



Efficiency Assessment of Fractional Models in R

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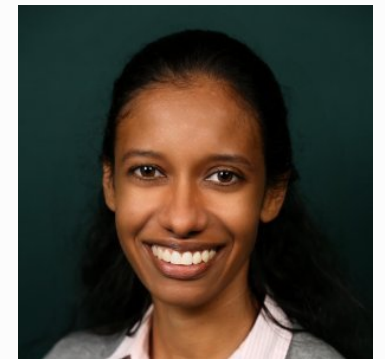
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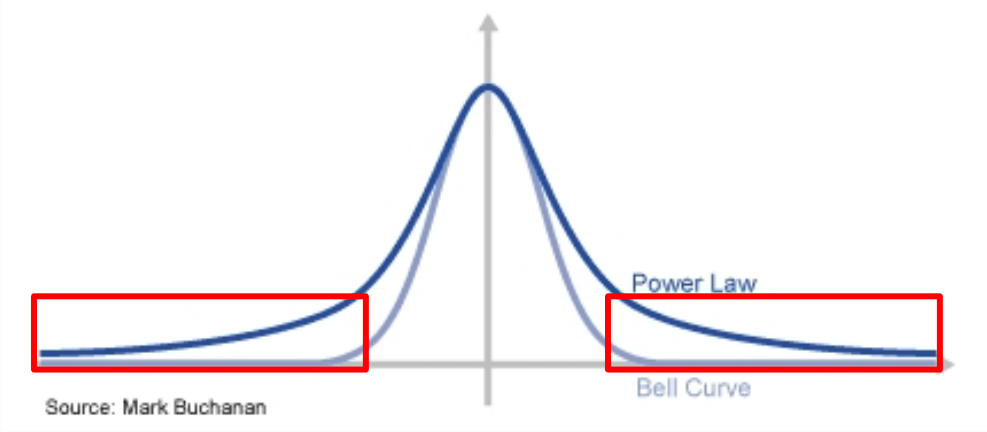
Camille Archer



