

# Masks2Metrics (M2M): A Matlab Toolbox for Gold Standard Morphometrics

Shadia Mikhael<sup>1</sup> and Calum Gray<sup>1, 2</sup>

<sup>1</sup> Edinburgh Imaging, The Chancellor's Building, 1st floor, FU303e, 49 Little France Crescent, Edinburgh, Scotland, UK, EH16 4SB. <sup>2</sup> Edinburgh Clinical Research Facility, University of Edinburgh, Edinburgh, UK.

DOI: [10.21105/joss.00436](https://doi.org/10.21105/joss.00436)

## Software

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

Submitted: 19 October 2017

Published: 12 February 2018

## Licence

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

## Summary

Human brains undergo morphometric changes over a lifetime, from conception through to birth, infancy, adolescence, adulthood, and old age (Thambisetty et al. (2010); Madan and Kensinger (2016)). This is further compounded by the changes associated with various brain pathologies such as tumours (e.g. Bauer et al. (2013)) and dementia (e.g., B. C. Dickerson et al. (2011)). It is therefore essential to accurately and scientifically characterise such changes by using an array of morphologic measurements, for a better understanding of the natural progression of ageing and disease (Mills et al. (2016); Madan (2017)). While many existing brain image analysis tools (e.g., FreeSurfer (Fischl et al. (2004); Desikan et al. (2006)), BrainSuite (Shattuck and Leahy (2002)), and BrainVISA (Kochunov et al. (2012))) automatically compute such data from a 3-dimensional (3D) brain image, they lack the ability to do so for the equivalent manually-traced regions of interest (ROIs). This is all the more significant as such ROIs are considered as the gold standard, thus making knowledge of their metrics essential.

We have developed an automated Matlab-based tool, Masks2Metrics (Mikhael and Gray (2017)), that calculates three metrics for a given ROI in a 3D image: thickness, volume and surface area. An ROI is defined by a pair of binary masks (in NIFTI file format) representing its outer and inner borders, each of which are drawn continuously along one direction (x-, y- or z-axis). In the specific case of brain images, when the ROI describes a gyrus, its paired masks would correspond to grey matter (GM) and white matter (WM) curves. The paired ROI NIFTI (.nii) masks are expected to be of the form `subj_roi_hem_gm/wmsegments.nii`. For example, a pair corresponding to subject 1's right SFG (superior frontal gyrus) would be `1_sfg_r_gm1.nii` and `1_sfg_r_wm1.nii`. A special feature of M2M is that multiple pairs, or segments, can be used rather than a single continuous ROI. These segments can be manually or automatically derived. The generated ROI metrics are grey matter thickness (GMth), grey matter volume (GMvol), and white matter surface area (WMSa), also classically calculated by popular existing automated tools (Fischl\_2000; Shattuck\_2002). Additionally, the ROI's corresponding mean Fréchet (Ursell (2013)) and mean Modified Hausdorff Distance (SasiKanth (2011)) are calculated and saved as matrices.

M2M is freely available on GitHub at <https://github.com/Edinburgh-Imaging/Masks2Metrics> under a GNU General Public License, along with external code that is called by the tool. It can be downloaded into 'Masks2Metrics' folder, added to the list of Matlab paths, and consequently run by calling 'masks2metrics' with the appropriate input and output parameters. As part of the tool's wiki, we provide a sample 3-segment ROI outlining part of a subject's superior frontal gyrus for demonstration purposes. The gyrus was manually segmented over a 3D image acquired by a Magnetic Resonance Imaging (MRI) machine.

This tool not only provides invaluable gold standard data for the brain imaging field, but equally so for any other field investigating morphometrics of manually and automatically-derived 3D ROIs represented as paired binary masks.

## Authors and Affiliations

Shadia Mikhael, Edinburgh Imaging, Centre for Clinical Brain Sciences, University of Edinburgh, Edinburgh, UK.

Calum Gray, Edinburgh Imaging and Edinburgh Clinical Research Facility, University of Edinburgh, Edinburgh, UK.

## Contributors

Maria del C. Valdés Hernández, Edinburgh Imaging, Centre for Clinical Brain Sciences, University of Edinburgh, Edinburgh, UK.

Corne, Hoogendoorn, Toshiba Medical Visualization Systems Edinburgh, Edinburgh, UK.

