

New gradient scheme of a time fractional Fokker-Planck equation with time independent forcing and its convergence analysis Abdallah Bradji

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Original Problem: Fokker-Planck Equation

The time fractional Fokker–Planck equation can be written as follows: $\partial_t u(\mathbf{x},t) - \nabla \cdot \left({}_{\mathrm{RL}} \partial_t^{1-\alpha} \kappa_\alpha \nabla u(\mathbf{x},t) - \mathbf{F}(\mathbf{x},t) {}_{\mathrm{RL}} \partial_t^{1-\alpha} u(\mathbf{x},t) \right) = g(\mathbf{x},t),$ (1)

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

> 7 > 0, 0 < α < 1, κ_{α} > 0

- \blacktriangleright F (the Force) and g are given functions.
- The operator $_{RL}\partial_t^{1-\alpha}u(t)$ is the Riemann–Liouville derivative defined by $\partial_t (\partial_t^{-\alpha} u(t))$ with $\partial_t^{-\alpha} u(t)$ is the fractional integral operator:

Approximation of the Caputo derivative

 $\partial_t^{\alpha} u(t_{n+1}) = \sum_{i=0}^n k \lambda_j^{n+1} \partial^1 u(t_{j+1}) + \mathbb{T}_1^{n+1}(u),$

where

$$\lambda_j^{n+1} = \frac{(n-j+1)^{1-\alpha} - (n-j)^{1-\alpha}}{k^{\alpha}\Gamma(2-\alpha)}$$

and

 $|\mathbb{T}_1^{n+1}(u)| \leq Ck^{2-\alpha}.$

(8)

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(14)

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

Initial condition is given by

$$u({oldsymbol x},0)=0, \qquad {oldsymbol x}\in \Omega.$$

Homogeneous Dirichlet boundary conditions are given by

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

Some Physics and others about the problem

- The Fokker–Planck equations describe:
- **1.** The time evolution of the probability density function of the position. **2.**The velocity of a particle.
- \blacktriangleright When α tends to one and the diffusion coefficient is constant, we get the standard Fokker–Planck equation.

Simple version for the Fokker-Planck equation

We assume that the driving force **F** is independent of time, $\mathbf{F} = \mathbf{F}(\mathbf{x})$. The equation can be written then as:

 $\partial_t u - {}_{\mathrm{RL}} \partial_t^{1-\alpha} \nabla \cdot (\kappa_\alpha \nabla u - \mathbf{F} u) = g.$

By acting the operator $\partial_t^{\alpha-1}$ on the both sides of the last equation yields

Properties of the approximation of the Caputo derivative

$$\frac{k^{-\alpha}}{\Gamma(2-\alpha)} = \lambda_n^{n+1} > \ldots > \lambda_0^{n+1} \ge \lambda_0 = \frac{T^{-\alpha}}{\Gamma(1-\alpha)}.$$

Formulation of a GS: First attempt

$$\sum_{j=0}^{n} k \lambda_{j}^{n+1} \left(\partial^{1} \Pi_{\mathcal{D}} u_{\mathcal{D}}^{j+1}, \Pi_{\mathcal{D}} v \right)_{L^{2}(\Omega)} + \kappa_{\alpha} \left(\nabla_{\mathcal{D}} u_{\mathcal{D}}^{n+1}, \nabla_{\mathcal{D}} v \right)_{L^{2}(\Omega)^{d}} - \left(\mathbf{F} \Pi_{\mathcal{D}} u_{\mathcal{D}}^{n+1}, \nabla_{\mathcal{D}} v \right)_{L^{2}(\Omega)^{d}} = (f(t_{n+1}), \Pi_{\mathcal{D}} v)_{L^{2}(\Omega)},$$
(11)

where $u_{\mathcal{D}}^0 = 0$.

Disadvantage: No discrete Coercivity.