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Fractional Models in Nature



Non-standard CDF with power-law tails

Source: Mark Buchanan

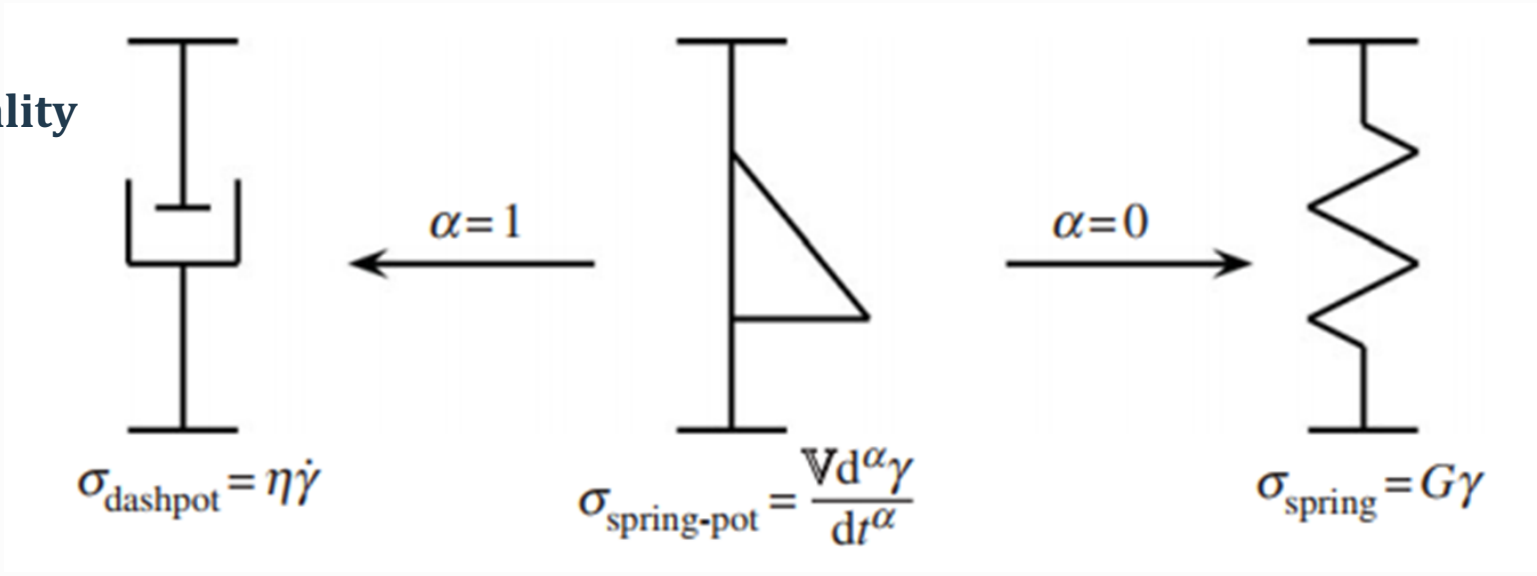
Fractional Derivative

$${}^{\text{RL}}\mathcal{D}_t^\nu u(t) = \frac{1}{\Gamma(1-\nu)} \frac{d}{dt} \int_0^t (t-s)^{-\nu} u(s) ds \quad (1)$$

Kernel captures power law

Integration captures history/non-locality

Interpolation operator



Methods in R

Projection (p-refinement)

$${}^{\text{RL}}\mathcal{D}_t^\nu u(t) = f(t), u(0) = 0$$

1. Expansion into P solution functions

$$u(t) = \sum_{p=1}^P \hat{u}_p \phi_p(t) = \sum_{p=1}^P u_p(t)$$

2. Project problem onto test functions

$$\int_0^T {}^{\text{RL}}\mathcal{D}_t^\nu [u_p(t)] v_j(t) dt = \int_0^T f(t) v_j(t) dt$$

3. Solve for coefficients

$$\mathbf{S}\vec{u} = \mathbf{F}$$

Discretization (h-refinement)

$${}^{\text{RL}}\mathcal{D}_t^\nu u(t) = f(t), u(0) = 0$$

1. Define discretization scheme

$${}^{\text{RL}}\mathcal{D}_t^\nu u(t) \cong \frac{\sum_{n=0}^{T/h} (-1)^n \binom{\alpha}{n} u(t - nh)}{h^\nu}$$

