Spatial orientation of microstructural features from 3D images of biological soft tissues

Aim

To develop an algorithm for determination of fiber orientations from 3D image datasets and present orientation data as a 3D vector field (in Mathematica)

Motivation

- Collagen fiber orientation provides an indication of the <u>functionality</u> and <u>performance</u> of the tissue. For example – skin wrinkling with age can be predicted using changes in the fiber density and orientations.
- A potential biomarker for <u>diagnosing</u> diseased tissues, assessing response to treatments, localizing injuries or monitoring engineered tissue development. For example – Osteoarthritis can be identified in early stages by detecting damage in collagen network.

Background

Orientation of fibers in 2D images are well studied and successful methods have been developed. Usually, these involve Fourier transforms, Hough transforms or Gradient filters.

3D orientation determination is a comparatively new area and not so well studied. Some of the existing methods include:

- Gaussian Orientation Space filters
- Fourier Transform of SHG images (Filter Bank method)
- Principal Component Analysis (Structure tensor, moment of inertia tensor)
- Weighted Vector Summation

Weighted vector summation is one of the latest methods (2015) and outperforms most other methods according to its author. This is used as the comparison point.

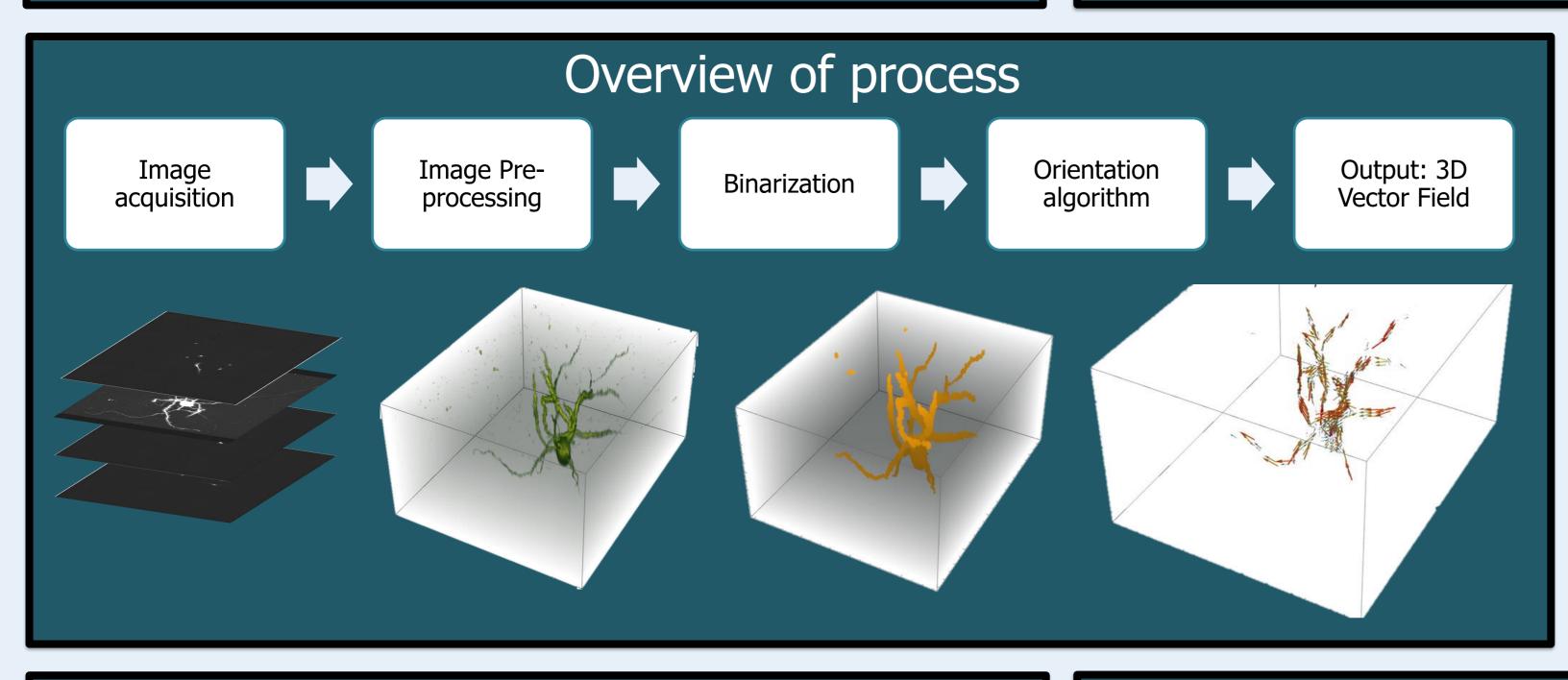


Image processing Noise filtering 3D Image construction Edge Detection: Canny, Sobel filters Binarization: Otsu's Global Thresholding, Hysteresis III <l>

