

A fully automatic pipeline for estimation of regional brain volume change using Jacobian integration

Introduction

Regional brain volume change is employed as a marker to assess disease progression and potential treatment effects within a clinical trial setting.

The Boundary Shift Integral (BSI; Freeborough, 1997) is commonly employed gold-standard method for estimating volume change. This method often requires manual or semi-automatic segmentations and is limited to regions with clear anatomical boundaries.

Here we investigate the use of a fully automatic Jacobian Integration (JI) approach for estimating volume change which has the potential to estimate volume change in regions lacking distinct anatomical bounds as required by the BSI pipeline

Methodology

We have implemented a fully-automatic pipeline for estimation of longitudinal volume change between T1W images for regions of interest within a whole brain parcellation schema.



Results

190 subjects with baseline and 1-year follow-up T1-weighted MR images were processed with the proposed JI pipeline and compared to BSI values for the lateral ventricles (VBSI), initialised with manual segmentations, 4 subjects were identified with failed registration.

When compared to VBSI we note significant correlation (r=0.909, p<<0.001) in reported percentage volume change from baseline for subjects processed with the JI pipeline. A mean difference in percentage volume of -0.12% (SD: 2.4%) was observed.



205 subjects with 1 year follow-up from the ADNI cohort were processed with JI pipeline and ventricle volume changes compared to VBSI values provided from ADNI. 4 subjects failed registration. A significant correlation (r=0.875 p<<0.001) was observed between JI and VBSI, with a mean percentage volume difference of 1.23%,



Comparing ventricle volumes between control (NC; N=57), cognitively impaired (MCI; N=100) and Alzheimer's disease (AD;N=44) groups reveals similar means and effect sizes.



Automatic detection of pipeline registration failure was performed via thresholding cross-correlation between non-linear registered images within a brain-masked area.

		JI		VBSI	
	Ν	mean	stdev	mean	stdev
NC	57	3.91	4.48	4.48	4.65
MCI	100	5.66	4.97	7.14	5.44
AD	44	7.99	4.94	9.49	5.79

References

- 1. Avants, B.B., Epstein, C.L., Grossman, M. and Gee, J.C., 2008. Symmetric diffeomorphic image registration with cross-correlation: evaluating automated labeling of elderly and neurodegenerative brain. *Medical image analysis*, 12(1), pp.26-41
- 2. Freeborough, P.A. and Fox, N.C., 1997. The boundary shift integral: an accurate and robust measure of cerebral volume changes from registered repeat MRI. *IEEE transactions on medical imaging*, *16*(5), pp.623-629.
- 3. Heckemann, R.A., Ledig, C., Gray, K.R., Aljabar, P., Rueckert, D., Hajnal, J.V. and Hammers, A., 2015. Brain extraction using label propagation and group agreement: Pincram. PloS one, 10(7), p.e0129211.
- 4. Ledig, C., Wolz, R., Aljabar, P., Lötjönen, J., Heckemann, R.A., Hammers, A. and Rueckert, D., 2012, May. Multi-class brain segmentation using atlas propagation and EM-based refinement. In 2012 9th IEEE International Symposium on Biomedical Imaging (ISBI) (pp. 896-899). IEEE. DBC

Conclusions

We present a fully automatic method for computing regional volume change employing Jacobian Integration, here reporting with comparable results to the commonly employed BSI methodology for lateral ventricle volume change.

This pipeline has the potential to estimate volume change for a range of automatically estimate ROIs in a scalable and efficient manner for clinical trial deployment, as currently under investigation.

- 5. Leung, K.K., Clarkson, M.J., Bartlett, J.W., Clegg, S., Jack Jr, C.R., Weiner, M.W., Fox, N.C., Ourselin, S. and Alzheimer's Disease Neuroimaging Initiative, 2010. Robust atrophy rate measurement in Alzheimer's disease using multi-site serial MRI: tissue-specific intensity normalization and parameter selection. Neuroimage, 50(2), pp.516-523.
- 6. Lewis, E.B. and Fox, N.C., 2004. Correction of differential intensity inhomogeneity in longitudinal MR images. *Neuroimage*, 23(1), pp.75-83.
- 7. Tustison, N.J., Avants, B.B., Cook, P.A., Zheng, Y., Egan, A., Yushkevich, P.A. and Gee, J.C., 2010. N4ITK: improved N3 bias correction. *IEEE transactions on* medical imaging, 29(6), p.1310.