2. Define initial condition

$$u(0) = 0$$

3. Iterate forward

Petrov-Galerkin Spectral Method

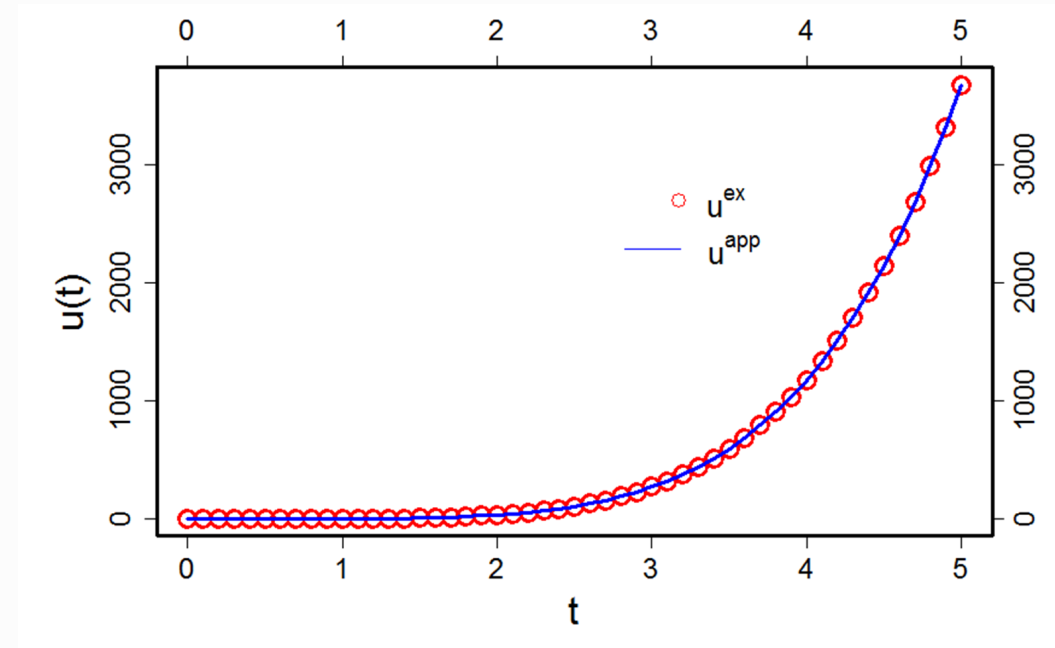
p-refinement method

$$u(t) = \sum_{p=1}^P \hat{u}_p (1 + \xi(t))^{\frac{\nu}{2}} P_{p-1}^{-\frac{\nu}{2}, \frac{\nu}{2}}(\xi(t))$$

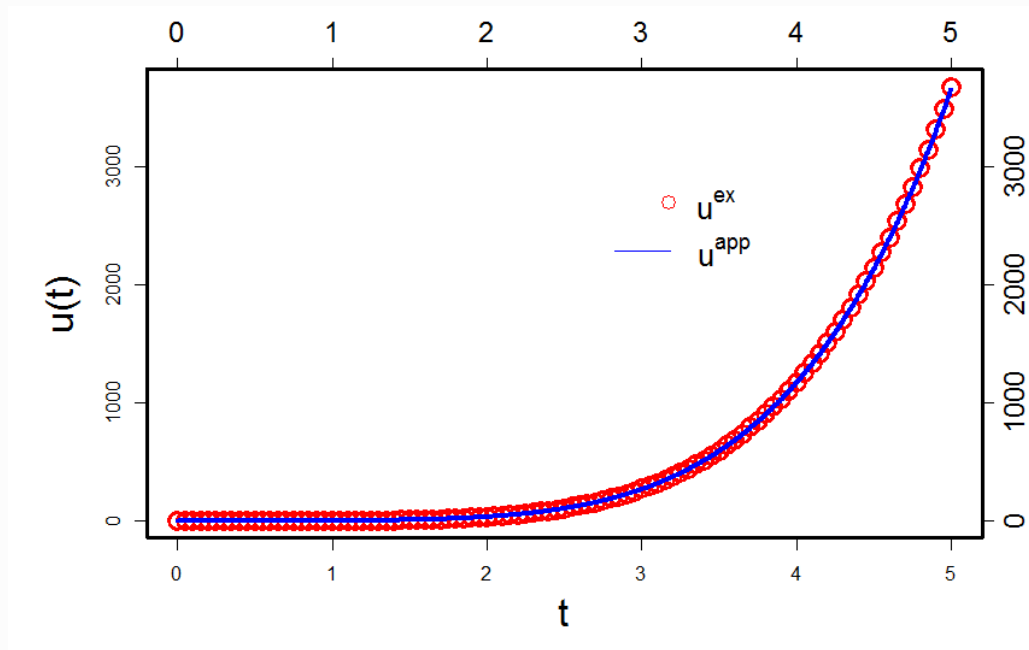
$$v_j(t) = (1 - \xi(t))^{\frac{\nu}{2}} P_{j-1}^{\frac{\nu}{2}, -\frac{\nu}{2}}(\xi(t))$$

