

# An $L^{\infty}(H^1)$ – error estimate for GSs applied to time fractional diffusion equations F. Benkhaldoun <sup>a</sup> and A. Bradji <sup>b,c</sup>



(2)

# Problem to be solved

Time Fractional Diffusion Equation:

 $\partial_t^{\alpha} u(\mathbf{x},t) - \nabla \cdot (\kappa \nabla u)(\mathbf{x},t) = f(\mathbf{x},t), \qquad (\mathbf{x},t) \in \Omega \times (0,T),$ (1)

where

 $\triangleright \Omega$  is an open polyhedral bounded subset in  $\mathbb{R}^d$ ,

 $ightarrow T > 0, 0 < \alpha < 1$  are given.

▶  $\kappa$  satisfies  $\kappa \in C^1(\overline{\Omega})$  and  $\kappa(\mathbf{x}) > \kappa_0 > 0$ . ► *f* is a given smooth function.

**Properties of the approximation of the Caputo derivative** 

Decreasing property:

$$\lambda_{j+1}^{n+1} > \lambda_j^{n+1} > \lambda_0 = \frac{1}{2T^{\alpha}\Gamma(1-\alpha)} \quad \text{and} \quad \sum_{j=0}^n k\lambda_j^{n+1} \le \frac{T^{1-\alpha}}{\Gamma(2-\alpha)}$$
(8)

Stability: For all 
$$(\beta^j)_{j=0}^{N+1} \in \mathbb{R}^{N+2}$$
 and for all  $n \in [0, N]$ :

The operator  $\partial_t^{\alpha}$  is the Caputo derivative defined by:

$$\partial_t^{\alpha} u(t) = rac{1}{\Gamma(1-\alpha)} \int_0^t (t-s)^{-\alpha} u_t(s) ds.$$

Initial condition is given by

$$u(\mathbf{X},0) = 0, \qquad \mathbf{X} \in \Omega.$$
 (3)

Homogeneous Dirichlet boundary conditions are given by

$$u(\mathbf{x},t) = 0, \qquad (\mathbf{x},t) \in \partial \Omega \times (0,T).$$
 (4)

## Application...

Fractional differential equations have been successfully used in theory and they appear in many areas of application, see [1].

## **Discretization in space: GDM framework introduced in [8]**

Let  $\Omega$  be an open domain of  $\mathbb{R}^d$ , with  $d \in \mathbb{N}^*$ . An approximate gradient discretization is given by  $\mathcal{D} = (X_{\mathcal{D},0}, \Pi_{\mathcal{D}}, \nabla_{\mathcal{D}})$ , where

- **1.** The set of discrete unknowns  $X_{\mathcal{D},0}$  is a finite dimensional vector space on  $\mathbb{R}$ .
- **2.** The linear mapping  $\Pi_{\mathcal{D}}: X_{\mathcal{D},0} \to L^2(\Omega)$  is the reconstruction of the approximate function.
- **3.** The gradient reconstruction  $\nabla_{\mathcal{D}} : X_{\mathcal{D},0} \to L^2(\Omega)^d$  is a linear mapping

$$\left(\sigma\beta^{n+1} + (1-\sigma)\beta^{n}\right)\sum_{j=0}\lambda_{j}^{n+1}(\beta^{j+1} - \beta^{j}) \ge \frac{1}{2}\sum_{j=0}\lambda_{j}^{n+1}\left((\beta^{j+1})^{2} - (\beta^{j})^{2}\right).$$
(9)

## **Formulation of a Gradient Scheme**

$$\sum_{j=0}^{n} k \lambda_{j}^{n+1} \left( \partial^{1} u_{\mathcal{D}}^{j+1}, \mathbf{v} \right)_{L^{2}(\Omega)} + \left( \nabla_{\mathcal{D}} u_{\mathcal{D}}^{n+\sigma}, \nabla_{\mathcal{D}} \mathbf{v} \right)_{L^{2}(\Omega)} = (f(t_{n+\sigma}), \mathbf{v})_{L^{2}(\Omega)}, \quad (10)$$

where  $u_{\mathcal{D}}^{0}$  is defined using a discrete projection.

New  $\mathbb{L}^{\infty}(H_0^1)$ –Error estimate for the Scheme (10)

- $\triangleright$  Error estimate in the discrete norm of  $\mathbb{L}^{\infty}(H_0^1)$ 
  - $\max_{n=N+1}^{n=N+1} \|\nabla u(t_n) \nabla_{\mathcal{D}} u_{\mathcal{D}}^n\|_{\mathbb{L}^2(\Omega)} \leq C(k^2 + \mathbb{E}_{\mathcal{D}}^k(u)),$ (11)

where  $\mathbb{E}_{\mathcal{D}}^{k}(u) = \max_{j \in \{0,1\}} \max_{n \in [j,N+1]} \mathbb{E}_{\mathcal{D}}(\partial^{j}u(t_{n}))$  and

 $\mathbb{E}_{\mathcal{D}}(\overline{u}) = \max\left(C_{\mathcal{D}}W_{\mathcal{D}}(\kappa\nabla\overline{u}) + (C_{\mathcal{D}}+1)S_{\mathcal{D}}(\overline{u}), W_{\mathcal{D}}(\kappa\nabla\overline{u}) + 2S_{\mathcal{D}}(\overline{u})\right).$ (12)

