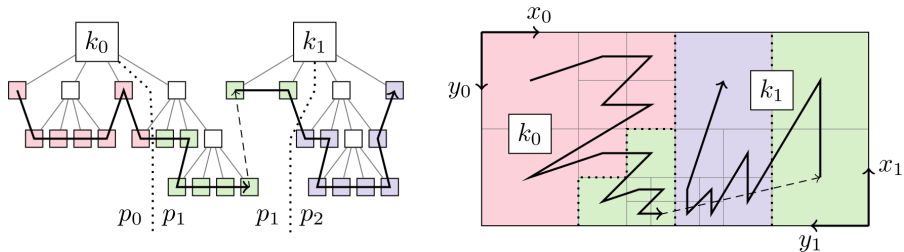
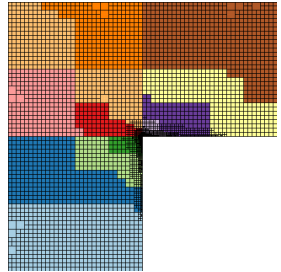
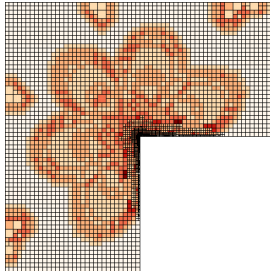
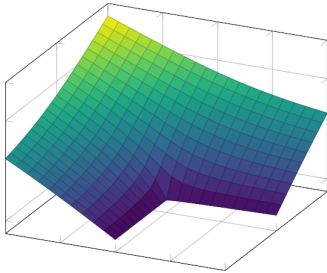


Figure: Correspondence of hierarchic tree structure and geometry, incorporating partitioning and the space-filling curve



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Details on algorithms

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Finite element method

- Solve variational equation from 'weak' formulation of the differential equation with bilinear form $a(u, v)$:

$$\exists u \in V : \forall v \in V : a(u, v) = f(v)$$

- Choose subspace V_h with basis w_i , out of which the approximate solution $u_h = \sum u_i w_i \in V_h$ will be constructed:

$$a(u_h, w_j) = \sum_i a(w_i, w_j) u_i = f(w_j) \rightarrow AU = F$$

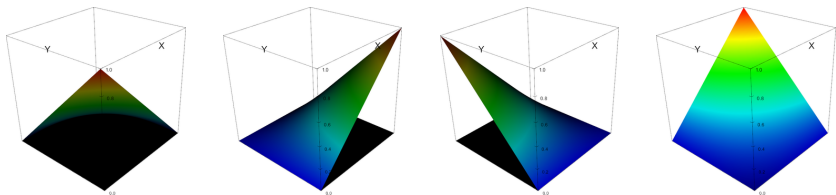


Figure: Q_1 elements in 2D

Laplace equation

Multiply with test function, integrate and apply Gauss theorem

$$\begin{aligned} -\nabla^2 u &= f \\ -\int_{\Omega} \nabla^2 u \varphi &= \int_{\Omega} \nabla u \nabla \varphi - \int_{\partial\Omega} (\mathbf{n} \cdot \nabla u) \varphi = \int_{\Omega} f \end{aligned}$$

$\varphi = 0$ on boundary

$$(\nabla u, \nabla \varphi) = (f, \varphi)$$

Discretized

$$\sum_j (\nabla \varphi_i, \nabla \varphi_j) u_j = (f, \varphi_i)$$

$$AU = F$$

Gaussian quadrature

$$A_{ij}^K = \int_K \nabla \varphi_i \cdot \nabla \varphi_j \approx \sum_q \nabla \varphi_i(\mathbf{x}_q^K) \cdot \nabla \varphi_j(\mathbf{x}_q^K) w_q^K,$$

$$F_i^K = \int_K \varphi_i f \approx \sum_q \varphi_i(\mathbf{x}_q^K) f(\mathbf{x}_q^K) w_q^K,$$

Choose finite elements upon adaptation

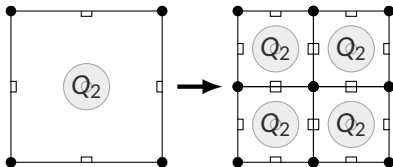


Figure: *hp* refinement

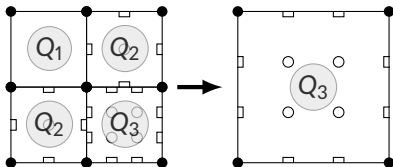


Figure: *hp* coarsening

Enumeration of degrees of freedom

- Numbering of DoFs necessary to build linear equation
- Parallelization and p-adaptive methods require different algorithms
 - See parallel [Bangerth et al., 2012] and hp [Bangerth and Kayser-Herold, 2009] papers for details

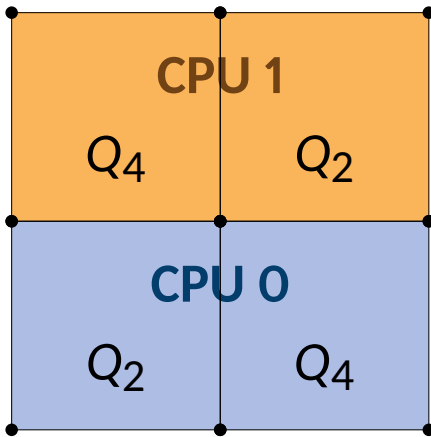
Combination of both algorithms **not trivial**

