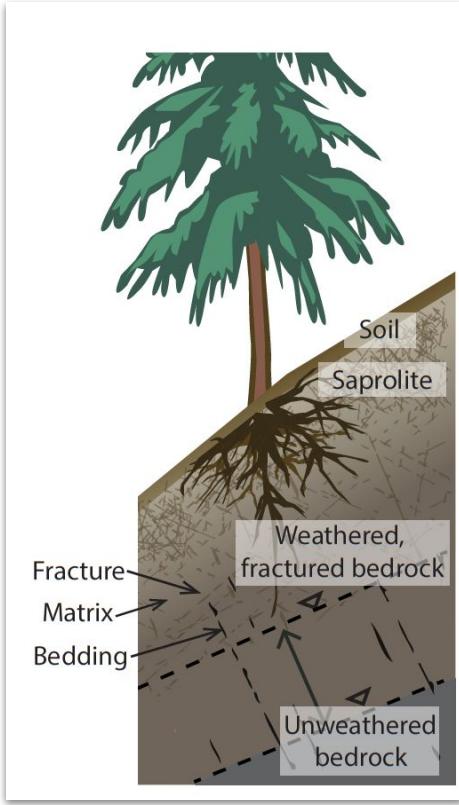


Fracture Measurement in CT

Omar Alamoudi and Logan Schmidt

Motivation: Water storage in unsaturated fractured bedrock is important



Eel River, CA Sandstone



McDonald Observatory, TX
Volcanics

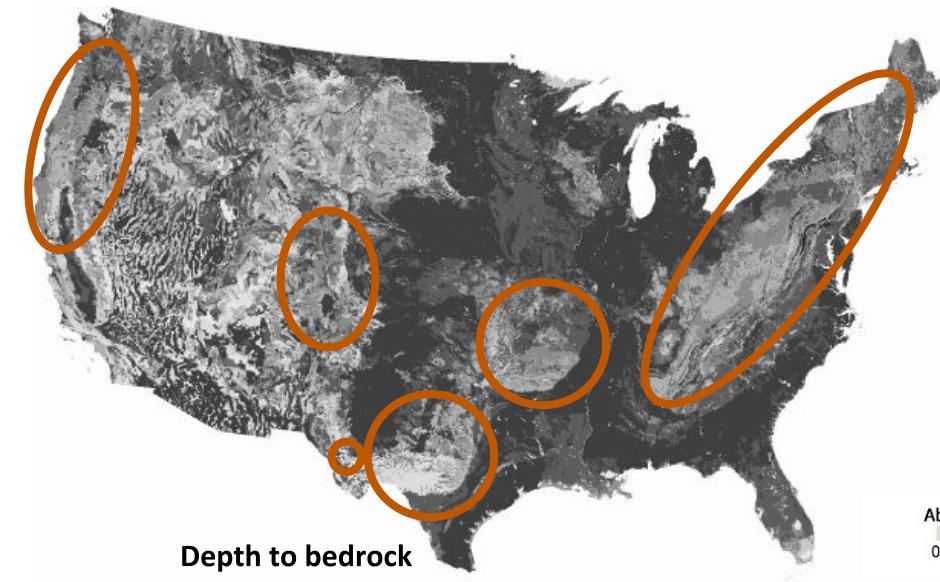


White Property, TX Karst

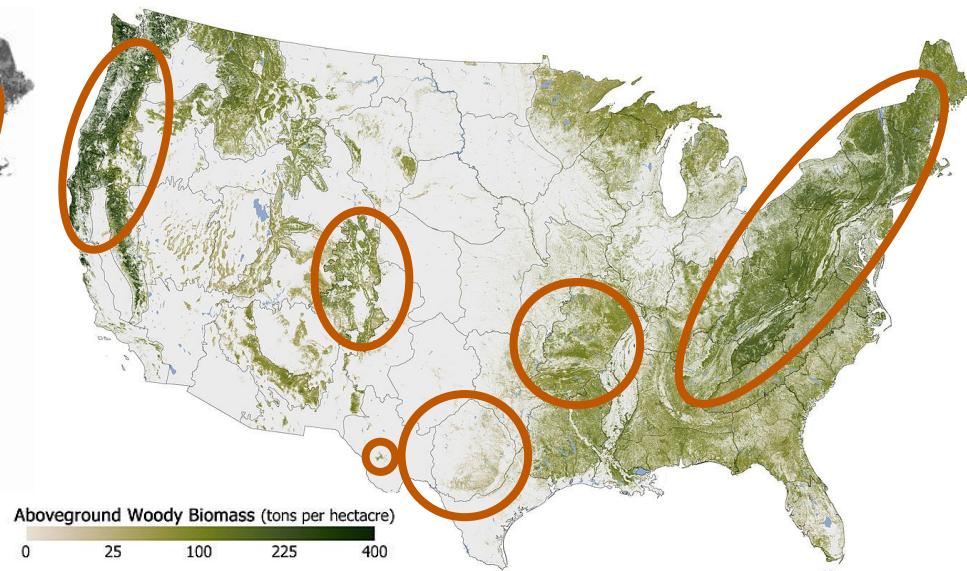


Gothic, CO Mudstone

Motivation: Water storage in unsaturated weathered bedrock is important

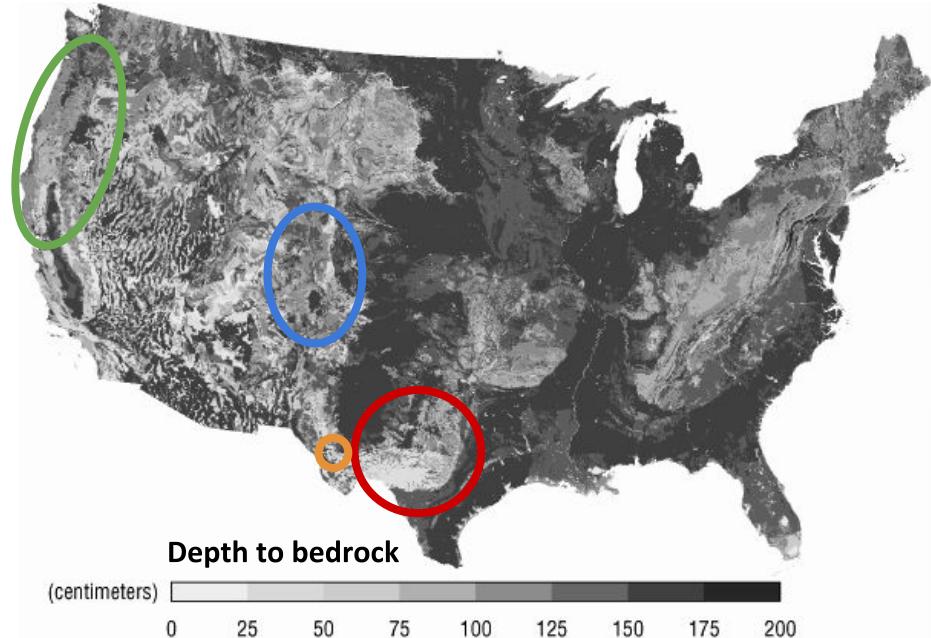


(Miller and White, 1998)



(NASA Earth Observatory map by Robert Simon)

Proposed Research Objectives



(Miller and White, 1998)



Eel River, CA Sandstone



McDonald Observatory, TX
Volcanics

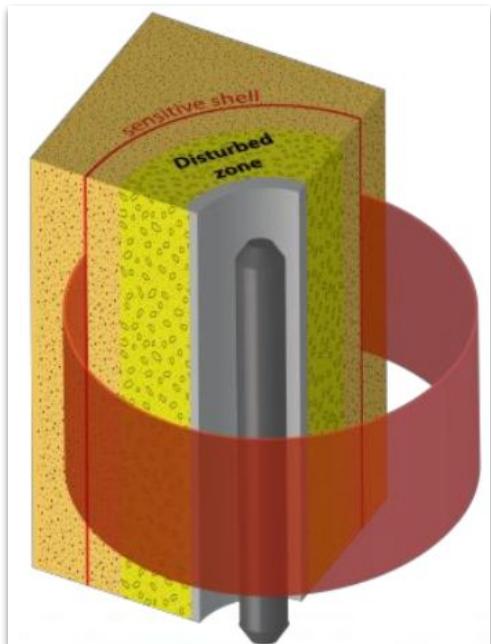


White Property, TX Karst



Gothic, CO Mudstone

Motivation: NMR is uniquely suited for hydrological investigations in the weathered bedrock vadose zone



(Vista Clara, Inc.)

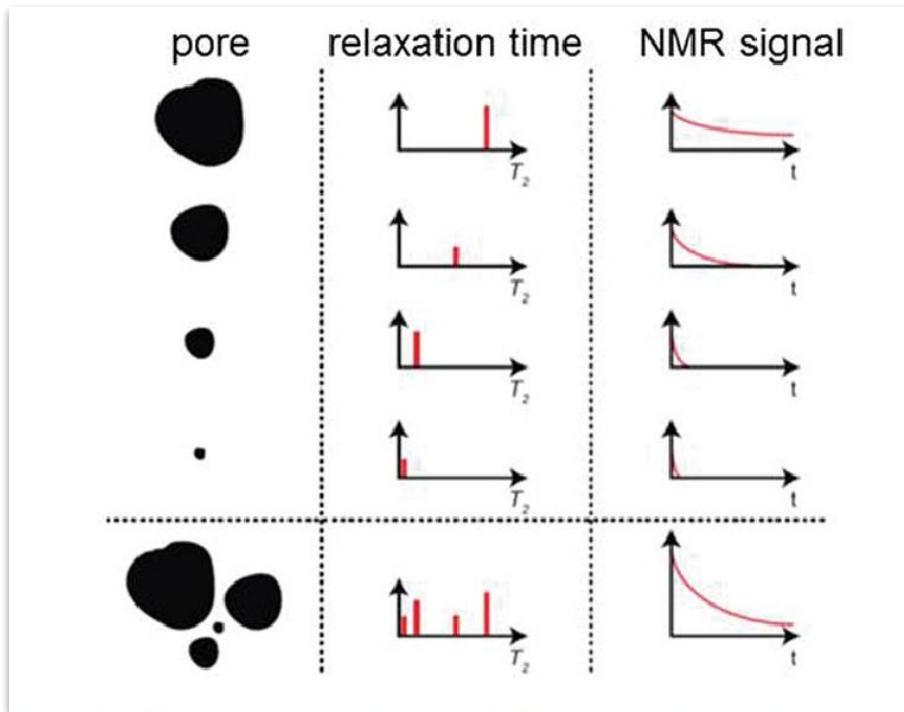
- Directly sensitive to water
- No site-specific calibration
- Matrix-independent water content measurement
- Fixed, well-defined volume of investigation
- High vertical resolution
- Decades of use in petroleum exploration
- Sensitive to pore environment

Motivation: NMR is uniquely suited for hydrological investigations in the weathered bedrock vadose zone

Initial amplitude directly proportional to water content

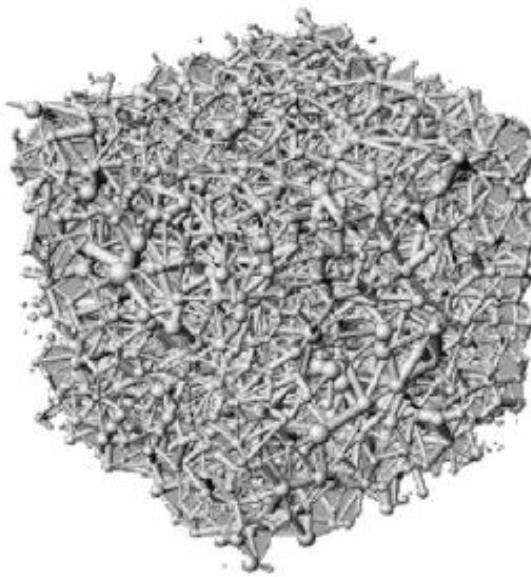
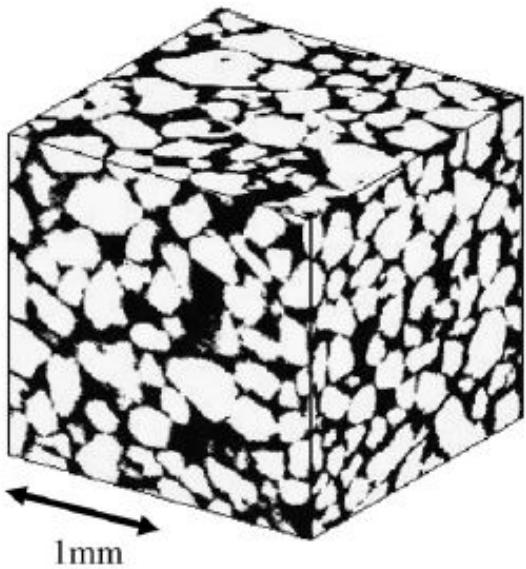
Signal decay (relaxation) influenced by pore environment

$$M(t) = M_0 \sum_{i=1}^n a_i e^{-t/T_{2i}}$$



(Coates et al., 1999)