$$\mathbf{S}\vec{u} = \mathbf{F}$$

\mathbf{S} is diagonal



Finite Difference Method

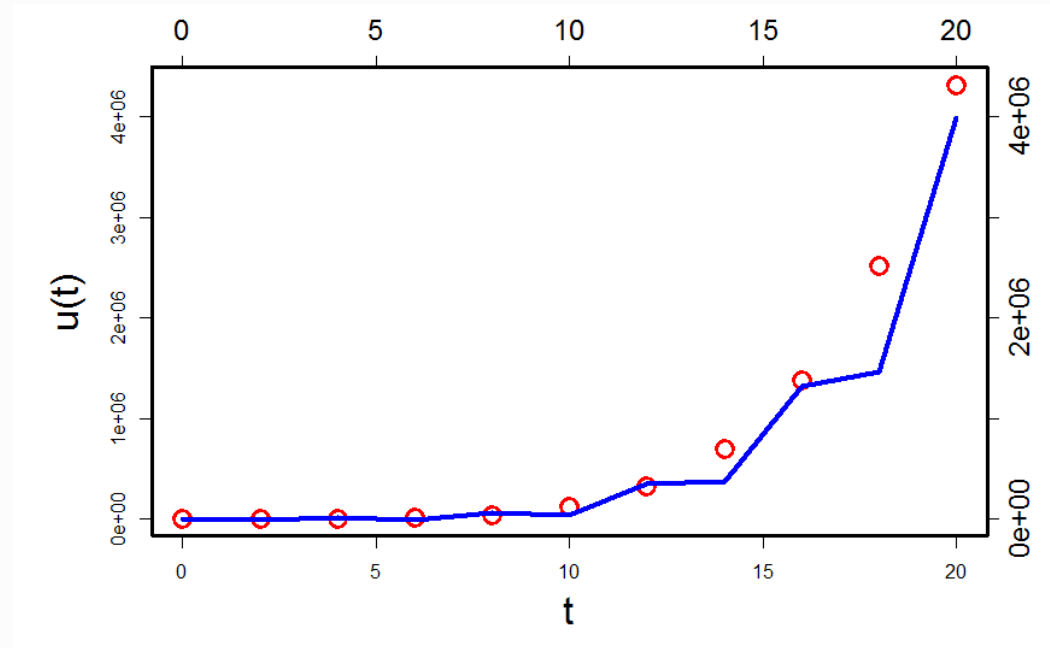


h-refinement method

$$\begin{aligned}
 &u(t_{k+1}) \\
 &= (1 - b_1)u(t_k) + \sum_{j=1}^{k-1} (b_j - b_{j+1})u(t_{k-j}) \\
 &+ b_k u(t_0) + \Delta t^\nu \Gamma(2 - \nu) f(t_{k+1})
 \end{aligned}$$

Simple iteration

Finite Element Method



hp-refinement method

$$\sum_{e=1}^{\epsilon-1} \sum_{p=1}^P \vec{u}_p^e H_{\epsilon,e,p} + \sum_{p=1}^P \vec{u}_p^\epsilon M_{\epsilon,p} = F_\epsilon$$