III: Canny edge

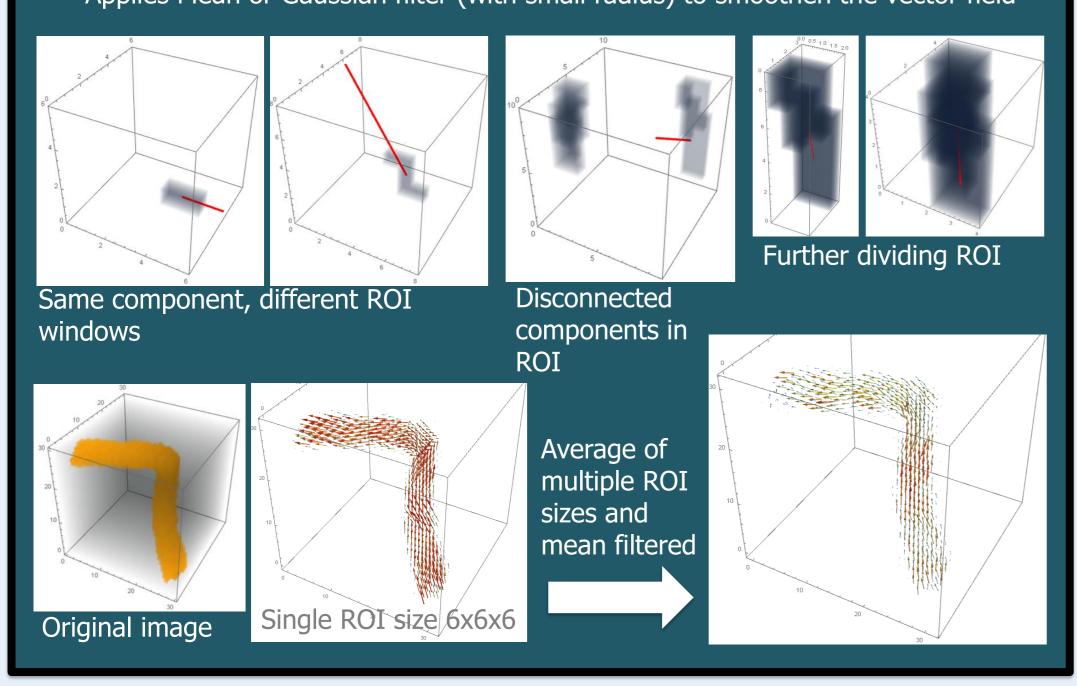
detection

Proposed Algorithm

- Based on Principal Component analysis of Moment of Inertia Tensor
- Binarized image divided into smaller sub volumes (Regions of Interests ROIs) and orientation found within each ROI

Features:

- Averages vector fields obtained from different ROI sizes (multiple passes)
- Checks for disconnected components and divides ROI further if needed
- Ignores fully white and fully black ROIs and perfectly cubic components in ROIs
- Does not iterate through all pixels. Assigns same vector to white pixels in a ROI
- Applies Mean or Gaussian filter (with small radius) to smoothen the vector field



Conclusion

 The proposed algorithm outperformed weighted vector summation both in terms of speed performance and accuracy *

Further research/improvement opportunities:

- Resultant 3D vector field could be used in a FEA study to fully appreciate a use case and refine algorithm.
- Optimizing current implementation for performance through parallelization

 Application of algorithms are learner data and a constant and
- Application of algorithm on larger datasets (compare with known orientations)
- Integration with other software systems for improved workflow.
- * The implementation of vector summation method in this study did not use parallelization (as intended by its author) and hence its speed performance can be improved drastically

Analysis/Results Accuracy vs Image Volume Speed performance vs. Image Volume 3D Weighted **Vector Summation** 3D Weighted **Vector Summation** Proposed Proposed algorithm algorithm Image Volume (voxels) Image Volume (voxels) Speed vs. Fiber Density Accuracy vs. Fibre Density Number of fibres, (fibre density) Number of fibres, (fibre density) Speed performance vs. Iteration Accuracy vs Fiber thickness factor Iteration factor Fiber thickness, % Example computer generated images used for analysis

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