

AFNI User Group Meeting

Anisotropic Smoothing

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Anisotropic Smoothing

- Image filtering operation that preferentially smooths one part of an image versus another.
- This operation is in contrast to the more standard mean, median and Gaussian smoothing operations.
- Method based on Ding (Weickert, et al and Gerig, et al). Diffusion-based filtering.

Anisotropic Smoothing

Pseudo-code for DWI data (SmoothDWI) (2D only)

U_0 is original image data

For $i=0$ to 10

- Compute Image diffusion tensor (D)
 - Smooth U_i a little with Gaussian (sigma=0.5)
 - Compute gradients, matrix of products of gradients $R = \begin{bmatrix} (du/dx)^2 & (du/dx)(du/dy) \\ (du/dx)(du/dy) & (du/dy)^2 \end{bmatrix}$
 - Smooth each component of R (sigma=1)
 - Compute eigenvalues (μ_1, μ_2) vectors for R
 - Compute $\phi_1 = 1 / (\mu_1 * s)$, $\phi_2 = 1 / (\mu_2 * s)$ where $s = 1/\mu_1 + 1/\mu_2$ (Ding method)
 - or $\phi_1 = 0.01 + 0.99 \exp [-0.01/(\mu_1 - \mu_2)^2]$, $\phi_2 = 0.01$ (exp method)
 - Compute $D = V \Phi V^T$ where $\Phi = \begin{bmatrix} \phi_1 & 0 \\ 0 & \phi_2 \end{bmatrix}$
- Smooth image dataset (DWI data) given D
 - Compute flux in image J_x, J_y
 - $J_x = E_{xx} du/dx + E_{xy} du/dy$
 - $J_y = E_{xy} du/dx + E_{yy} du/dy$where $E = \begin{bmatrix} D_{xx}-D_m & D_{xy} \\ D_{xy} & D_{yy}-D_m \end{bmatrix}$ and $D_m = \text{mean diffusivity}$
 - $G_{pqn} = dJ_x/dx + dJ_y/dy = \text{anisotropic part of the smoothing}$
 - $U_{p,q,n+1} = U_{p,q,n} + \Delta t (F_{pqn} + G_{pqn})$
 - where $F = D_m / \Delta x^2 * U_{\text{smooth}} = \text{isotropic smoothing}$
 - and $\Delta t = D_{\text{max}}/4$

End loop

3danisosmooth

Usage: **3danisosmooth** [options] dataset
Smooth a DWI dataset using anisotropic smoothing.

The output dataset is preferentially smoothed in similar areas
may use a sub-brick selection list, as in program 3dcalc.

Options :

- prefix** pname = Use 'pname' for output dataset prefix name.
- iters** nnn = compute nnn iterations (default=10)
- 2D** = smooth a slice at a time
- 3D** = smooth through slices. Can not be combined with 2D option
- mask** dset = use dset as mask to include/exclude voxels
- automask** = automatically compute mask for dataset
Can not be combined with -mask
- viewer** = show central axial slice image every iteration.
Starts aiv program internally.

- nosmooth** = do not do intermediate smoothing of gradients
- sigma1** n.nnn = assign Gaussian smoothing sigma before gradient computation for calculation of structure tensor.
Default = 0.5
- sigma2** n.nnn = assign Gaussian smoothing sigma after gradient matrix computation for calculation of structure tensor.
Default = 1.0
- deltat** n.nnn = assign pseudotime step.
Default = 0.25
- savetempdata** = save temporary datasets each iteration. Dataset prefixes are Gradient, Eigens, phi and Dtensor.
Each is overwritten each iteration.
- phiding** = use Ding method for computing phi (default)
- phiexp** = use exponential method for computing phi

- help** = print this help screen

Gradient Filter Kernels

2D kernels

du/dx

$-a$	0	a
$-b$	0	b
$-a$	0	a

du/dy

$-a$	$-b$	$-a$
0	0	0
a	b	a

3D kernels at $p-1, p+1$

a	b	a
b	c	b
a	b	a

where

$$a = 0.02, b = 0.06, c = 0.18$$

where $a=3/16, b= 10/16$

Isotropic smoothing kernel

a	b	a
b	c	b
a	b	a

2D

$$a = 1/6, \quad b = 2/3, \quad c = -10/3$$

$$a = 0.4, \quad b = 2.0/15.0, \quad c = 1.0/60.0,$$
$$d = (-6.0 * a) - (12.0 * b) - (8.0 * c) = -4.13$$

