

The Gradient Discretisation Method

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- 1 Presentation of the gradient discretisation method: linear stationary diffusion**
 - From FEM to GDM
 - Measures of gradient scheme accuracy, error estimate
- 2 Gradient schemes for non-linear models**
 - Semi-linear equation
 - Quasi-linear equations (and time-stepping)
- 3 Do gradient discretisations exist?**
 - A few examples
 - Proving coercivity, limit-conformity, compactness: polytopal toolboxes
 - Proving consistency: local linearly exact GD
- 4 About time-dependent problems**
- 5 New results obtained through gradient schemes**

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Model problem

$$\begin{cases} -\operatorname{div}(\Lambda \nabla \bar{u}) = f & \text{in } \Omega, \\ \bar{u} = 0 & \text{on } \partial\Omega. \end{cases}$$

- Ω open bounded in \mathbb{R}^d ,
- $\Lambda : \Omega \rightarrow M_d(\mathbb{R})$ bounded uniformly coercive,
- $f \in L^2(\Omega)$.

Weak formulation

Find $\bar{u} \in H_0^1(\Omega)$ s.t., for all $\bar{v} \in H_0^1(\Omega)$,
$$\int_{\Omega} \Lambda \nabla \bar{u} \cdot \nabla \bar{v} = \int_{\Omega} f \bar{v}.$$

- ▶ Existence and uniqueness of the weak solution (*Riesz representation theorem, or Lax-Milgram*).

Conforming Finite Element Method (e.g., \mathbb{P}_1)

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Computational issue: $H_0^1(\Omega)$ is an infinite-dimensional space.
 \rightsquigarrow Cannot be understood/manipulated by computer.

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Easy solution: replace $H_0^1(\Omega)$ by a finite-dimensional subspace $E = V_h$.

Find $u_h \in V_h$ s.t., for all $v_h \in V_h$,
$$\int_{\Omega} \Lambda \nabla u_h \cdot \nabla v_h = \int_{\Omega} f v_h.$$

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$$\int_{\Omega} \Lambda \nabla u_h \cdot \nabla v_h = \int_{\Omega} f v_h.$$

► If $u_h = \sum_i U_i \phi_i$ on a basis $(\phi_i)_i$ of V_h , this leads to the square system

$$AU = B$$

where $A_{ij} = \int_{\Omega} \Lambda \nabla \phi_i \cdot \nabla \phi_j$ and $B_i = \int_{\Omega} f \phi_i$.

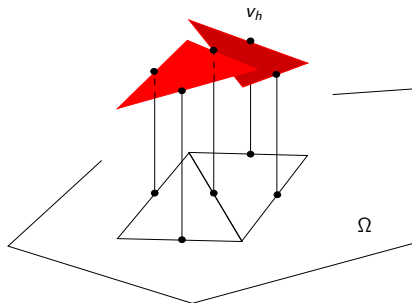
Non-conforming FEM (e.g., non-conforming \mathbb{P}_1 /Crouzet-Raviart)

Non-conforming approximation: V_h is a space of functions, but not a subspace of $H_0^1(\Omega)$.

\rightsquigarrow Need to define ∇v_h if $v_h \in V_h$?

Example: non-conforming \mathbb{P}_1 :

$V_h = \{v_h : \Omega \rightarrow \mathbb{R} : v_h \text{ piecewise linear, continuous at edge midpoints, zero at boundary edge midpoints}\}$



Non-conforming FEM (e.g., non-conforming \mathbb{P}_1 /Crouzet-Raviart)

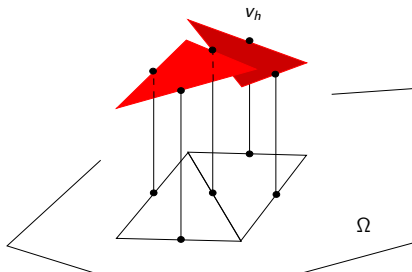
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Example: non-conforming \mathbb{P}_1 :

► ∇u_h replaced with broken gradient $\nabla_h u_h$, computed in each cell.

$$\text{Find } u_h \in V_h \text{ s.t., for all } v_h \in V_h, \quad \int_{\Omega} \Lambda \nabla_h u_h \cdot \nabla_h v_h = \int_{\Omega} f v_h.$$



Mass-lumping

Consider diffusion-reaction (from, e.g., Euler time-stepping):

$$-\operatorname{div}(\Lambda \nabla \bar{u}) + \bar{u} = f.$$

Issues:

- ▶ Requires computation of $M_{ij} = \int_{\Omega} \phi_i \phi_j$: non-diagonal mass-matrix (costly for explicit time-stepping).
- ▶ Also problematic for non-linear models $-\operatorname{div}(\Lambda \nabla \bar{u}) + \beta(\bar{u}) = f$ (e.g. Richards' equation).

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Mass-lumping

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Better vision: reconstruct piecewise constant functions $\Pi_h u_h$ and $\Pi_h v_h$ from u_h and v_h .

$$\text{Find } u_h \in V_h \text{ s.t., for all } v_h \in V_h, \\ \int_{\Omega} \Lambda \nabla_h u_h \cdot \nabla_h v_h + \int_{\Omega} \Pi_h u_h \Pi_h v_h = \int_{\Omega} f \Pi_h v_h.$$

The gradient discretisation method (GDM) in a nutshell

In the weak formulation of the PDE, replace continuous space and operators (gradient, function) by discrete space and reconstructed operators.

- ▶ Set of discrete space and reconstructed operators: *gradient discretisation* (GD).
- ▶ A large number of possible gradient discretisations.

Gradient discretisation and gradient scheme

Definition (GD)

A gradient discretisation is $\mathcal{D} = (X_{\mathcal{D},0}, \Pi_{\mathcal{D}}, \nabla_{\mathcal{D}})$ with

- $X_{\mathcal{D},0}$ finite dimensional space (of degrees of freedom), taking into account the Dirichlet BC,
- $\Pi_{\mathcal{D}} : X_{\mathcal{D},0} \rightarrow L^2(\Omega)$ reconstruction of function,
- $\nabla_{\mathcal{D}} : X_{\mathcal{D},0} \rightarrow L^2(\Omega)^d$ reconstructed gradient,

such that $v \rightarrow \|\nabla_{\mathcal{D}} v\|_{L^2(\Omega)^d}$ is a norm on $X_{\mathcal{D},0}$.