Part 3 Question: What is the water potential of rock moisture

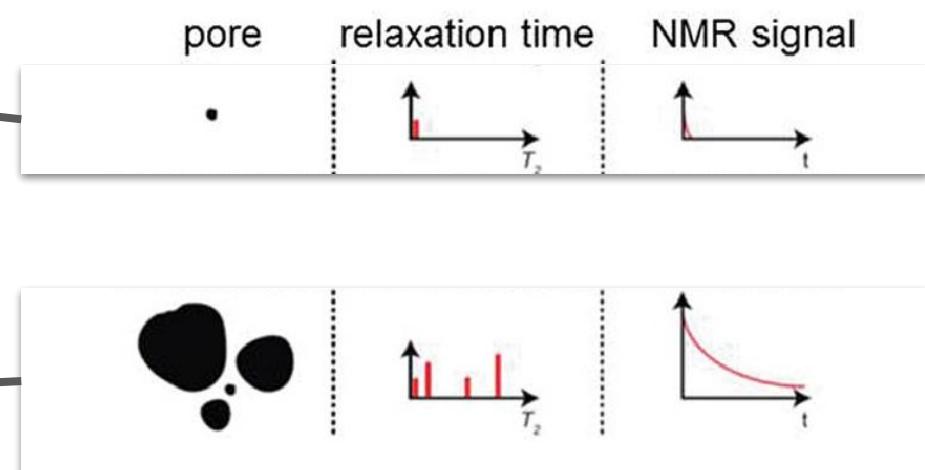
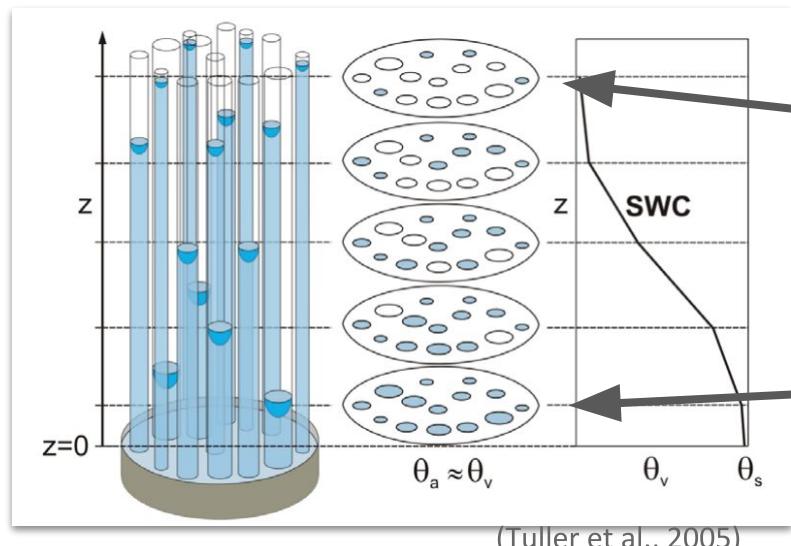


(Talabi et al., 2009)

Part 3 Methods: NMR relaxation is controlled by pore radius and can be related to water potential

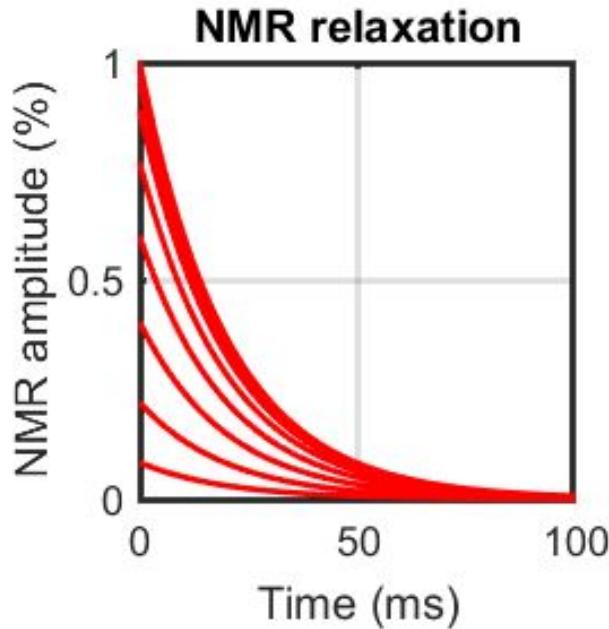
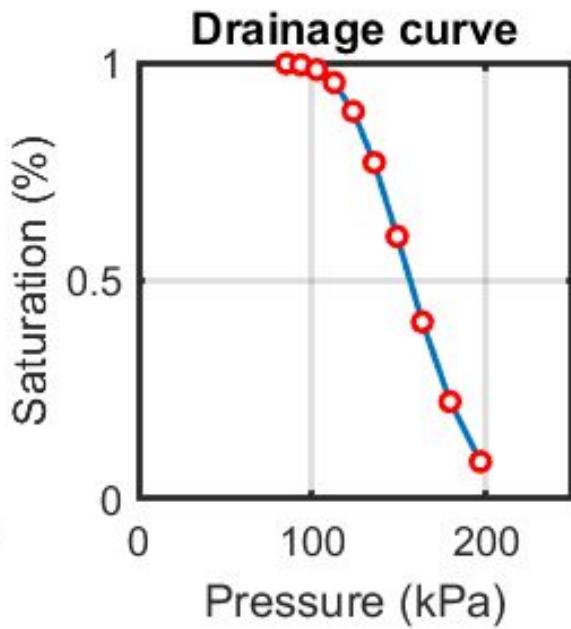
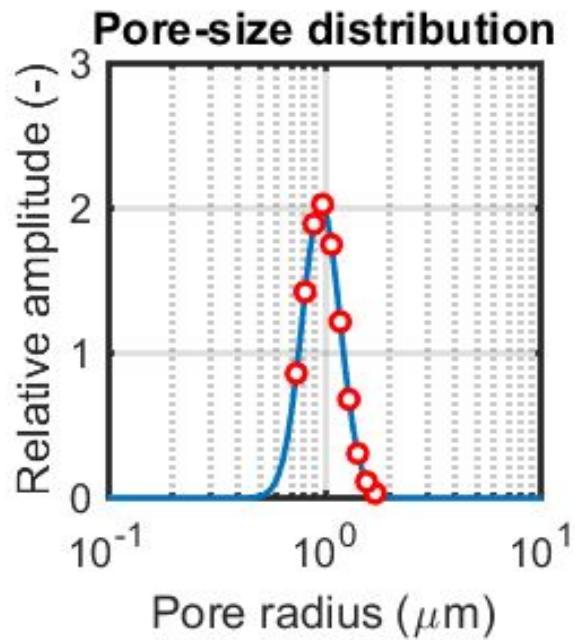
$$\Delta P \propto \frac{\sigma}{r} \quad \frac{1}{T_2} = \rho_2 \left(\frac{\alpha}{r} \right)$$

$$M(t) = M_0 \sum_i M_i e^{(-t/T_{2,i})}$$

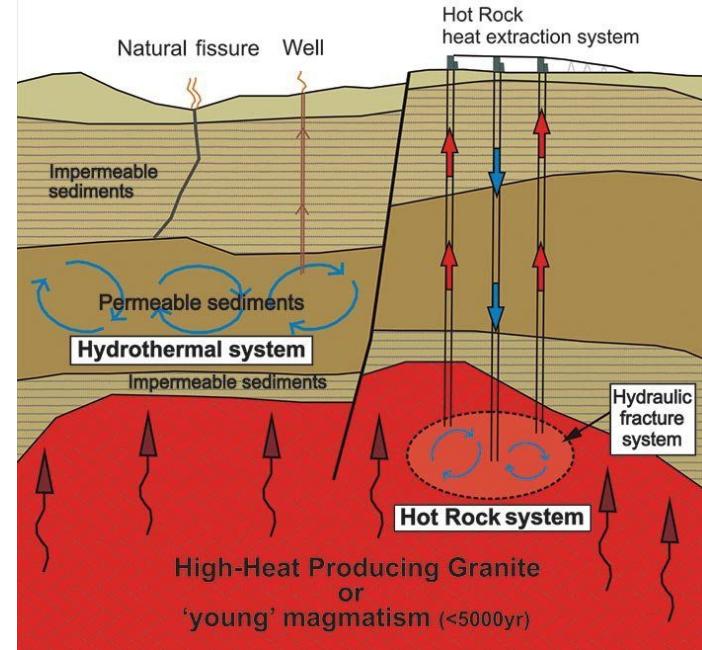
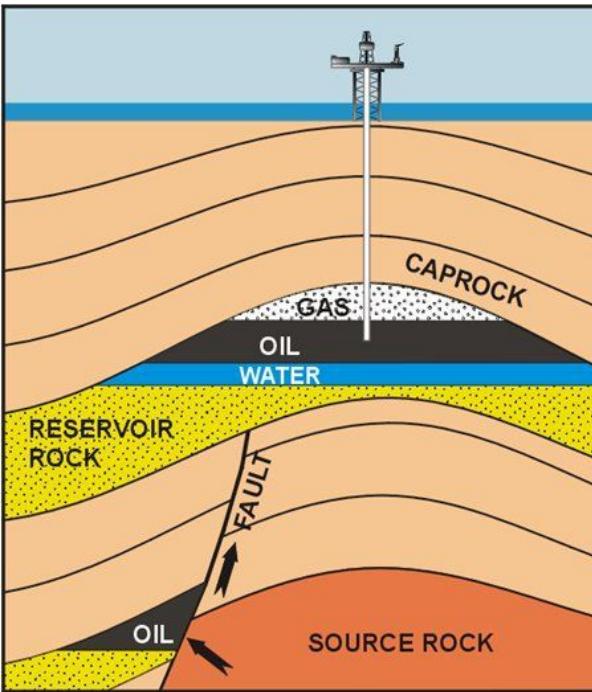
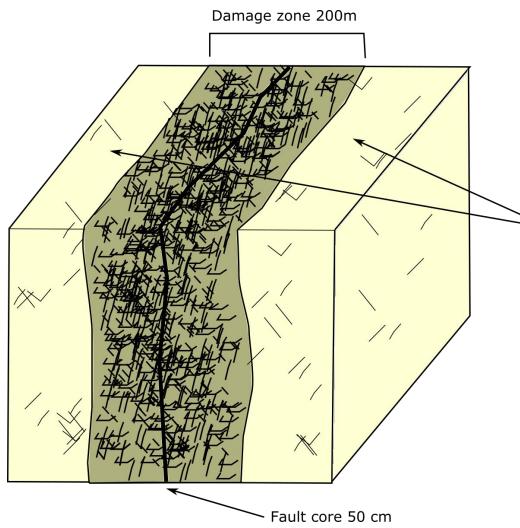


(Coates et al., 1999)

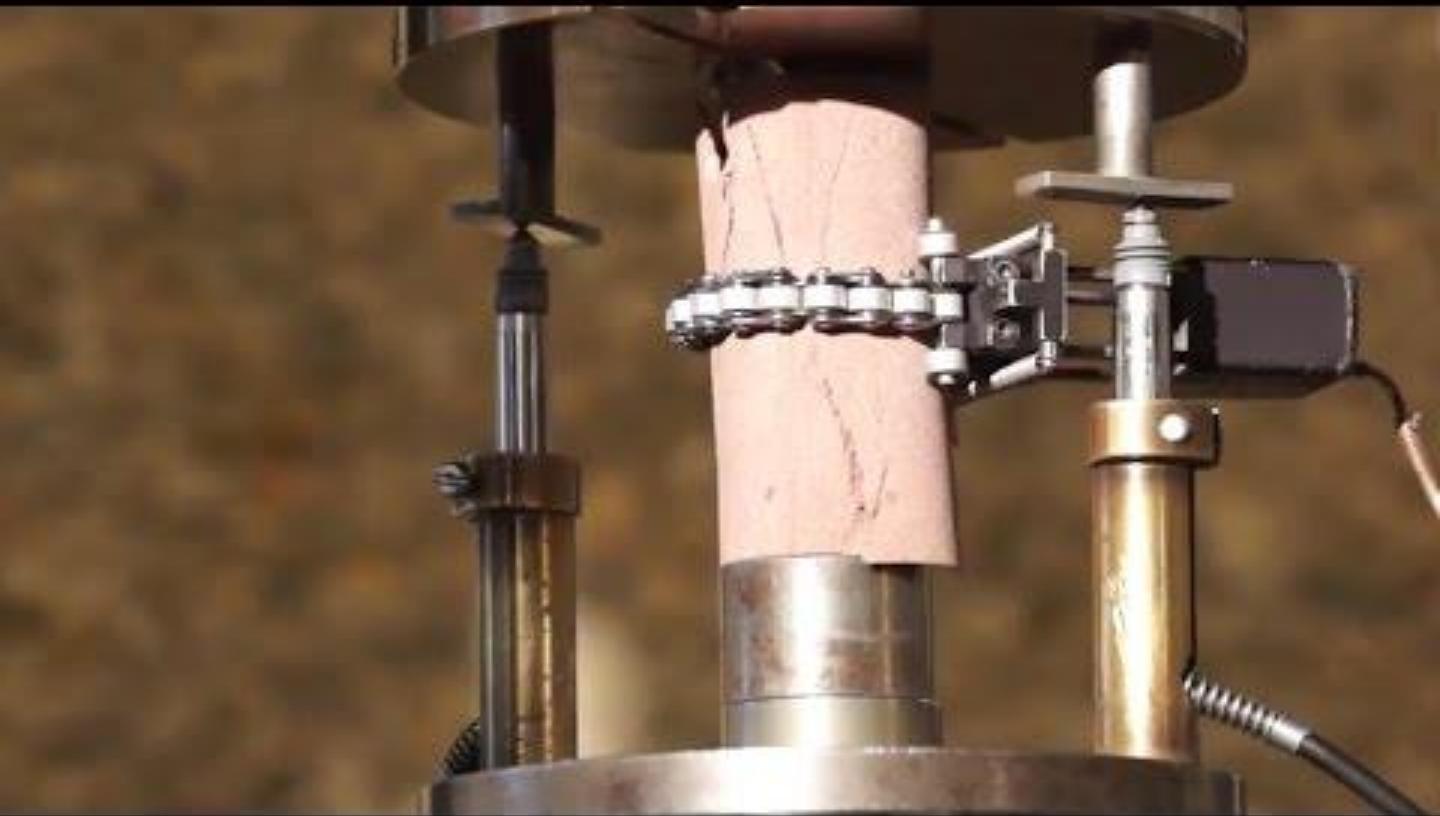
Part 3 Methods: NMR relaxation can be related to water potential



