## Total Fractional-Order Variation Model

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This review is based on the paper 'A Total Fractional-Order Variation Model for Image Restoration with Non-homogeneous Boundary Conditions and its Numerical Solution' by Jianping Zhang, Ke Chen (2015).

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# **Outline**

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# A First Look

### $\blacktriangleright$  Total  $\alpha$ –order variation model:

$$
\min_{u \in BV^{\alpha}(\Omega)} \{ E(u) := TV^{\alpha}(u) + \frac{\lambda}{2} F(u) \}, \ F(u) = \int_{\Omega} |u - z|^2 dx.
$$

Where

$$
TV^{\alpha}(u):=\sup_{\phi\in K}\int_{\Omega}(-udiv^{\alpha}\phi)dx, \text{ with } div^{\alpha}\phi=\sum_{i=1}^d\frac{\partial\phi_i}{\partial x_i^{\alpha}}
$$

 $\blacktriangleright$  When  $u \in W_1^{\alpha}(\Omega)$ ,  $TV^{\alpha}(u) = \int_{\Omega} |\nabla^{\alpha} u| dx$ .

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## Fractional Derivative

### Theorem

The Abels integral equation, with,

$$
\frac{1}{\Gamma(\alpha)} \int_0^x \frac{\psi(\tau)}{(x-\tau)^{1-\alpha}} d\tau = f(x), \ x > 0
$$

has the solution given by the formula

$$
\psi(x) = \frac{1}{\tau(1-\alpha)} \frac{d}{dx} \int_0^x \frac{f(\tau)}{(x-\tau)} d\tau, \ x > 0.
$$

With the help of this theorem, we can then define the left, right sided and the central Riemann-Liouville (R-L) derivative:

$$
D_{[a,x]}^{\alpha} f(x) = \frac{1}{\Gamma(n-\alpha)} \left(\frac{d}{dx}\right)^n \int_a^x \frac{f(\tau)}{(x-\tau)^{\alpha-n+1}}
$$

$$
D_{[x,b]}^{\alpha} f(x) = \frac{(-1)^n}{\Gamma(n-\alpha)} \left(\frac{d}{dx}\right)^n \int_a^x \frac{f(\tau)}{(\tau-x)^{\alpha-n+1}}
$$

$$
D_{[a,b]}^{\alpha} f(x) = \frac{1}{2} \left(D_{[a,x]}^{\alpha} f(x) + (-1)^n D_{[x,b]}^{\alpha} f(x)\right)
$$
<sup>4</sup>

## Other Fractional Derivative Definitions

 $\triangleright$  Grunwald-Letnikov (G-L) left-sided derivative is denoted by:

$$
GD_{[a,x]}^{\alpha} f(x) = \lim_{h \to 0} \frac{1}{h^{\alpha}} \sum_{j=0}^{\left[\frac{x-a}{h}\right]} (-1)^j {\alpha \choose j} f(x - jh),
$$
  
where 
$$
{\alpha \choose j} = \frac{\alpha(\alpha - 1)...(\alpha - j + 1)}{j!},
$$

 $\blacktriangleright$  The third definition is the Caputo order  $\alpha$  derivative defined by

$$
CD_{[a,x]}^{\alpha} f(x) = \frac{1}{\Gamma(n-\alpha)} \int_{a}^{x} \frac{f^{(n)}(\tau)}{(x-\tau)^{\alpha-n+1}}
$$

$$
CD_{[x,b]}^{\alpha} f(x) = \frac{(-1)^{n}}{\Gamma(n-\alpha)} \int_{a}^{x} \frac{f^{(n)}(\tau)}{(\tau-x)^{\alpha-n+1}}
$$

$$
CD_{[a,b]}^{\alpha} f(x) = \frac{1}{2} (CD_{[a,x]}^{\alpha} f(x) + (-1)^{n} CD_{[x,b]}^{\alpha} f(x))
$$

# **Properties**

### $\blacktriangleright$  Linearity.

$$
D_{[a,x]}^{\alpha}(pf(x)+qg(x))=pD_{[a,x]}^{\alpha}f(x)+qD_{[a,x]}^{\alpha}g(x)
$$