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Definition (GS)

If \mathcal{D} is a GD, the corresponding gradient scheme for $-\operatorname{div}(\Lambda \nabla \bar{u}) = f$ with homogeneous Dirichlet BC is

Find $u_{\mathcal{D}} \in X_{\mathcal{D},0}$ s.t., for all $v_{\mathcal{D}} \in X_{\mathcal{D},0}$,

$$\int_{\Omega} \Lambda \nabla_{\mathcal{D}} u_{\mathcal{D}} \cdot \nabla_{\mathcal{D}} v_{\mathcal{D}} = \int_{\Omega} f \Pi_{\mathcal{D}} v_{\mathcal{D}}.$$

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3 measures of accuracy

Measure of coercivity (discrete Poincaré's constant)

$$C_D = \max_{v_D \in X_{D,0} \setminus \{0\}} \frac{\|\Pi_D v_D\|_{L^2}}{\|\nabla_D v_D\|_{L^2}}.$$

Measure of consistency (“interpolation error” in FEM vocabulary)

$$S_D(\varphi) = \min_{v_D \in X_{D,0}} (\|\Pi_D v_D - \varphi\|_{L^2} + \|\nabla_D v_D - \nabla \varphi\|_{L^2}).$$

Measure of limit-conformity

$$W_D(\psi) = \max_{v_D \in X_{D,0} \setminus \{0\}} \frac{1}{\|\nabla_D v_D\|_{L^2}} \left| \int_{\Omega} \nabla_D v_D \cdot \psi + \Pi_D v_D \operatorname{div} \psi \right|.$$

Error estimate

$$\begin{aligned} \|\Pi_{\mathcal{D}} u_{\mathcal{D}} - \bar{u}\|_{L^2} + \|\nabla_{\mathcal{D}} u_{\mathcal{D}} - \nabla \bar{u}\|_{L^2} \\ \leq C(1 + C_{\mathcal{D}}) [S_{\mathcal{D}}(\bar{u}) + W_{\mathcal{D}}(\wedge \nabla \bar{u})]. \end{aligned}$$

Error estimate

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Convergence: if a sequence $(\mathcal{D}_m)_{m \in \mathbb{N}}$ of gradient discretisations is

(P1) **Coercive**: $(C_{\mathcal{D}_m})_{m \in \mathbb{N}}$ bounded,

(P2) **Consistent**: $S_{\mathcal{D}_m}(\varphi) \rightarrow 0$ for all $\varphi \in H_0^1(\Omega)$,

(P3) **Limit-conforming**: $W_{\mathcal{D}_m}(\psi) \rightarrow 0$ for all $\psi \in H_{\text{div}}(\Omega)$,

then $\Pi_{\mathcal{D}_m} u_m \rightarrow \bar{u}$ and $\nabla_{\mathcal{D}_m} u_m \rightarrow \nabla \bar{u}$ in L^2 .

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No more work to do to write a GS

Semi-linear model:

$$-\operatorname{div}(\Lambda(\bar{u})\nabla\bar{u}) = f \quad \text{with homogeneous Dirichlet BC.}$$

No more work to do to write a GS

Semi-linear model: Weak form:

Find $\bar{u} \in H_0^1(\Omega)$ such that, for all $\bar{v} \in H_0^1(\Omega)$,

$$\int_{\Omega} \Lambda(\bar{u}) \nabla \bar{u} \cdot \nabla \bar{v} = \int_{\Omega} f \bar{v}.$$

No more work to do to write a GS

Semi-linear model: Weak form:

$$\text{Find } \bar{u} \in H_0^1(\Omega) \text{ such that, for all } \bar{v} \in H_0^1(\Omega), \\ \int_{\Omega} \Lambda(\bar{u}) \nabla \bar{u} \cdot \nabla \bar{v} = \int_{\Omega} f \bar{v}.$$

Gradient scheme: if \mathcal{D} is a GD,

$$\text{Find } u_{\mathcal{D}} \in X_{\mathcal{D},0} \text{ s.t., for all } v_{\mathcal{D}} \in X_{\mathcal{D},0}, \\ \int_{\Omega} \Lambda(\Pi_{\mathcal{D}} u_{\mathcal{D}}) \nabla_{\mathcal{D}} u_{\mathcal{D}} \cdot \nabla_{\mathcal{D}} v_{\mathcal{D}} = \int_{\Omega} f \Pi_{\mathcal{D}} v_{\mathcal{D}}.$$

Convergence result

- ▶ To deal with low-order non-linearities, $(\mathcal{D}_m)_{m \in \mathbb{N}}$ must be (P4) Compact: (*discrete Rellich theorem*) for all $v_m \in X_{\mathcal{D}_m, 0}$ such that $(\|\nabla_{\mathcal{D}_m} v_m\|_{L^2})_{m \in \mathbb{N}}$ is bounded, the sequence $(\Pi_{\mathcal{D}_m} v_m)_{m \in \mathbb{N}}$ is relatively compact in L^2 .

Convergence result

- To deal with low-order non-linearities, $(\mathcal{D}_m)_{m \in \mathbb{N}}$ must be
- (P4) **Compact**: (*discrete Rellich theorem*) for all $v_m \in X_{\mathcal{D}_m,0}$ such that $(\|\nabla_{\mathcal{D}_m} v_m\|_{L^2})_{m \in \mathbb{N}}$ is bounded, the sequence $(\Pi_{\mathcal{D}_m} v_m)_{m \in \mathbb{N}}$ is relatively compact in L^2 .

Theorem (Convergence of the GDM for semi-linear equations)

If $(\mathcal{D}_m)_{m \in \mathbb{N}}$ is coercive, consistent, limit-conforming and compact, then, up to a subsequence,

$$\Pi_{\mathcal{D}_m} u_m \rightarrow \bar{u} \quad \text{and} \quad \nabla_{\mathcal{D}_m} u_m \rightarrow \nabla \bar{u} \quad \text{strongly in } L^2(\Omega).$$

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With $\beta(s)s \geq 0$,

$$-\operatorname{div}(\Lambda \nabla \bar{u}) + \beta(\bar{u}) = f.$$

► Weak form:

$$\bar{u} \in H_0^1(\Omega) \text{ s.t. } \forall v \in H_0^1(\Omega), \int_{\Omega} \Lambda \nabla \bar{u} \cdot \nabla v + \int_{\Omega} \beta(\bar{u})v = \int_{\Omega} f v.$$

Model

$$\bar{u} \in H_0^1(\Omega) \text{ s.t. } \forall v \in H_0^1(\Omega), \int_{\Omega} \Lambda \nabla \bar{u} \cdot \nabla v + \int_{\Omega} \beta(\bar{u}) v = \int_{\Omega} f v.$$

Application of GDM:

► Option 1:

$$u \in X_{\mathcal{D},0} \text{ s.t. } \forall v \in X_{\mathcal{D},0}, \\ \int_{\Omega} \Lambda \nabla_{\mathcal{D}} u \cdot \nabla_{\mathcal{D}} v + \int_{\Omega} \beta(\Pi_{\mathcal{D}} u) \Pi_{\mathcal{D}} v = \int_{\Omega} f \Pi_{\mathcal{D}} v.$$

$$\bar{u} \in H_0^1(\Omega) \text{ s.t. } \forall v \in H_0^1(\Omega), \int_{\Omega} \Lambda \nabla \bar{u} \cdot \nabla v + \int_{\Omega} \beta(\bar{u}) v = \int_{\Omega} f v.$$

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► Option 2:

$$u \in X_{\mathcal{D},0} \text{ s.t. } \forall v \in X_{\mathcal{D},0}, \\ \int_{\Omega} \Lambda \nabla_{\mathcal{D}} u \cdot \nabla_{\mathcal{D}} v + \int_{\Omega} \Pi_{\mathcal{D}} \beta(u) \Pi_{\mathcal{D}} v = \int_{\Omega} f \Pi_{\mathcal{D}} v$$

with $\beta(u_{\mathcal{D}}) \in X_{\mathcal{D},0}$ constructed DOF by DOF.