b	a	b
a	d	a
b	a	b

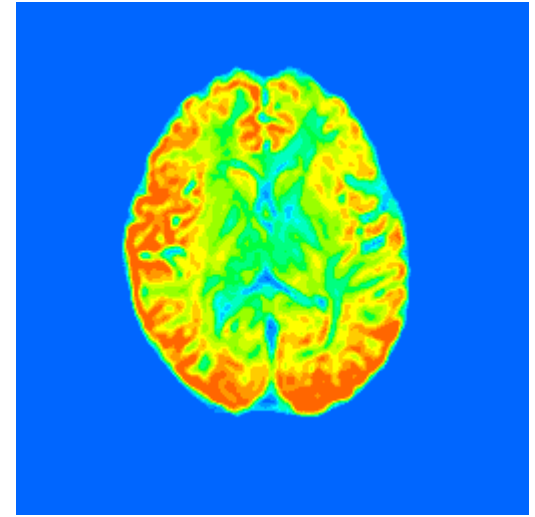
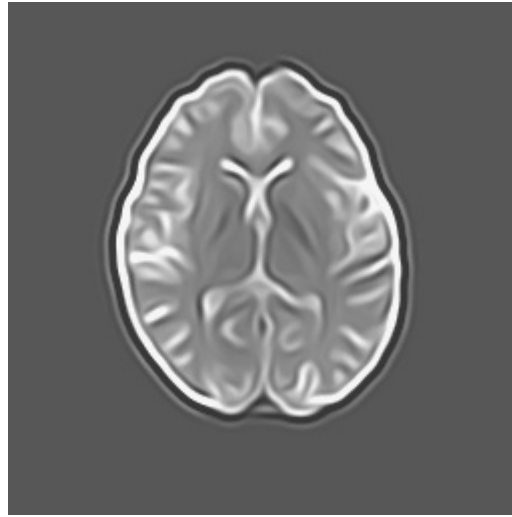
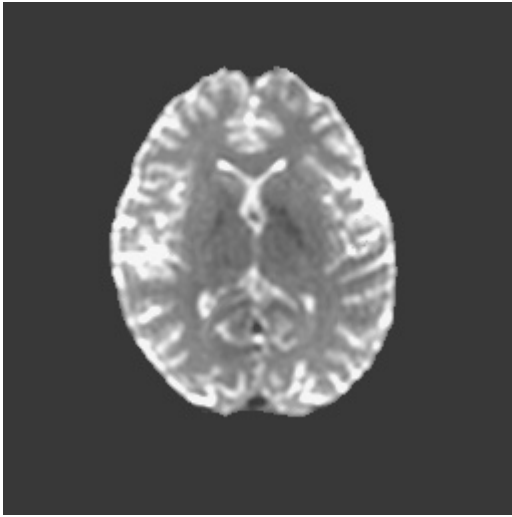
at slice p

c	b	c
b	a	b
c	b	c

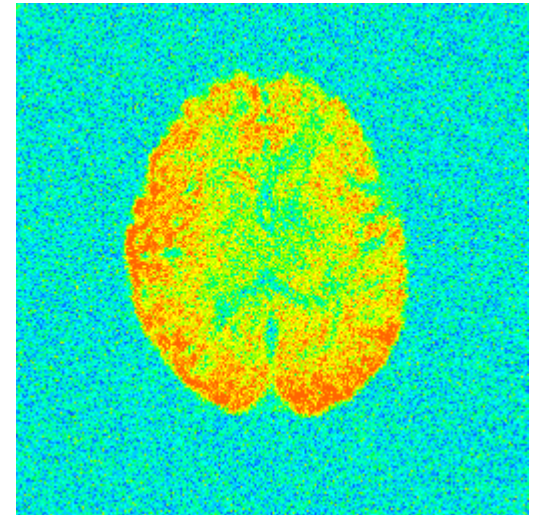
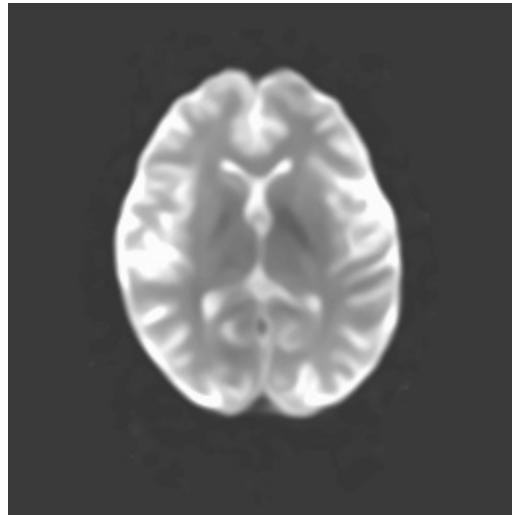
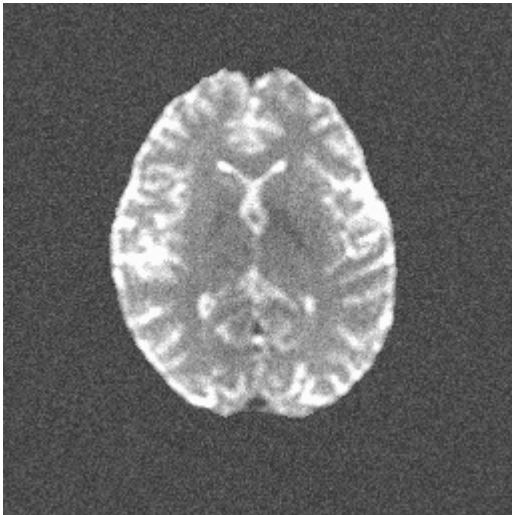
3D