Cyril R. Pernet, Edinburgh Imaging, Centre for Clinical Brain Sciences, University of Edinburgh, Edinburgh, UK.

## References

Bauer, Stefan, Roland Wiest, Lutz-P Nolte, and Mauricio Reyes. 2013. “A Survey of Mri-Based Medical Image Analysis for Brain Tumor Studies” 58 (June):R97–R129. <https://doi.org/10.1088/0031-9155/58/13/R97>.

Desikan, Rahul S., Florent Ségonne, Bruce Fischl, Brian T. Quinn, Bradford C. Dickerson, Deborah Blacker, Randy L. Buckner, et al. 2006. “An Automated Labeling System for Subdividing the Human Cerebral Cortex on Mri Scans into Gyral Based Regions of Interest.” *NeuroImage* 31 (3):968–80. <https://doi.org/https://doi.org/10.1016/j.neuroimage.2006.01.021>.

Dickerson, B. C., T. R. Stoub, R. C. Shah, R. A. Sperling, R. J. Killiany, M. S. Albert, B. T. Hyman, D. Blacker, and L. deToledo-Morrell. 2011. “Alzheimer-signature MRI biomarker predicts AD dementia in cognitively normal adults.” *Neurology* 76 (16):1395–1402. <https://doi.org/10.1212/wnl.0b013e3182166e96>.

Fischl, Bruce, André van der Kouwe, Christophe Destrieux, Eric Halgren, Florent Ségonne, David H. Salat, Evelina Busa, et al. 2004. “Automatically Parcellating the Human Cerebral Cortex.” *Cerebral Cortex* 14 (1):11–22. <https://doi.org/10.1093/cercor/bhg087>.

Kochunov, Peter, William Rogers, Jean-Francois Mangin, and Jack Lancaster. 2012. “A Library of Cortical Morphology Analysis Tools to Study Development, Aging and Genetics of Cerebral Cortex.” *Neuroinformatics* 10 (1):81–96. <https://doi.org/10.1007/s12021-011-9127-9>.

Madan, Christopher R. 2017. “Advances in Studying Brain Morphology: The Benefits of Open-Access Data.” *Frontiers in Human Neuroscience* 11:405. <https://doi.org/10.3389/fnhum.2017.00405>.

Madan, Christopher R., and Elizabeth A. Kensinger. 2016. “Cortical Complexity as a Measure of Age-Related Brain Atrophy.” *NeuroImage* 134 (Supplement C):617–29. <https://doi.org/https://doi.org/10.1016/j.neuroimage.2016.04.029>.

Mikhael, Shadia, and Calum Gray. 2017. “Masks2Metrics.” 2017. <https://github.com/Edinburgh-Imaging/Masks2Metrics>.

Mills, Kathryn L., Anne-Lise Goddings, Megan M. Herting, Rosa Meuwese, Sarah-Jayne Blakemore, Eveline A. Crone, Ronald E. Dahl, et al. 2016. “Structural Brain Development Between Childhood and Adulthood: Convergence Across Four Longitudinal Samples.” *NeuroImage* 141 (Supplement C):273–81. <https://doi.org/https://doi.org/10.1016/j.neuroimage.2016.07.044>.

SasiKanth, S. 2011. “Modified Hausdorff Distance.” 2011. <http://uk.mathworks.com/matlabcentral/fileexchange/29968-modified-hausdorff-distance>.

Shattuck, David W, and Richard M Leahy. 2002. “BrainSuite: An Automated Cortical Surface Identification Tool.” *Medical Image Analysis* 6 (2):129–42. [https://doi.org/https://doi.org/10.1016/S1361-8415\(02\)00054-3](https://doi.org/https://doi.org/10.1016/S1361-8415(02)00054-3).

Thambisetty, Madhav, Jing Wan, Aaron Carass, Yang An, Jerry L. Prince, and Susan M. Resnick. 2010. “Longitudinal Changes in Cortical Thickness Associated with Normal Aging.” *NeuroImage* 52 (4):1215–23. <https://doi.org/https://doi.org/10.1016/j.neuroimage.2010.04.258>.

Ursell, Tristan. 2013. “Frechet Distance Calculator.” 2013. <https://uk.mathworks.com/matlabcentral/fileexchange/41956-frechet-distance-calculator>.