

# Medical Image Analysis and Processing

## Image Noise Filtering

## Anisotropic Diffusion Filter

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Distance/online Course: Session 14 (Episode #1)

Date: 13 April 2021, 24<sup>th</sup> Farvardin 1400



# Contents

## › Discrete Implementation

# Discrete Implementation

- › Image is discrete, our needs:

$$\frac{\partial u}{\partial t}, \frac{\partial u}{\partial x_1}, \frac{\partial u}{\partial x_2}$$

- ›  $u(x_1, x_2, t) = u(ih, jh, n\Delta t) \rightsquigarrow u_{i,j}^n$

- › Fundamental approximation ( $h = 1$ ):

$$\frac{\partial u}{\partial x_1} \cong \frac{u(x_1, x_2, t) - u(x_1 - h, x_2, t)}{h}$$

$$\frac{\partial u}{\partial x_1} \cong \frac{u(x_1 + h, x_2, t) - u(x_1, x_2, t)}{h}$$

$$\frac{\partial u}{\partial x_1} \cong \frac{u(x_1 + h, x_2, t) - u(x_1 - h, x_2, t)}{2h}$$

# Discrete Implementation

› Derivatives with respect to time:

$$\frac{\partial u}{\partial t} \cong \frac{u(x_1, x_2, t + \Delta t) - u(x_1, x_2, t)}{\Delta t} = \frac{u_{i,j}^{n+1} - u_{i,j}^n}{\Delta t}$$

› General Diffusion filter:

$$\frac{\partial u}{\partial t} = F(u, \nabla u) \Rightarrow \frac{u_{i,j}^{n+1} - u_{i,j}^n}{\Delta t} = F_{i,j}^n$$

$$u_{i,j}^{n+1} = u_{i,j}^n + \Delta t \times F_{i,j}^n$$

›  $u_{i,j}^0$ : Initial guess (noisy image or its gaussian smoothed)

# Discrete Implementation

- › We cover isotropic diffusion

$$\frac{\partial u}{\partial t} = \frac{\partial}{\partial x_1} (g(|\nabla u|^2) u_{x_1}) + \frac{\partial}{\partial x_2} (g(|\nabla u|^2) u_{x_2})$$

- › Notation and fundamental approximation:

$$› g(|\nabla u|^2) \cong g \left( \left( \frac{u_{i+1,j} - u_{i-1,j}}{2} \right)^2 + \left( \frac{u_{i,j+1} - u_{i,j-1}}{2} \right)^2 \right) = g_{i,j}$$

$$› \frac{\partial u}{\partial x_1} \cong \partial_{x_1}^* (u_{i,j}) = u_{i+0.5,j} - u_{i-0.5,j} \text{ (Half pixel, estimated via interpolation)}$$

$$› \frac{\partial u}{\partial x_2} \cong \partial_{x_2}^* (u_{i,j}) = u_{i,j+0.5} - u_{i,j-0.5}$$

# Discrete Implementation

› We cover isotropic diffusion

$$\frac{\partial u}{\partial t} = \frac{\partial}{\partial x_1} (g(|\nabla u|^2) u_{x_1}) + \frac{\partial}{\partial x_2} (g(|\nabla u|^2) u_{x_2})$$

$$\frac{\partial u}{\partial t}$$

$$= \delta_{x_1}^* \left( g_{i,j} (u_{i+0.5,j} - u_{i-0.5,j}) \right) + \delta_{x_2}^* \left( g_{i,j} (u_{i,j+0.5} - u_{i,j-0.5}) \right)$$

$$= g_{i+0.5,j} (u_{i+1,j} - u_{i,j}) - g_{i-0.5,j} (u_{i,j} - u_{i-1,j}) \\ + g_{i,j+0.5} (u_{i,j+1} - u_{i,j}) - g_{i,j-0.5} (u_{i,j} - u_{i,j-1})$$

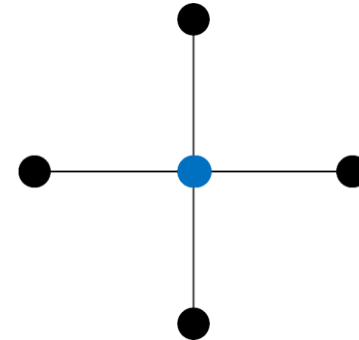
# Discrete Implementation

› Half pixel estimation estimated with simple interpolation:

$$g_{i\pm 0.5, j\pm 0.5} = \frac{g_{i\pm 1, j\pm 1} + g_{i, j}}{2}$$

› After substitution and simplification:

$$\frac{u_{i,j}^{n+1} - u_{i,j}^n}{\Delta t} = \frac{1}{2} \sum_{(k,l)=(i\pm 1, j\pm 1)} (g_{k,l}^n + g_{i,j}^n) (u_{k,l}^n - u_{i,j}^n)$$