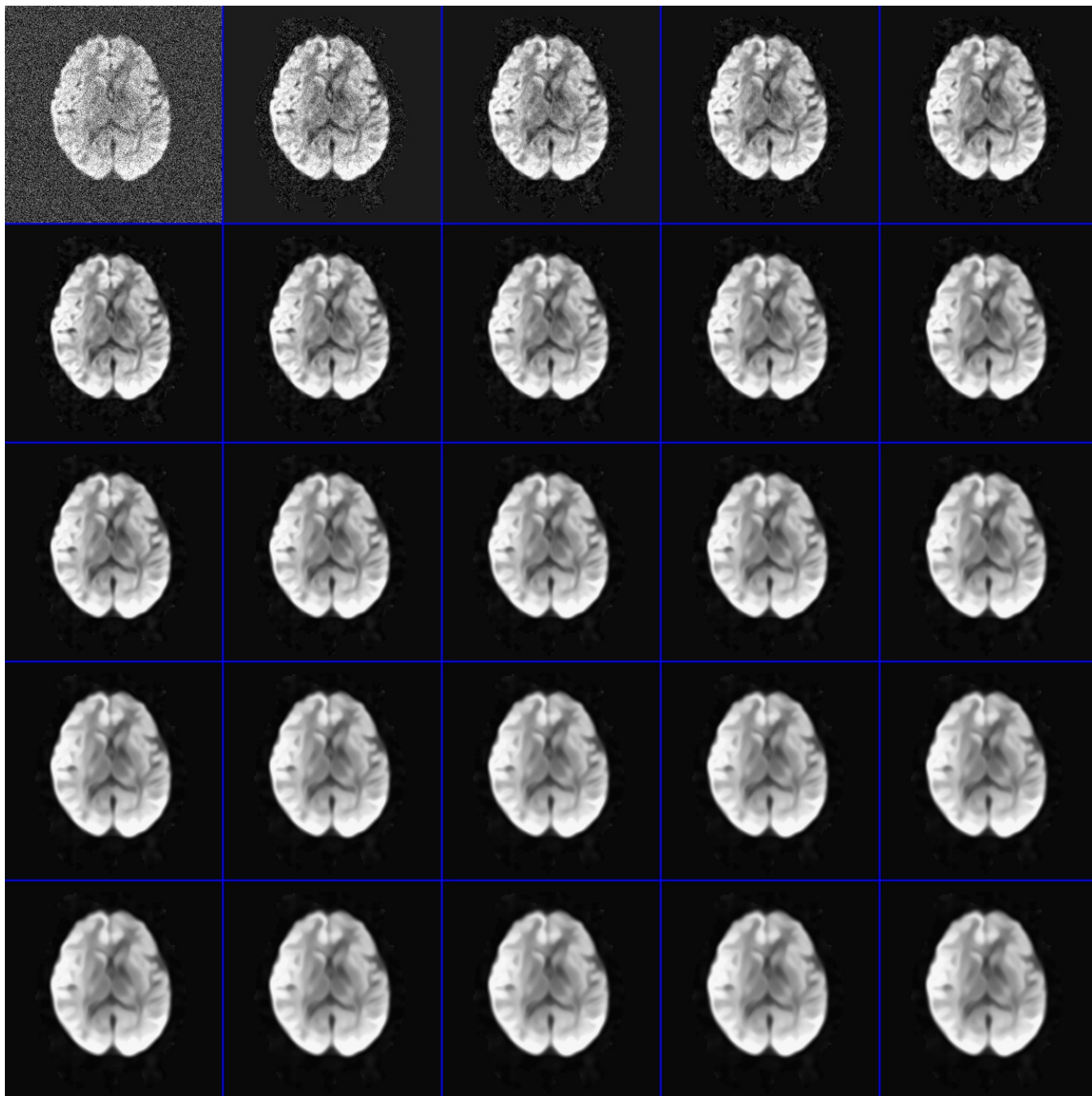
at slice p-1,
p+1

DWI Images

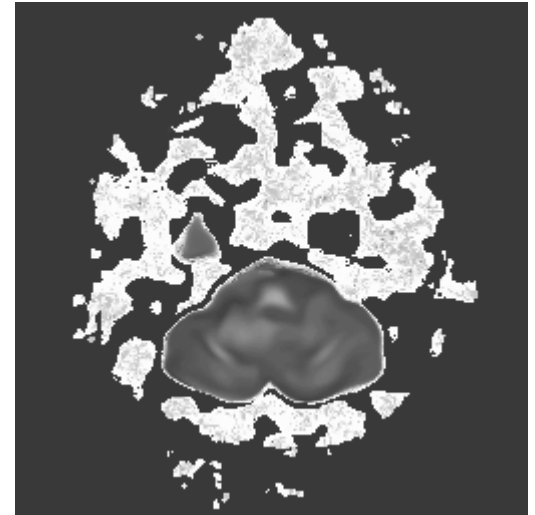
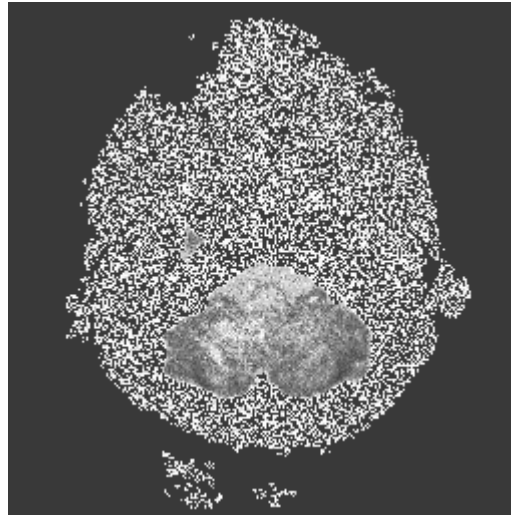
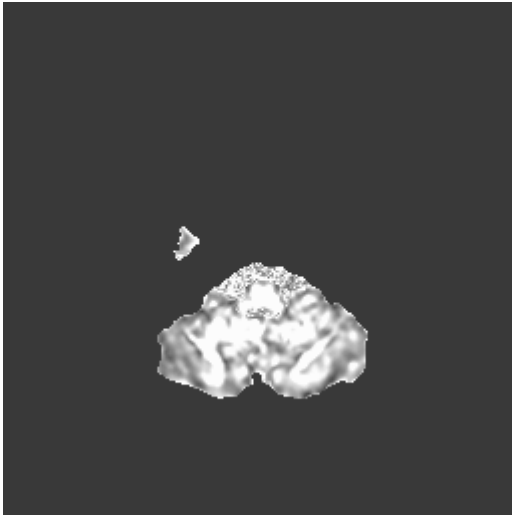