$$M_G \vec{u}_G = F_G$$

Originally for elliptical problems

Convergence issues

Comparison

Pros and cons for various methods

Pros

Cons

PGSM

- Captures singular & smooth behavior
- High, fast accuracy

- Hard to implement
- Limited to smooth functions

FDM

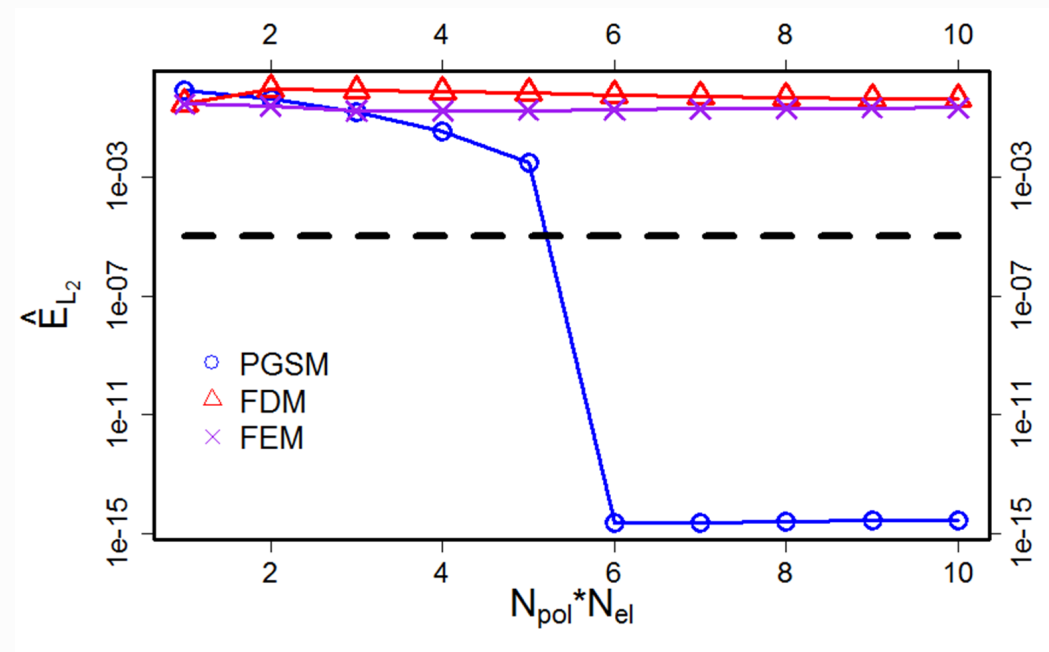
- Generalizable
- Easy to Implement

- Large history iteration
- High accuracy unfeasible

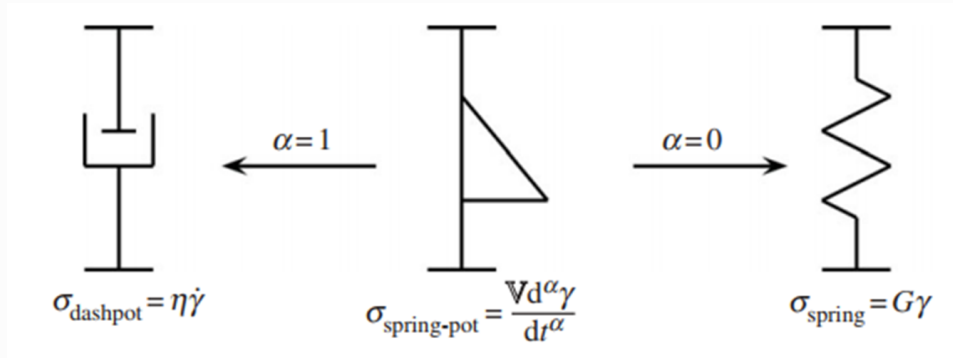
FEM

- Adaptive to irregular domains or singularities

- Large history matrix

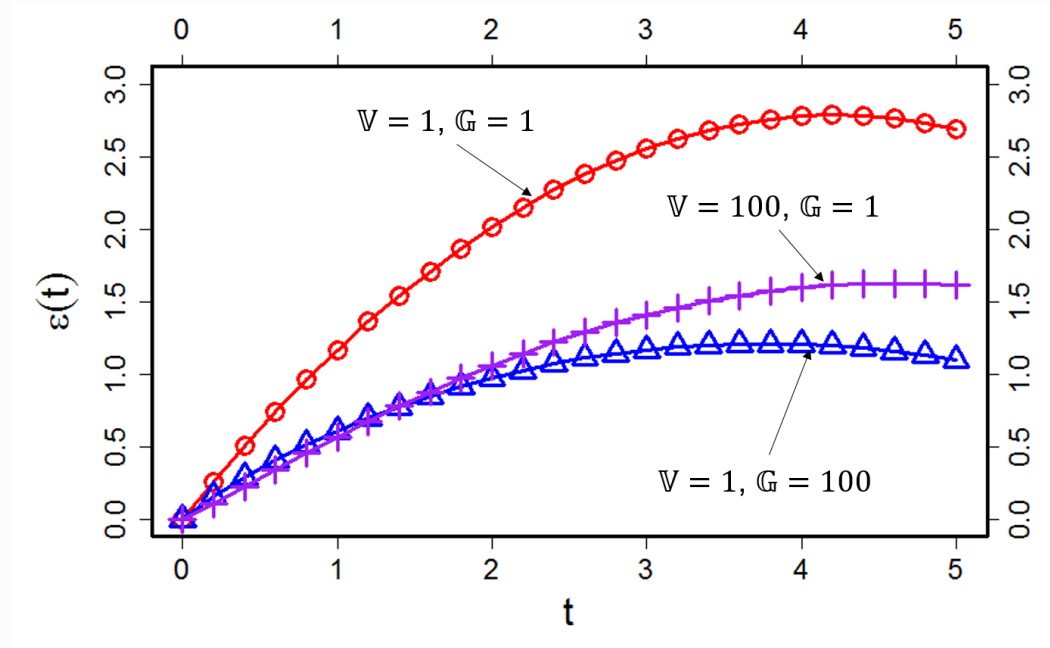


Fractional Maxwell Application



$${}^{\text{RL}}\mathcal{D}_t^\nu \epsilon(t) = \frac{1}{V} \sigma(t) + \frac{1}{G} {}^{\text{RL}}\mathcal{D}_t^{\nu-\mu} \sigma(t)$$

Resembles ${}^{\text{RL}}\mathcal{D}_t^\nu u(t) = f(t)$



Discussion

Selection of method depends on application

- Data with power-law singularity/models → PGSM
