

¹ Cardiotensor: A Python Library for Orientation Analysis and Tractography in 3D Cardiac Imaging

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¹⁰ Summary

¹¹ Understanding the architecture of the human heart requires analysis of its microstructural organization across scales. With the advent of high-resolution imaging techniques such as synchrotron-based tomography, it has become possible to visualize entire hearts at micron-scale resolution. However, translating these large, complex volumetric datasets into interpretable, quantitative descriptors of cardiac organization remains a major challenge. Here we present ¹⁴ cardiotensor, an open-source Python package designed to quantify 3D cardiomyocyte orientation ¹⁵ in whole- or partial-heart imaging datasets. It provides efficient, scalable implementations of ¹⁶ structure tensor analysis, enabling extraction of directional metrics such as helical angle (HA), ¹⁷ intrusion angle (IA), and fractional anisotropy (FA). The package supports datasets reaching ¹⁸ teravoxel-scale and is optimized for high-performance computing environments, including ¹⁹ parallel and chunk-based processing pipelines. In addition, cardiotensor includes tractography ²⁰ functionality to reconstruct continuous cardiomyocyte trajectories. This enables multi-scale ²¹ myoaggregate visualization down to the myocyte level, depending on resolution. These ²² capabilities enable detailed structural mapping of cardiac tissue, supporting the assessment of ²³ anatomical continuity and regional organization.

²⁶ Statement of Need

²⁷ Despite major advances in high-resolution 3D imaging, there is a lack of open-source tools to ²⁸ analyze cardiomyocyte orientation in large volumetric datasets. Most established frameworks ²⁹ were developed for diffusion tensor MRI (DT-MRI), where orientation is inferred from local ³⁰ diffusion of water. Examples include MRtrix3 ([Tournier et al., 2019](#)), DIPY ([Garyfallidis et al., 2014](#)), and DSI Studio ([Yeh, 2025](#)). While powerful for diffusion-based neuro and cardiac ³¹ applications ([Mekkaoui et al., 2017](#)), these packages are not designed to handle direct image- ³² gradient-based orientation estimation or the teravoxel-scale datasets produced by synchrotron ³³ tomography, micro-CT, or 3D optical microscopy.

³⁵ For non-diffusion imaging modalities, such as micro-CT ([Reichardt et al., 2020](#)), optical ³⁶ microscopy ([Dileep et al., 2023; Garcia-Canadilla et al., 2022](#)), and synchrotron tomography ³⁷ ([Brunet et al., 2024; Dejea et al., 2019](#)), researchers have historically relied on custom structure ³⁸ tensor implementations to estimate myoaggregate orientation directly from image intensity ³⁹ gradients. However, most of these are in-house codes, often unpublished or not scalable. ⁴⁰ Existing tools like OrientationJ (Fiji) and OrientationPy (Python) enable 2D and 3D structure ⁴¹ tensor analysis ([Navaee et al., 2023](#)), but are not optimized for teravoxel-scale datasets, do not

42 compute classical cardiac microstructure descriptors such as HA and IA, and do not support
 43 tractography for myoaggregate orientation mapping.

44 Cardiotensor addresses this gap by providing an open-source Python package specifically
 45 tailored to structure tensor analysis of large cardiac volumes. Rather than relying on diffusion
 46 modeling, cardiotensor infers tissue orientation directly from image intensity gradients, making
 47 it applicable across a wide range of modalities and scales. Previous studies have demonstrated
 48 strong agreement between structure tensor-based orientation and DT-MRI-derived metrics
 49 when applied to the same human hearts (Teh et al., 2016). The package supports full pipelines
 50 from raw image stacks to myocyte orientation maps and tractography. Its architecture is
 51 optimized for large datasets, using chunked and parallel processing suitable for high-performance
 52 computing environments. The overall processing workflow, from input volumes to orientation
 53 maps and tractography, is summarized in Figure 1.

54 Cardiotensor has already been successfully applied in published work to characterize 3D
 55 cardiomyocyte architecture in healthy and diseased human hearts using synchrotron tomography
 56 (Brunet et al., 2024) to datasets over a terabyte in size. While cardiotensor was conceived
 57 for cardiac imaging, the package is modality- and tissue-agnostic. Any volumetric dataset
 58 exhibiting coherent fibrous microstructure can be analyzed, including brain white matter,
 59 skeletal muscle, and tendon. This generality makes the library useful for both cardiovascular
 60 and broader anatomical or histological studies.

