

1

2 **Proposed Methodology for Reducing Bias in Structural MRI Analysis in the Presence of Lesions:**

3 **Data from a Pediatric Traumatic Brain Injury Cohort**

4 Daniel Griffiths-King¹, Adam Shephard¹, Jan Novak¹, Cathy Catroppa^{2,3}, Vicki A. Anderson^{2,3} Amanda

5 G. Wood^{1,2,4}

6 ¹College of Health and Life Sciences & Aston Institute of Health and Neurodevelopment, Aston
7 University, Birmingham, B4 7ET UK

8 ²Brain and Mind Research, Clinical Sciences, Murdoch Children's Research Institute, Melbourne,
9 Australia

10 ³Department of Psychology, Royal Children's Hospital, Melbourne, Australia

11 ⁴School of Psychology, Faculty of Health, Melbourne Burwood Campus, Deakin University, Geelong,
12 Victoria, Australia

13

14

15 Correspondence to:

16 Dr Daniel Griffiths-King

17 Email: d.griffiths-king@aston.ac.uk

18

19 **Word Count:**

20 **Figures:** 3; **Tables:** 1

21 **Supplementary Material:** Supplementary Methods, Supplementary Results, 1 Supplementary Figure

22 **Keywords:** traumatic brain injury, lesion, MRI, FreeSurfer, neuroanatomy

23

24

25

Abstract

26 Traumatic brain injury can lead to multiple pathologic features, including brain lesions, which are
27 visible on magnetic resonance imaging (MRI). These resulting heterogenous lesions can present a
28 difficulty for several standard approaches to neuroimaging, resulting in bias and error in subsequent
29 quantitative measurements. Thus, cases presenting with lesions on MRI may be excluded from
30 analyses, biasing samples across the research field. We outline a potential solution to this issue in
31 the case of Freesurfer, a popular neuroimaging tool for surface-based segmentation of brain tissue
32 from structural MRI. The proposed solution involves two-steps, a) Pre-processing: Enantiomorphic
33 Lesion-Filling and b) Post-processing: Lesion Labelling. We applied this methodology to 14 pediatric
34 TBI cases which presented with lesions on T1w MRI. Following qualitative inspection of these cases
35 after implementation of the approach, 8 out of 14 cases were retained as being of sufficient quality.
36 In brief, we have presented here an adapted pipeline for processing structural MRI (sMRI) of
37 patients who have experienced a TBI using the Freesurfer software package. This approach aims to
38 mitigate potential lesion-induced biases that exist beyond the locality of the pathological tissue,
39 even in the contralesioned hemisphere.

40

41 Proposed Methodology for Reducing Bias in Structural MRI Analysis in the Presence of Lesions: Data
42 from a Pediatric Traumatic Brain Injury Cohort

43 Structural MRI (sMRI) can be utilised to estimate functionally relevant brain ‘damage’ after a
44 traumatic brain injury (TBI), primarily through the quantification of the morphometry of brain
45 regions (see [1] for review). These approaches may be more sensitive to subtle effects of injury on
46 the brain compared to routine visual reporting by neuroradiologists. Therefore, these methods may
47 better allow the understanding of the neuroanatomical basis of later impairment.

48 The accuracy of these methods, however, are biased by errors introduced during the automated-
49 processing of sMRI containing gross anatomical lesions and/or pathology. Essentially, morphometric
50 measures generated for these cases may not be biologically valid for two main reasons; a) gross
51 pathology (such as encephalomalacic regions) or pathological voxel intensities (such as gliosis or
52 oedema) can either render boundaries undetectable or discontinuous [2-5], or b) systematic biases
53 introduced by the presence of pathology on the Freesurfer pipeline (i.e. contralesional hemisphere
54 biases [6] or atlas registration biases [3, 4]). These potential errors make it difficult to ascertain
55 whether differences between control and patient morphology are due to an injury-related pathology
56 or due to inaccuracies in morphometric measures specific to patients with gross lesions [1].