25 iterations

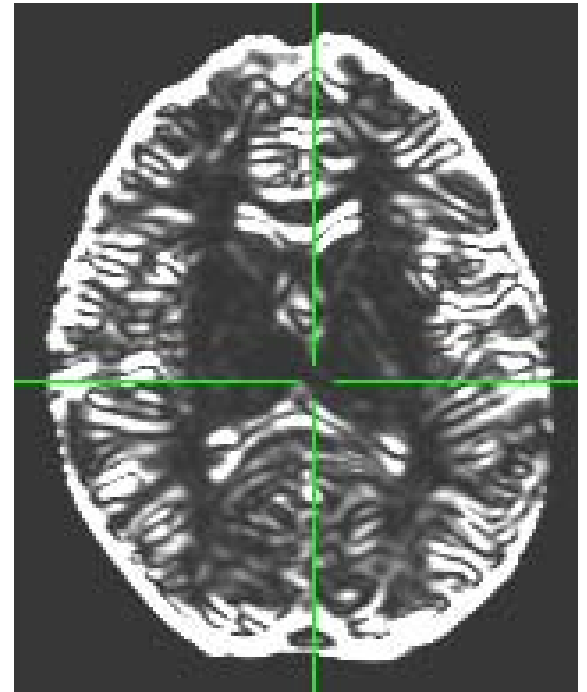
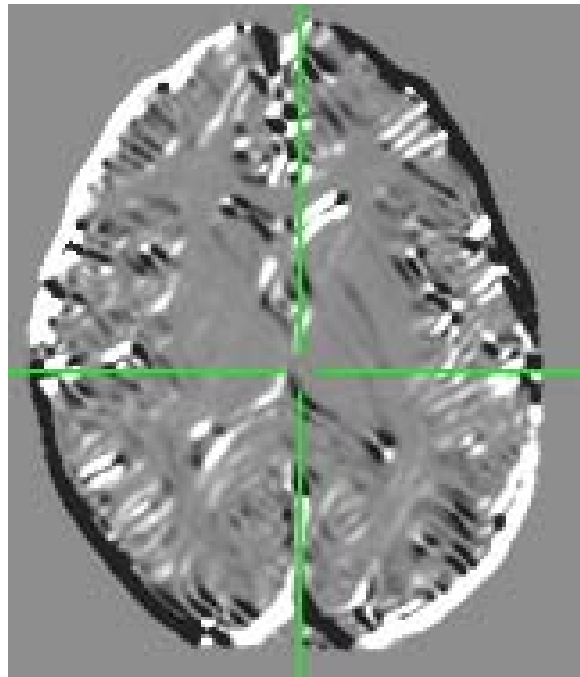
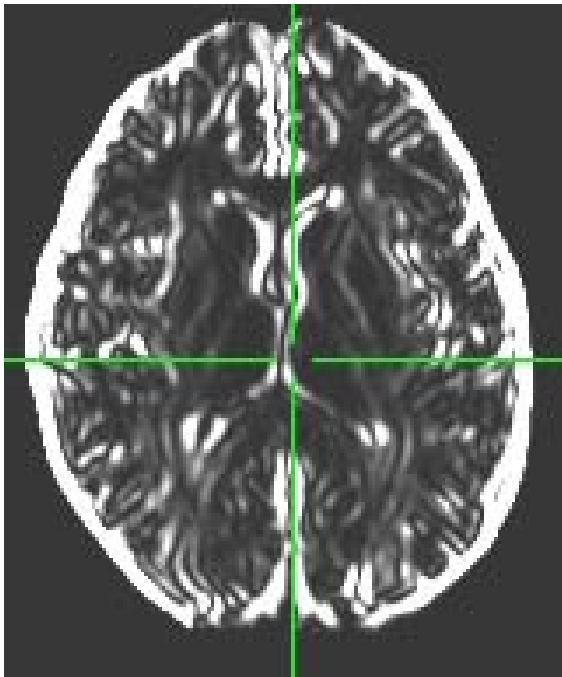