Stability vs. Computability

Stability: with Option 1, since $\beta(s)s \geq 0$, with $v_D = u_D$,

$$\int_{\Omega} \beta(\Pi_D u_D) \Pi_D u_D \geq 0.$$

► Problem: $\beta(\Pi_D u)$ non-linear function of $\Pi_D u$, no exact quadrature.

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Computability: with Option 2

$$\int_{\Omega} \Pi_D \beta(u_D) \Pi_D v_D$$

is fully computable since Π_D usually reconstructs piecewise polynomial functions.

► Problem: $\Pi_D \beta(u_D) \Pi_D u_D \geq 0$? Stability?

Fifth and last property

To have $\Pi_{\mathcal{D}}\beta(v_{\mathcal{D}}) = \beta(\Pi_{\mathcal{D}}v_{\mathcal{D}})$:

(P5) *Piecewise constant reconstruction*: there is a basis $(e_i)_{i \in I}$ of $X_{\mathcal{D},0}$ and a partition $(\omega_i)_i$ of Ω such that, if $v = \sum_i v_i e_i \in X_{\mathcal{D},0}$,

$$\forall i \in I, \quad (\Pi_{\mathcal{D}}v_{\mathcal{D}})|_{\omega_i} = v_i.$$

5 properties to rule them all, and in the light of conciseness bind them

- (P1) *Coercivity*
- (P2) *Consistency*
- (P3) *Limit-conformity*
- (P4) *Compactness*
- (P5) *Piecewise constant reconstruction*

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Conforming Galerkin (incl. \mathbb{P}_k FE)

V finite dimensional subspace of $H_0^1(\Omega)$, $(\phi_i)_{i \in I}$ basis of V .

- $X_{\mathcal{D},0} = \{v_{\mathcal{D}} = (v_i)_{i \in I}\}$,
- $\Pi_{\mathcal{D}} v_{\mathcal{D}} = \sum_{i \in I} v_i \phi_i$,
- $\nabla_{\mathcal{D}} v_{\mathcal{D}} = \nabla(\Pi_{\mathcal{D}} v_{\mathcal{D}}) = \sum_{i \in I} v_i \nabla \phi_i$.

► Remark: $C_{\mathcal{D}} \leq C_P$ (Poincaré constant in H_0^1) and $W_{\mathcal{D}} \equiv 0$ (conforming method).

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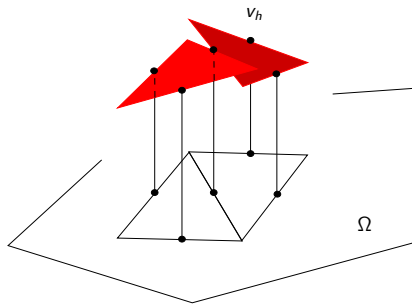
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▶ Mass-lumped method: change $\Pi_{\mathcal{D}}$.

Non-conforming \mathbb{P}_1

\mathcal{T}_h triangulation, $(\phi_\sigma)_{\sigma \in \mathcal{E}_{\text{int}}}$ non-conforming \mathbb{P}_1 basis (one element per interior edge).

- $\mathcal{X}_{\mathcal{D},0} = \{v_{\mathcal{D}} = (v_\sigma)_{\sigma \in \mathcal{E}_{\text{int}}}\}$,
- $\Pi_{\mathcal{D}} v_{\mathcal{D}} = \sum_{\sigma \in \mathcal{E}_{\text{int}}} v_\sigma \phi_\sigma$,
- $\nabla_{\mathcal{D}} v_{\mathcal{D}} = \nabla_h(\Pi_{\mathcal{D}} v_{\mathcal{D}})$ broken gradient.



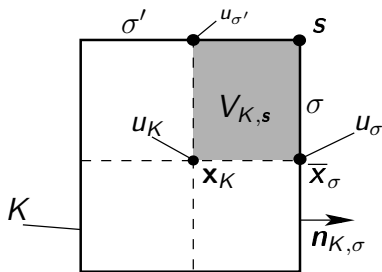
► Mass-lumped non-conforming \mathbb{P}_1 : change $\Pi_{\mathcal{D}}$.

Finite volume/finite difference scheme ($\Lambda = \text{Id}$)

\mathcal{T} rectangular mesh.

- $X_{\mathcal{D},0} = \{v = ((v_K)_{K \in \mathcal{M}}, (v_\sigma)_{\sigma \in \mathcal{E}_{\text{int}}})\}$, cell and interior edge DOFs.
- $\Pi_{\mathcal{D}}v = v_K$ in $K \in \mathcal{M}$ (piecewise constant),
- On $V_{K,s}$ as in the figure,

$$(\nabla_{\mathcal{D}}v)|_{V_{K,s}} = \frac{v_\sigma - v_K}{d(\bar{\mathbf{x}}_\sigma, \bar{\mathbf{x}}_K)} \mathbf{n}_{K,\sigma} + \frac{v_{\sigma'} - v_K}{d(\bar{\mathbf{x}}_{\sigma'}, \bar{\mathbf{x}}_K)} \mathbf{n}_{K,\sigma'}.$$



Also...

- Mixed \mathbb{RT}_k finite elements,
- Multi-point flux approximation-O (MPFA-O) finite volumes on cartesian or simplicial meshes,
- Discrete duality finite volumes,
- Hybrid mimetic mixed (HMM) schemes, including mixed-hybrid mimetic finite difference (MFD) schemes,
- Nodal mimetic finite difference methods,
- Vertex approximate gradient (VAG) schemes...

Take-home message: what can GDM do for you?

- (S1) Develop/take your favourite method, say \mathcal{FM} , for linear diffusion stationary equation (E),
- (S2) Identify a gradient discretisation $\mathcal{D}_{\mathcal{FM}}$ such that the correspond gradient scheme for (E) is precisely \mathcal{FM} ,
- (S3) Prove that $\mathcal{D}_{\mathcal{FM}}$ satisfies the *coercivity*, *consistency*, *limit-conformity*, *compactness* and has, perhaps, *piecewise constant reconstruction*,

\rightsquigarrow through $\mathcal{D}_{\mathcal{FM}}$, \mathcal{FM} can be applied to any model studied in the framework of gradient scheme, and yields a converging scheme (without additional work).