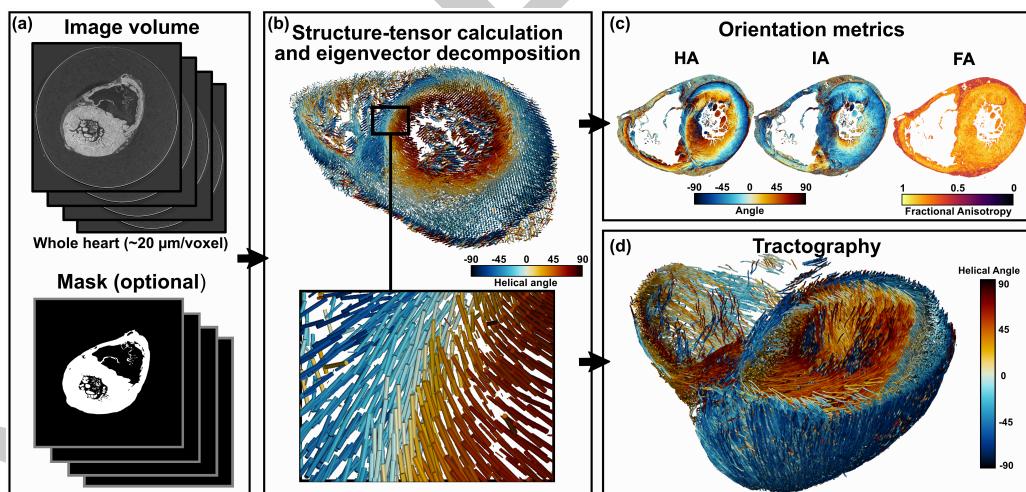


Figure 1: Cardiotensor pipeline for 3D cardiac orientation analysis and tractography. (a) Input whole- or partial-heart volume with optional myocardial mask. (b) Local cardiomyocyte orientation estimated via 3D structure tensor and eigenvector decomposition. The third eigenvector field (smallest eigenvalue) is visualized as arrows color-coded by helical angle (HA); inset shows structure tensor orientation in the ventricular septum. (c) Transformation to a cylindrical coordinate system enables computation of voxel-wise helical angle (HA), intrusion angle (IA), and fractional anisotropy (FA) maps. (d) Streamline tractography reconstructs continuous cardiomyocyte trajectories, color-coded by HA.

61 Implementation

62 Cardiotensor is implemented in Python and designed to efficiently process very large 3D cardiac
 63 imaging datasets. It relies primarily on NumPy (Van Der Walt et al., 2011) for numerical
 64 computation, with I/O accelerated by tifffile (Gohlke, 2025), Glymur (Evans, 2025), and
 65 OpenCV (Bradski, 2000). Dask (Rocklin, 2015) is used exclusively to parallelize file reading,
 66 while the core computations rely on Python's multiprocessing module for local parallelism.
 67 The package builds on the structure-tensor library (Jeppesen et al., 2021) to calculate the 3D
 68 structure tensor and eigenvector decomposition.

69 The package supports multiple use cases:

- 70 ▪ Command-line workflows, which automate batch processing from a configuration file of
71 terabyte-scale heart volumes and produce results as live plots or files saved to disk.
- 72 ▪ Embedded use in larger Python analysis workflows, enabling flexible scripting and scalable
73 execution on cluster environments.

74 Efficient computation is achieved through a chunk-based processing strategy with padding,
75 which avoids edge artifacts. This architecture allows parallelization across computing clusters by
76 splitting volumes into independent jobs, enabling cardiotensor to process whole-heart volumes
77 in hours rather than days while maintaining practical memory requirements.

78 **Architecture**

79 Cardiotensor is organized into five main modules, designed for clarity and scalability:

- 80 ▪ **orientation**: Computes local cardiomyocyte (or other texture feature) orientation using
81 a chunked 3D structure tensor pipeline, including eigenvalue decomposition, cylindrical
82 coordinate rotation, and calculation of helical angle (HA), intrusion angle (IA), and
83 fractional anisotropy (FA).
- 84 ▪ **tractography**: Generates and filters streamlines tracing cardiomyocyte trajectories from
85 the orientation field for myoaggregate-level reconstruction and analysis.
- 86 ▪ **analysis**: Provides a GUI for regional quantification and plotting transmural profile.
- 87 ▪ **visualization**: Supports interactive 3D visualization of vector fields and streamlines,
88 HA color-coding, and export to VTK/ParaView for large-scale rendering.
- 89 ▪ **utils**: Contains general utilities for I/O, image preprocessing, configuration parsing,
90 and vector math, supporting the entire pipeline.

91 This modular architecture ensures reproducibility, maintainability, and easy integration into
92 larger cardiac imaging workflows.

93 **Documentation and Usage**

94 The documentation for cardiotensor is available online at:

95 <https://josephbrunet.github.io/cardiotensor>

96 The main components of the documentation are:

- 97 ▪ Step-by-step walkthroughs for installation, first steps, and a guided example covering all
98 available commands. A small example dataset and its corresponding mask are provided
99 with the package.
- 100 ▪ In-depth explanations of the core algorithms used in cardiotensor, including structure
101 tensor theory, helical angle calculation, fractional anisotropy (FA), and tractography
102 integration.
- 103 ▪ Reference guides for the command-line interface, configuration file format, and public
104 API.

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