DWI/DT Images



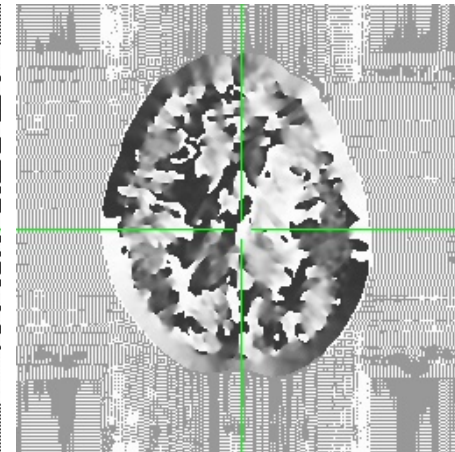
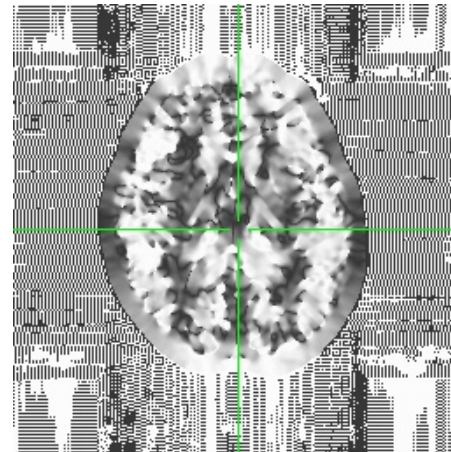
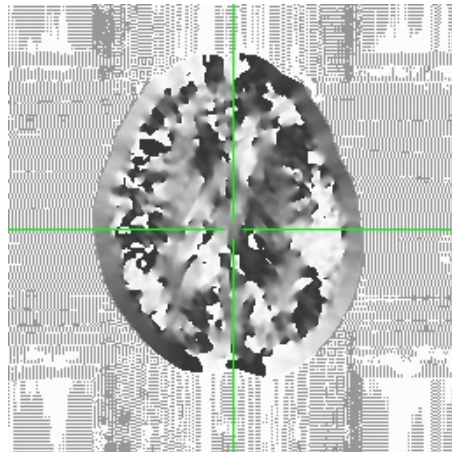
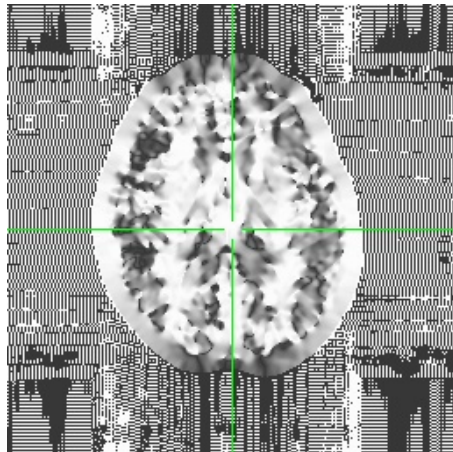
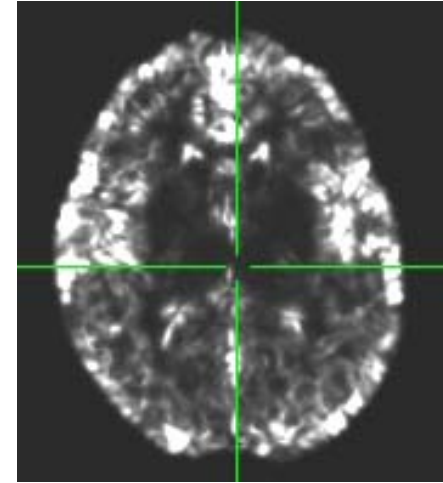
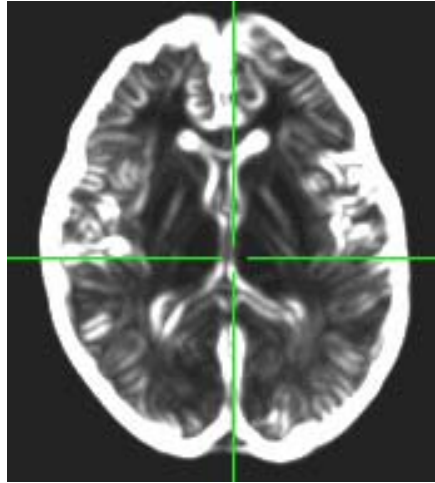
DWI Intermediate Images



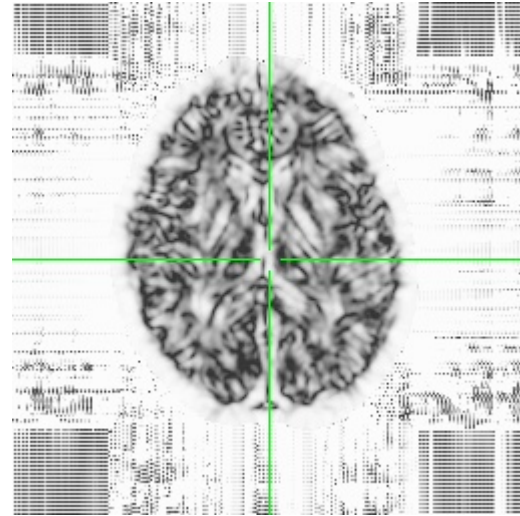
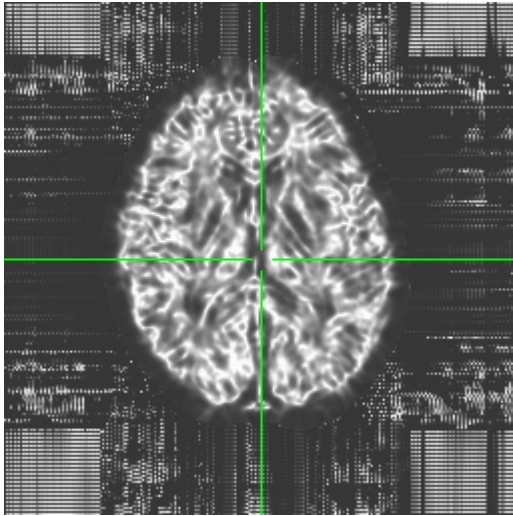
Gradient D_{xx}^2 , D_{xy} , D_{yy}^2