- $\blacktriangleright$  Singularity. Assume that $D^{\alpha}_{[a,x]}$  is one of the above three fractional-order derivative operators. For any non-integer  $\alpha > 0$  and  $x > a$ , there exists a non-constant value function  $f(\tau)$  in  $(a, x]$  such that  $D_{[a,x]}^\alpha f(x).$  For example,  $f(\tau)=(x-2\tau)(x-\tau)^\alpha$  will give  $D^\alpha_{[0,x]} \widetilde{f}(x) = 0$  in Abel's inverse transform.
- ▶ Boundary conditions: we need to assume zero Dirichlet boundary condition, otherwise there is a singularity at the end point. However, if we have nonzero Dirichlet boundary conditions, one solution could be to extract off a linear approximation  $g(x)$  (that coincides with f at  $x = a, b$ ) and to consider  $D_{[a,x]}^{\alpha}(f(x) - g(x)).$

# Theoretical Results

We may skip the third part of the paper (3 pages), which clarifies some theoretical/classical results. Basically, it says  $W^\alpha_p$  embedded with  $\|\cdot\|_{W^\alpha_p}$ is a Banach space (this is basic guarantee of any iterative approximation algorithms) and the proposed model has a unique minimizer (convexity of energy functional  $E(u)$ ).

# Boundary Conditions

In order to use deal with non-zero Dirichlet boundary conditions (which is almost always the case in real world), an auxiliary function is introduced.

In 1D case, if  $u(0) = a$ ,  $u(1) = b$ , then introduce  $e(x) = a(1-x) + bx$  and take  $\hat{u}(x) = u(x) - e(x)$ , then we achieve

$$
\hat{u}(0) = \hat{u}(1) = \hat{u}'(0) = \hat{u}'(1) = 0.
$$

 $\triangleright$  In 2D, assuming that the four corners of the solution are given or accurately estimated:

$$
u(0,0) = a, u(0,1) = b, u(1,0) = c, u(1,1) = d.
$$

Similarly we can construct a bilinear function,

 $e_1(x, y) = a + (c - a)x + (b - a)y + (d + a - c - b)xy$ , then  $\hat{u}(x, y) = u(x, y) - e_1(x, y)$  will have zero values on all 4 corners. If boundary conditions  $u(0, y) = a_1(y)$ ,  $u(1, y) = a_2(y)$ ,  $u(x, 0) = b_1(x)$ ,  $u(x, 1) = b_2(x)$  at  $\partial\Omega$  are known as a priori, then we define  $e_2(x,y) = ((1-x)\hat{a}(y) + x\hat{a}(y)) + ((1-y)\hat{b}(x) + y\hat{b}(x)),$ then we achieve

$$
\hat{u}(x,y)|_{\partial\Omega} = 0, \text{ where } \hat{u}(x,y) = u(x,y) - e_1(x,y) - e_2(x,y).
$$

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# **Discretization**

### Key Point: Discrete version of fractional derivative. It turns out the x-direction

$$
D_{[a,b]}^{\alpha} f(x_k, y_l) = \frac{\delta_0^{\alpha} f(x_k, y_l)}{h^{\alpha}} + O(h) = \frac{1}{2} \left( \frac{\delta_-^{\alpha} f(x_k, y_l)}{h^{\alpha}} + \frac{\delta_+^{\alpha} f(x_k, y_l)}{h^{\alpha}} \right) + O(h)
$$
  
= 
$$
\frac{1}{2} (h^{-\alpha} \sum_{j=0}^{k+1} w_j^{\alpha} f_{k-j+1}^l + h^{-\alpha} \sum_{j=0}^{N-k+2} w_j^{\alpha} f_{k+j-1}^l) + O(h)
$$

is applicable to both R-L and C derivatives, where  $f_s^l = f(x_s, y_l)$ ,  $w_j^{(\alpha)} = (-1)^j {\alpha \choose j}$ j  $\bigg), j = 0, 1, ..., N+1.$ 

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## Matrix Version

Denote  $w=w_0^\alpha+w_2^\alpha$ , then we can write  $D_{[a,b]}^\alpha f(x_k,y_l)$  into matrix:



Review the model:

$$
\min_{u,d}\{\int_{\Omega}|d|dx+\frac{\lambda}{2}F(u)\},\text{ s.t. }F(u)=\int_{\Omega}|u-z|^{2}dx,d=\nabla^{\alpha}(u).
$$

and discrete matrix version of TV:

$$
\min_{u} \frac{\lambda}{2} \|u - z\|^2 + \|D_1 u\| + \|D_2 u\|
$$

Now, it's clear, whenever you see  $D$  just replace it with  $B!$ <br>Algorithms [Algorithms](#page-8-0) 11 Discussions?

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