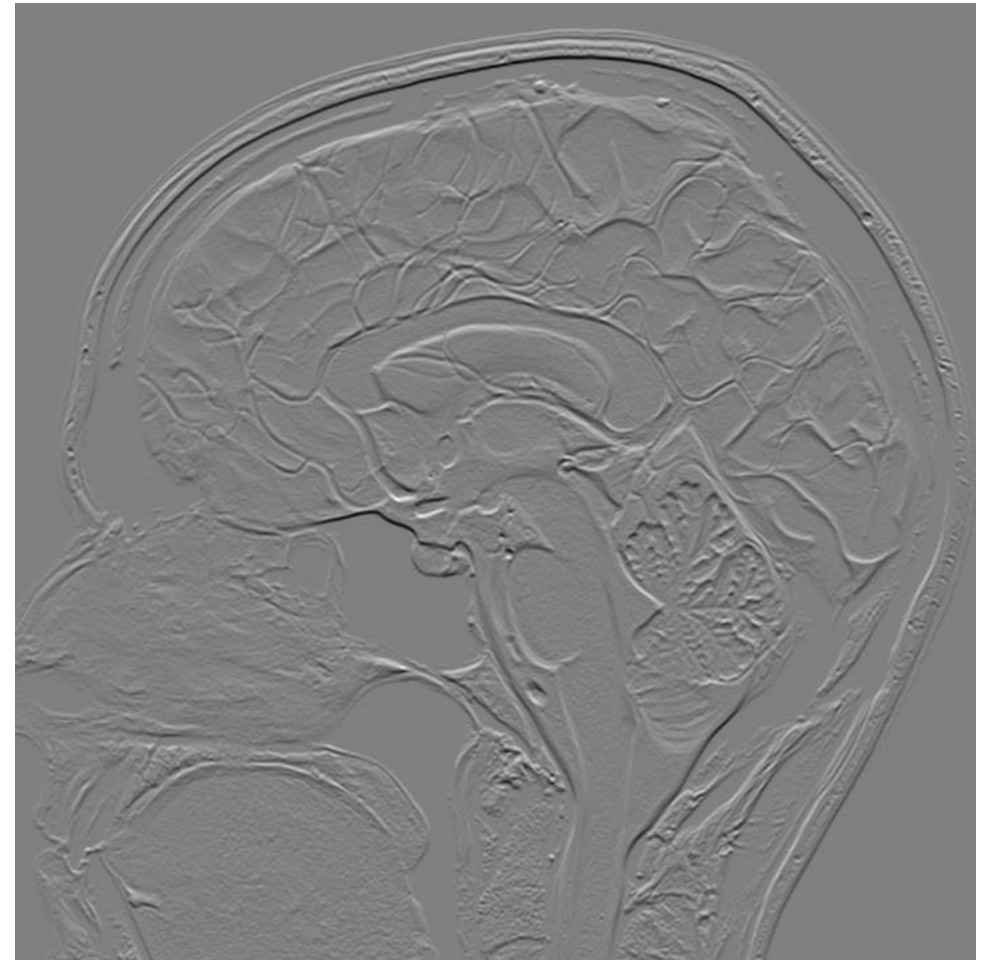
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## Example – Input Image