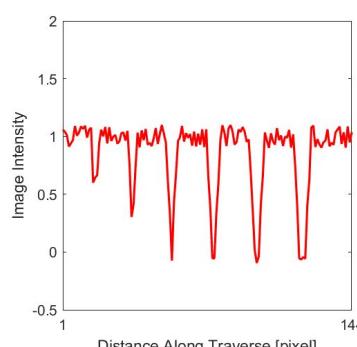
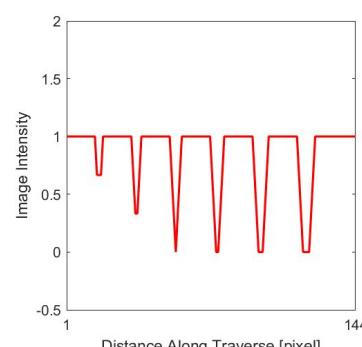
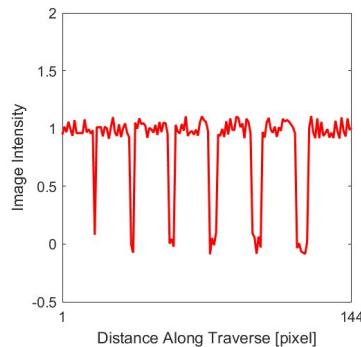
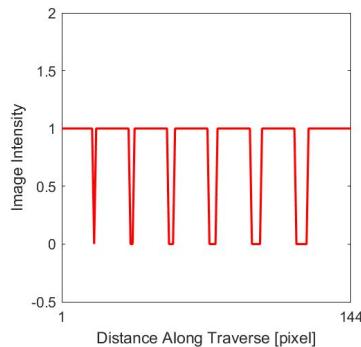
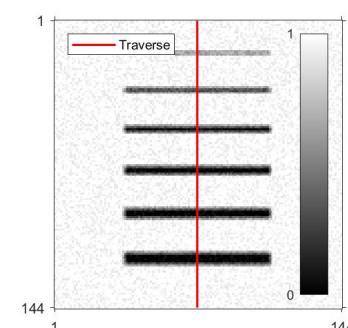
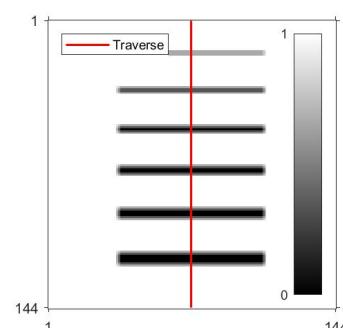
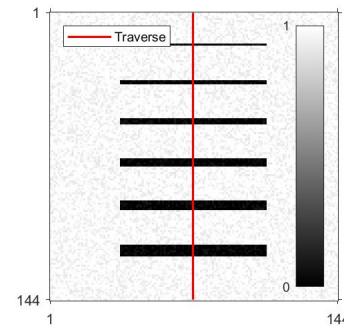
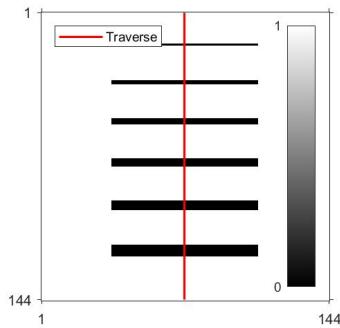
Motivation



Motivation



Methods: Generating Synthetic Fractures



Methods: Fracture Detection (The Hessian Method)

$$g(x) = ae^{-\frac{1}{2}\left(\frac{x-b}{c}\right)^2}$$

$$g'(x) = \left(-\frac{a}{c^2}\right)(x - b)e^{-\frac{1}{2}\left(\frac{x-b}{c}\right)^2}$$

$$g''(x) = \frac{a \left(x^2 - 2bx - c^2 + b^2\right) e^{-\frac{(x-b)^2}{2c^2}}}{c^4}$$

$$g(x, y) = A \exp \left[-\frac{1}{2} \left(\left(\frac{x - x_0}{c_x} \right)^2 + \left(\frac{y - y_0}{c_y} \right)^2 \right) \right]$$

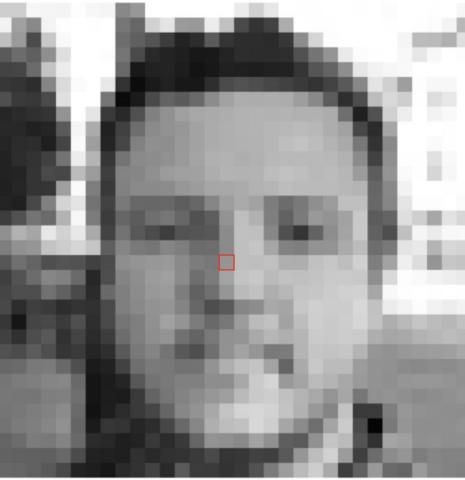
Methods: Computing The Hessian

$$H = \begin{bmatrix} I_{xx} & I_{xy} & I_{xz} \\ I_{yx} & I_{yy} & I_{yz} \\ I_{zx} & I_{zy} & I_{zz} \end{bmatrix}$$

$$I_{ij} = I(x, y, z) * (\alpha G_{ij}(x, y, z, s))$$

where $i, j = x, y, z$

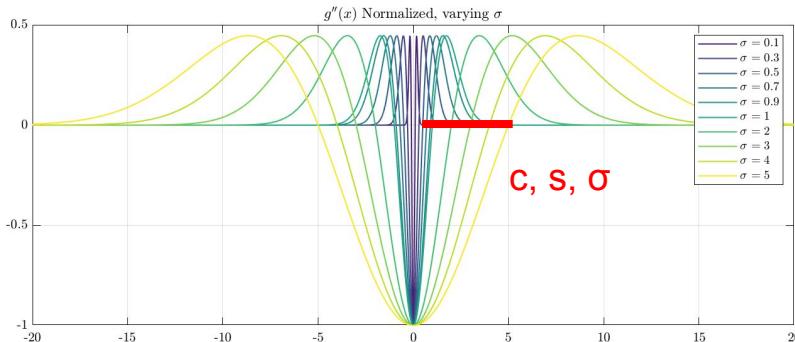
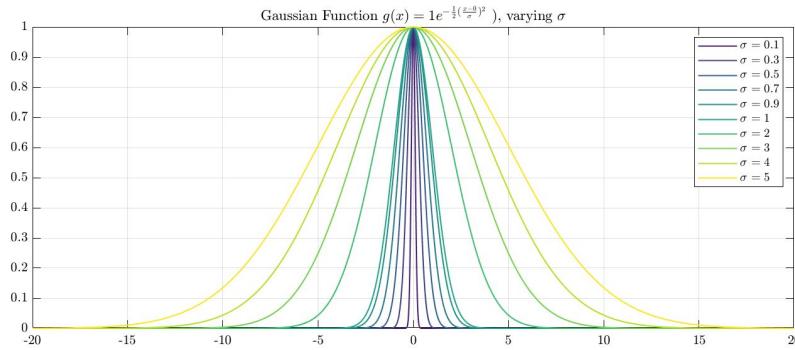
Methods: Computing The Hessian



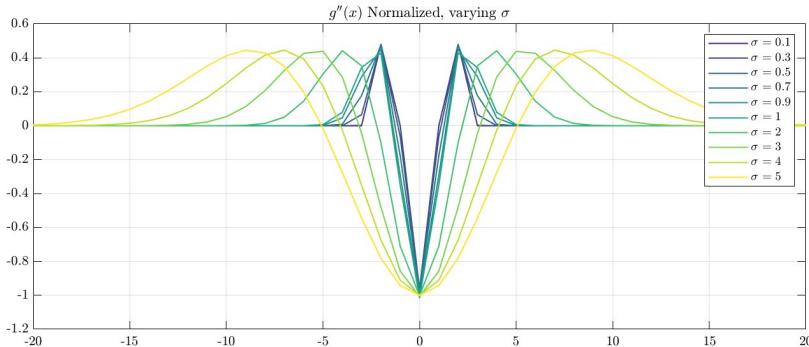
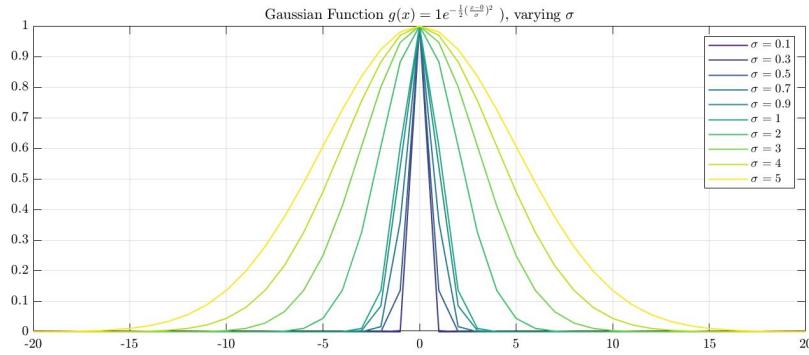
Let's walk through applying the following 3x3 **sharpen** kernel to the image of a face from above.

$$\left(\begin{array}{ccc} 0 & -1 & 0 \\ -1 & 5 & -1 \\ 0 & -1 & 0 \end{array} \right)$$