- Why not number DoFs locally and handle relations with constraints?
 - Number of DoFs would vary with number of processors
 - Larger matrices & vectors, thus higher memory demand
- Need for algorithm independent of number of processors
 - Results independent of number of processors
 - Proof that solvers work
 - Simplifies debugging

Enumeration algorithm

- Developed 6-phase algorithm
 - Requires two ghost exchanges
- 1 Local enumeration of DoFs
- 2 Invalidate DoFs on ghost interfaces to processors with lower rank
- 3 Unification of DoFs on local domain **and** ghost interfaces
 - Ownership of DoFs clarified
- 4 Global re-enumeration of DoFs
 - Local DoF indices set
- 5 Exchange of locally owned DoFs
- 6 Merge DoFs on ghost interfaces
 - Global DoF indices set

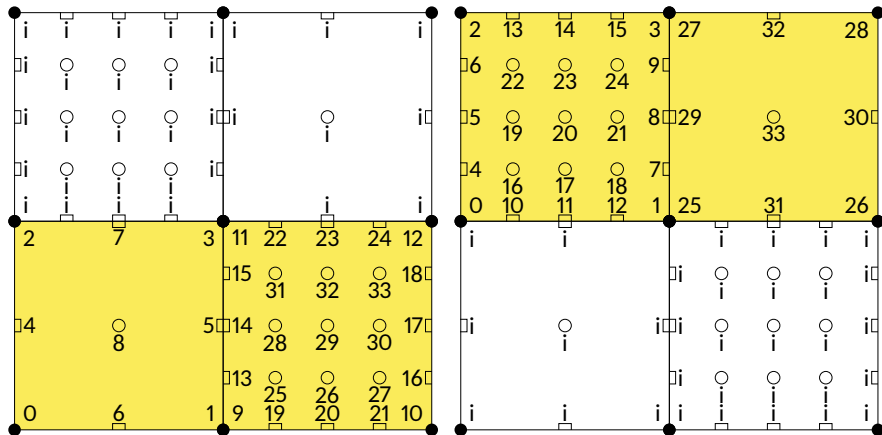
Example of application of enumeration algorithm



(Phase 1) Local enumeration

Figure: Example of application of our enumeration algorithm for degrees of freedom

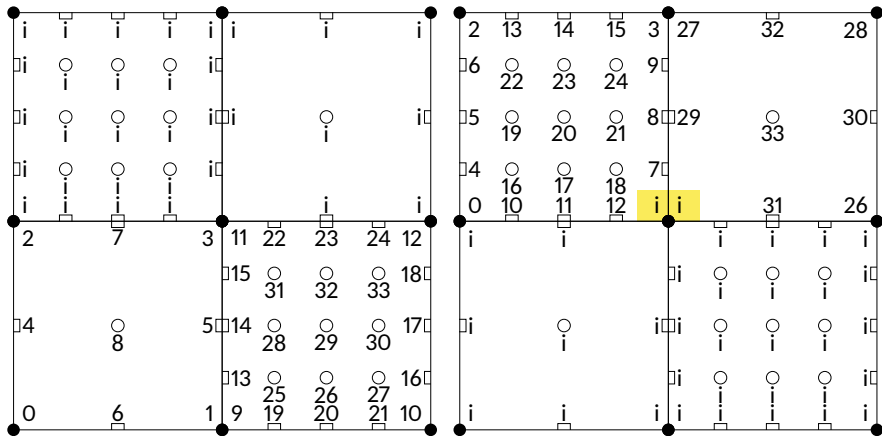
Example of application of enumeration algorithm



(Phase 1) Local enumeration

Figure: Example of application of our enumeration algorithm for degrees of freedom

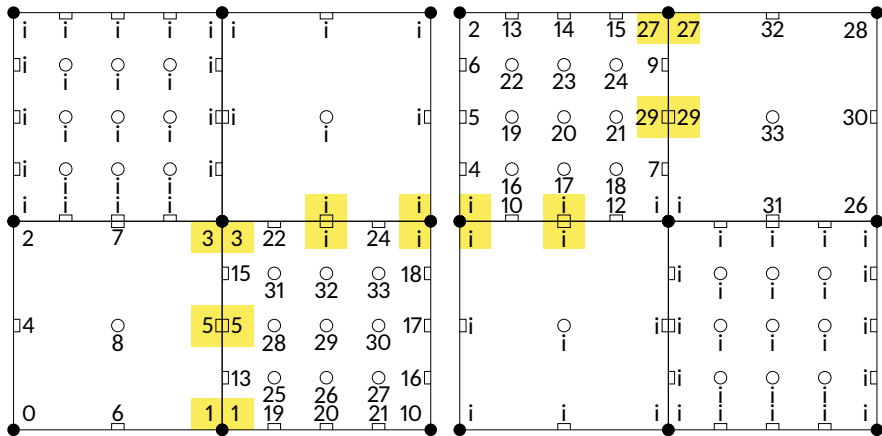
Example of application of enumeration algorithm



(Phase 2) Invalidation

Figure: Example of application of our enumeration algorithm for degrees of freedom