DWI Intermediate Images

Eigenvalues, vectors

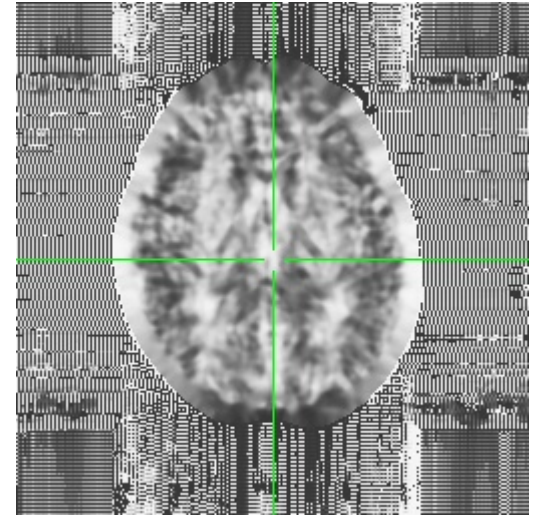
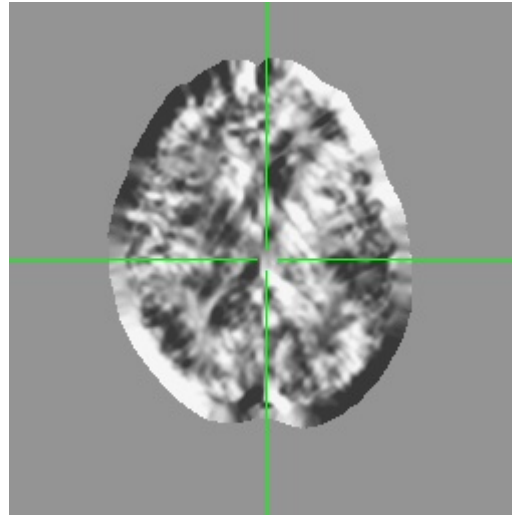
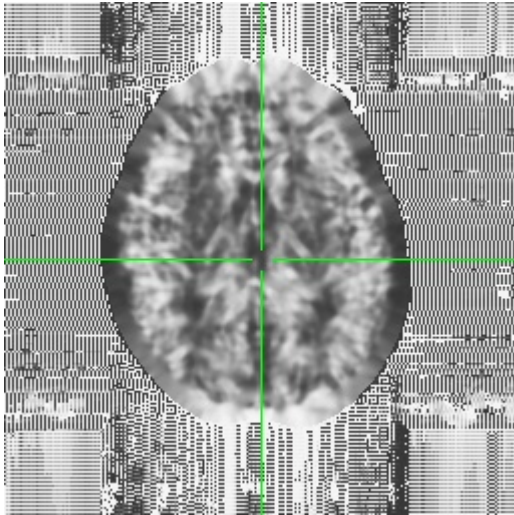


DWI Intermediate Images



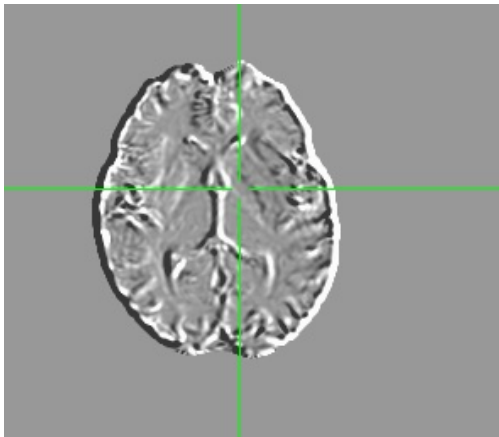
Phi Values

DWI Intermediate Images

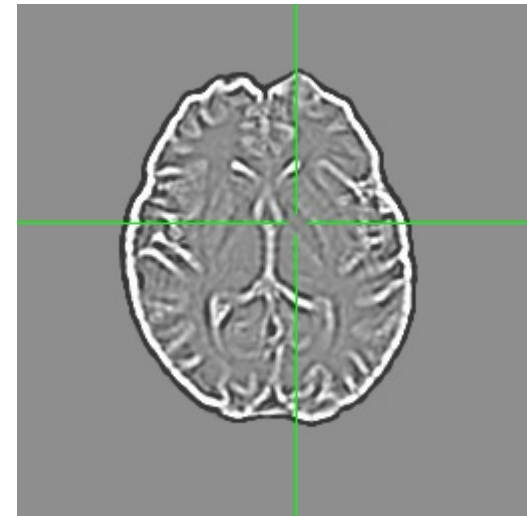
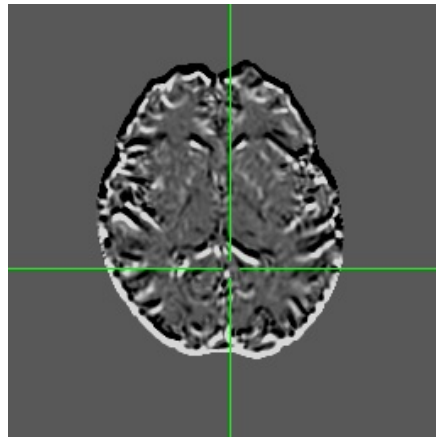