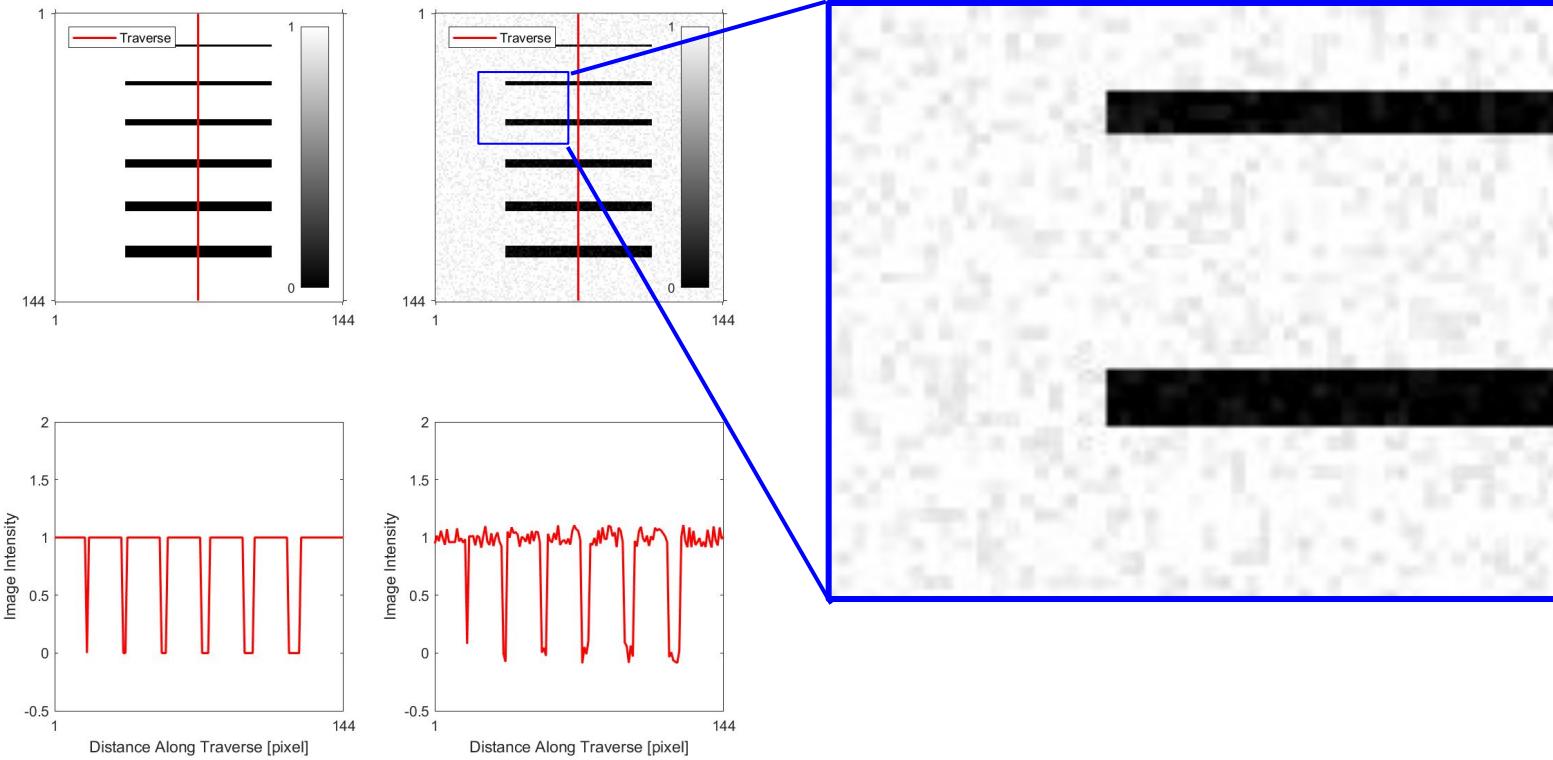
Methods: Continuous G_xx



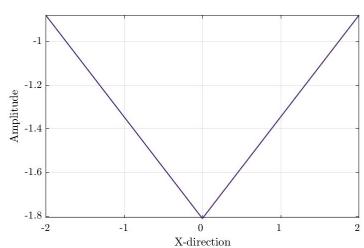
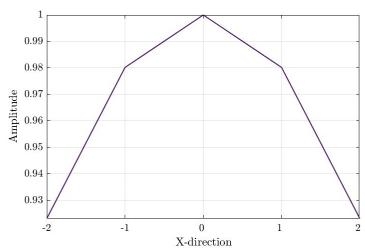
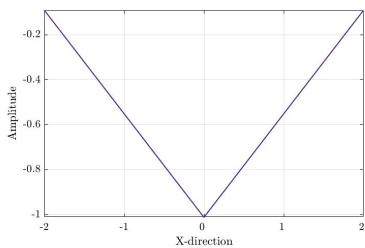
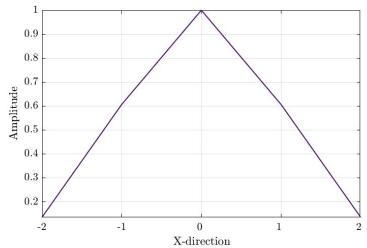
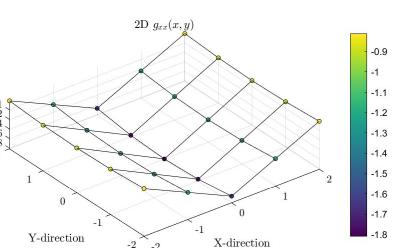
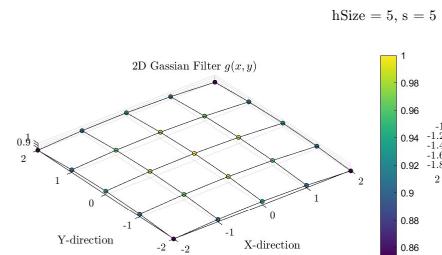
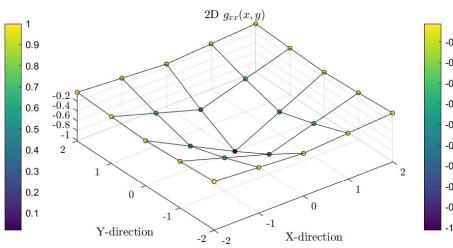
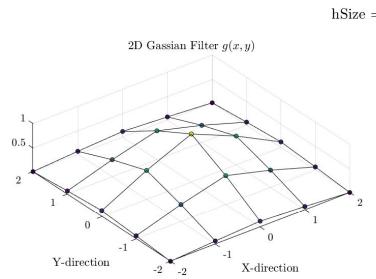
Methods: Discrete G_xx



Method: Images are discrete (and 3D)

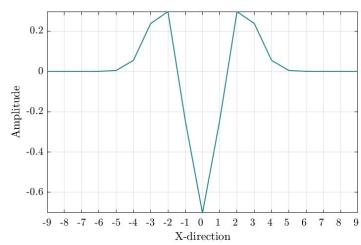
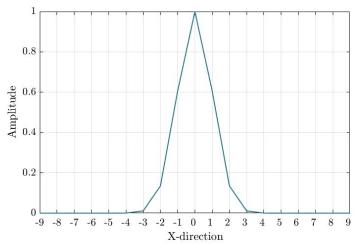
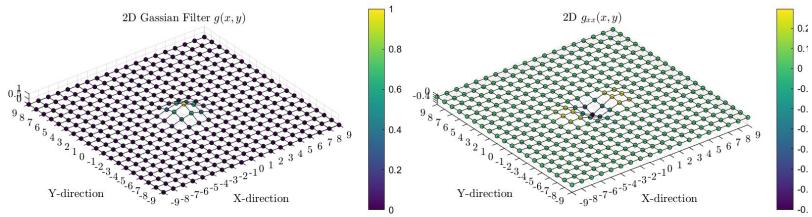


Methods: Discrete G_{xx} in 2D, shape is not appropriate

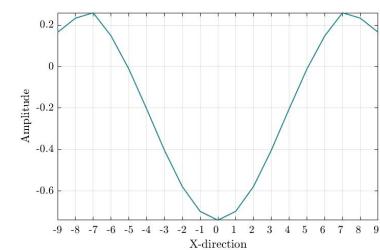
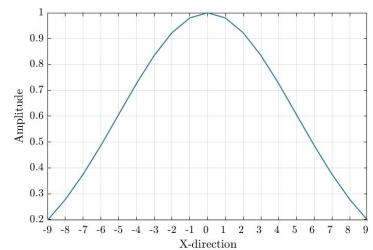
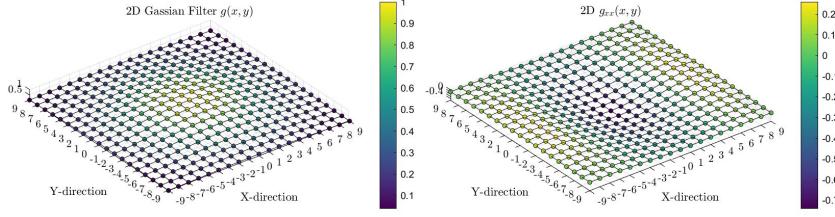


Methods: Discrete G_{xx} in 2D, shape looks good

hSize = 19, s = 1



hSize = 19, s = 5



Methods: Computing the Hessian

$$H = \begin{bmatrix} I_{xx} & I_{xy} & I_{xz} \\ I_{yx} & I_{yy} & I_{yz} \\ I_{zx} & I_{zy} & I_{zz} \end{bmatrix}$$

$$I_{ij} = I(x, y, z) * (\alpha G_{ij}(x, y, z, s))$$

where $i, j = x, y, z$

Methods: How is this useful?

2D		3D			Orientation pattern
λ_1	λ_2	λ_1	λ_2	λ_3	Noisy, no orientation pattern
N	N	N	N	N	Plate-like structure (dark)*
		L	L	H+	Plate-like structure (bright)
		L	L	H-	Tubular structure (dark)
L	H+	L	H+	H+	Tubular structure (bright)
L	H-	L	H-	H-	Blob-like structure (dark)
H+	H+	H+	H+	H+	Blob-like structure (bright)
H-	H-	H-	H-	H-	

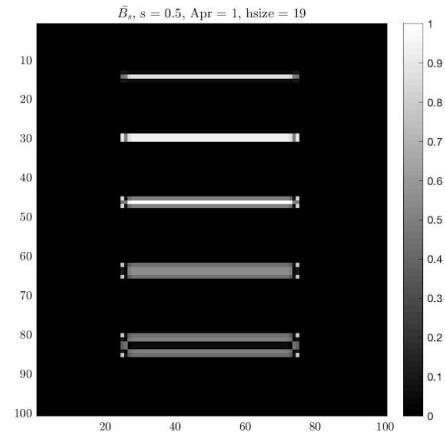
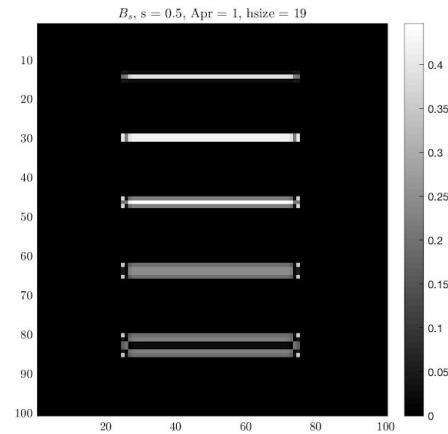
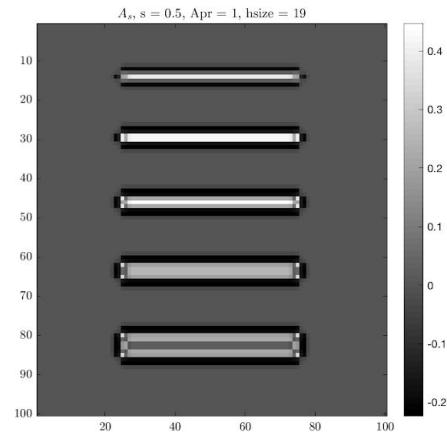
$|\lambda_1| \leq |\lambda_2| \leq |\lambda_3|$. H=high value; L=low value; N=noise; +/-=positive or negative eigenvalue. Dark=dark feature on a bright background. Bright=bright feature on a dark background. The fracture representation we are interested in is indicated with a *. After [Frangi et al. \(1998\)](#).

Methods: Computing the “Hessian”

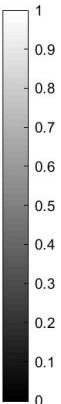
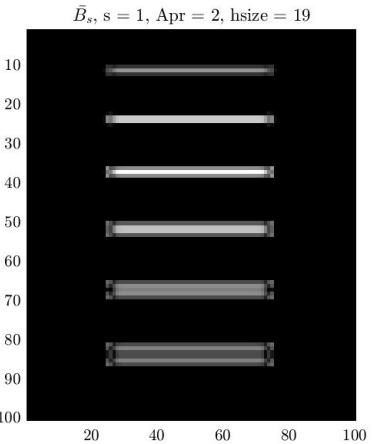
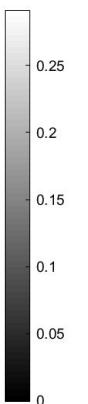
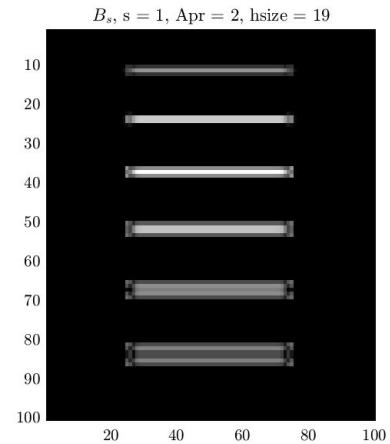
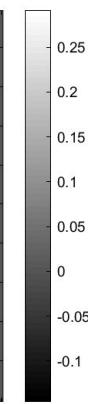
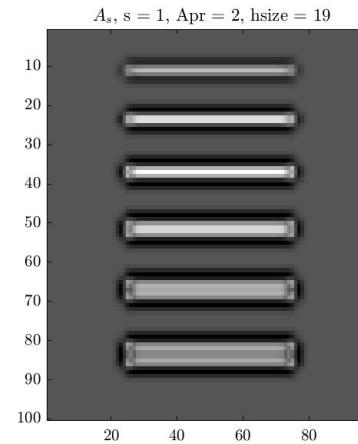
Algorithm

- For each s (scaling factor $\sim \frac{1}{2}$ aperture of interest) in our case [1,2,3,4,6]
- Compute all Hessian elements by convolution for each s
- Loop over all voxels and compute the eigenvalues and eigenvectors
- Sort the eigenvalues (from smallest to largest)
- *For each voxel compute $A_s = \lambda_3 - |\lambda_2| - |\lambda_1|$*
- *Compute $B_s = A_s$ if $A_s > 0$, or 0 otherwise*
- *Normalize B_s such that $0 < B_s < 1$*
- Determine C_s such that $C_s = 1$ if $A_x > 1 - tol$, or 0 otherwise
- The sum up all of C_s for each s value

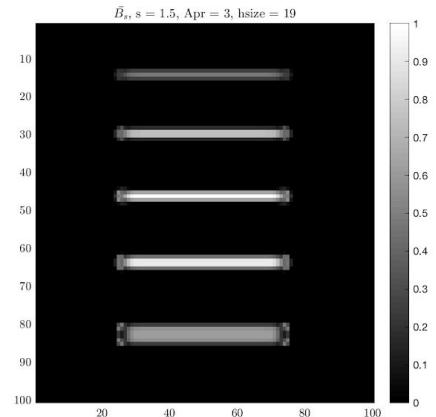
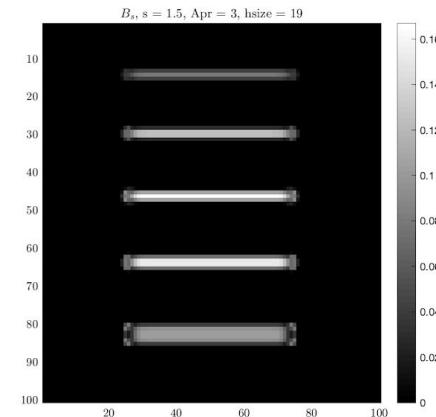
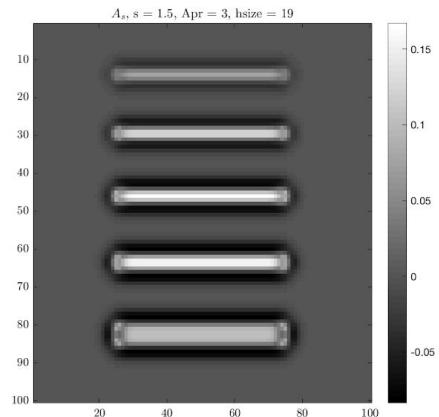
Method: the values for A_s, B_s, and Normalized B_s