57 Historically, studies of paediatric TBI (pTBI) have excluded cases with major pathology present on
58 MRI (for instance [7]) due to these potential processing biases. However, this limits the utility of
59 previous research with the exclusion of these patients risking a systematic bias in sampling. Given
60 that the location and extent of focal lesions to the brain following a pTBI are seemingly insufficient
61 to fully explain post-injury neuropsychological deficits [8] (i.e. following early brain injury impaired
62 executive function occurs irrespective of injury factors such as lesion location [9, 10]) Inclusion of
63 these lesion cases in research may increase accuracy of prognostic quantitative models and ensure
64 they generalise to the full spectrum of pathology [1, 2, 5, 6]. Therefore, approaches and/or
65 methodologies that are robust to the presence of lesions are necessary for future studies.

66 In a recent paper, Diamond and colleagues [2, 5] identified and outlined a potential methodology
67 with which to ‘optimise’ structural segmentation of sMRI for patients with TBI. This utilised the
68 Freesurfer pipeline, an automated approach to the surface-based structural segmentation of T1w
69 MRI. Diamond and colleagues’ [2, 5] approach involves the manual labelling of tissue where the
70 reconstructed surfaces pass through cortical lesions. However, this post-processing approach, which
71 results in very focal edits to the surface reconstruction, does not address global algorithmic biases
72 indicated by the presence of lesions. For instance, in a recent study, we identified that the presence
73 of simulated lesion pathology, resulted in a small but systematic bias in the contralesional

74 hemisphere, and the magnitude of this bias is seemingly associated with voxel intensities within this
75 pathology [6]. Diamond and colleagues' approach [2, 5] will not account for this bias.

76 In the current paper, we highlight a potential adjustment to the Freesurfer pipeline to mitigate some
77 of the observed-issues in surface-based parcellation of the cortex in the presence of traumatic
78 lesions, particularly the bias in the surface-placement of the contralateral hemisphere to the lesion.

79 **Methods**

80 The data used in the current experiment are a subset of an existing dataset of children who have
81 experienced a TBI between the ages of five and 16 years of age. 114 patients with pTBI were
82 recruited between 2007 and 2010 into a study on 'Prevention and Treatment of Social Problems
83 Following TBI in Children and Adolescents'. More detailed descriptions have been published
84 elsewhere [11-13]. In brief, children with TBI were recruited on presentation to the Melbourne Royal
85 Children's Hospital's emergency department. Patients were eligible if they: i) were aged between
86 five and 16 years at the time of injury, ii) had recorded evidence of both a closed-head injury and
87 also two post-concussive symptoms (such as headaches, dizziness, nausea, irritability, poor
88 concentration), iii) had sufficient detail within medical records to determine injury severity (e.g.,
89 Glasgow Coma Scale (GCS; Teasdale and Jennett [14]), neurological and radiological findings), iv) had
90 no prior history of neurological or neurodevelopmental disorder, non-accidental injuries or previous
91 TBI, and v) were English speaking.

92 **MRI Acquisition**

93 MRI were acquired sub-acutely after injury (<90 days post-injury). MRI were acquired at 3T on a
94 Siemens Trio scanner (Siemens Medical Systems, Erlangen, Germany) using a 32-channel matrix
95 head coil. The acquisition included a sagittal three-dimensional (3D) MPRAGE [TR = 1900 ms; TE =
96 2.15 ms; IR prep = 900 ms; parallel imaging factor (GRAPPA) 2; flip angle 9 degrees; BW 200 Hz/Px;
97 176 slices; resolution 1 × 1 × 1 mm] and sagittal 3D T2-FLAIR non-selective inversion preparation
98 SPACE (Sampling Perfection with Application-optimised Contrast using different flip-angle Evolution)
99 [TR = 6000 ms; TE = 405 ms; inversion time (TI) = 2100 ms; water excitation; GRAPPA Pat2; 176
100 slices; 1 × 1 × 1 mm resolution matched in alignment to the 3D T1w sequence].