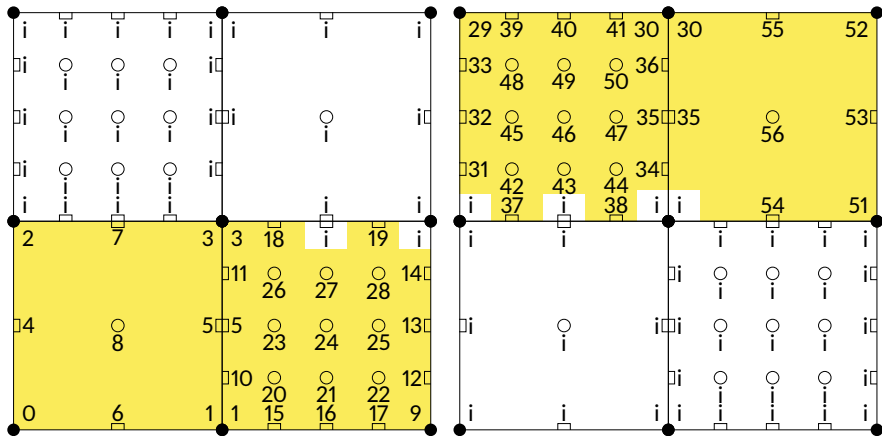
Example of application of enumeration algorithm



(Phase 3) Unification

Figure: Example of application of our enumeration algorithm for degrees of freedom

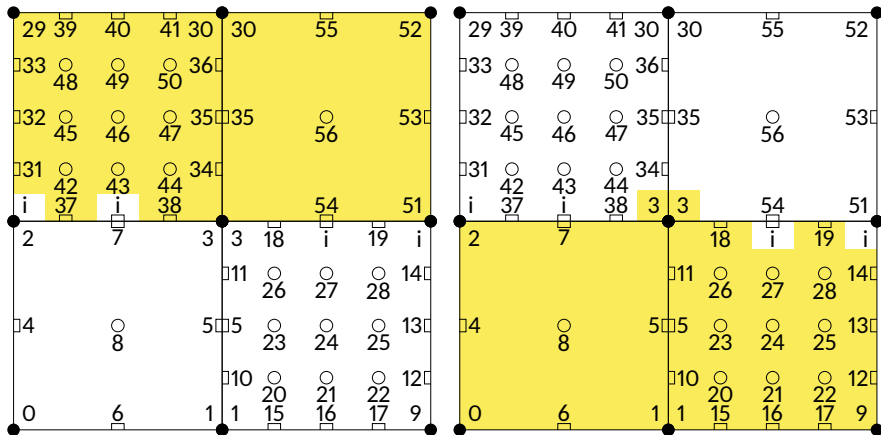
Example of application of enumeration algorithm



(Phase 4) Re-enumeration

Figure: Example of application of our enumeration algorithm for degrees of freedom

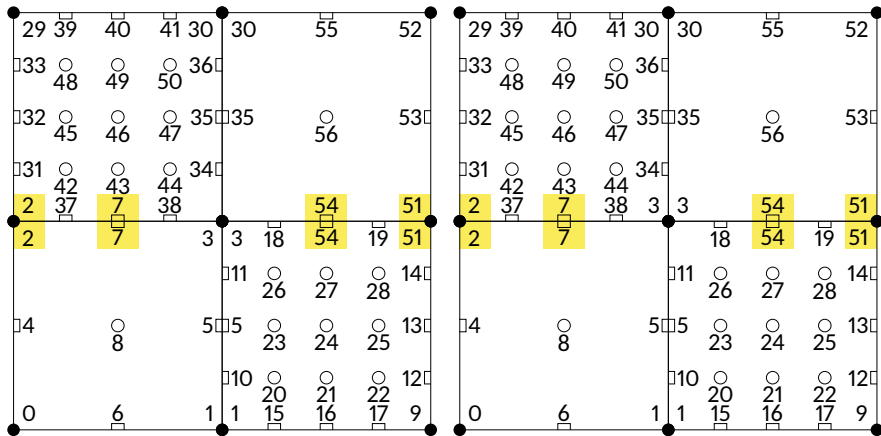
Example of application of enumeration algorithm



(Phase 5) Ghost exchange

Figure: Example of application of our enumeration algorithm for degrees of freedom

Example of application of enumeration algorithm



(Phase 6) Merge

Figure: Example of application of our enumeration algorithm for degrees of freedom

Data transfer across subdomains

- On distributed triangulations, each subdomain needs access to relevant fraction of global quantities
- Changes on cell ownership requires transfer of these quantities
 - Active finite element indices
 - Solution
 - ...
- With p-adaptive methods, per cell data sizes may differ

Communication between involved processors required

- Algorithm should be generic, i.e. independent of scenario
 - 1 Creation of **memory buffers**
 - 2 Transfer data of **fixed** and **variable** size

Structure of memory buffers

- Register functions that prepare data to be transferred via callback
- **On each locally owned cell:**
 - 1 Pack data for each registered callback individually
 - 2 Store memory sizes of each callback's data pack
 - 3 Store memory size of cell's complete data pack
- **After each cell is processed:**
 - 4 Move each cell's packed data into contiguous buffer

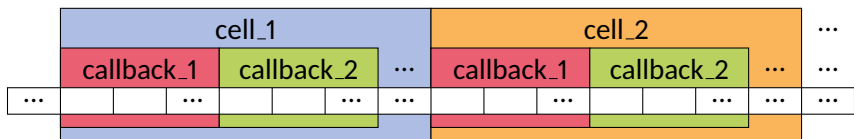


Figure: Division of contiguous memory chunk

Data consignment

- Treat fixed and variable size data separately
 - Each transfer algorithm optimized for their specific task
 - Potentially slower variable size transfer will only be used when necessary

1 Fixed size data:

- Refrain from using compression
- Additionally pack CellStatus information
- Gathering callback sizes on first packed cell will suffice
 - Communicate callback sizes to all processors