- This results improves the one of [5] which dealt with only  $\mathbb{L}^{\infty}(L^2)$ -Error estimate.
- ► The convergence order in time of the GS (10) is better than the one of [3] which dealt with a SUSHI scheme of only order  $k^{2-\alpha}$  in time.

which reconstructs, from an element of  $X_{\mathcal{D},0}$ , a "gradient" (vector-valued function) over  $\Omega$ . The gradient reconstruction must be chosen such that  $\|\nabla_{\mathcal{D}} \cdot \|_{L^2(\Omega)^d}$  is a norm on  $X_{\mathcal{D},0}$ . Let us define the bi-linear form  $\langle \cdot, \cdot \rangle_{\mathcal{D},\kappa}$ given by

$$\langle u, v \rangle_{\mathcal{D},\kappa} = \int_{\Omega} \kappa(\mathbf{x}) \nabla_{\mathcal{D}} u(\mathbf{x}) \cdot \nabla_{\mathcal{D}} v(\mathbf{x}) d\mathbf{x}, \quad \forall (u, v) \in X_{\mathcal{D},0} \times X_{\mathcal{D},0}.$$
 (5)

In order to analyse the convergence of the gradient schemes, we consider the following parameters related to the mesh  $\mathcal{D}$ :

**1.** The coercivity of the discretization is measured via the constant  $C_{\mathcal{D}}$ given by

$$C_{\mathcal{D}} = \max_{\boldsymbol{v} \in X_{\mathcal{D},0} \setminus \{\boldsymbol{0}\}} \frac{\| \Pi_{\mathcal{D}} \boldsymbol{v} \|_{L^{2}(\Omega)}}{\| \nabla_{\mathcal{D}} \boldsymbol{v} \|_{L^{2}(\Omega)^{d}}}.$$

This yields the Poincaré inequality:  $\|\Pi_{\mathcal{D}} v\|_{L^2(\Omega)} \leq C_{\mathcal{D}} \|\nabla_{\mathcal{D}} v\|_{L^2(\Omega)^d}$ . 2. The consistency is measured through the interpolation error function  $S_{\mathcal{D}}: H_0^1(\Omega) \to [0, +\infty)$  defined by, for all  $\varphi \in H_0^1(\Omega)$ 

$$S_{\mathcal{D}}(\varphi) = \min_{\boldsymbol{v} \in X_{\mathcal{D},0}} \left( \|\Pi_{\mathcal{D}}\boldsymbol{v} - \varphi\|_{L^{2}(\Omega)}^{2} + \|\nabla_{\mathcal{D}}\boldsymbol{v} - \nabla\varphi\|_{L^{2}(\Omega)^{d}}^{2} \right)^{\frac{1}{2}}.$$

**3.** The limit-conformity is measured through the conformity error function  $W_{\mathcal{D}}: H_{\text{div}}(\Omega) \to [0, +\infty)$  defined by, for all  $\varphi \in H_{\text{div}}(\Omega)$ 

$$W_{\mathcal{D}}(\varphi) = \max_{u \in X_{\mathcal{D},0} \setminus \{0\}} \frac{1}{\|\nabla_{\mathcal{D}} u\|_{L^2(\Omega)^d}} \left| \int_{\Omega} \left( \nabla_{\mathcal{D}} u(\boldsymbol{x}) \cdot \varphi(\boldsymbol{x}) + \Pi_{\mathcal{D}} u(\boldsymbol{x}) \operatorname{div}\varphi(\boldsymbol{x}) \right) d\boldsymbol{x} \right|.$$

# Additional assumption on $\|\Pi_{\mathcal{D}} \cdot \|_{L^{2}(\Omega)}$

#### Main ideas on the proof

 $\blacktriangleright$  We introduce an approximation for the operator  $-\nabla(\kappa \nabla \cdot)$  similar to the discrete Laplace operator introduced in [8].

# A well-developed discrete a priori estimate...

## In Progress

Extension to GDM without the stated assumption on  $\|\Pi_{\mathcal{D}} \cdot \|_{L^2(\Omega)}$ . This work is the subject of [2].

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We assume, in addition, that the generic mesh  $\mathcal{D} = (X_{\mathcal{D},0}, \Pi_{\mathcal{D}}, \nabla_{\mathcal{D}})$  is chosen such that  $\|\Pi_{\mathcal{D}} \cdot \|_{L^{2}(\Omega)}$  is a norm on  $X_{\mathcal{D},0}$ . This includes, for instance, Conforming FEMs, Cell-Centered SUSHI, and MFEMs.

## Time discretization and discrete temporal derivative

The discretization of [0, T] is performed with a constant time step where  $N \in \mathbb{N}^*$ . Mesh points in time:  $t_n = nk$ ,  $\forall n \in [0, N + 1]$ . The discrete temporal derivative given by  $\partial^1 v^n = \frac{v^n - v^{n-1}}{v}$ .

## **Approximation of the Caputo derivative, see [1]**

$$\partial_t^{\alpha} u(t_{n+\sigma}) = \sum_{j=0}^n k \lambda_j^{n+1} \partial^1 u(t_{j+1}) + \mathbb{T}_1^{n+1}(u),$$
(6)  
where  $\sigma = 1 - \alpha/2, \, \lambda_j^{n+1}$  are developed in [1], and  
 $|\mathbb{T}_1^{n+1}(u)| \le Ck^{3-\alpha}.$ 
(7)

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