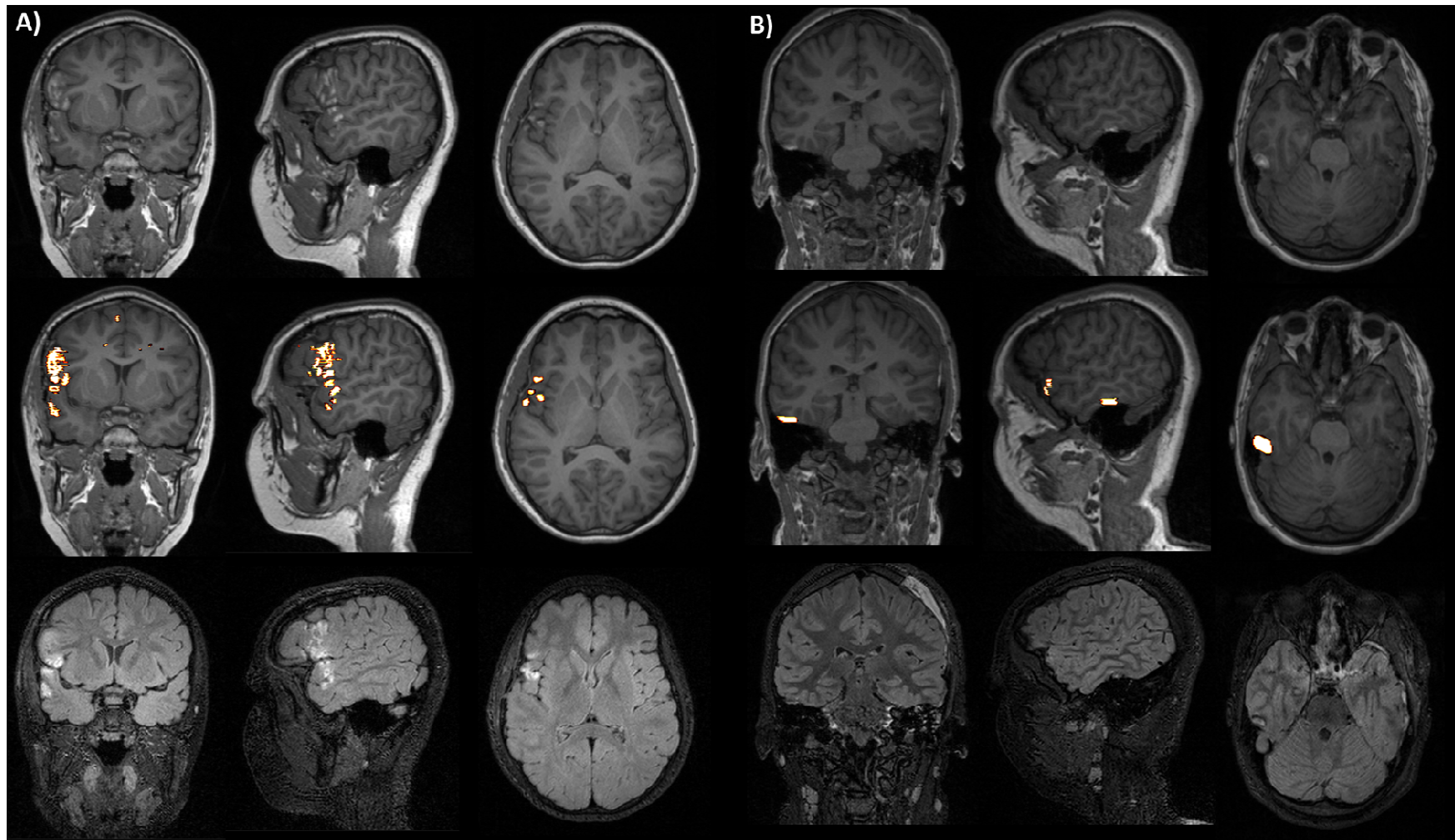
101 **Lesion Delineation and production of lesion masks**

102 A trained rater (JN) visually inspected participant's MRI for pathology, scrolling through contiguous
103 axial slices of the 3D T1w and FLAIR images independently. Identified lesions were segmented
104 manually (by JN) by drawing binary lesion masks on each of the T1w and FLAIR MRI scans using the

105 ROI editor tool in MRtrix3.0 [15]. A second rater (AS) visually confirmed these masks. For this study,
106 only the lesion masks drawn on the T1w MRI scan were utilised/necessary.

107 TBI lesions are typically extremely heterogenous in appearance on MRI scans [16]. Increasingly,
108 white matter hyperintensities (WMH i.e. Leukoaraiaosis) and enlarged perivascular spaces (EPVS i.e.
109 Virchow-Robin spaces) are recognised as potential biomarkers for an increased risk of later emerging
110 diseases/diagnoses [16-19]. Consequently, we also segmented these abnormalities.