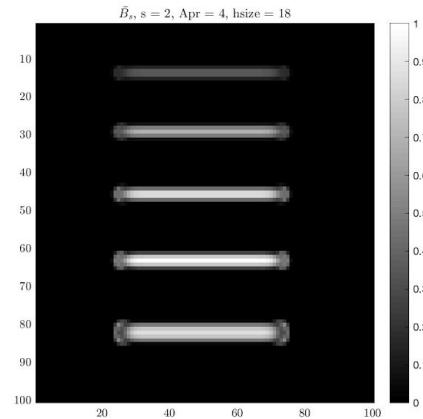
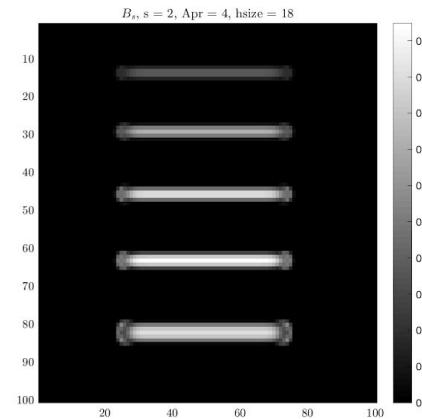
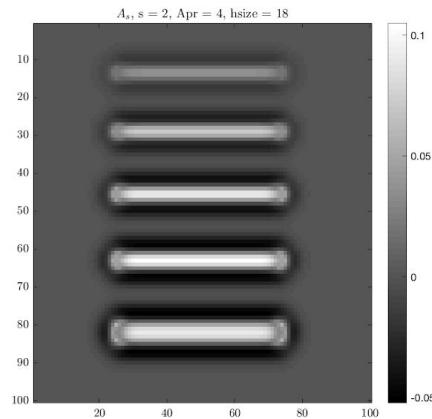
Method: the values for A_s, B_s, and Normalized B_s



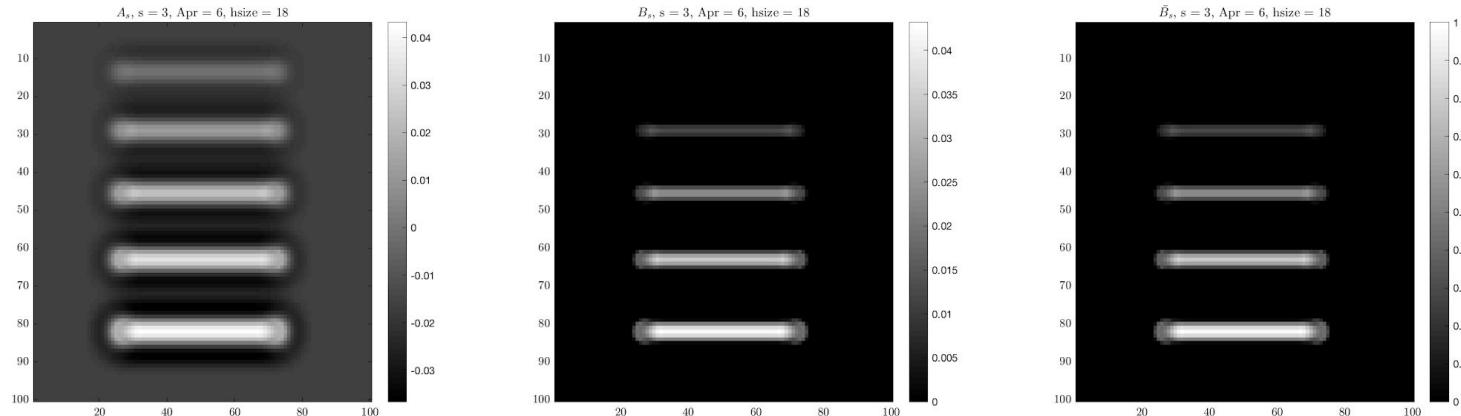
Method: the values for A_s, B_s, and Normalized B_s



Method: the values for A_s, B_s, and Normalized B_s

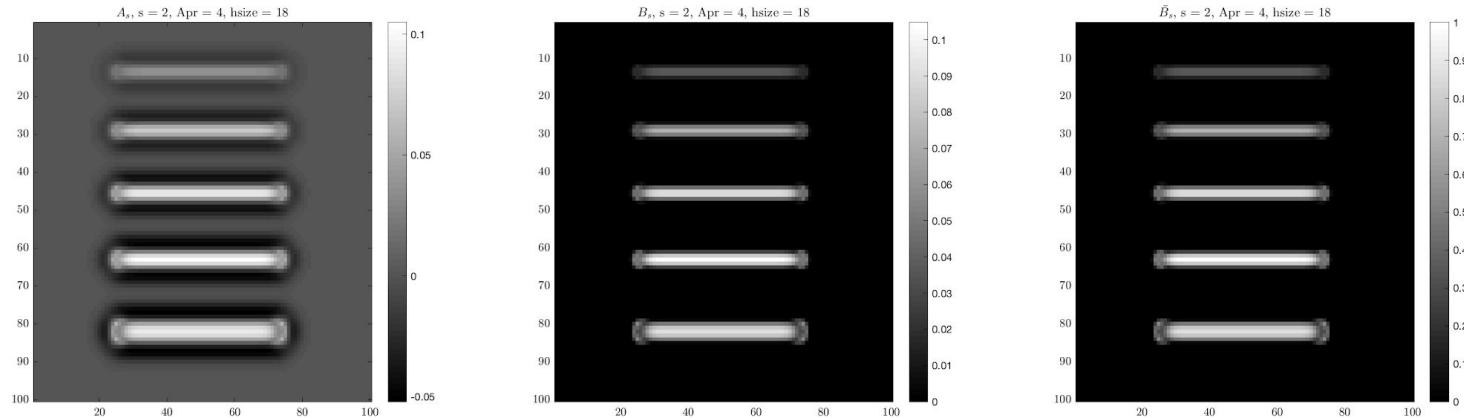


Method: the values for A_s, B_s, and Normalized B_s

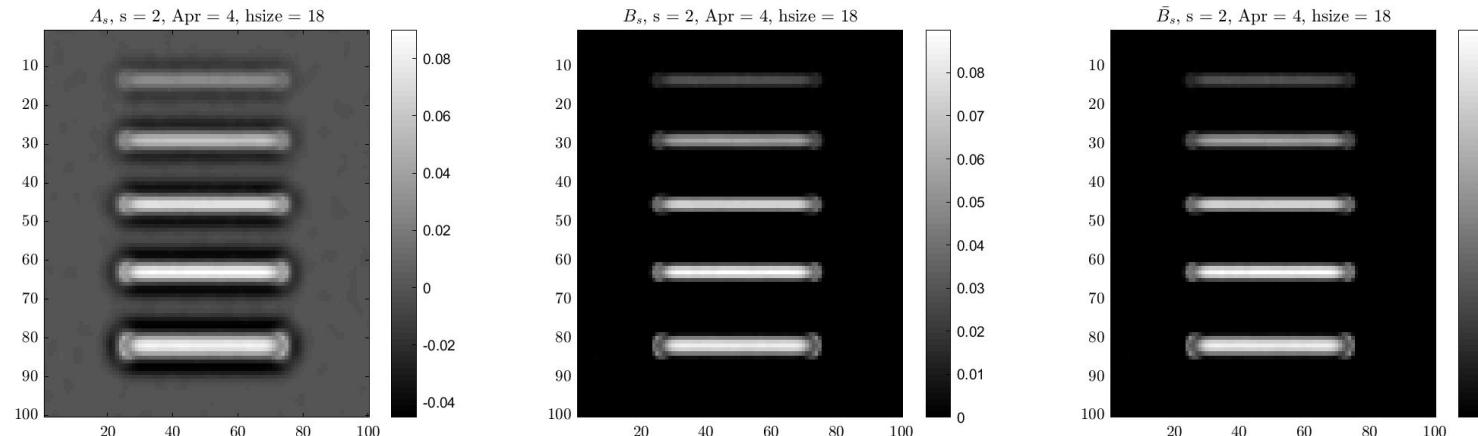


Method: the values for A_s, B_s, and Normalized B_s

Crisp
image



Blurry
and
noisy
image

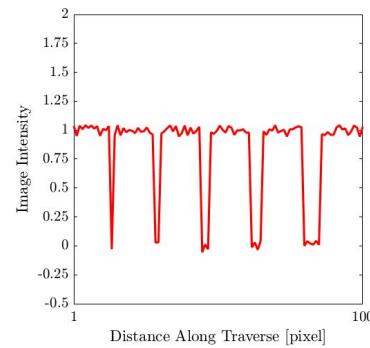
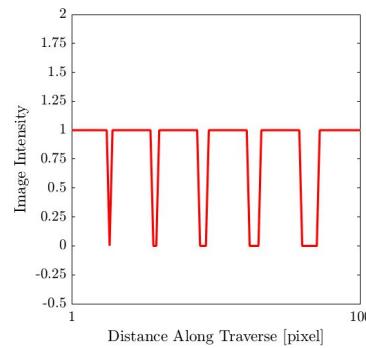
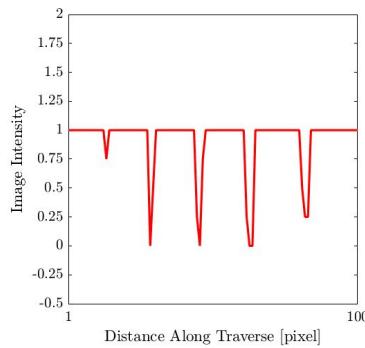
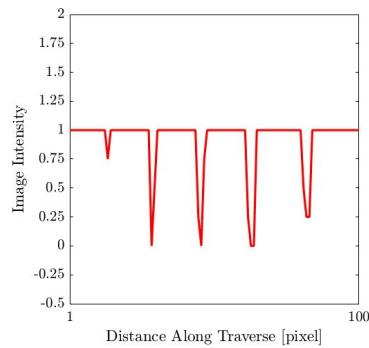
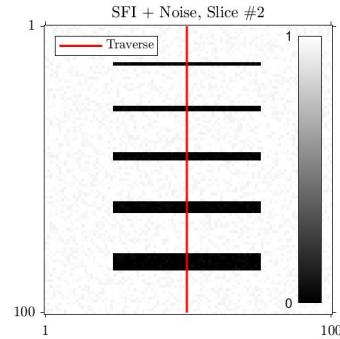
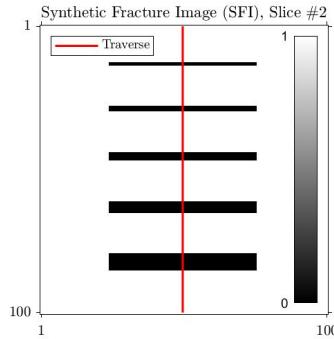
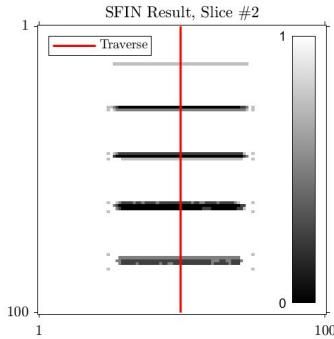
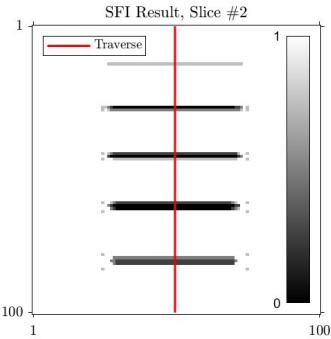


Methods: Computing the “Hessian”