Exemple of models: stationary and transient Leray–Lions (p -Laplace), doubly degenerate parabolic, Stokes, Navier-Stokes, variational inequalities, diphasic flows in fractured networks...

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Polytopal toolbox

Set of scheme-independent results, for polytopal meshes, to prove (P1), (P3), (P4) (coercivity, limit-conformity, compactness).

Polytopal toolbox: discrete objects

- $\mathfrak{T} = (\mathcal{M}, \mathcal{E}, \mathcal{V}, \mathcal{P})$ polytopal mesh: cells, faces, vertices, cell “centers”.
- Face and cell DOF gathered in $X_{\mathfrak{T},0}$:

$$X_{\mathfrak{T},0} = \{v = ((v_K)_{K \in \mathcal{M}}, (v_\sigma)_{\sigma \in \mathcal{E}}) \text{ with } v_\sigma = 0 \text{ if } \sigma \subset \partial\Omega\}.$$

Polytopal toolbox: discrete objects

- $\mathfrak{T} = (\mathcal{M}, \mathcal{E}, \mathcal{V}, \mathcal{P})$ polytopal mesh: cells, faces, vertices, cell “centers”.
- Face and cell DOF gathered in $X_{\mathfrak{T},0}$:

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- $\Pi_{\mathfrak{T}} : X_{\mathfrak{T},0} \rightarrow L^\infty(\Omega)$ defined by

$$(\Pi_{\mathfrak{T}} v)|_K = v_K \text{ for all } K \in \mathcal{M},$$

- $\bar{\nabla}_{\mathfrak{T}} : X_{\mathfrak{T},0} \rightarrow (L^\infty(\Omega))^d$ defined by

$$(\bar{\nabla}_{\mathfrak{T}} v)|_K = \frac{1}{|K|} \sum_{\sigma \in \mathcal{E}_K} |\sigma| v_\sigma \mathbf{n}_{K,\sigma} \text{ for all } K \in \mathcal{M}.$$

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- “ H_0^1 -norm” on $X_{\mathfrak{T},0}$:

$$\|v\|_{\mathfrak{T}} = \left(\sum_{K \in \mathcal{M}} \sum_{\sigma \in \mathcal{E}_K} |\sigma| d_{K,\sigma} \left| \frac{v_\sigma - v_K}{d_{K,\sigma}} \right|^2 \right)^{1/2}.$$

Polytopal toolbox: discrete functional analysis

Under standard regularity assumptions on \mathcal{T} :

- Poincaré inequality:

$$\|\Pi_{\mathcal{T}} v\|_{L^2(\Omega)} \leq C \|v\|_{\mathcal{T}}.$$

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$$\left| \int_{\Omega} (\bar{\nabla}_{\mathfrak{T}} v \cdot \varphi + \Pi_{\mathfrak{T}} v \operatorname{div}(\varphi)) \right| \leq C \|\nabla \varphi\|_{L^2(\Omega)^d} \|v\|_{\mathfrak{T}} h_{\mathcal{M}}.$$

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$$\left| \int_{\Omega} (\overline{\nabla}_{\mathfrak{T}} v \cdot \varphi + \Pi_{\mathfrak{T}} v \operatorname{div}(\varphi)) \right| \leq C \|\nabla \varphi\|_{L^2(\Omega)^d} \|v\|_{\mathfrak{T}} h_{\mathcal{M}}.$$

- Discrete Rellich theorem: if $(\|v_m\|_{\mathfrak{T}_m})_{m \in \mathbb{N}}$ is bounded then $(\Pi_{\mathfrak{T}_m} v_m)_{m \in \mathbb{N}}$ is relatively compact in $L^2(\Omega)$.

Polytopal toolbox: control of a GD

Definition (Control of a GD by a polytopal toolbox)

A control of \mathcal{D} by \mathfrak{T} is a linear mapping $\Phi : X_{\mathcal{D},0} \rightarrow X_{\mathfrak{T},0}$.

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► Regularity factors:

$$\|\Phi\|_{\mathcal{D},\mathfrak{T}} = \max_{v \in X_{\mathcal{D},0} \setminus \{0\}} \frac{\|\Phi(v)\|_{\mathfrak{T}}}{\|\nabla_{\mathcal{D}} v\|_{L^2}},$$

$$\omega^{\Pi}(\mathcal{D}, \mathfrak{T}, \Phi) = \max_{v \in X_{\mathcal{D},0} \setminus \{0\}} \frac{\|\Pi_{\mathcal{D}} v - \Pi_{\mathfrak{T}} \Phi(v)\|_{L^2(\Omega)}}{\|\nabla_{\mathcal{D}} v\|_{L^2}},$$

$$\omega^{\nabla}(\mathcal{D}, \mathfrak{T}, \Phi) = \max_{v \in X_{\mathcal{D},0} \setminus \{0\}} \frac{\left(\sum_{K \in \mathcal{M}} |K|^{-1} \left| \int_K [\nabla_{\mathcal{D}} v - \bar{\nabla}_{\mathfrak{T}} \Phi(v)] \right|^2 \right)^{\frac{1}{2}}}{\|\nabla_{\mathcal{D}} v\|_{L^2}}.$$

Polytopal toolbox: control of a GD

Estimates through control:

$$C_{\mathcal{D}} \leq \omega^{\Pi}(\mathcal{D}, \mathfrak{T}, \Phi) + C\|\Phi\|_{\mathcal{D}, \mathfrak{T}}.$$

$$W_{\mathcal{D}}(\varphi) \leq \|\varphi\|_{H^1(\Omega)^d} \left[Ch_{\mathcal{M}}(1 + \|\Phi\|_{\mathcal{D}, \mathfrak{T}}) + \omega^{\Pi}(\mathcal{D}, \mathfrak{T}, \Phi) + \omega^{\nabla}(\mathcal{D}, \mathfrak{T}, \Phi) \right].$$

Polytopal toolbox: control of a GD

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$$W_{\mathcal{D}}(\varphi) \leq \|\varphi\|_{H^1(\Omega)^d} \left[Ch_{\mathcal{M}}(1 + \|\Phi\|_{\mathcal{D}, \mathfrak{T}}) + \omega^{\Pi}(\mathcal{D}, \mathfrak{T}, \Phi) + \omega^{\nabla}(\mathcal{D}, \mathfrak{T}, \Phi) \right].$$

Consequence for a sequence $(\mathcal{D}_m)_m$ of GD: if Φ_m control of \mathcal{D}_m by \mathfrak{T}_m such that

$$\begin{aligned} h_{\mathcal{M}_m} &\rightarrow 0, & \sup_{m \in \mathbb{N}} \|\Phi_m\|_{\mathcal{D}_m, \mathfrak{T}_m} &< +\infty, \\ \lim_{m \rightarrow \infty} \omega^{\Pi}(\mathcal{D}_m, \mathfrak{T}_m, \Phi_m) &= 0, & \lim_{m \rightarrow \infty} \omega^{\nabla}(\mathcal{D}_m, \mathfrak{T}_m, \Phi_m) &= 0, \end{aligned}$$

then $(\mathcal{D}_m)_m$ is coercive, limit-conforming and compact.