111 For lesion segmentation, the following criteria were applied; i) abnormality visible on >3 contiguous
112 axial slices (i.e. ≥ 1.5 mm), ii) visible WMHs should appear hyperintense on FLAIR and hypointense on
113 T1w MRI [19], iii) visible EPVS should appear hypointense on both T1w and FLAIR MRI, and be
114 tubular shaped depending on lesion orientation [19]. Examples of these lesion masks can be seen in
115 Figure 1A+B.



116

117 Figure 1. sMRI of two (A + B) patients with TBI with lesions. Top row: Unedited T1w image with visible gross pathology. Middle row: T1w image with
118 overlaid binary lesion mask (mask is interpolated for visual purposes). Bottom row: Unedited FLAIR image.

119 **Proposed pipeline for processing sMRI with visible pathology**

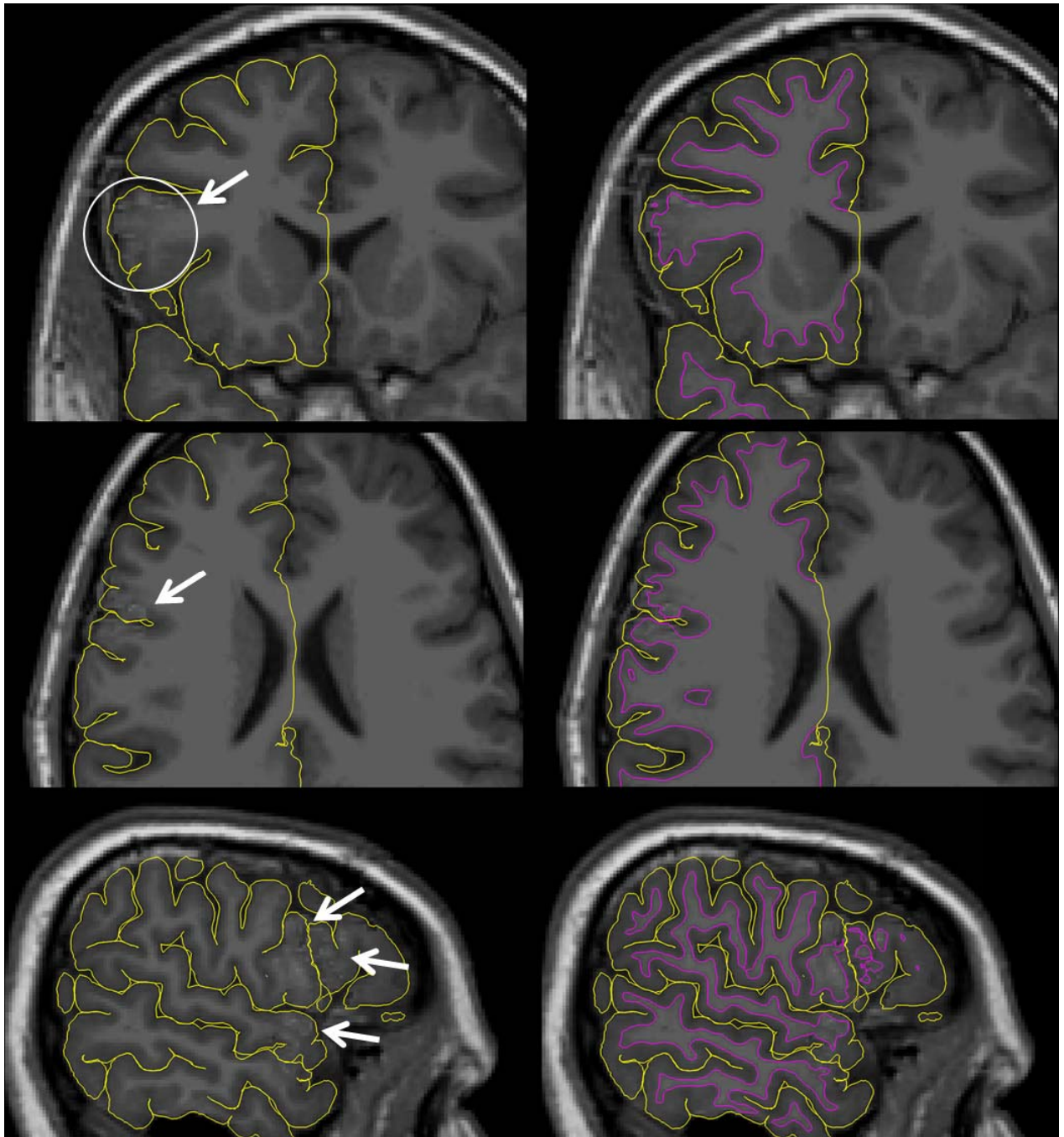
120 The current paper utilises a new approach to Freesurfer segmentation in the presence of focal
121 lesions in the cortical GM ribbon, involving both pre- and post- processing procedures.