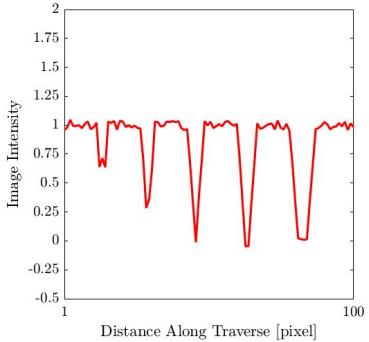
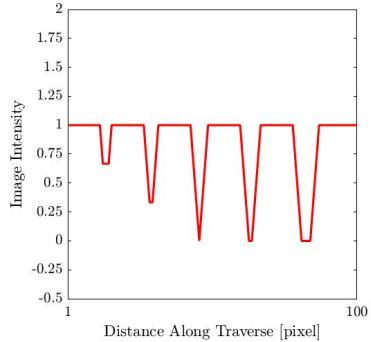
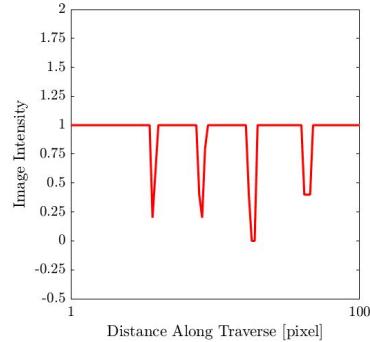
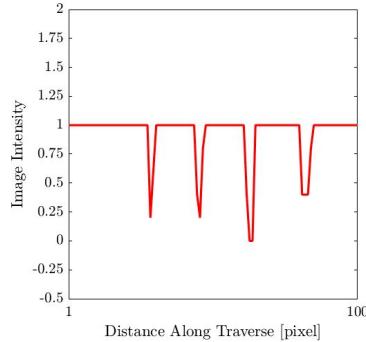
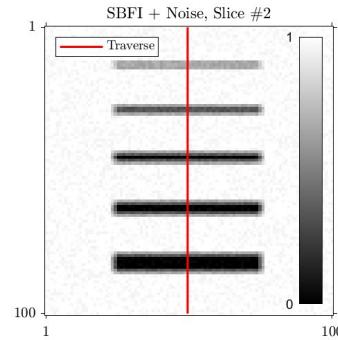
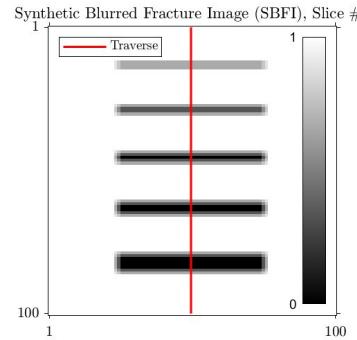
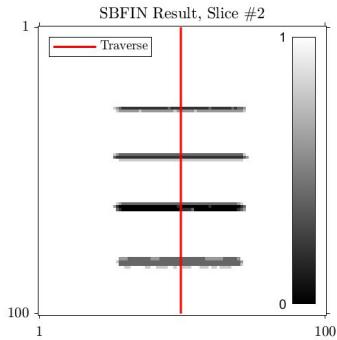
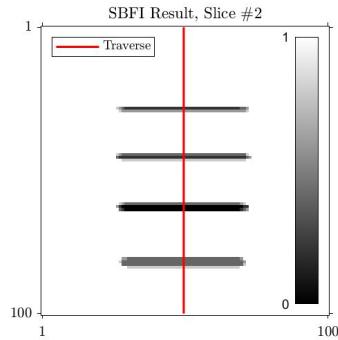
Algorithm

- For each s (scaling factor $\sim \frac{1}{2}$ aperture of interest) in our case [1,2,3,4,6]
- Compute all Hessian elements by convolution for each s
- Loop over all voxels and compute the eigenvalues and eigenvectors
- Sort the eigenvalues (from smallest to largest)
- For each voxel compute $A_s = \lambda_3 - |\lambda_2| - |\lambda_1|$
- Compute $B_s = A_s$ if $A_s > 0$, or 0 otherwise
- Normalize B_s such that $0 < B_s < 1$
- Determine C_s such that $C_s = 1$ if $A_x > 1 - tol$, or 0 otherwise
- The sum up all of C_s for each s value

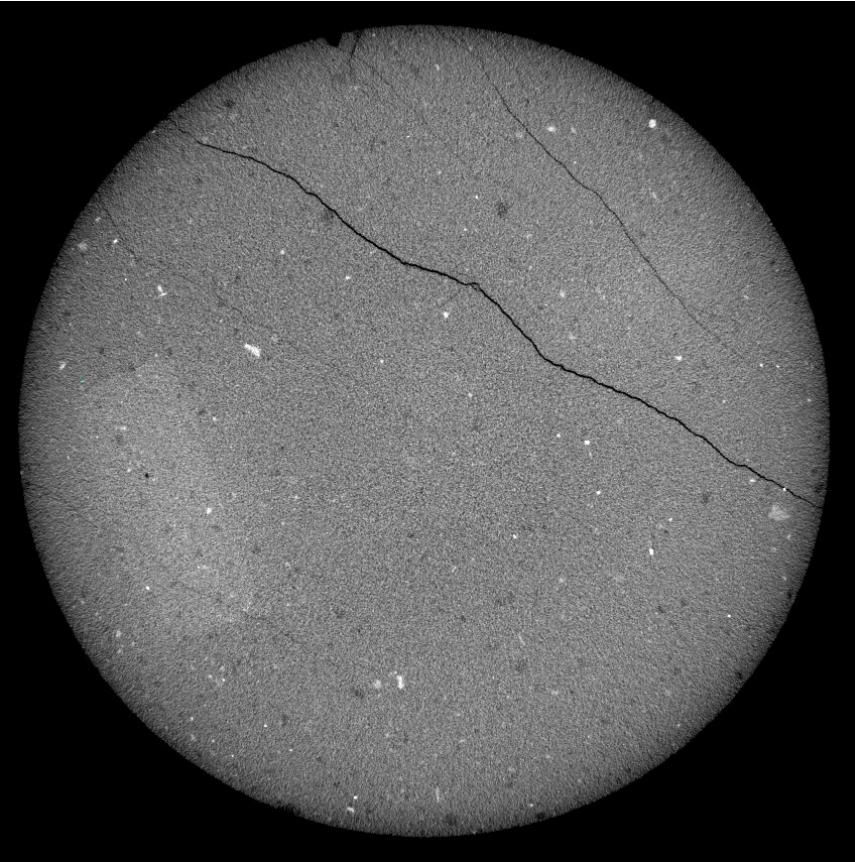
Methods: Hessian Result (crisp images)



Methods: Hessian Result (blurry images)



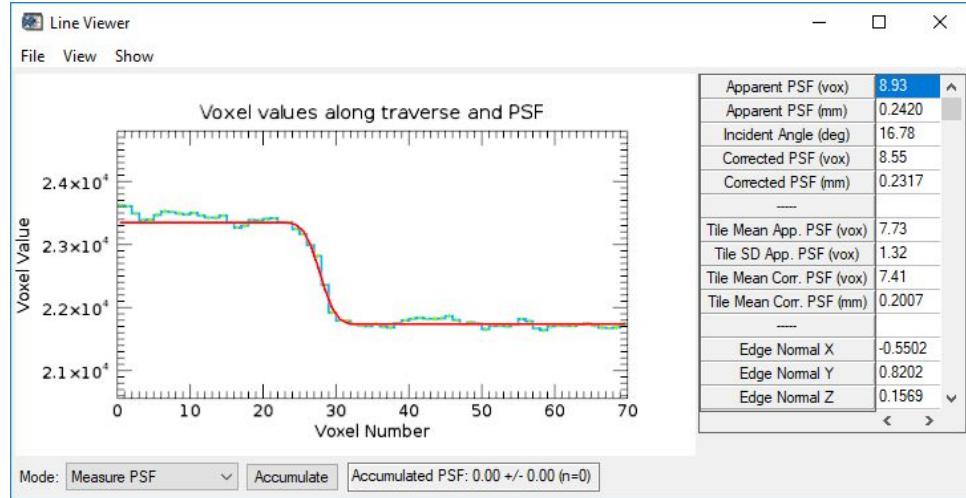
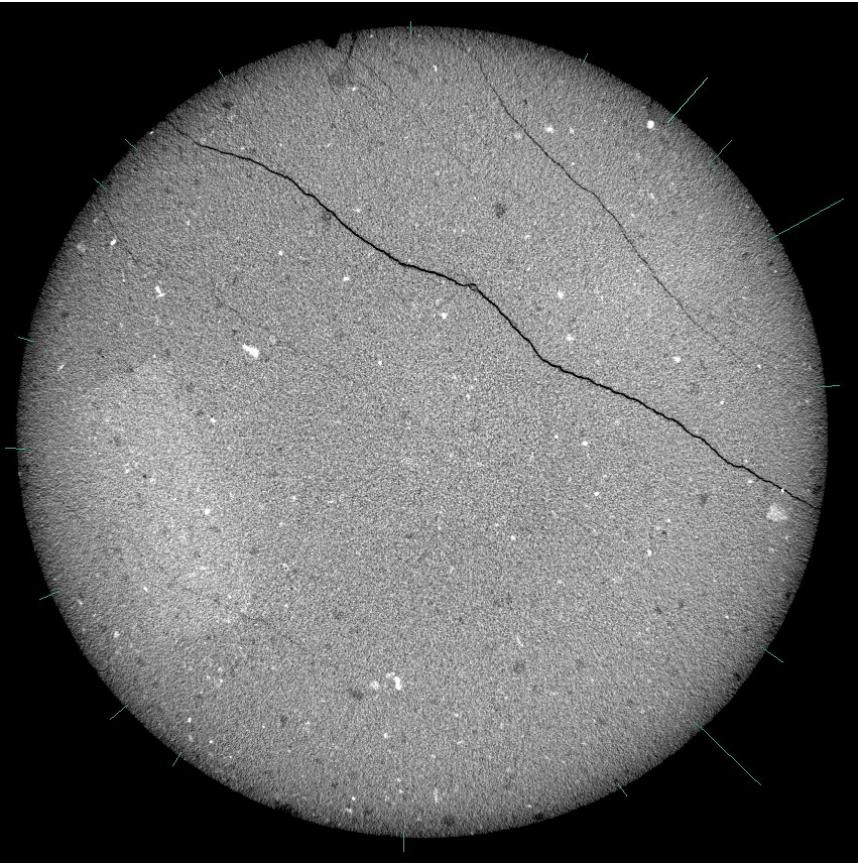
Materials