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Local linearly exact (LLE) gradient discretisations

Definition (LLE GD)

Let \mathfrak{T} be a mesh. A GD \mathcal{D} is local linearly exact if $X_{\mathcal{D},0} = \mathbb{R}^I$ and, for each cell K , there is $I_K \subset I$ s.t.

$$\forall \mathbf{x} \in K, \Pi_{\mathcal{D}} v(\mathbf{x}) = \sum_{i \in I_K} v_i \alpha_K^i(\mathbf{x}) \quad \text{and} \quad \nabla_{\mathcal{D}} v(\mathbf{x}) = \sum_{i \in I_K} v_i \mathcal{G}_K^i(\mathbf{x})$$

where, for some $(\mathbf{x}_i)_{i \in I_K}$ close to K ,

$$\forall \mathbf{x} \in K, \sum_{i \in I_K} \alpha_K^i(\mathbf{x}) = 1 \quad \text{and} \quad \forall q_1 \in \mathbb{P}_1, \sum_{i \in I_K} q_1(\mathbf{x}_i) \mathcal{G}_K^i(\mathbf{x}) = \nabla q_1.$$

- A parameter $\text{reg}_{\text{LLE}}(\mathcal{D})$ measures, for all $K \in \mathcal{M}$,
- (i) how far $(\mathbf{x}_i)_{i \in I_K}$ are from K ,
 - (ii) scaled $L^2(K)$ norms of α_K^i and \mathcal{G}_K^i .

Local linearly exact (LLE) gradient discretisations

Theorem (LLE GD are consistent (i.e. satisfy (P2)))

If $(\mathcal{D}_m)_m$ are LLE GD associated with meshes $(\mathcal{T}_m)_m$ such that $h_{\mathcal{M}_m} \rightarrow 0$ and $(\text{reg}_{\text{LLE}}(\mathcal{D}_m))_m$ is bounded, then $(\mathcal{D}_m)_m$ is consistent.

Entire proof of (P1)–(P4) for non-conforming \mathbb{P}_1

Proof of the property (P) for non-conforming \mathbb{P}_1 gradient discretisations. We drop the index m from time to time for sake of legibility, and all constants below do not depend on m or the considered cells/edges. Let us define a control of \mathcal{D} by \mathcal{T} in the sense of Definition 2.29, where \mathcal{T} is the simplicial mesh associated to \mathcal{D} , with $\mathbf{x}_K = \bar{\mathbf{x}}_K = \frac{1}{d+1} \sum_{\sigma \in \mathcal{E}_K} \bar{\mathbf{x}}_\sigma$ the centres of gravity of the cells K . We define the linear (injective) mappings $\Phi : X_{\mathcal{D},m,0} \rightarrow X_{\mathcal{T},m,0}$ by $\Phi(u)_K = \frac{1}{d+1} \sum_{\sigma \in \mathcal{E}_K} u_\sigma = \Pi_{\mathcal{D}} u(\mathbf{x}_K)$ and $\Phi(u)_\sigma = u_\sigma = \Pi_{\mathcal{D}} u(\bar{\mathbf{x}}_\sigma)$.

Since $\Phi(u)_K = \Pi_{\mathcal{D}} u(\mathbf{x}_K)$ and $\mathcal{G}_K u = \nabla(\Pi_{\mathcal{D}} u)$ in K , we get

$$\Phi(u)_\sigma - \Phi(u)_K = \mathcal{G}_K u \cdot (\bar{\mathbf{x}}_\sigma - \mathbf{x}_K). \quad (3.2)$$

Therefore, since $\frac{|\bar{\mathbf{x}}_\sigma - \mathbf{x}_K|}{d_{K,\sigma}} \leq \frac{h_K}{d_{K,\sigma}} \leq \theta_{\mathcal{T}}$,

$$\sum_{\sigma \in \mathcal{E}_K} |\sigma| d_{K,\sigma} \left| \frac{\Phi(u)_\sigma - \Phi(u)_K}{d_{K,\sigma}} \right|^p \leq \theta_{\mathcal{T}}^p d|K| |\mathcal{G}_K u|^p.$$

This implies (2.34). We now observe that the affine function α_σ reaches its extremal values at the vertices of K . It is easy to see that $\alpha_\sigma(\mathbf{v}_\sigma) = 1 - d$, where \mathbf{v}_σ is the vertex opposite to the face σ , and that $\alpha_\sigma(\mathbf{v}_{\sigma'}) = 1$ for all $\sigma' \neq \sigma$. Therefore, for $\mathbf{x} \in K$,

$$|\Pi_{\mathcal{D}} u(\mathbf{x}) - \Phi(u)_K| = \left| \sum_{\sigma \in \mathcal{E}_K} (\Phi(u)_\sigma - \Phi(u)_K) \alpha_\sigma(\mathbf{x}) \right| \leq (d+1) \max(1, d-1) \max_{\sigma \in \mathcal{E}_K} |\mathcal{G}_K u \cdot (\bar{\mathbf{x}}_\sigma - \mathbf{x}_K)|.$$

This inequality implies $\omega^\Pi(\mathcal{D}, \mathcal{T}, \Phi) \leq (d+1) \max(1, d-1) h_{\mathcal{M}}$ and therefore (2.35) holds. Finally, recalling that $\Pi_{\mathcal{D}} u$ is affine in each simplex K and that $\nabla_{\mathcal{T}}$ is exact on interpolants of affine functions (cf. Lemma 2.28), we see that $\nabla_{\mathcal{D}} u = \nabla_{\mathcal{T}} \Phi(u)$ in Ω . Hence $\omega^\nabla(\mathcal{D}, \mathcal{T}, \Phi) = 0$ and (2.36) holds. Proposition 2.31 therefore shows that $(\mathcal{D}_m)_{m \in \mathbb{N}}$ is coercive in the sense of Definition 2.2, limit-conforming in the sense of Definition 2.4, and compact in the sense of Definition 2.5.