122 **a) Pre-processing: Enantiomorphic Lesion-Filling**

123 Lesion masks were used for pre-processing of MRI, to perform anatomically-informed lesion-filling,
124 using the enantiomorphic approach of Nachev, Coulthard [20]. Briefly, this approach robustly
125 registers the lesioned hemisphere to the contra-lesioned hemisphere and ‘fills’ the lesioned voxels
126 (indicated by the lesion-mask) with subject-level, ‘healthy-appearing’ signal intensities from the
127 homologous region in the contra-lesional hemisphere. The output is an MRI with approximately-
128 typical T1w voxel-intensities, in place of the lesioned tissue. This step was conducted using the
129 normalisation tool of the BCBlab (Brain Connectivity and Behaviour) [21]. We only performed these
130 lesion-filling processes for those cases with frank GM lesions. Some recent evidence suggests that
131 filling approaches for white matter lesions results in no changes to Freesurfer derived volume
132 estimates [22]. This, and the fact that geometric inaccuracies due to WMH can be corrected using
133 manual editing approaches as per Freesurfer’s guidelines, means that we focus on an approach to
134 tackle GM lesion.

135 The enanteomorphically filled T1w image was then processed using the standard Freesurfer (6.0)
136 cortical surface segmentation pipeline (using the –FLAIRpial commands) [23]. By processing this
137 image rather than the original T1w MRI, we mitigate potential contrast-induced errors that may
138 contribute to lesion-induced error/bias in structural segmentation, even in the contralesional
139 hemisphere. An example of the resultant surfaces can be seen in Figure 2.

140



141

142 Figure 2. Freesurfer plotted surfaces overlaid on enanteomorphically filled T1w sMRI. Arrows
143 highlight the filled areas. First column: Only pial surface visualised. Second column: Both pial and
144 white surfaces visualised. The original lesion is identified in the MRI displayed in Figure 1A.

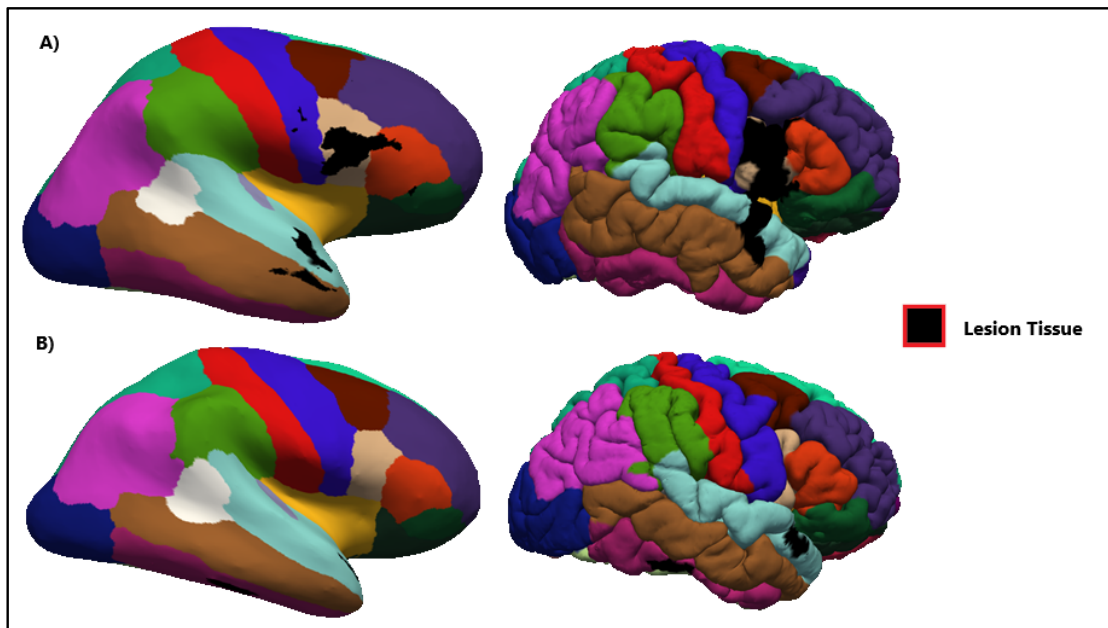
145 b) Post-processing: Lesion Labelling