Example:  $\nabla u: (\partial u / \partial x_1, \partial u / \partial x_2)$



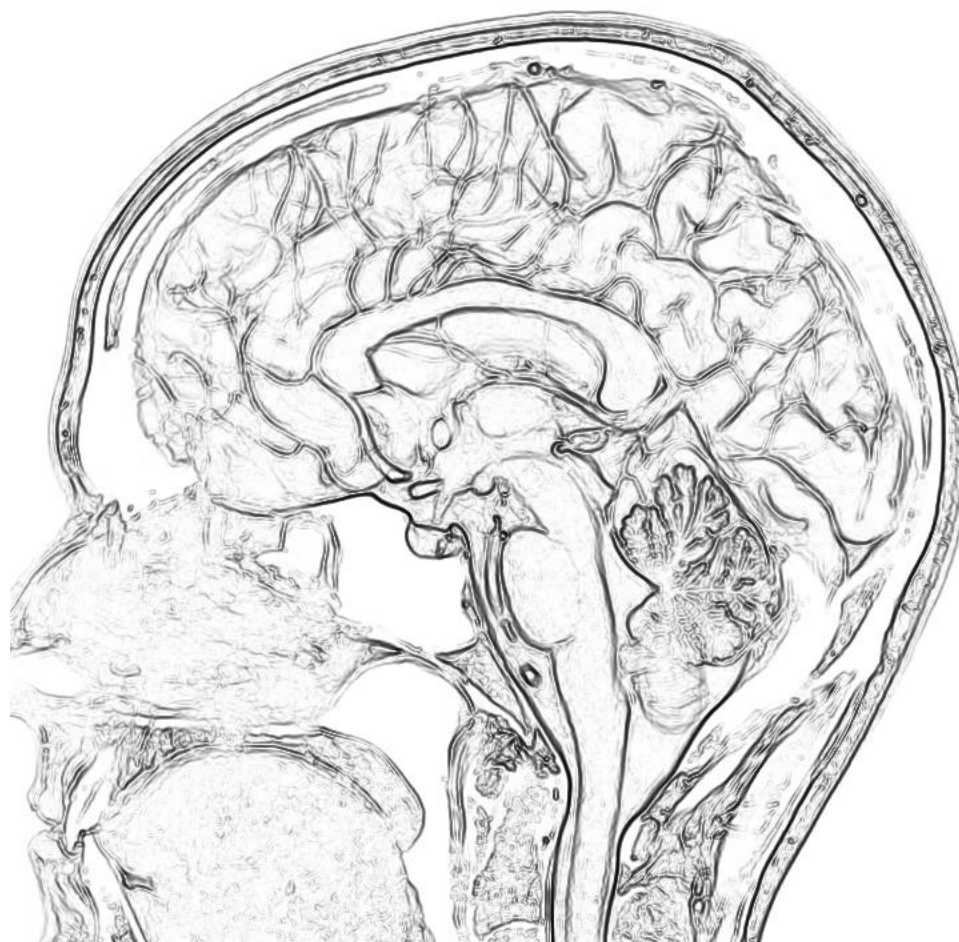
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Example:  $|\nabla u|^2$



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Example:  $G(|\nabla u|^2)$

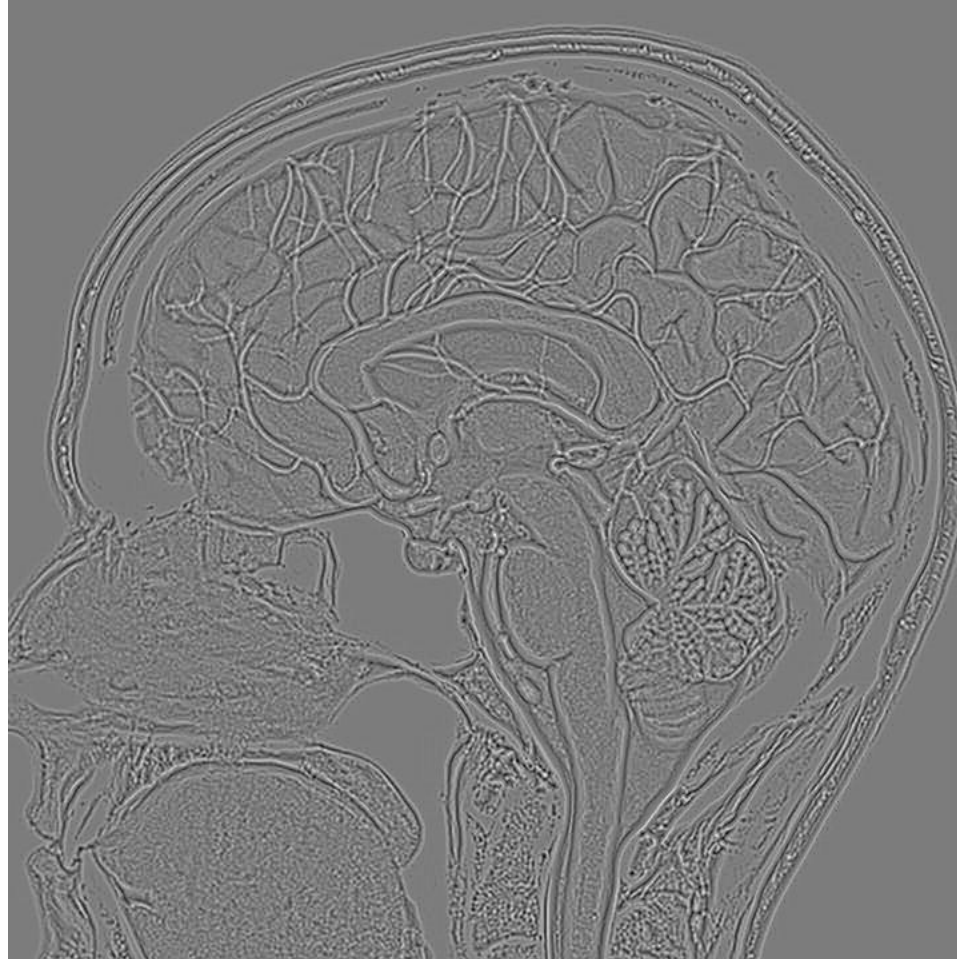




Example  $G\nabla u$ :  $(G \times (\partial u / \partial x_1), G \times (\partial u / \partial x_2))$