Since non-conforming \mathbb{P}_1 gradient discretisations are LLE gradient discretisations, the consistency of $(\mathcal{D}_m)_{m \in \mathbb{N}}$ follows from Proposition 2.14 by noticing that $\text{reg}_{\text{LLE}}(\mathcal{D}_m)$ is controlled by $\theta_{\mathcal{T}_m}$. \square

Entire proof of (P1)–(P4) for MPFA-O

Proof of the property (P) for MPFA-O gradient discretisations. We drop the indices m for sake of legibility. We consider the polytopal mesh $\mathcal{T} = (\mathcal{M}, \mathcal{E}', \mathcal{P}, \mathcal{V}')$ where the sets $(\mathcal{M}, \mathcal{P})$ are those of the original polytopal mesh, $\mathcal{E}' = \{\sigma_v; \sigma \in \mathcal{E}, v \in \mathcal{V}\}$, and \mathcal{V}' is the set of all vertices of the elements of \mathcal{E}' . We define a control of \mathcal{D} by \mathcal{T} (in the sense of Definition 2.29) as the isomorphism $\Phi : X_{\mathcal{D},0} \rightarrow X_{\mathcal{T},0}$ given by $\Phi(u)_K = u_K$ and $\Phi(u)_{\sigma_v} = u_{(\sigma,v)}$. We observe that

$$\int_K |\nabla_{\mathcal{D}} u(\mathbf{x})|^p d\mathbf{x} \geq C_3 \sum_{\sigma \in \mathcal{E}_K} \sum_{v \in \mathcal{V}_\sigma} |\sigma_v| d_{K,\sigma} \left| \frac{u_{(\sigma,v)} - u_K}{d_{K,\sigma}} \right|^p,$$

with $C_3 = 1$ for parallelepipedic meshes, and $C_3 > 0$ depends on an upper bound of the regularity of the mesh for simplicial meshes. Therefore $\|\nabla_{\mathcal{D}} u\|_{L^p(\Omega)^d}^p \geq C_3 \|\Phi(u)\|_{\mathcal{T},0,p}^p$ and (2.34) is proved. Since $\Pi_{\mathcal{D}} u = \Pi_{\mathcal{T}} \Phi(u)$, we get $\omega^{\Pi}(\mathcal{D}, \mathcal{T}, \Phi) = 0$, which proves (2.35). Finally, we have

$$\int_K \nabla_{\mathcal{D}} u(\mathbf{x}) d\mathbf{x} = \sum_{\sigma \in \mathcal{E}_K} \sum_{v \in \mathcal{V}_\sigma} |\sigma_v| (u_{\sigma,v} - u_K) \mathbf{n}_{K,\sigma} = \sum_{\sigma' \in \mathcal{E}'_K} |\sigma'| (\Phi(u)_{\sigma'} - \Phi(u)_K) \mathbf{n}_{K,\sigma'} = |K| \nabla_{\mathcal{T}} \Phi(u)|_K.$$

This shows that $\omega^{\nabla}(\mathcal{D}, \mathcal{T}, \Phi) = 0$, which establishes (2.36). Proposition 2.31 therefore shows that $(\mathcal{D}_m)_{m \in \mathbb{N}}$ is coercive in the sense of Definition 2.2, limit-conforming in the sense of Definition 2.4, and compact in the sense of Definition 2.5.

It is proved in [40,41] that the definitions of the approximation points S give the LLE property in both the Cartesian and simplicial cases. Hence, the consistency of $(\mathcal{D}_m)_{m \in \mathbb{N}}$ follows from Proposition 2.14. \square

Entire proof of (P1)–(P4) for HMM

Proof of the property (P) for HMM gradient discretisations. Let $(\mathcal{D}_m)_{m \in \mathbb{N}}$ be HMM gradient discretisations built on polytopal meshes $(\mathcal{T}_m)_{m \in \mathbb{N}}$, and let us define a control of \mathcal{D}_m by \mathcal{T}_m in the sense of Definition 2.29. We drop the index m from time to time. Since $X_{\mathcal{D},0} = X_{\mathcal{T},0}$, we can take $\Phi = \text{Id}$. Estimate (2.34) is given by (3.14). Relation (2.35) follows immediately since $\omega^\Pi(\mathcal{D}, \mathcal{T}, \Phi) = 0$, owing to $\Pi_{\mathcal{D}}u = \Pi_{\mathcal{T}}u = \Pi_{\mathcal{T}}\Phi(u)$. Recalling that $|D_{K,\sigma}| = \frac{|\sigma|d_{K,\sigma}}{d}$ we have

$$\int_K \nabla_{\mathcal{D}}u(\mathbf{x})d\mathbf{x} = |K|\nabla_K u + \frac{1}{\sqrt{d}} \sum_{\sigma \in \mathcal{E}_K} |\sigma|[\mathcal{L}_K R_K(Q_K(u))]_{\sigma} \mathbf{n}_{K,\sigma}. \quad (3.15)$$

The definition of R_K and the property $\sum_{\sigma \in \mathcal{E}_K} |\sigma| \mathbf{n}_{K,\sigma} (\bar{\mathbf{x}}_{\sigma} - \mathbf{x}_K)^T = |K|\text{Id}$ (a consequence of Stokes' formula) show that for any $\eta \in \text{Im}(R_K)$ we have $\sum_{\sigma \in \mathcal{E}_K} |\sigma| \eta_{\sigma} \mathbf{n}_{K,\sigma} = 0$. Hence, since $\text{Im}(\mathcal{L}_K) = \text{Im}(R_K)$, (3.15) gives

$$\int_K \nabla_{\mathcal{D}}u(\mathbf{x})d\mathbf{x} = |K|\nabla_K u = |K|\nabla_{\mathcal{T}}\Phi(u)|_K,$$

which shows that $\omega^\nabla(\mathcal{D}, \mathcal{T}, \Phi) = 0$, and thus that (2.36) holds. The coercivity, limit-conformity and compactness of $(\mathcal{D}_m)_{m \in \mathbb{N}}$ therefore follow from Proposition 2.31. Since HMM gradient discretisations are LLE gradient discretisations, the consistency of $(\mathcal{D}_m)_{m \in \mathbb{N}}$ readily follows from Proposition 2.14, after noticing that the regularity assumption on $(\mathcal{D}_m)_{m \in \mathbb{N}}$ gives a bound on $(\text{reg}_{\text{LLE}}(\mathcal{D}_m))_{m \in \mathbb{N}}$. \square

Entire proof of (P1)–(P4) for nodal MFD

Proof of the property (P) for the nMFD gradient discretisation. As in previous proofs, we drop indices m from time to time. We define a control Φ of \mathcal{D} by \mathcal{T} , in the sense of Definition 2.29, by

$$\forall K \in \mathcal{M}, \Phi(u)_K = u_K = \frac{1}{|K|} \sum_{v \in \mathcal{V}_K} \omega_K^v u_v \quad \text{and} \quad \forall \sigma \in \mathcal{E}, \Phi(u)_\sigma = \frac{1}{|\sigma|} \sum_{v \in \mathcal{V}_\sigma} \omega_\sigma^v u_v. \quad (3.28)$$