146 Post-processing of the produced surface segmentations was also conducted. Lesion masks were
147 projected onto the generated cortical surface vertices and the projected lesion ROI was filled (to
148 avoid holes due to voxel-vertex mismatches). These approaches were adapted from scripts made
149 available by the Multi-centre Epilepsy Lesion Detection (MELD) project [24, 25].

150 Individual-subject surface parcellations were masked using these surface projected lesion masks.
151 Thus, region labels completely or partially occluded by lesion tissue were overwritten with the lesion
152 label. Morphometric measures (such as cortical thickness, volume, etc) were calculated using
153 standard Freesurfer approaches but, due to relabelling, no measures will be taken from tissue which
154 is a) lesioned within the original image and b) filled with estimated voxel intensities in the
155 enantomorphically filled T1w images. For those regions that are completely occluded by the lesion
156 label, morphological measures are reported as zero however, these can be recoded as 'not a
157 number' (NaN) to ensure that they are not included in analyses and bias results.

158 The output of this pipeline is therefore cortical morphometric estimates for ROIs not contaminated
159 by lesion tissue or the wider error associated with the processing of lesioned T1w images. A visual
160 depiction of this can be seen in Figure 3.

161



162

163 Figure 3. ROI parcellation based upon the Desikan-Killany atlas [26] for two lesion cases projected
164 onto both the inflated (first column) and regular (second column) surface models for each subject.
165 The lesion label can also be seen Subjects A) and B) relate to the corresponding subjects in Figure 1.

166 **Quality Assessments**

167 Quality was visually assessed for all cases, based upon the delineation of both pial and white matter
168 surfaces generated by Freesurfer. This allowed identification of cases where manual edits needed to
169 be undertaken, and were carried out per standard Freesurfer protocols.

170 **Results**

171 **Lesion Identification**

172 Of the pTBI cases (N=114), we identified n=14 as having visible pathology definable as lesion tissue
173 within the cortical GM ribbon. These cases were selected and underwent our lesion correction
174 procedure

175 **QA**

176 On initial visual inspection of the output of the lesion-correction procedure, two cases were initially
177 excluded, due to the bilaterality of contusions in near-homologous regions, leading to unsuccessful
178 lesion filling using the enanteomorphic approach. Five cases required no manual edits, although two
179 of these had poor reconstruction focal in relation to the filled lesion tissue. However, as per
180 Diamond and colleagues [2, 5], to reduce subjectivity in manual edits for these areas, no edits were
181 undertaken, as these would later be labelled as lesion tissue. Seven cases underwent manual edits to
182 improve surface placement. Of these, three cases were acceptable after editing. The four cases
183 which were rejected were due to surface reconstruction issues that were related to motion present
184 within the image, rather than due to the lesion correction procedure. The final number of cases
185 processed with the lesion correction procedure was eight (out of 14).

186 **Post-Hoc Volume analyses**

187 We conducted post-hoc analyses to assess whether the lesion correction methodology impacted
188 cortical volumes and cortical thickness in the contralesioned hemisphere. Differences in both
189 measures were found in both the contralesioned and lesioned hemispheres when comparing cases
190 which have been corrected with the lesion pipeline versus those that have not been corrected with
191 the lesion pipeline. Further details and figures can be found in supplementary materials. This
192 suggests that that this method may in fact be correcting contralesioned hemisphere biases
193 introduced by gross GM pathology found in [6].

194

195 **Discussion**

196 The current paper highlights an alternative method to Diamond and colleagues [2, 5] for optimising
197 the structural segmentation of sMRI in individuals with pTBI, specifically designed to mitigate the
198 more global segmentation biases which are introduced in these lesioned cases [6]. This approach is
199 being utilised in our lab enable the inclusion of lesioned cases into studies of the neuroanatomical
200 correlates of cognitive impairment post paediatric TBI. The benefit of such methodologies is to
201 simultaneously a) increase power to detect relevant case-control differences or brain-behaviour
202 relationships and b) increase generalisability of findings to a wider spectrum of pathology. These
203 benefits come about via the ability to include cases with visible pathology on MRI which previously
204 would be precluded from analysis pipelines.