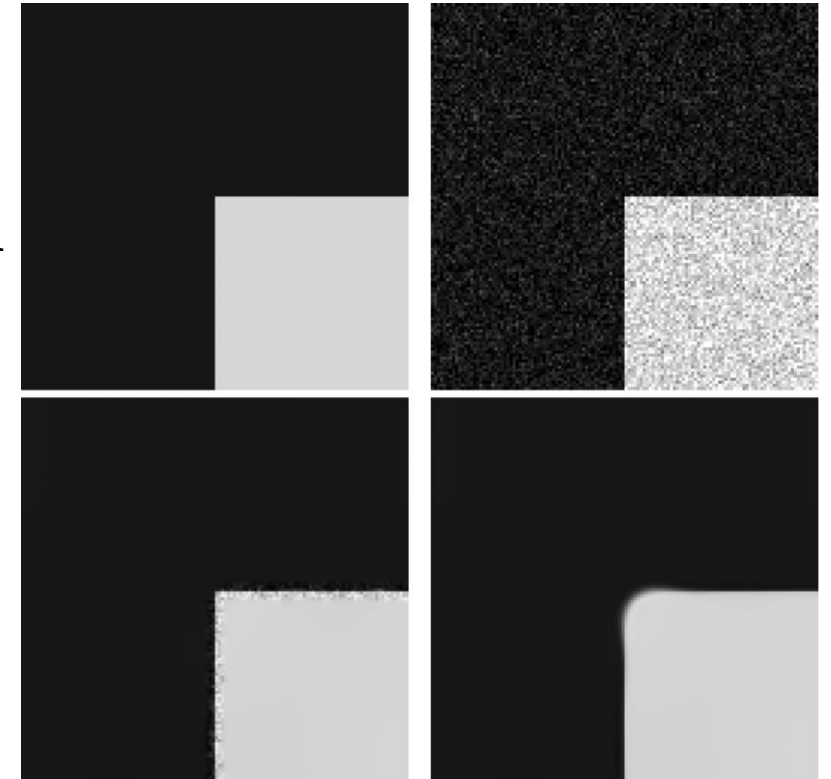


Example:  $\operatorname{div}(G(|\nabla u|^2)\nabla u)$



## Example – Denoising (Phantom)

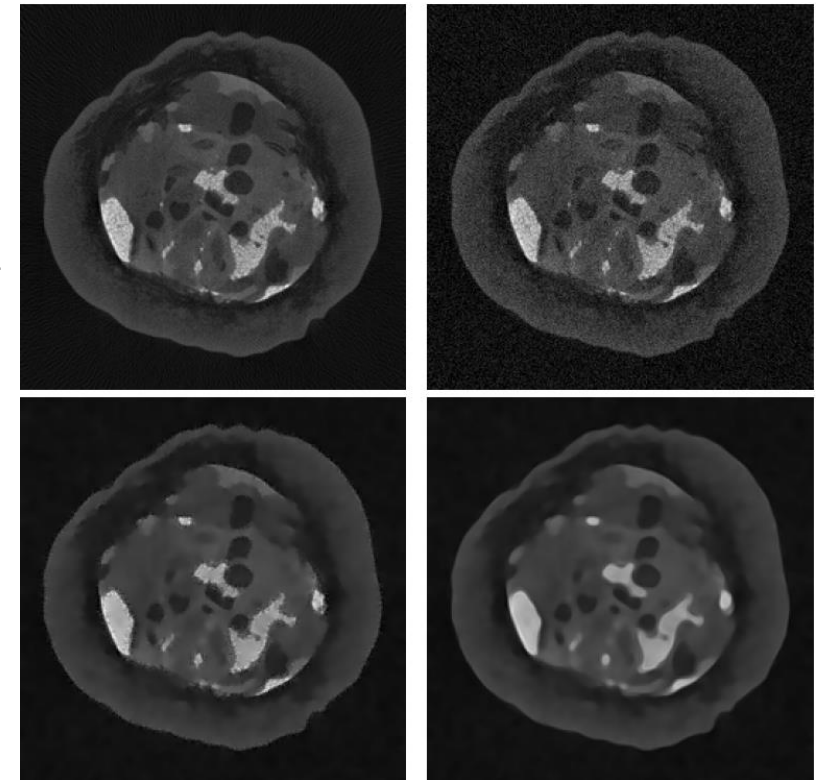
- › Top Left: Original
- › Top Right: Noisy
- › Bottom Left: Isotropic diffusion
- › Bottom Right: Anisotropic diffusion





## Example – Denoising (Real)

- › Top Left: Original
- › Top Right: Noisy
- › Bottom Left: Isotropic diffusion
- › Bottom Right: Anisotropic diffusion



## Example – Artistic Effect





# The End

› AnY QuEsTiOn?