D Tensor: D_{xx} , D_{xy} , D_{yy}

DWI Intermediate Images

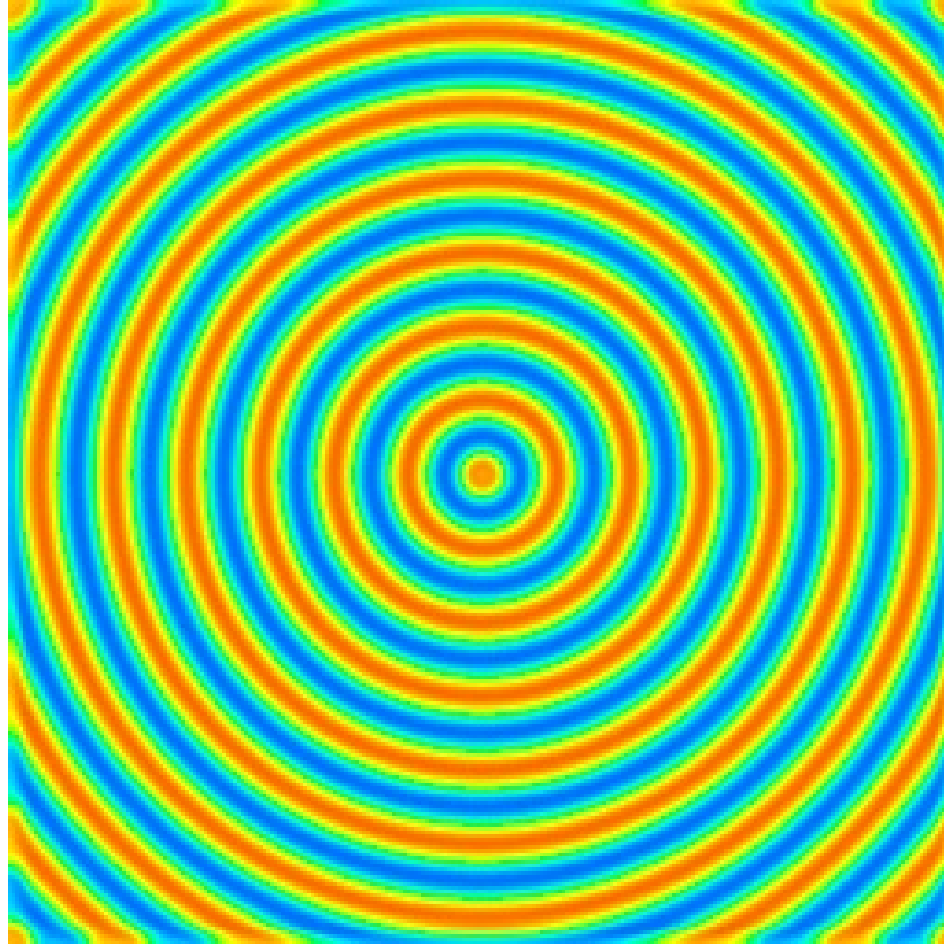


Flux in x
and y



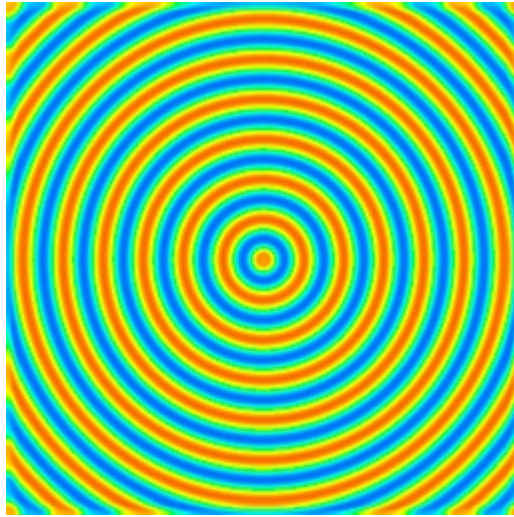
G matrix -
anisotropic part

Cosine Circles

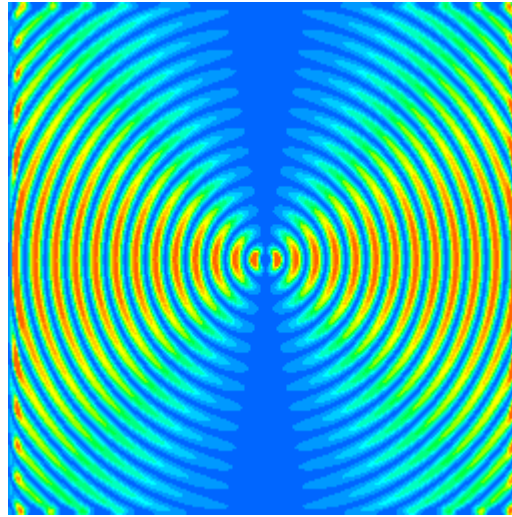


After 25 iterations

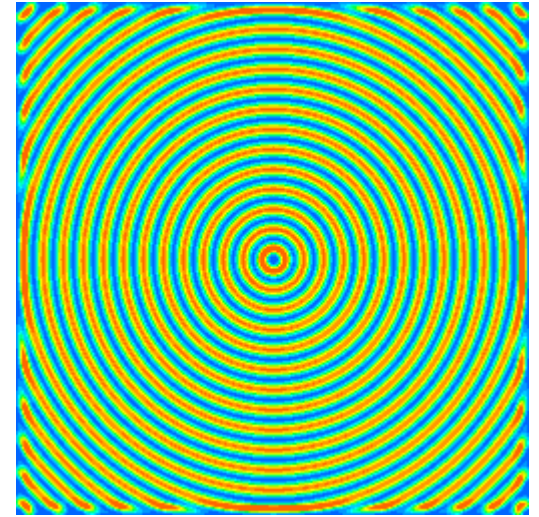
Cosine Circles



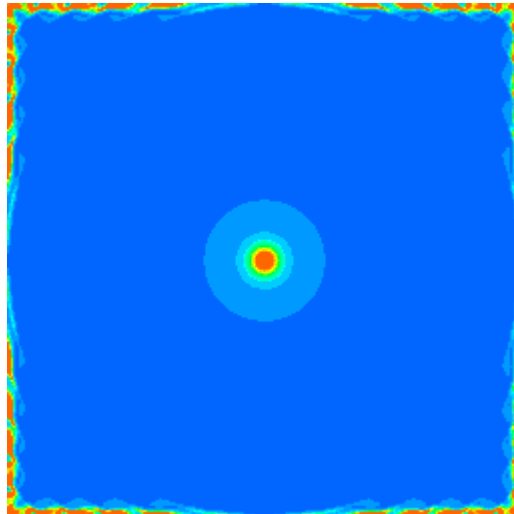
25 iterations



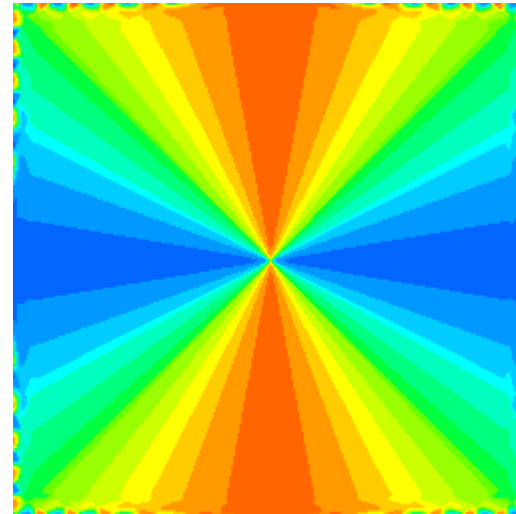
Gradient D_{xx}^2 , D_{xy} , D_{yy}^2



Eigenvalues, vectors

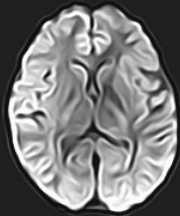
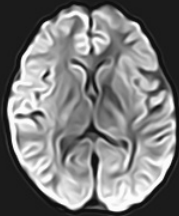
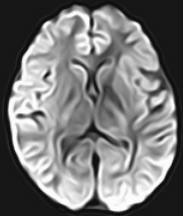
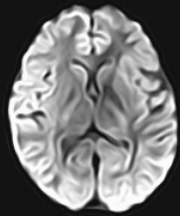
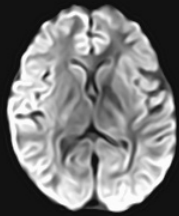
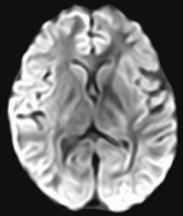
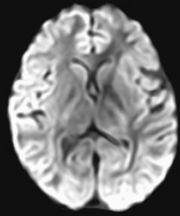
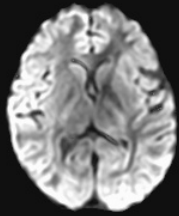
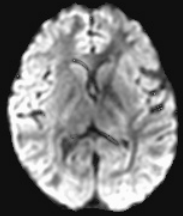


phi values

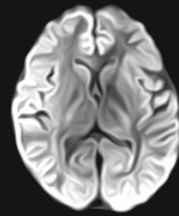
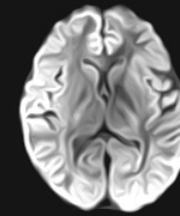
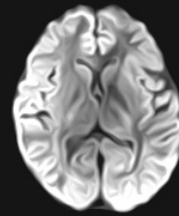
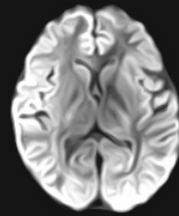
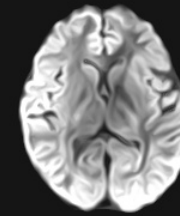
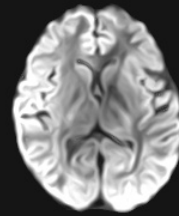
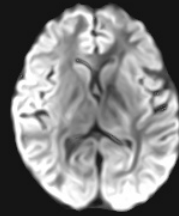
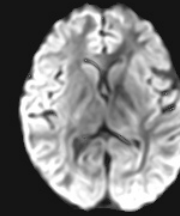
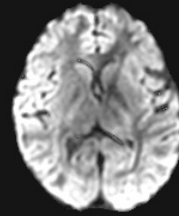


D tensor D_{xx} , D_{xy} , D_{yy}

Phi value calculation



Ding



Exponential
 $c_2 = 0.01$

Issues

- Performance and Memory
 - 3:30 for 10 iterations, 21 seconds per iteration with DWI 256x256x41 x 22 sub-bricks (2D)
 - ~ 1 sec/sub-brick/iteration ~ 0.025 sec/slice/iteration
 - 9:00 for 10 iterations, 54 seconds per iteration, 2.5 sec/sub-brick/iteration, 0.06 sec/slice/iteration (3D),

 - 5:54 for 10 iterations, 35 seconds per iteration with T2 512x512x112 single brick data (2D)