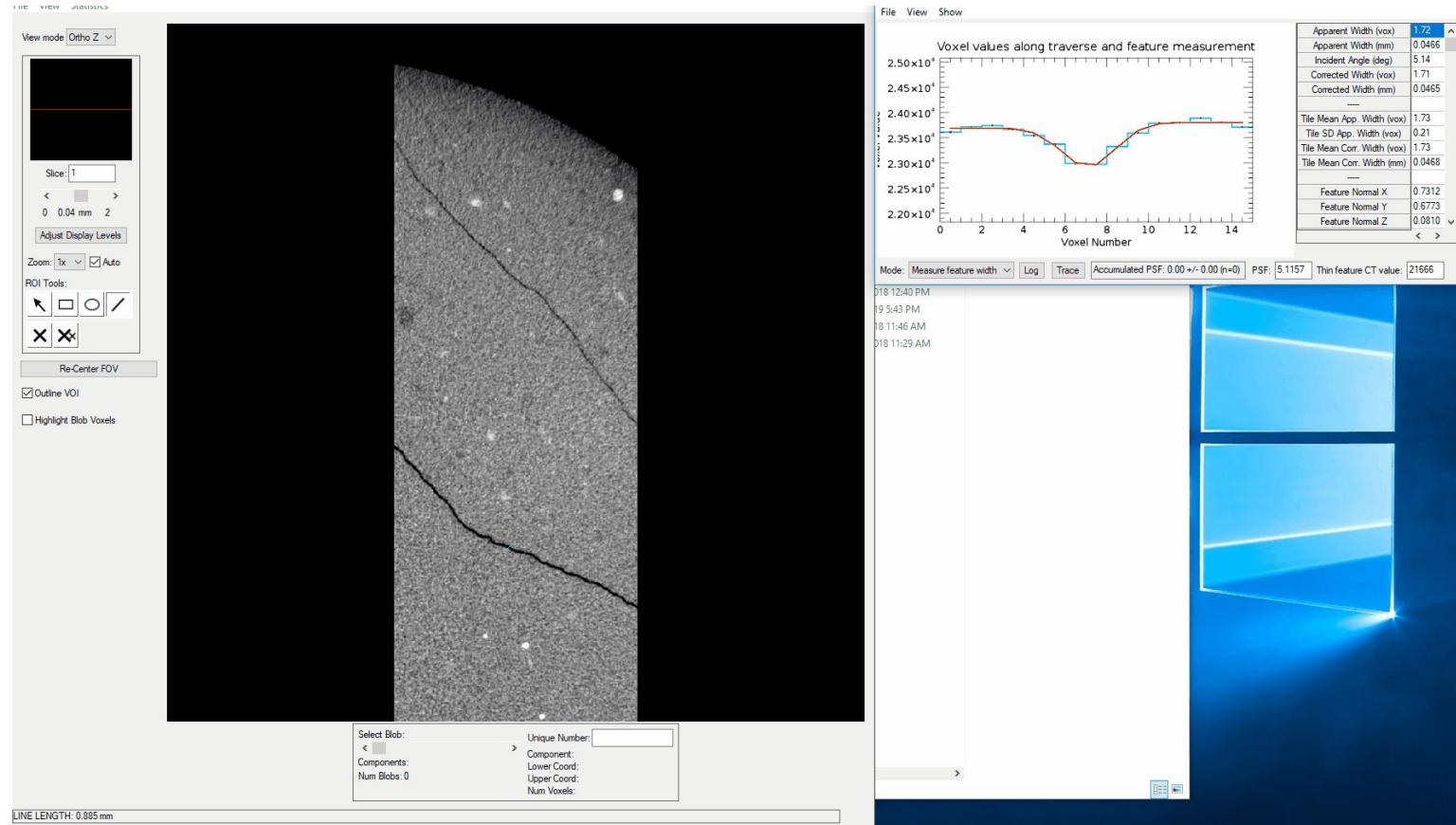
Argillite mudrock

Core hand-drilled from creekbed

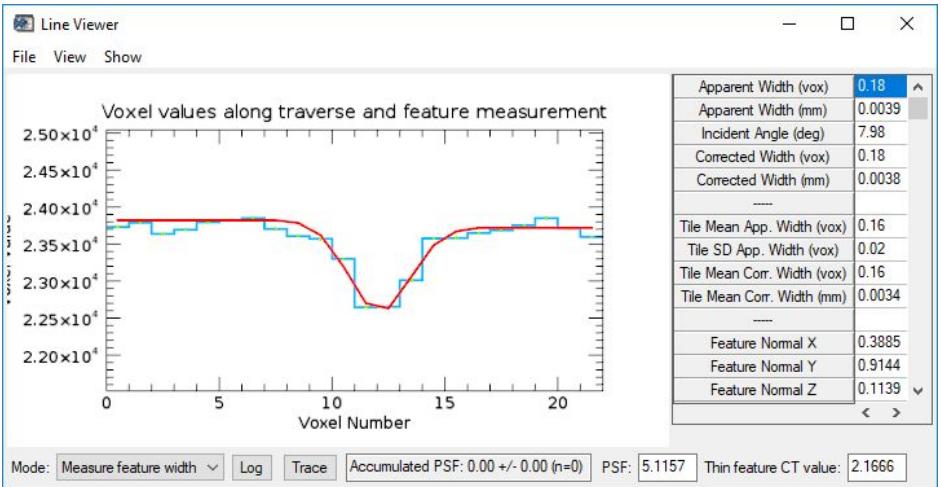
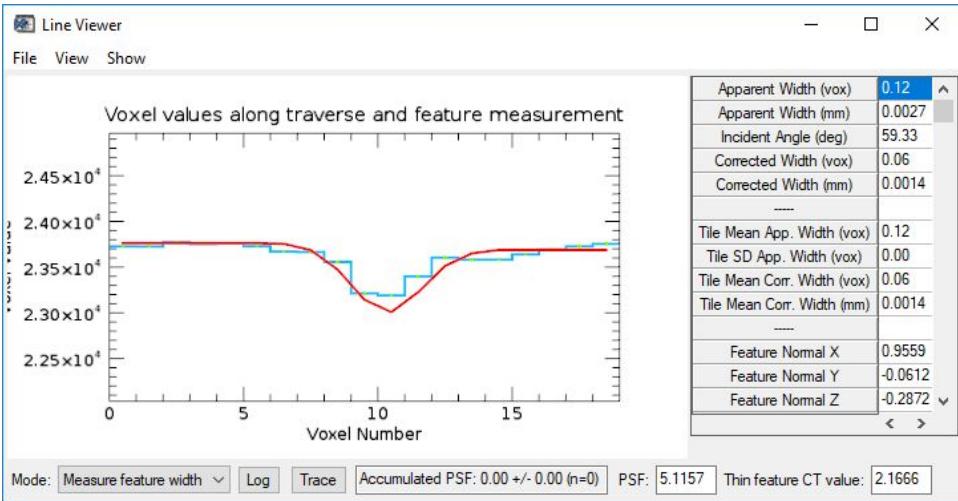
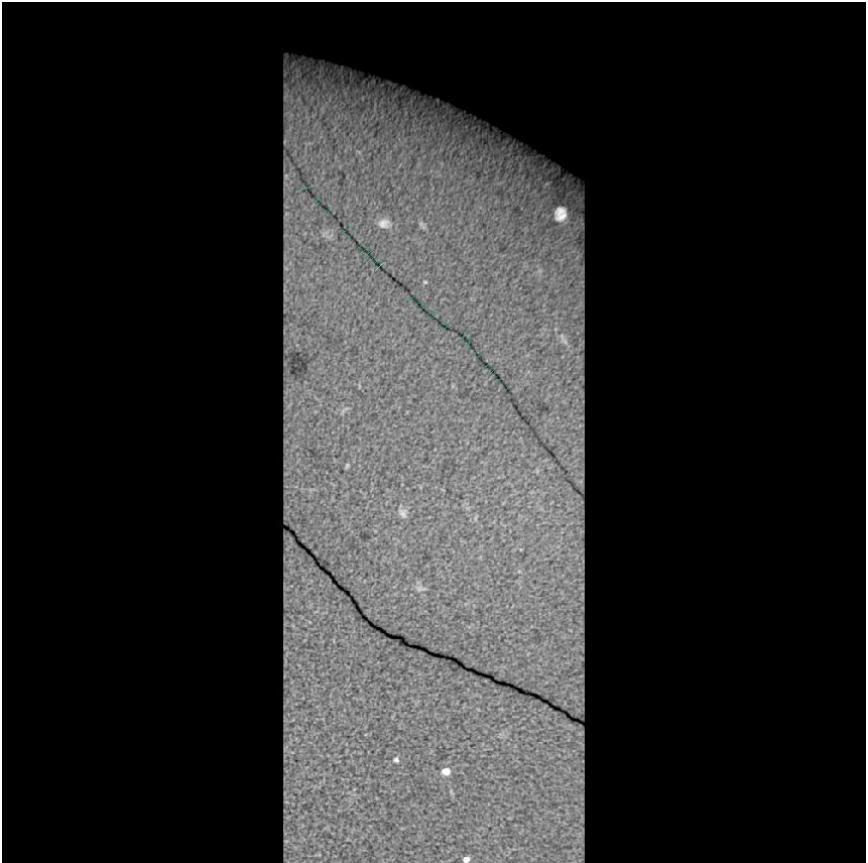
Point Spread Function in Blob3D



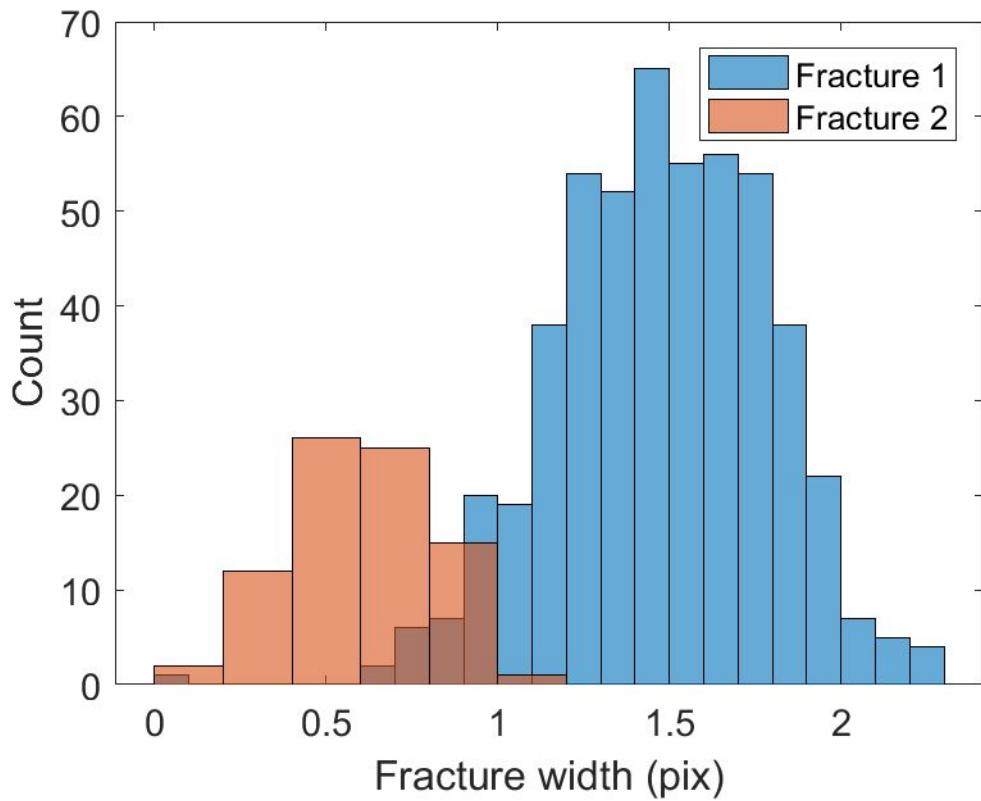
No segmentation: Blob3D



No segmentation: Blob3D



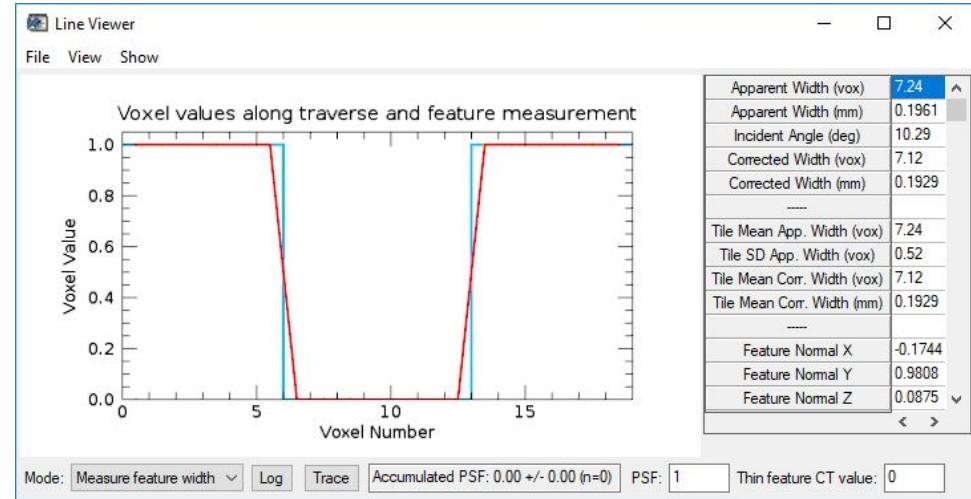
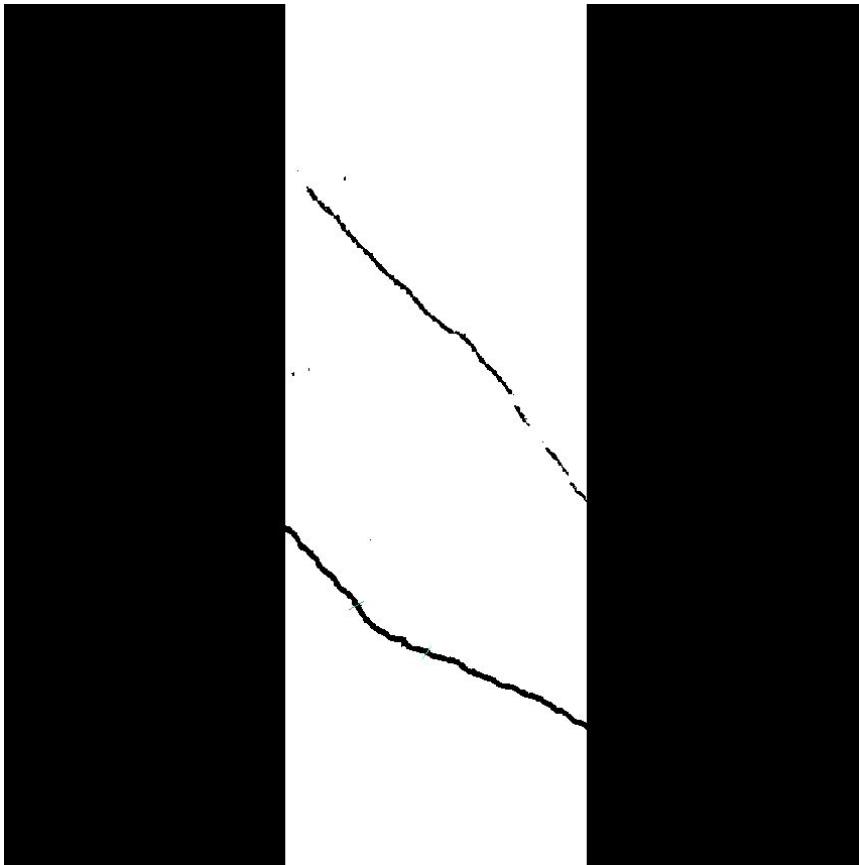
No segmentation: Blob3D



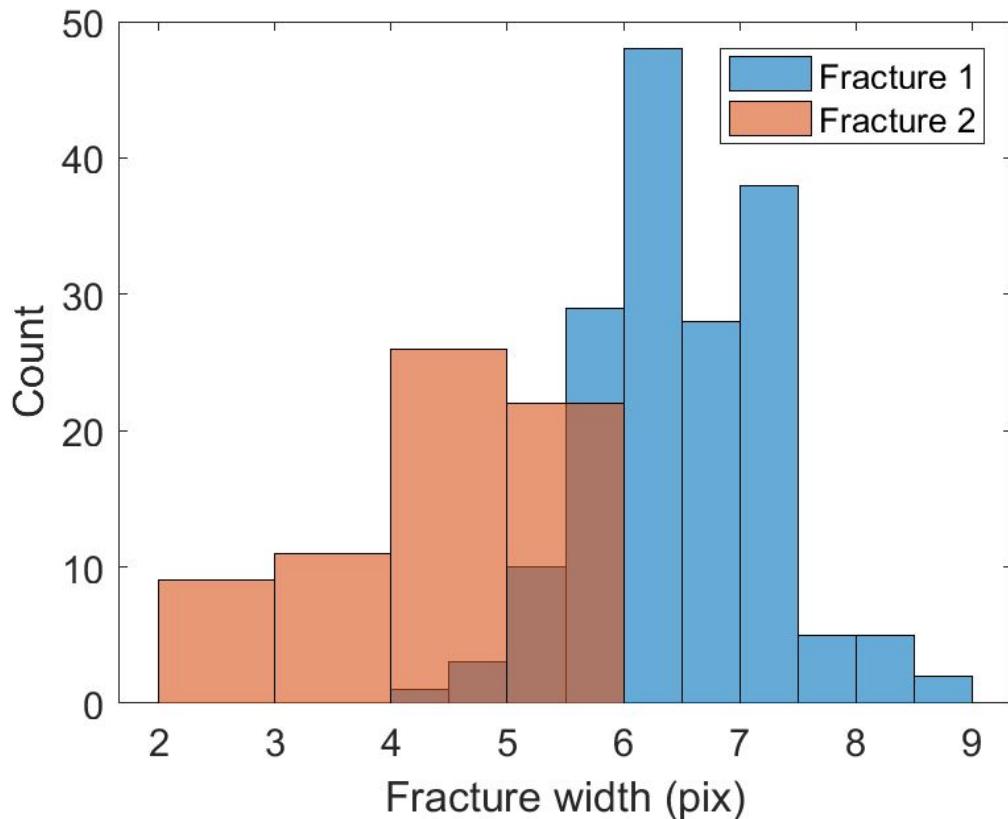
Fracture 1 mean: 1.5+0.3

Fracture 2 mean: 0.6+0.2

Trainable segmentation: Blob3D



Trainable segmentation: Blob3D

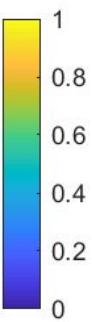
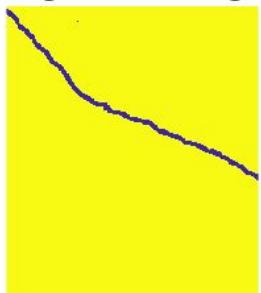