Let us prove (2.34). Since $\sum_{v \in \mathcal{V}_\sigma} \omega_\sigma^v = |\sigma|$ we have $\Phi(u)_\sigma - \Phi(u)_K = \frac{1}{|\sigma|} \sum_{v \in \mathcal{V}_\sigma} \omega_\sigma^v (u_v - u_K)$. Therefore, using Jensen's inequality and the fact that $\frac{1}{d_{K,\sigma}} \leq \frac{\theta_{\mathcal{T}}}{h_K}$ we find

$$\begin{aligned} \sum_{\sigma \in \mathcal{E}_K} |\sigma| d_{K,\sigma} \left| \frac{\Phi(u)_\sigma - \Phi(u)_K}{d_{K,\sigma}} \right|^p &\leq \sum_{\sigma \in \mathcal{E}_K} d_{K,\sigma} \sum_{v \in \mathcal{V}_\sigma} \omega_\sigma^v \left| \frac{u_v - u_K}{d_{K,\sigma}} \right|^p \leq \theta_{\mathcal{T}}^p \sum_{\sigma \in \mathcal{E}_K} d_{K,\sigma} \sum_{v \in \mathcal{V}_\sigma} \omega_\sigma^v \left| \frac{u_v - u_K}{h_K} \right|^p \\ &\leq \theta_{\mathcal{T}}^p \sum_{v \in \mathcal{V}_K} \left(\sum_{\sigma \in \mathcal{E}_{K,v}} d_{K,\sigma} \omega_\sigma^v \right) \left| \frac{u_v - u_K}{h_K} \right|^p = \theta_{\mathcal{T}}^p d \sum_{v \in \mathcal{V}_K} |V_{K,v}| \left| \frac{u_v - u_K}{h_K} \right|^p. \end{aligned}$$

We conclude the proof of (2.34) thanks to (3.26). Since $\Pi_{\mathcal{D}} u = \Pi_{\mathcal{T}} \Phi(u)$, we have $\omega^\Pi(\mathcal{D}, \mathcal{T}, \Phi) = 0$ and (2.35) follows. For $K \in \mathcal{M}$ we have $\nabla_K u = (\nabla_{\mathcal{T}} \Phi(u))|_K$. Therefore

$$\int_K \nabla_{\mathcal{D}} u(\mathbf{x}) d\mathbf{x} = |K| (\nabla_{\mathcal{T}} \Phi(u))|_K + \frac{1}{d} \sum_{v \in \mathcal{V}_K} [\mathcal{L}_K R_K(Q_K(v))]_v \sum_{\sigma \in \mathcal{E}_{K,v}} \omega_\sigma^v \mathbf{n}_{K,\sigma}. \quad (3.29)$$

Similarly as for the HMM method, for any $\eta \in \text{Im}(R_K)$ we have $\sum_{v \in \mathcal{V}_K} \eta_v \sum_{\sigma \in \mathcal{E}_{K,v}} \omega_\sigma^v \mathbf{n}_{K,\sigma} = 0$. Hence, the last term in (3.29) vanishes and (2.36) holds since $\omega^\nabla(\mathcal{D}, \mathcal{T}, \Phi) = 0$. Hence the hypotheses of Proposition 2.31 are verified, which shows that $(\mathcal{D}_m)_{m \in \mathbb{N}}$ is coercive, limit-conforming and compact.

By noticing that $\text{reg}_{\text{LLE}}(\mathcal{D}_m)$ remains bounded by regularity assumption on $(\mathcal{D}_m)_{m \in \mathbb{N}}$, the consistency of $(\mathcal{D}_m)_{m \in \mathbb{N}}$ is an immediate consequence of Proposition 2.14 since nMFD gradient discretisations are LLE gradient discretisations. \square

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A catch-all model

$$\begin{cases} \partial_t \beta(\bar{u}) - \operatorname{div}(\mathbf{a}(\mathbf{x}, \nabla \zeta(\bar{u}))) = f & \text{in } \Omega \times (0, T), \\ \zeta(\bar{u}) = 0 & \text{on } \partial\Omega \times (0, T), \\ \beta(\bar{u}) = \beta(u_{\text{ini}}) & \text{at } t = 0. \end{cases}$$

- $\mathbf{a}(\mathbf{x}, \xi) = |\xi|^{p-2}\xi$, $\beta(s) = \zeta(s) = s$:
transient **p -Laplace** (also for generic Leray–Lions operator),
- $\mathbf{a}(\mathbf{x}, \xi) = \Lambda(\mathbf{x})\xi$, $\beta(s) = s$, ζ non-decreasing:
Stefan's model of melting material (\bar{u} = enthalpy, $\zeta(\bar{u})$ = temperature),
- $\mathbf{a}(\mathbf{x}, \xi) = \Lambda(\mathbf{x})\xi$, $\zeta(s) = s$, β non-decreasing:
Richards' equation of underground water flow (\bar{u} = pressure, $\beta(\bar{u})$ = water content).

Application of the GDM

Weak formulation: find \bar{u} in the proper space s.t.

$$\int_0^T \langle \partial_t \beta(\bar{u}), v \rangle + \int_0^T \int_{\Omega} a(\mathbf{x}, \nabla \zeta(\bar{u})) \cdot \nabla v = \int_0^T \int_{\Omega} f v, \quad \forall v.$$

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Consider \mathcal{D} , time steps $0 = t_1 < t_2 < \dots < t_N = T$, and interpolator $I_{\mathcal{D}} : L^2(\Omega) \rightarrow X_{\mathcal{D},0}$:

Find $u_{\mathcal{D}} = (u^n)_{n=0,\dots,N} \in X_{\mathcal{D},0}^{N+1}$ s.t. $u^0 = I_{\mathcal{D}} u_{\text{ini}}$ and, for all $n = 0, \dots, N-1$ and all $v_{\mathcal{D}} \in X_{\mathcal{D},0}$,

$$\begin{aligned} \int_{\Omega} \delta_{\mathcal{D}}^{n+\frac{1}{2}} \beta(u_{\mathcal{D}}) \Pi_{\mathcal{D}} v_{\mathcal{D}} + \int_{\Omega} a(\mathbf{x}, \nabla_{\mathcal{D}}[\zeta(u_{\mathcal{D}}^{n+1})]) \cdot \nabla_{\mathcal{D}} v_{\mathcal{D}} \\ = \frac{1}{\delta t} \int_{n\delta t}^{(n+1)\delta t} \int_{\Omega} f \Pi_{\mathcal{D}} v_{\mathcal{D}} \end{aligned}$$

Convergence analysis tools

- ▶ Space–time Kolmogorov with (discrete) varying spaces,
- ▶ Discrete Aubin–Simon theorem with (discrete) varying spaces,
- ▶ Discontinuous weak Ascoli–Arzela.
- ▶ Discrete “compensated compactness” theorem (convergence of $\int f_n g_n$ from time-derivatives estimates on f_n and space-derivative estimates on g_n).

All these results already fully developed for GD, can be used off-the-shelf.

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A few of them

- ▶ Unified analysis framework for host of methods (no need to re-invent the wheel).
- ▶ Proper generic definition and treatment of mass-lumping.
- ▶ Generic treatment of barycentric elimination of DOFs.
- ▶ Development of generic discrete functional analysis results, for the GDM and more.
- ▶ Uniform-in-time strong $L^2(\Omega)$ convergence results for degenerate parabolic equations, without regularity assumptions.