205 As outlined above, our approach differs from that of Diamond and colleagues [2, 5] in one major
206 aspect. This is specifically the pre-processing of cases using an enantomorphic filling approach. This
207 was to tackle global biases in morphometric measures beyond the site of specific pathology (such as
208 the contralesioned hemisphere) which may not be reduced by local correction methods. However, it
209 may be argued that Diamond and colleagues' [2, 5] approach better meets the first benefit outlined
210 above, to increase statistical power through inclusion of cases. This is because, whilst our approach
211 led to the inclusion of 8/14 cases which would have otherwise be excluded in future analyses,
212 Diamond and colleagues [5] retained 87/98 MRIs for which Freesurfer surfaces were successfully
213 generated and corrected using their methodology. However, this is not a direct comparison. It is
214 important to remember that, in this study, cases were also excluded were removed for typical
215 reasons not associated with the lesion, in this case motion artefact. This is unsurprising given the
216 fact that this study utilised a paediatric population where movement artefact is more common [27],
217 whereas Diamond [2, 5] investigated an adult cohort.

218 Whilst Freesurfer is primarily an automated tool for the processing of sMRI and generating surface-
219 based models of cortical morphometry, these surface models can be and are frequently utilised
220 further in the analysis pipeline of functional MRI and diffusion MRI studies of TBI. Therefore,
221 effective methods to ensure inclusion of cases with visible lesions on MRI and reduce the biases that
222 these lesions can introduce into these automated pipelines will have wide-reaching implications for
223 the field. It is also important to note that, whilst TBI pathology is particularly heterogeneous, the
224 effects of visible pathology on neuroimaging pipelines is not limited to TBI and thus the methods
225 outlined here may also find use in other neurological disease (i.e. multiple sclerosis or
226 tuberous sclerosis).

227 This pipeline has been used in our lab in published works [28] allowing the inclusion of cases that
228 typically may not have been able to be reliably included in research paradigms. A recent paper has
229 outlined a similar approach termed "virtual brain grafting (VBG)" [29], which has subsequently been

230 utilised in structural connectomic approaches to understanding the individualised effects of TBI on
231 the brain [30]. Given the VGB approach was validated on “synthetic” MRI, it would be prudent to
232 test the two methodologies head-to-head in future research.

233 **Limitations**

234 One limitation of the current approach is its ability to handle bilateral lesions in homologous brain
235 regions, put simply lesions to both hemispheres in approximately similar anatomical locations. This is
236 because the enantiomorphic filling approach would attempt to ‘fill’ abnormal voxel intensities (the
237 lesion) with abnormal, lesion voxel intensities from the ipsilateral hemisphere, rather than healthy
238 tissue as intended. Two cases were excluded from the current study for this reason. Adopters of this
239 approach must also be aware that, for the largest lesions, statistical analysis approaches, that
240 robustly deal with missing data may be required. Fully occluded ROIs will be labelled as NaN, which
241 is likely to be treated as missing data. to include them as zero makes an implicit assumption about
242 the underlying tissue, that it is functionally and neuroanatomically irrelevant. These are important
243 methodological considerations for the application of these approaches.

244 One potential issue with both proposed approaches is the reliance on accurate segmentations or
245 ‘masks’ of lesion present on MRI. This can be difficult and time consuming, requires considerable
246 expertise to be considered as ‘gold standard’ in the field (normally a trained neuroradiologist) with
247 no ‘ground truth’ with which to truly assess performance. This can make approaches requiring such
248 lesion masks prohibitive to; larger studies with a greater number of lesions to segment; labs without
249 such neuroradiological expertise and most importantly clinical applications/practice where
250 considerable ‘pre-processing’ would be required for new and incoming cases.

251 It is difficult to ascertain whether the approach outlined by Diamond and colleagues, or the
252 approach outlined in the current study best optimises the Freesurfer pipeline for use in TBI cases
253 with visible pathology, as there exists no ‘ground truth’ in these circumstances. Both approaches
254 likely go some way in addressing the potentially biases introduces when Freesurfer is used to
255 process MRI of patients with TBI and visible gross pathology. However, we would argue that our
256 approaches goes further, trying to mitigate further biases which we have observed previously in
257 these types of analyses [6]. Future work directly compare these methodologies using a more
258 pragmatic approach, such as evaluating which method allows us to recover the most accurate
259 predictions of cognitive functioning post-injury, or even predict injury-severity. These may be more
260 clinically useful assessments of these methodologies in the absence of ‘ground truth’.