Formulation of a GS: Second attempt-The right choice, see [1]



$$\partial_t^{\alpha} u - \nabla \cdot (\kappa_{\alpha} \nabla u - \mathbf{F} u) = f, \qquad (5)$$

where $f = \partial_t^{\alpha-1} g$ and $\partial_t^{\alpha} u$ is the Caputo derivative of order α given by
 $\partial_t^{\alpha} u(t) = \frac{1}{\Gamma(1-\alpha)} \int_0^t (t-s)^{-\alpha} u_t(s) ds. \qquad (6)$

Definition of an approximate gradient discretization, cf. [2]

Approximate gradient discretization \mathcal{D} is defined by

 $\mathcal{D} = (X_{\mathcal{D},0}, \Pi_{\mathcal{D}}, \nabla_{\mathcal{D}})$

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 which reconstructs, from an element of $X_{D,0}$, a "gradient" (vector-valued function) over Ω . The gradient reconstruction must be chosen such that

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

Parameters of an approximate gradient discretization \mathcal{D}

1. Coercivity. Poincaré inequality:

$$+ \frac{1}{2} \left(\mathbf{F} \cdot \nabla_{\mathcal{D}} u_{\mathcal{D}}^{n+1}, \Pi_{\mathcal{D}} v \right)_{L^{2}(\Omega)} - \frac{1}{2} \left(\mathbf{F} \Pi_{\mathcal{D}} u_{\mathcal{D}}^{n+1}, \nabla_{\mathcal{D}} v \right)_{L^{2}(\Omega)^{d}} + \frac{1}{2} \left(\operatorname{div}(\mathbf{F}) \Pi_{\mathcal{D}} u_{\mathcal{D}}^{n+1}, \Pi_{\mathcal{D}} v \right)_{L^{2}(\Omega)} = (f(t_{n+1}), \Pi_{\mathcal{D}} v)_{L^{2}(\Omega)},$$

$$\text{ where } u_{\mathcal{D}}^{0} = 0.$$

$$(12)$$

$\mathbb{L}^{\infty}(L^2)$ and $\mathbb{L}^2(H_0^1)$ –Error estimates for the GS (12)

$$\sum_{n=0}^{N+1} \|\Pi_{\mathcal{D}} u_{\mathcal{D}}^{n} - u(t_{n})\|_{L^{2}(\Omega)} + \left(\sum_{n=0}^{N} k \|\nabla_{\mathcal{D}} u_{\mathcal{D}}^{n} - \nabla u(t_{n})\|_{L^{2}(\Omega)^{d}}^{2}\right)^{\frac{1}{2}} \leq C(1+C_{\mathcal{D}}) \left(\mathbb{E}_{\mathcal{D}}^{k}(u) + k^{2-\alpha}\right), \quad (13)$$

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

$$\mathbb{E}_{\mathcal{D}}(\overline{u}) = (1 + C_{\mathcal{D}}) \left(W_{\mathcal{D}}(\nabla \overline{u}) + W_{\mathcal{D}}(\mathbf{F}\overline{u}) \right) + (1 + C_{\mathcal{D}} + C_{\mathcal{D}}^2) S_{\mathcal{D}}(\overline{u}).$$
(15)

Main idea on the proof

A well-developed discrete a priori estimate.

 $\|\Pi_{\mathcal{D}} v\|_{L^{2}(\Omega)} \leq C_{\mathcal{D}} \|\nabla_{\mathcal{D}} v\|_{L^{2}(\Omega)^{d}}, \quad \forall v \in X_{\mathcal{D},0}.$

2. Strong consistency.

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

3. Dual consistency

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

Time discretization and discrete temporal derivative

The discretization of [0, T] is performed with a constant time step

$$k=rac{T}{N+1},$$

where $N \in \mathbb{N}^{\star}$

$$t_n = nk, \forall n \in \llbracket 0, N + 1
bracket.$$

The discrete temporal derivative given by

$$\partial^1 v^n = rac{v^n - v^{n-1}}{k}.$$

In Progress

(7)

The force **F** is dependent on time, i.e. $\mathbf{F} = \mathbf{F}(\mathbf{x}, t)$.

Second order time accurate.

References

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