

Impact of bias-correction and skull-stripping pipelines on spatial normalisation using SPM5 in a phantom model

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Introduction: In recent years, statistical parametric mapping (SPM) [1] has been widely used for the technique of optimised voxel-based morphometry (VBM). The VBM technique localises differences in structural magnetic resonance (MR) images between patient populations on a voxel-by-voxel basis [2]. Several studies have shown that VBM can be improved by skull-stripping [3-9] and correcting for intensity non-uniformity (radio frequency bias) [10, 11] in images prior to analysis. The latest SPM release, SPM5, enables spatial normalisation to standard space, tissue segmentation and bias correction to be combined within the same model [12]. The purpose of the present study was to investigate the impact that the pre-processing methods enumerated below have on normalised gray-matter (GM) segments in comparison to each other and to that derived from default SPM5 settings:

Skull-stripping. (i) Fully-automated hybrid watershed algorithm (HWA) using atlas information [6]; (ii) manually optimised brain surface extractor (BSE) [3] and (iii) fully-automated brain extraction tool v.2 (BET2) [4] with fractional intensity threshold, f , set to 0.4 and 0.5 (vertical gradient, g , set to 0).

Bias-correction. (i) Non-parametric non-uniform intensity normalisation (N3) [10] and (ii) bias field corrector (BFC) [3], both fully-automated.

Methods and Results: The pre-processing pipelines were evaluated on a T₁-weighted MRI BrainWeb phantom (3% noise, 40% bias) [13] as shown in Fig. 1. A gold-standard dataset was obtained by manually delineating the cortical surface on the ground-truth image (0% bias). All volumes were skull-stripped and then intensity-

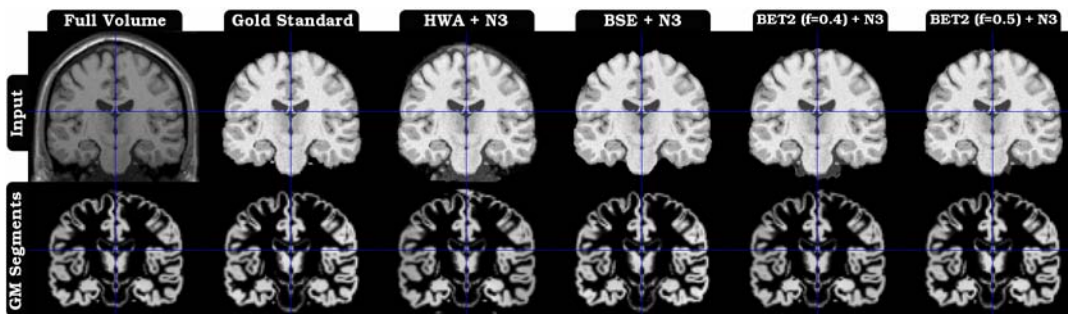


Figure 1: Input volumes inserted in SPM5 and resulting GM segments for the full volume, gold standard and each pre-processing pipeline.

Method	BC off		BC on	
	N3	BFC	N3	
Full Volume	—	—	—	4.37
HWA	4.26	7.34	4.43	4.40
BSE	4.22	4.60	4.49	4.44
BET2, $f=0.4$	4.23	5.57	4.49	4.41
BET2, $f=0.5$	4.22	5.51	4.46	4.42

Table 1: RMS error of the bias-corrected images for the full-volume and pre-processing methods.

Method	J	FN (%)	FP (%)	ΔN (%)
Full Volume	0.84	9.0	8.3	-0.7
HWA + N3	0.80	8.3	15.2	6.9
BSE + N3	0.90	6.4	4.4	-2.0
BET2, $f=0.4$ + N3	0.86	5.5	9.5	3.9
BET2, $f=0.5$ + N3	0.89	4.8	6.9	2.1

Table 2: Jaccard similarity coefficient, false negative rates, false positive rates and differences in number of voxels for the normalised GM segments obtained from the full-volume and pre-processing methods.

This is the opposite effect to that for HWA, but has a similar impact on the FN rate after warping to the template. (iii) BET2 was the most consistent method; it kept the lowest FN rates, and FN, FP and ΔN could be lowered by increasing f . Table 2 also shows that the full-volume method performed better than HWA, but ostensibly worse than the BSE, BET2 + N3 pipelines.

Discussion and Conclusions: This study evaluated the performance of SPM5 using different methods to pre-process structural MRI data. Running BET2 to skull-strip and N3 to bias-correct the BrainWeb phantom performed especially well. BSE + N3 also performed better than the full-volume and HWA + N3 method, but its high specificity came at the price of GM tissue removal that, in turn, could adversely impact on statistical analyses. It was also demonstrated that removing brain tissue or including non-brain voxels have a negative effect on the warping process. In summary, this phantom study suggests that BET2 + N3 and BSE + N3 pre-processing may offer greater normalisation accuracy compared to SPM5 alone.

References: [1] Available from <http://www.fil.ion.ucl.ac.uk/spm>; [2] Ashburner J. et al., Neuroimage 8:1105 (1997); [3] Shattuck, D.W. et al., Neuroimage 13: 856 (2001); [4] Smith S.M. Hum. Brain Mapp. 17: 143 (2002); [5] Boesen K. et al., Neuroimage 22: 1255 (2004); [6] Ségonne F. et al., Neuroimage 22: 1060 (2004); [7] Rex D.E. et al., Neuroimage 23: 625 (2004); [8] Zhuang A.H. et al., Neuroimage 32: 79 (2006); [9] Fennema-Notestine C. et al., Hum. Brain Mapp. 27: 99 (2006); [10] Sled J.G. et al., IEEE Trans. Med. Imag. 17: 87 (1998); [11] Arnold J.B. et al., Neuroimage 13: 931 (2001); [12] Ashburner J. et al., Neuroimage 26: 839 (2005); [13] Available from http://www.bic.mni.mcgill.ca/brainweb/selection_normal.html.