Fracture 1 mean: 6.5+0.8

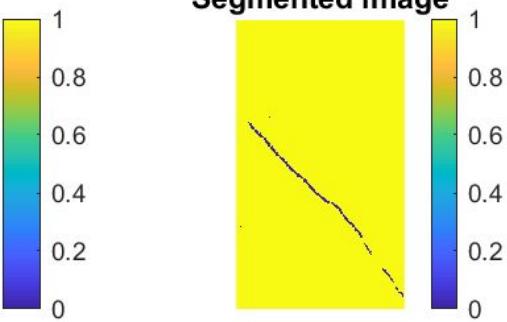
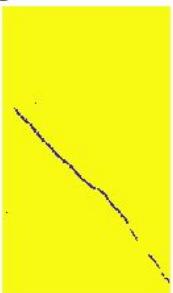
Fracture 2 mean: 4.5+1.0

Trainable segmentation: Distance function

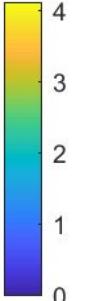
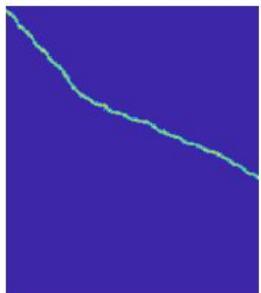
Segmented image



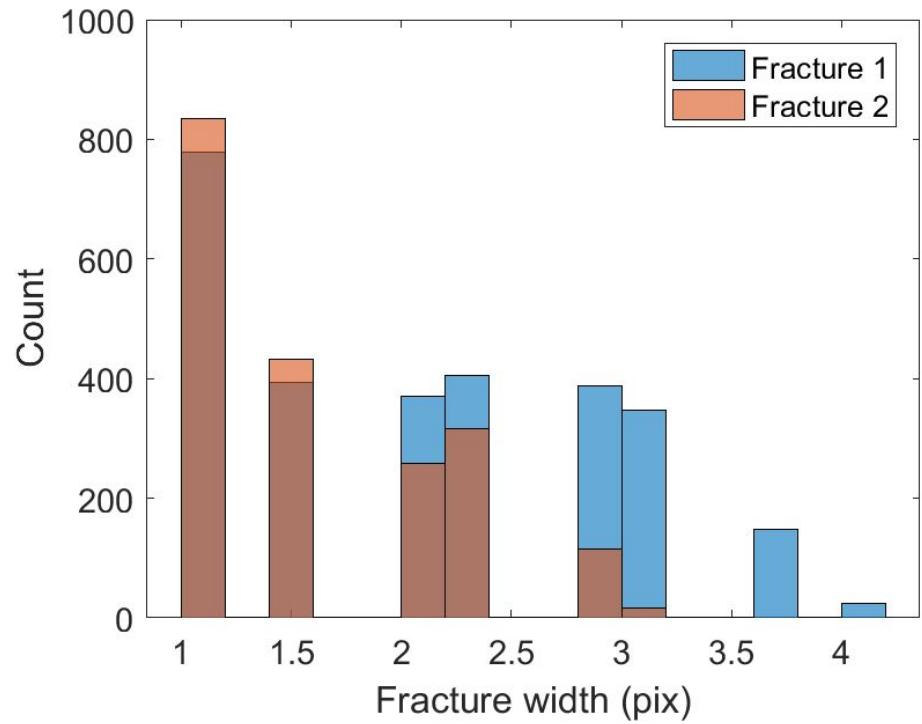
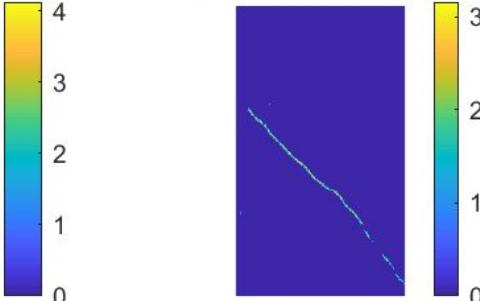
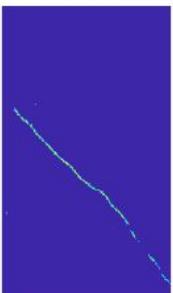
Segmented image



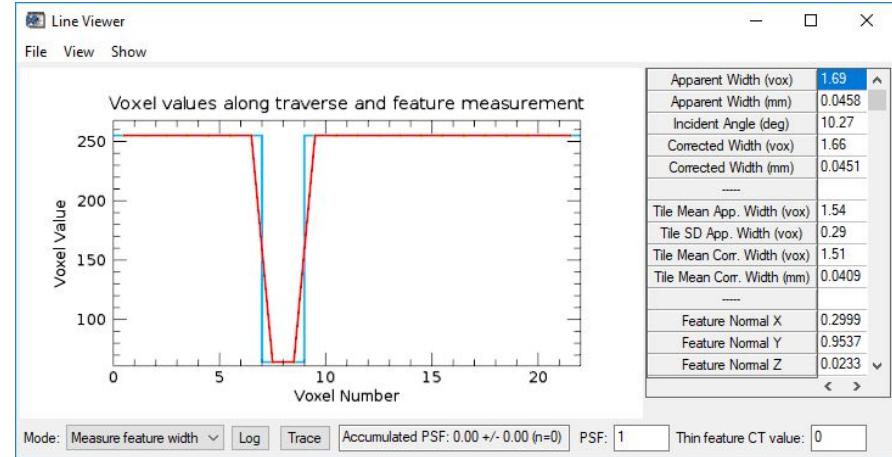
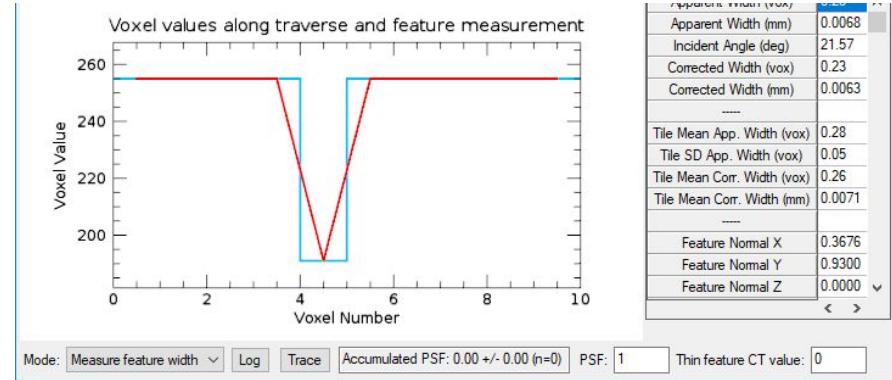
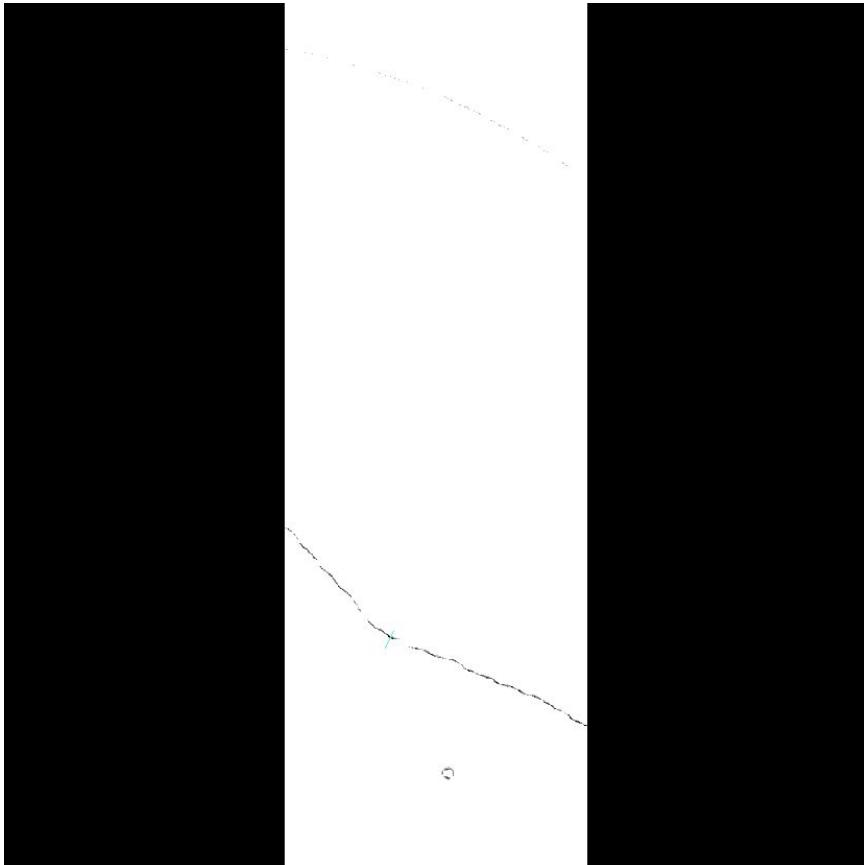
Distance transform



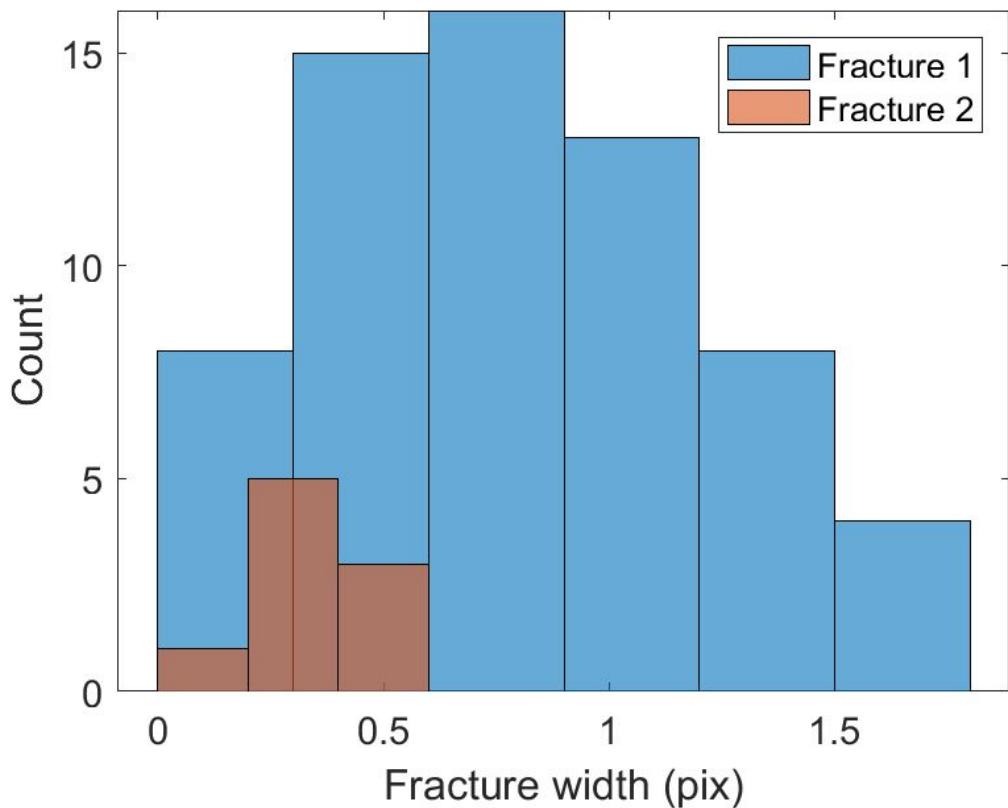
Distance transform



Hessian segmentation: distance function



Hessian segmentation: distance function



Fracture 1 mean: 0.8+-0.4

Fracture 2 mean: 0.3+-0.1

Hessian segmentation: distance function

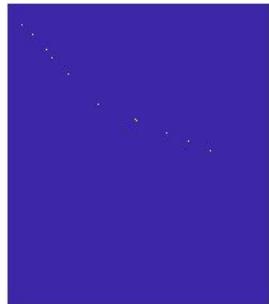
Segmented image



Segmented image



Distance transform



Distance transform