2 Variable size data:

- Compression allowed (using ZLIB)
- Size of each callback's data differs from cell to cell
 - Register additional callback for fixed size data transfer

Data transfer

- We have contiguous memory chunks for data transfer during repartitioning, refinement/coarsening, serialization
 - Program may be resumed with a different number of processors

- Data consignment **independent** of transfer algorithms used for repartitioning, refinement/coarsening, serialization
 - Use non-blocking MPI communication for all operations
 - deal.II utilizes interface to p4est [Burstedde, 2018]

FEM error behavior

- Error behavior for hp -FEM is well understood [Babuška and Suri, 1990]

$$\|\nabla (u - u_{hp})\|_{H^1(\Omega)} \leq C \frac{h^p}{p^{m-1}} \|u\|_{H^m(\Omega)}$$

- Exponential convergence rate possible with hp -adaptation [Babuška and Guo, 1996]

$$\|\nabla (u - u_{hp})\|_{H^1(\Omega)} \leq C \exp(-b N_{\text{dofs}}^\alpha)$$

with $\alpha = 1/3$ in 2D and $\alpha = 1/5$ in 3D

Kelly error estimation

- Error estimator by [Kelly et al., 1983]
- Meant for generalized Poisson equation $-\nabla \cdot (a\nabla u) = f$
- Proved to be useful in other applications as well

$$\|e_{hp}\|_{H_1(\Omega)}^2 \leq c \sum_{K \in \Omega} \eta_K^2$$

$$\eta_K^2 = \sum_{F \in \partial K} c_F \int_F \left[a \frac{\partial u_{hp}}{\partial n} \right]^2 do$$

$$\left[a \frac{\partial u_{hp}}{\partial n} \right] = a \frac{\partial u_{hp}}{\partial n_K} \Big|_K + a \frac{\partial u_{hp}}{\partial n_J} \Big|_J$$

Refinement history

- Verify prediction's accuracy
- Decide on either h- or p-adaptation on marked cells

Keep **h** small: $\eta_K > \eta_{K,\text{pred}}$

Keep **p** large: $\eta_K \leq \eta_{K,\text{pred}}$

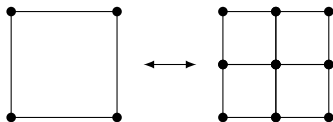


Figure: *h*-adaptation

Figure: Error prediction algorithm based on [Melenk and Wohlmuth, 2001]

Adaptation type	Prediction formula
no adaptation	$\eta_{K,\text{pred}} = \eta_K \gamma_n$
<i>p</i> -adaptation	$\eta_{K,\text{pred}} = \eta_K \gamma_p^{(p_{K,\text{future}} - p_K)}$
<i>hp</i> -refinement	$(\eta_{K_c,\text{pred}})^2 = n_{K_c}^{-1} \left(\eta_{K_p} \gamma_h 0.5^{p_{K_c,\text{future}}} \gamma_p^{(p_{K_c,\text{future}} - p_{K_p})} \right)^2$
<i>hp</i> -coarsening	$(\eta_{K_p,\text{pred}})^2 = \sum_{K_c} \left(\eta_{K_c} (\gamma_h 0.5^{p_{K_p,\text{future}}})^{-1} \gamma_p^{(p_{K_p,\text{future}} - p_{K_c})} \right)^2$

Smoothness estimation

- Different representations of finite element approximation

$$u_{hp}(x) = \sum_i u_i \varphi_i(x) = \sum_k c_k P_k(x)$$

- Choose orthogonal basis $\{P_k\}$ of increasing frequency
- Uniquely attribute frequency of oscillations to each basis function

Orthogonal basis

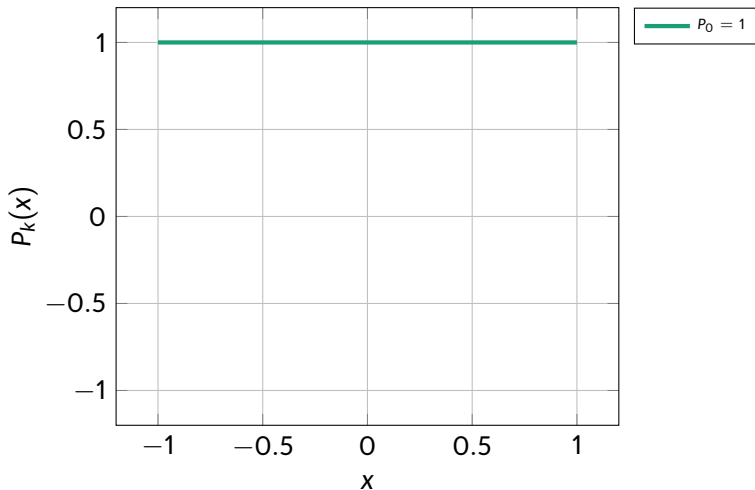
- Legendre: Orthogonal basis of polynomials

$$P_k(x) = \frac{1}{2^k k!} \frac{d^k}{dx^k} ((x^2 - 1)^k)$$
$$\langle P_k, P_l \rangle = \int P_k(x) P_l(x) dx = \frac{2}{2k + 1} \delta_{kl}$$