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- ▶ **Super-convergence for TPFA finite volumes.**

Improved L^2 estimate for GS

$$\begin{aligned} -\operatorname{div}(A\nabla\bar{u}) &= f \text{ in } \Omega, & \bar{u} &= 0 \text{ on } \partial\Omega \\ -\operatorname{div}(A\nabla\bar{w}) &= \frac{\Pi_{\mathcal{D}}u - \bar{u}}{\|\Pi_{\mathcal{D}}u - \bar{u}\|_2} \text{ in } \Omega, & \bar{w} &= 0 \text{ on } \partial\Omega. \end{aligned}$$

Theorem (Improved L^2 estimate for GS)

For all $P_{\mathcal{D}}\bar{u} \in X_{\mathcal{D},0}$,

$$\begin{aligned} \|\Pi_{\mathcal{D}}u - \bar{u}\|_2 &\lesssim [h^{-1}I_h(\bar{u}, P_{\mathcal{D}}\bar{u})]^2 + [S_{\mathcal{D}}(\bar{u}) + W_{\mathcal{D}}(A\nabla\bar{u})]^2 \\ &\quad + I_h(\bar{u}, P_{\mathcal{D}}\bar{u}) + |\widetilde{W}_{\mathcal{D}}(A\nabla\bar{u}, P_{\mathcal{D}}\bar{w})| + \text{symmetric in } \bar{w} \end{aligned}$$

where

$$I_h(\phi, P_{\mathcal{D}}\phi) = \|\Pi_{\mathcal{D}}(P_{\mathcal{D}}\phi) - \phi\|_2 + h\|\nabla_{\mathcal{D}}(P_{\mathcal{D}}\phi) - \nabla\phi\|_2,$$

$$\widetilde{W}(\psi, P_{\mathcal{D}}\phi) = \int_{\Omega} \nabla_{\mathcal{D}}(P_{\mathcal{D}}\phi) \cdot \psi + \Pi_{\mathcal{D}}(P_{\mathcal{D}}\phi)\operatorname{div}\psi.$$

Application to Hybrid Mimetic Mixed schemes

- ▶ HMM = family of schemes gathering Hybrid FV, hybrid Mimetic Finite Differences, and Mixed FV.
- ▶ HMM is a GDM, based on a polytopal mesh \mathfrak{T} and such that

$$\begin{aligned} X_{\mathcal{D},0} &= X_{\mathfrak{T},0}, & \Pi_{\mathcal{D}}v &= \Pi_{\mathfrak{T}}v \text{ (piecewise constant) ,} \\ \nabla_{\mathcal{D}}v &= \overline{\nabla}_{\mathfrak{T}}v + \text{ stabilisation.} \end{aligned}$$

Theorem (Super-convergence for HMM)

Assume optimal H^2 regularity for the PDE. If the cell “centers” \mathcal{P} are, on average on local patches of cells, close to the centers of mass of the cells, then, for $u_{\mathcal{D}}$ solution to HMM,

$$\|\Pi_{\mathcal{D}}u_{\mathcal{D}} - \bar{u}\|_2 = \mathcal{O}(h^2\|f\|_{H^1}).$$

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- ▶ Starts by a modified HMM scheme that *always* super-converges.

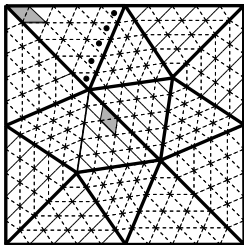
$$\forall \mathbf{x} \in K, \Pi_{\mathcal{D}}^*v(\mathbf{x}) = \Pi_{\mathcal{D}}v(\mathbf{x}) + \overline{\nabla}_{\mathcal{D}}v(\mathbf{x}) \cdot (\mathbf{x} - \mathbf{x}_K).$$

Super-convergence for TPFA

- ▶ On triangular meshes with $A(\mathbf{x}) = a(\mathbf{x})\text{Id}$, TPFA is an HMM, with $\mathbf{x}_K = \text{circumcenter of } K$.

Super-convergence for TPFA

- ▶ On triangular meshes with $A(\mathbf{x}) = a(\mathbf{x})\text{Id}$, TPFA is an HMM, with $\mathbf{x}_K = \text{circumcenter of } K$.
- ▶ For TPFA on triangular meshes as used in benchmarks, local compensation always occurs (up to a small portion).



\mathcal{T}_0	$\partial\mathcal{T}$	\mathcal{T}_0	$\partial\mathcal{T}$	\mathcal{T}_0^\bullet
\perp^0	\perp^0	\perp^0	\perp^0	\perp^0
\mathcal{T}_0	$\partial\mathcal{T}$	\mathcal{T}_0	$\partial\mathcal{T}$	\mathcal{T}_0^\bullet
\perp^0	\perp^0	\perp^0	\perp^0	\perp^0
\mathcal{T}_0^\bullet	$\partial\mathcal{T}^\bullet$	\mathcal{T}_0^\bullet	$\partial\mathcal{T}^\bullet$	\mathcal{T}_0^\bullet

\mathcal{T}_0	\mathcal{T}_0	\mathcal{T}_0	\mathcal{T}_0	\mathcal{T}_0
\mathcal{T}_0	\mathcal{T}_0	\mathcal{T}_0	\mathcal{T}_0	\mathcal{T}_0
\mathcal{T}_0	\mathcal{T}_0	\mathcal{T}_0	\mathcal{T}_0	\mathcal{T}_0
\mathcal{T}_0	\mathcal{T}_0	\mathcal{T}_0	\mathcal{T}_0	\mathcal{T}_0
\mathcal{T}_0	\mathcal{T}_0	\mathcal{T}_0	\mathcal{T}_0	\mathcal{T}_0

Super-convergence for TPFA

Theorem (Super-convergence for TPFA on triangles)

In all classical triangular meshes used in benchmarking, with \mathbf{x}_K circumcenter of K , under optimal H^2 regularity,

$$\|u_h - \bar{u}_{\mathcal{P}}\|_2 = \mathcal{O}(h^{2-\varepsilon} \|f\|_{H^1}),$$

where u_h is the solution to the TPFA FV scheme and $\bar{u}_{\mathcal{P}}$ is the piecewise constant function equal to $\bar{u}(\mathbf{x}_K)$ on K .

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Long-standing (20+ years) conjecture, only obtained by first abstracting from the specificities of the schemes.

Conclusion

- ▶ GDM = generic analysis framework for many numerical methods and many diffusion models (linear and non-linear).
- ▶ Easy proof that a given method fits into the GDM.
- ▶ Proof of convergence based on 3-5 properties.
- ▶ Discrete functional analysis results readily usable in the GDM.
- ▶ Led to novel results, thanks to a level of abstraction that liberates from the specificities of each scheme.

Main reference

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Thanks.