- Non-intuitive data/models → FDM
- Fragmented data/domains → FEM

Key features for various methods

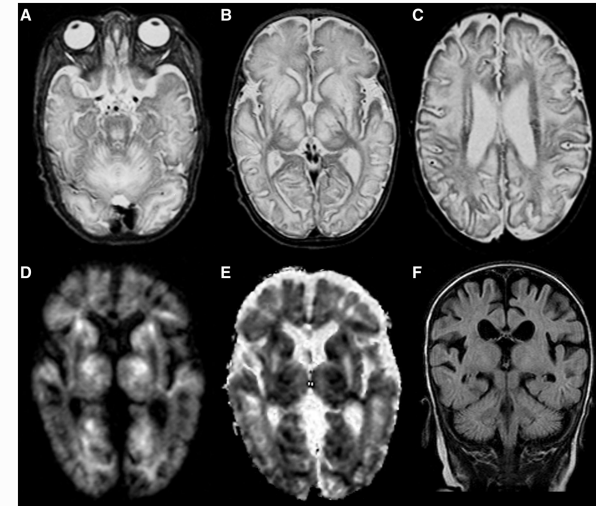
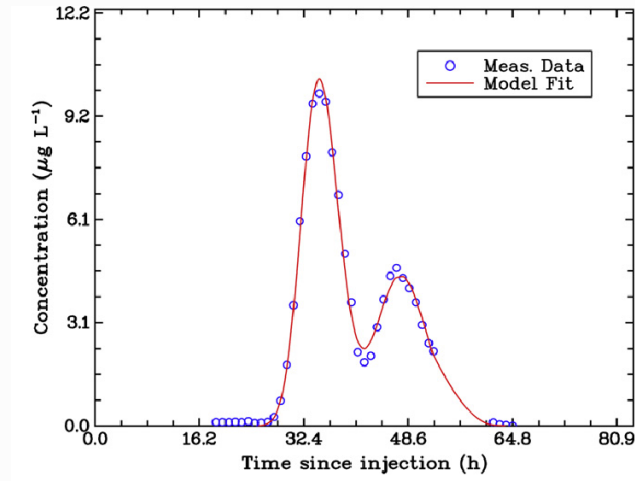
Method	Key Feature
PGSM	Diagonal linear system
FDM	Generalizable form
FEM	Adaptive to irregularities

Future Work

- Extension to fractional advection-dispersion equation

$${}_0\mathcal{D}_t^{2\tau}u + \sum_{i=1}^d \left[(1 - \theta_i)_{a_i} \mathcal{D}_{x_i}^{2\mu_i} + (\theta_i)_{x_i} \mathcal{D}_{b_i}^{2\mu_i} \right] u = \sum_{i=1}^d \left[(1 - \phi_i)_{a_i} \mathcal{D}_{x_i}^{2\nu_i} + (\phi_i)_{x_i} \mathcal{D}_{b_i}^{2\nu_i} \right] u - \gamma u + f$$

- Inclusion of error estimates against observable data
- Take advantage of R platform for observable data comparisons
- Fine tune CPU time analysis



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