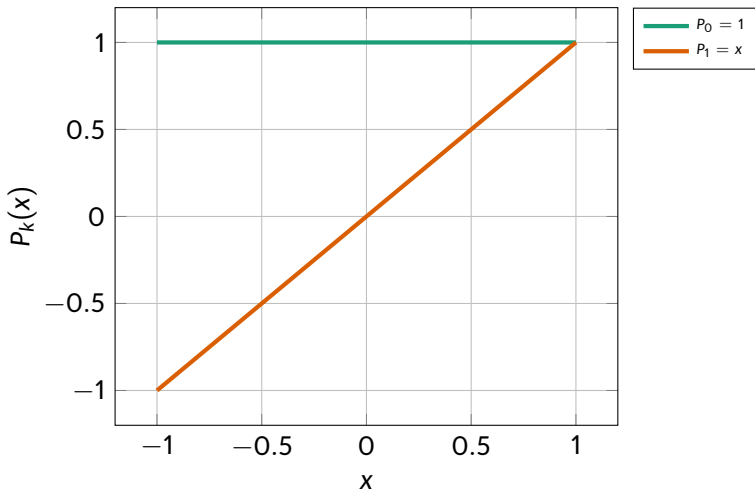
- Fourier: Orthogonal basis of sinusoids

$$P_k(x) = \exp(-i 2\pi k x)$$
$$\langle P_k, P_l \rangle = \int P_k(x) P_l^*(x) dx = \delta_{kl}$$

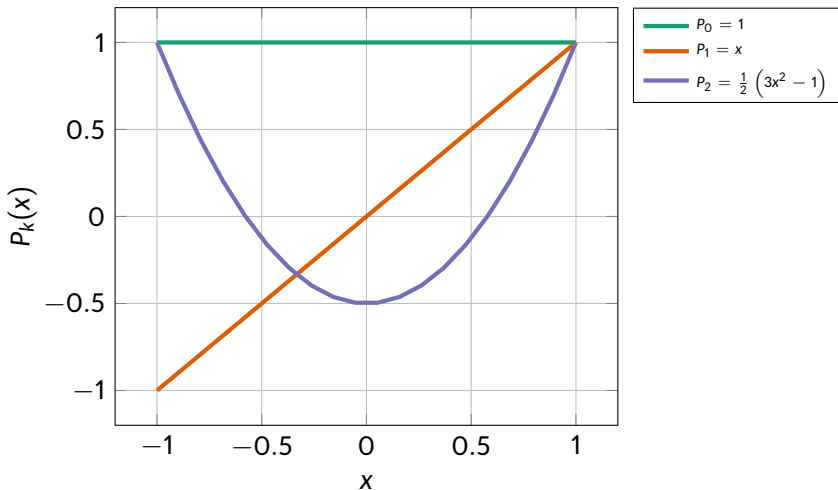
Legendre polynomials



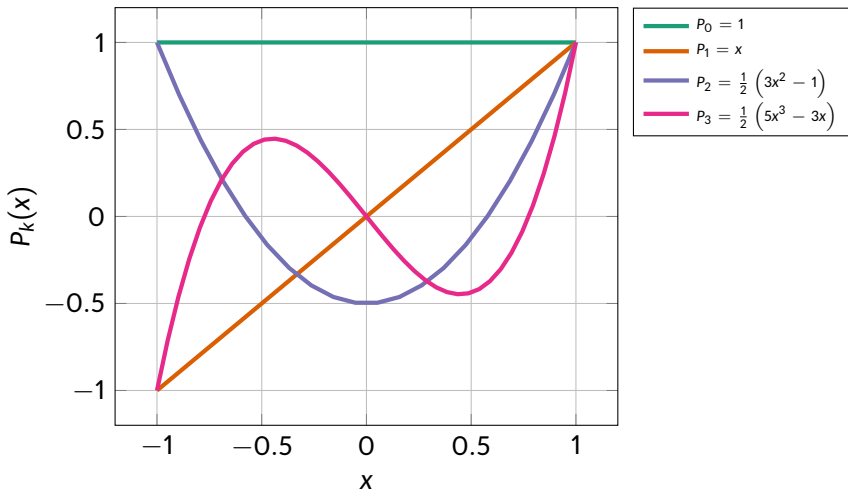
Legendre polynomials



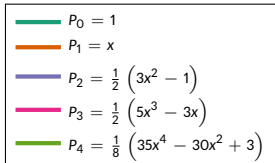
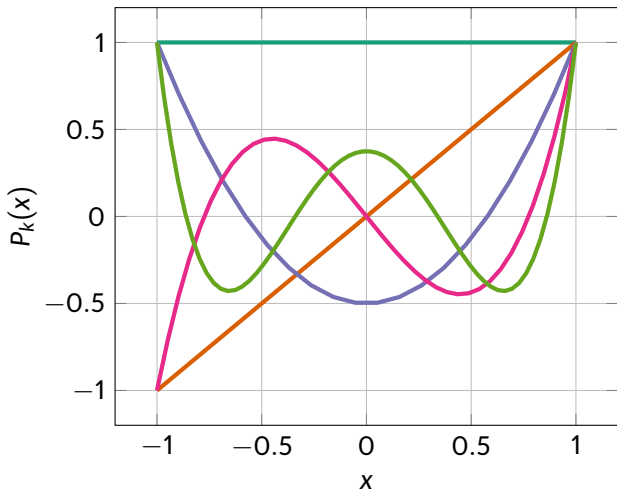
Legendre polynomials



