

# graynet: single-subject morphometric networks for neuroscience connectivity applications

Pradeep Reddy Raamana<sup>1</sup> and Stephen C. Strother<sup>1, 2</sup>

<sup>1</sup> Rotman Research Institute, Baycrest Health Sciences, Toronto, ON, Canada <sup>2</sup> Department of Medical Biophysics, University of Toronto, Toronto, ON, Canada

DOI: [10.21105/joss.00924](https://doi.org/10.21105/joss.00924)

## Software

- [Review](#) ↗
- [Repository](#) ↗
- [Archive](#) ↗

Submitted: 25 August 2018

Published: 08 September 2018

## License

Authors of papers retain copyright and release the work under a Creative Commons Attribution 4.0 International License ([CC-BY](#)).

## Summary

Connectivity, and network-level features in general, have proven to be valuable tools in several aspects of neuroscience research. Although network analysis is rooted in analysis of functional MRI data, it has recently gained traction in the analyses of morphometric features such as cortical thickness (Evans 2013). Such networks of anatomical covariance (derived based on distributions of features across a group of subjects) provided insight into changes caused by various brain disorders. When we individualize this approach to enable extraction of single-subject network features, they further enriched insights into abnormalities due to disease (Tijms et al. 2012, Raamana et al. (2015), Palaniyappan et al. (2015), Xu et al. (2017)). Moreover, these network-level features demonstrated potential for prognostic applications (Raamana et al. 2015, Raamana et al. (2014)), in addition to being robust to changes in scale and edge weight metrics (Raamana and Strother 2017b).

However, deriving these network-level features from input T1w-MRI data is non-trivial. With this fully-open-source and pure-python library **graynet**, we attempt to make this task easier and extend it to support all currently available morphometric features. Currently, it interfaces directly with the outputs produced by Freesurfer (Fischl 2012) and supports vertex-wise data. We plan to extend this to support volumetric atlases and voxel-wise features such as gray matter density. Together with many convenience scripts (e.g. to launch jobs on the high-performance cluster and assemble the outputs produced), we believe **graynet** makes an useful addition to the neuroimaging in python open source ecosystem.

**graynet** is dependent on the following libraries: **nibabel** (Brett et al. 2016), **networkx** (Hagberg, Schult, and Swart 2005), **numpy** (Oliphant 2007, Walt, Colbert, and Varoquaux (2011)) and **hiwenet** (Raamana and Strother 2017a).

## Acknowledgement

Pradeep Reddy Raamana is grateful for the support of the Canadian Biomarker Integration Network for Depression (CAN-BIND) and Ontario Neurodegenerative Disease Research Initiative (ONDRI), which are two integrated discovery programs of the Ontario Brain Institute (OBI), Canada. OBI is an independent non-profit corporation, funded partially by the Ontario government. The opinions, results, and conclusions are those of the authors and no endorsement by the OBI is intended or should be inferred.

## References

- Brett, Matthew, Michael Hanke, Ben Cipollini, Marc-Alexandre Côté, Chris Markiewicz, Stephan Gerhard, Eric Larson, et al. 2016. “Nibabel: Access a Cacophony of Neuro-Imaging File Formats, Version 2.1. 0.” *Zenodo*. <https://doi.org/10.5281/zenodo.591597>.
- Evans, Alan C. 2013. “Networks of Anatomical Covariance.” *Neuroimage* 80. Elsevier:489–504. <https://doi.org/10.1016/j.neuroimage.2013.05.054>.
- Fischl, Bruce. 2012. “FreeSurfer.” *Neuroimage* 62 (2). Elsevier:774–81. <https://doi.org/10.1016/j.neuroimage.2012.01.021>.
- Hagberg, A, D Schult, and P Swart. 2005. “Networkx: Python Software for the Analysis of Networks.” *Mathematical Modeling and Analysis, Los Alamos National Laboratory*.
- Oliphant, Travis E. 2007. “Python for Scientific Computing.” *Computing in Science & Engineering* 9 (3). IEEE. <https://doi.org/10.1109/MCSE.2007.58>.
- Palaniyappan, Lena, Bert Park, Vijender Balain, Raj Dangi, and Peter Liddle. 2015. “Abnormalities in Structural Covariance of Cortical Gyrification in Schizophrenia.” *Brain Structure and Function* 220 (4). Springer:2059–71. <https://doi.org/10.1007/s00429-014-0772-2>.
- Raamana, Pradeep Reddy, and Stephen C Strother. 2017a. “Histogram-Weighted Networks for Feature Extraction, Connectivity and Advanced Analysis in Neuroscience.” *Journal of Open Source Software* 2 (19). <https://doi.org/10.21105/joss.00380>.
- Raamana, Pradeep Reddy, and Stephen C. Strother. 2017b. “Impact of Spatial Scale and Edge Weight on Predictive Power of Cortical Thickness Networks.” *bioRxiv*. Cold Spring Harbor Labs Journals. <https://doi.org/10.1101/170381>.
- Raamana, Pradeep Reddy, Michael W Weiner, Lei Wang, Mirza Faisal Beg, Alzheimer’s Disease Neuroimaging Initiative, and others. 2015. “Thickness Network Features for Prognostic Applications in Dementia.” *Neurobiology of Aging* 36. Elsevier:S91–S102. <https://doi.org/10.1016/j.neurobiolaging.2014.05.040>.
- Raamana, Pradeep Reddy, Wei Wen, Nicole A Kochan, Henry Brodaty, Perminder S Sachdev, Lei Wang, and Mirza Faisal Beg. 2014. “Novel Thicknet Features for the Discrimination of Amnesic MCI Subtypes.” *NeuroImage: Clinical* 6. Elsevier:284–95. <https://doi.org/10.1016/j.nicl.2014.09.005>.
- Tijms, Betty M, Peggy Seriès, David J Willshaw, and Stephen M Lawrie. 2012. “Similarity-Based Extraction of Individual Networks from Gray Matter Mri Scans.” *Cerebral Cortex* 22 (7). Oxford University Press:1530–41. <https://doi.org/10.1093/cercor/bhr221>.
- Walt, Stéfan van der, S Chris Colbert, and Gael Varoquaux. 2011. “The Numpy Array: A Structure for Efficient Numerical Computation.” *Computing in Science & Engineering* 13 (2). IEEE:22–30. <https://doi.org/10.1109/MCSE.2011.37>.
- Xu, Jinping, Jiuquan Zhang, Jinlei Zhang, Yue Wang, Yanling Zhang, Jian Wang, Guanglin Li, Qingmao Hu, and Yuanchao Zhang. 2017. “Abnormalities in Structural Covariance of Cortical Gyrification in Parkinson’s Disease.” *Frontiers in Neuroanatomy* 11. Frontiers Media SA. <https://doi.org/10.3389/fnana.2017.00012>.