 - ~ 0.3 sec/slice/iteration
 - 16:22 for 10 iterations, 98 seconds per iteration, 0.88 sec/slice/iteration (3D)
 - $2n + 6$ sub-bricks for 2D, $2n+12$ (3D), n =number of sub-bricks (almost 1GB)
 - Improvements made in
 - Gaussian smoothing (spatial kernels instead)
 - more efficient spatial kernels for gradient and other smoothing kernel in algorithm
 - eigenvalue solver specific for symmetric 2x2 and 3x3
 - mask operations – edge of mask gets special treatment
 - Larger Delta T step (instability possible)
 - Cheaper D tensor and alternative phis - eigenvalue alternative (Ding, Matlab)
- Edges, Masks, Anisotropy
 - Edges show problems after many iterations
 - Masks treated equivalently to edges in program
 - Voxels treated “isotropically” ($dx=dy=dz$)
- Overfiltering and end points
 - Truth
 - Shock filter (stop determination)

References

Z Ding, JC Gore, AW Anderson, Reduction of Noise in Diffusion Tensor Images Using Anisotropic Smoothing, Mag. Res. Med., 53:485-490, 2005

J Weickert, H Schar, A Scheme for Coherence-Enhancing Diffusion Filtering with Optimized Rotation Invariance, CVGPR Group Technical Report at the Department of Mathematics and Computer Science, University of Mannheim, Germany, TR 4/2000.

J. Weickert, H. Schar. A scheme for coherence-enhancing diffusion filtering with optimized rotation invariance. J Visual Communication and Image Representation, Special Issue On Partial Differential Equations In Image Processing, Comp Vision Computer Graphics, pages 103-118, 2002.

Gerig, G., Kubler, O., Kikinis, R., Jolesz, F., Nonlinear anisotropic filtering of MRI data, IEEE Trans. Med. Imaging 11 (2), 221-232, 1992.