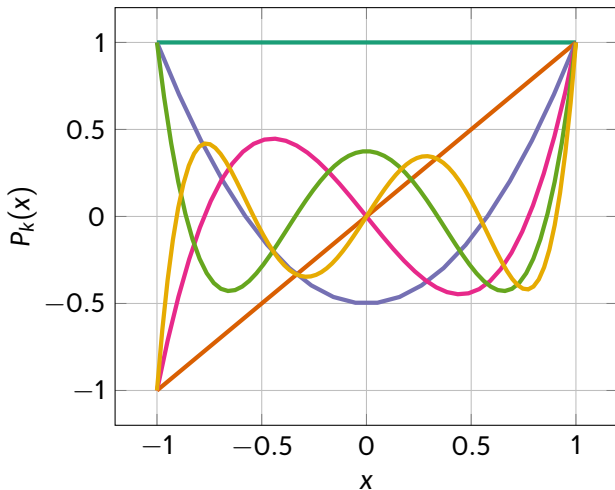
Legendre polynomials



Legendre polynomials

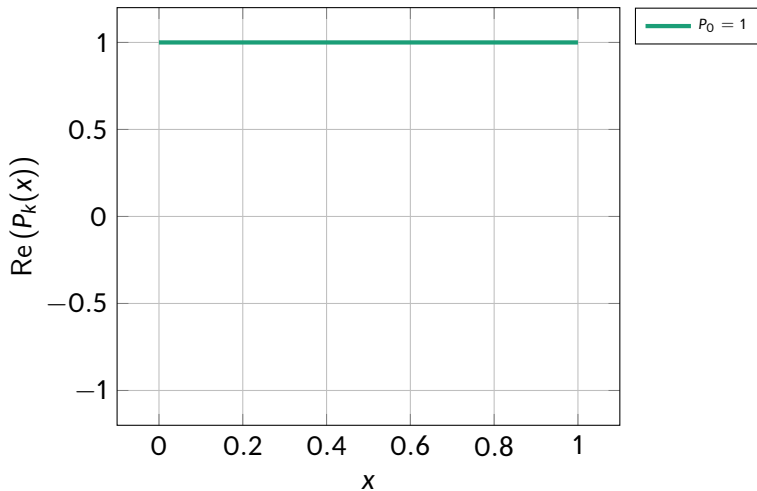


Legendre polynomials

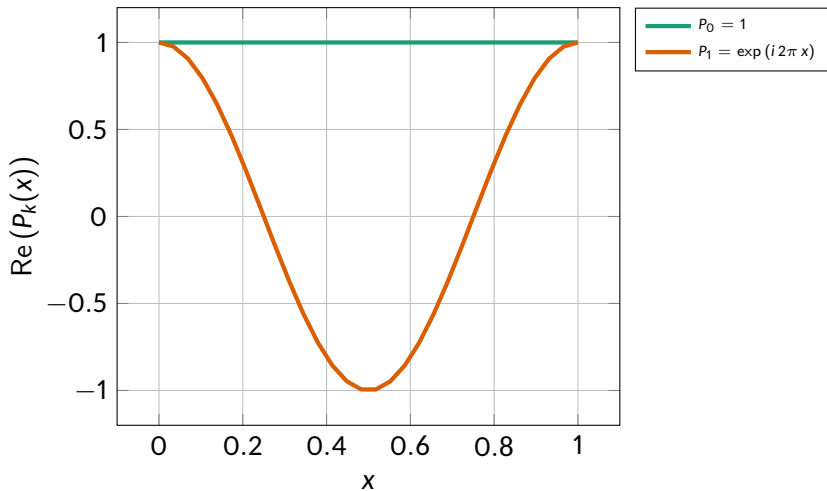


- $P_0 = 1$
- $P_1 = x$
- $P_2 = \frac{1}{2} (3x^2 - 1)$
- $P_3 = \frac{1}{2} (5x^3 - 3x)$
- $P_4 = \frac{1}{8} (35x^4 - 30x^2 + 3)$
- $P_5 = \frac{1}{8} (63x^5 - 70x^3 + 15x)$

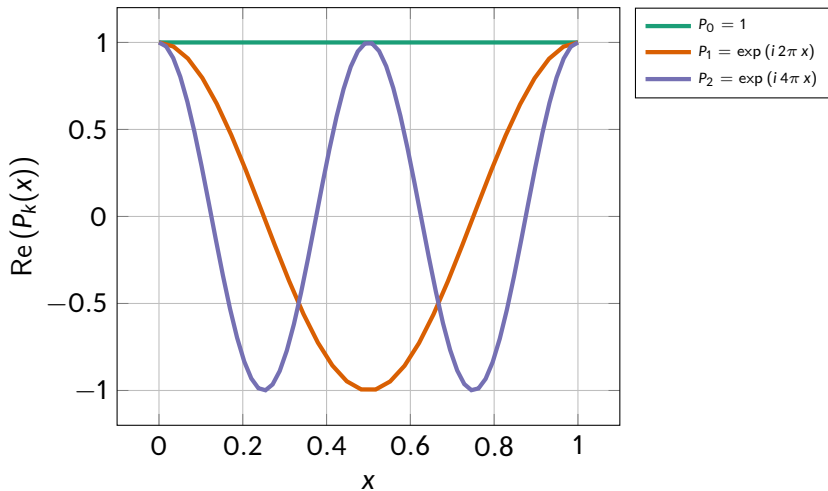
Fourier sinusoids



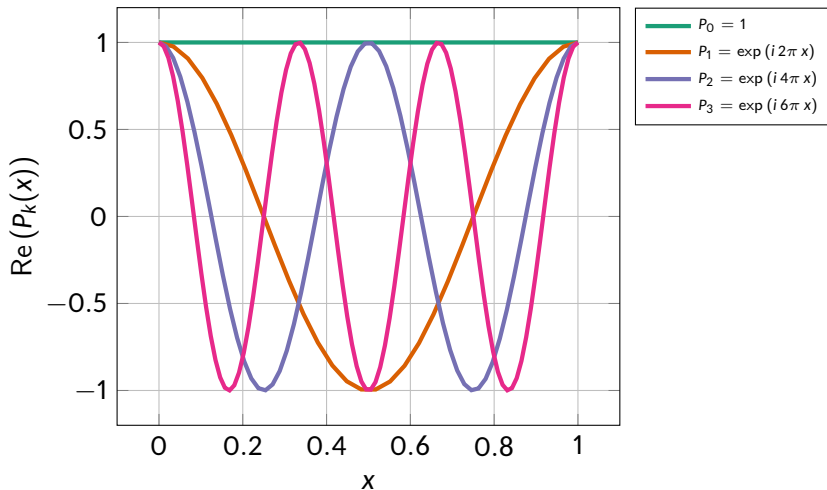
Fourier sinusoids



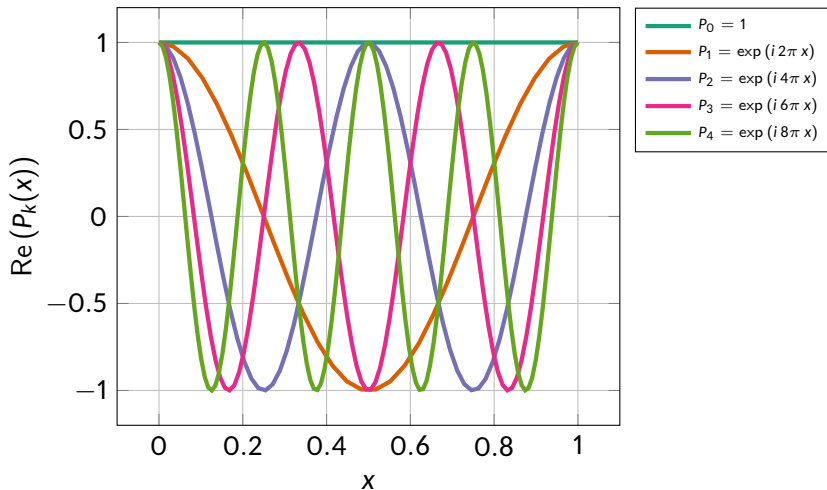
Fourier sinusoids



Fourier sinusoids



Fourier sinusoids



Decay of Legendre coefficients

- Determine coefficients $\{c_k\}$ on every cell K :

$$c_k = \int_K u_{hp}(x) P_k(x) dx$$

- FE approximation is analytic if coefficients decay exponentially [Eibner and Melenk, 2007]

$$|c_k| \leq C \exp(-\sigma|k|)$$

- Determine decay rate σ via least-squares fit

$$\ln(|c_k|) \sim \tilde{C} - \sigma|k|$$

Decay rates σ considered as smoothness estimates

Keep p large if σ is large



Decay of Fourier coefficients

- FE approximation is part of Hilbert space \mathcal{H}^s if integral exists

$$\int_K |\nabla^s u_{hp}(x)|^2 dx = (2\pi)^{2s} \sum_k |c_k|^2 k^{2s} < \infty$$

$$\Rightarrow |c_k| = \mathcal{O}(k^{-\sigma-\epsilon}) \quad \text{with} \quad \sigma = s + \dim/2$$

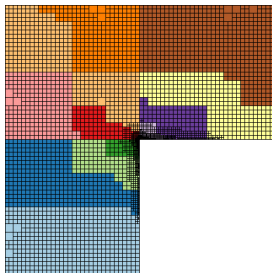
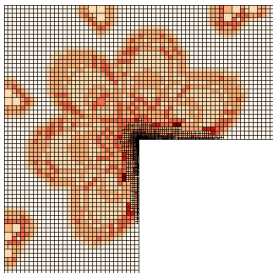
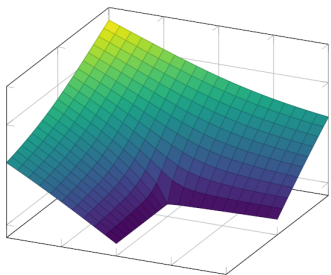
- Determine coefficients $\{c_k\}$ and their decay via least-squares fit on every cell K :

$$c_k = \int_K u_{hp}(x) P_k^*(x) dx \quad \ln(|c_k|) \sim C - \sigma \ln(|k|)$$

Decay rates σ considered as smoothness estimates

Keep p large if σ is large





MASSIVELY PARALLEL HP-ADAPTIVE FEM

More results

June 05, 2020 | Marc Fehling | m.fehling@fz-juelich.de |

Error convergence

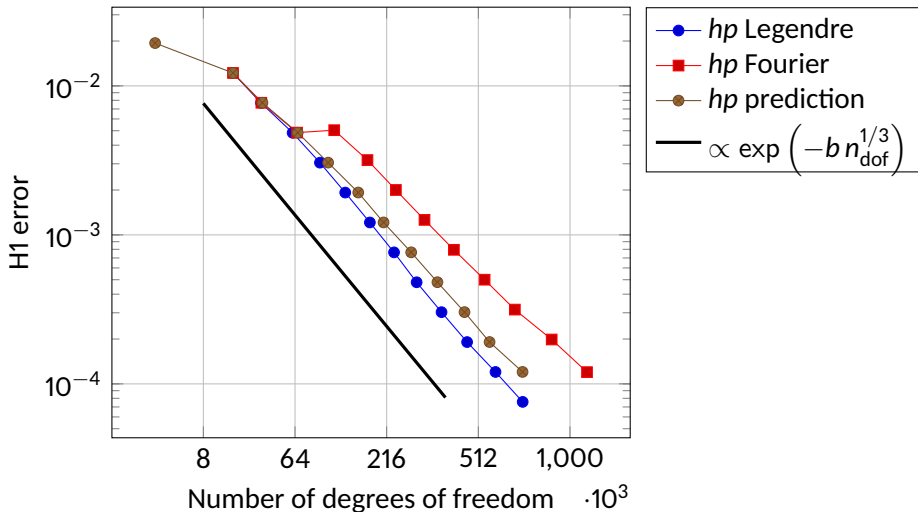


Figure: Customly scaled error convergence plot

Error vs runtime

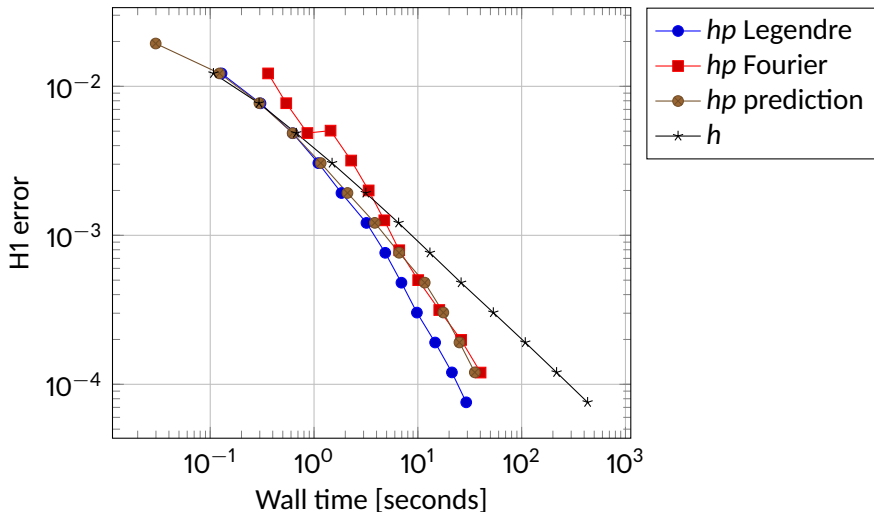


Figure: Error against walltime

Weighting exponent

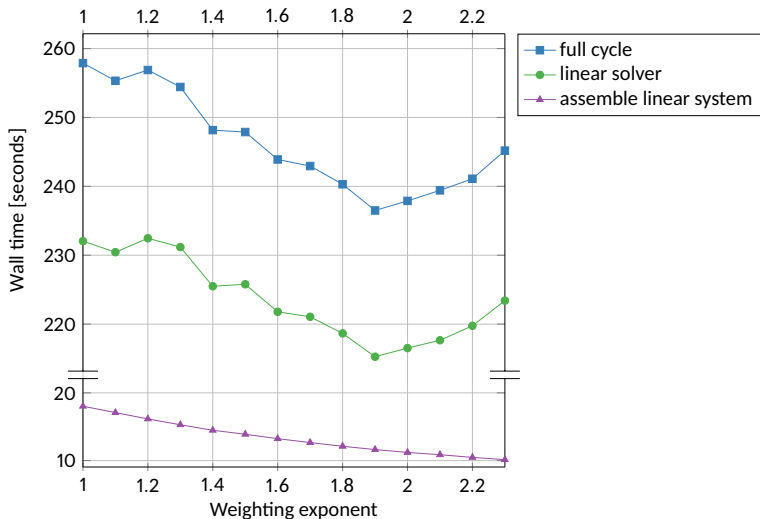


Figure: Wall times for load balancing with varying weighting exponents

Scaling on successive refinement

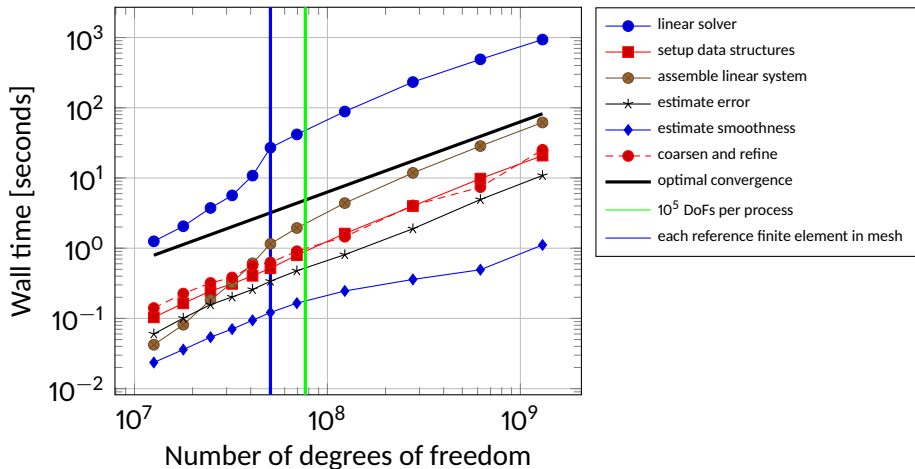


Figure: Scaling on successively refined grids