261 In brief, we have presented here an adapted pipeline for processing sMRI of patients who have
262 experienced a TBI using the Freesurfer software package. This approach aims to mitigate potential

263 lesion-induced biases that exist beyond the locality of the pathological tissue, even in the
264 contralesioned hemisphere.

265

266 **Acknowledgments and Funding**

267 The authors thank Dr Jan Novak and Dr Adam Shephard for their technical assistance in providing
268 the lesion masks necessary for this study. This work was supported by a European Research Council
269 (ERC) - Consolidator Grant (ERC-CoG) to A.G.W [grant number 682734]. This work was conducted
270 whilst DGK was supported by a studentship from Aston University, School of Life and Health Science.
271 and a Birmingham Childrens' Hospital Research Foundation Grant (BCHRF) to Dr Sukhvir Wright and
272 A.G.W. DGK is currently funded by a grant from Aston University, College of Health and Life Sciences
273 to J.N.

274

References

275

- 276 1. King, D.J., et al., *A systematic review of cross-sectional differences and longitudinal changes*
277 *to the morphometry of the brain following paediatric traumatic brain injury*. *Neuroimage*
278 *Clin*, 2019. **23**: p. 101844.
- 279 2. Edlow, B., et al., *Optimizing the Accuracy of Cortical Volumetric Analysis in Traumatic Brain*
280 *Injury*. *Journal of Neurotrauma*, 2019. **36**(13): p. A62-A62.
- 281 3. Goh, S.Y., et al., *Neuroinformatics challenges to the structural, connectomic, functional and*
282 *electrophysiological multimodal imaging of human traumatic brain injury*. *Front*
283 *Neuroinform*, 2014. **8**: p. 19.
- 284 4. Irimia, A., et al., *Neuroimaging of structural pathology and connectomics in traumatic brain*
285 *injury: Toward personalized outcome prediction*. *Neuroimage Clin*, 2012. **1**(1): p. 1-17.
- 286 5. Diamond, B.R., et al., *Optimizing the accuracy of cortical volumetric analysis in traumatic*
287 *brain injury*. *MethodsX*, 2020. **7**: p. 100994.
- 288 6. King, D.J., et al., *Lesion Induced Error on Automated Measures of Brain Volume: Data From a*
289 *Pediatric Traumatic Brain Injury Cohort*. *Front Neurosci*, 2020. **14**: p. 491478.
- 290 7. Serra-Grabulosa, J.M., et al., *Cerebral correlates of declarative memory dysfunctions in early*
291 *traumatic brain injury*. *Journal of Neurology Neurosurgery and Psychiatry*, 2005. **76**(1): p.
292 129-131.
- 293 8. Bigler, E., *Quantitative magnetic resonance imaging in traumatic brain injury*. *The Journal of*
294 *head trauma rehabilitation*, 2001. **16**(2): p. 117-134.
- 295 9. Anderson, V., et al., *Children's executive functions: are they poorer after very early brain*
296 *insult*. *Neuropsychologia*, 2010. **48**(7): p. 2041-50.
- 297 10. Jacobs, R., A.S. Harvey, and V. Anderson, *Are executive skills primarily mediated by the*
298 *prefrontal cortex in childhood? Examination of focal brain lesions in childhood*. *Cortex*, 2011.
299 **47**(7): p. 808-24.
- 300 11. Anderson, V., et al., *Social competence at 6 months following childhood traumatic brain*
301 *injury*. *J Int Neuropsychol Soc*, 2013. **19**(5): p. 539-50.
- 302 12. Anderson, V., et al., *Social Competence at Two Years after Childhood Traumatic Brain Injury*.
303 *J Neurotrauma*, 2017. **34**(14): p. 2261-2271.
- 304 13. Catroppa, C., et al., *Social and Behavioral Outcomes following Childhood Traumatic Brain*
305 *Injury: What Predicts Outcome at 12 Months Post-Insult?* *J Neurotrauma*, 2017. **34**(7): p.
306 1439-1447.
- 307 14. Teasdale, G. and B. Jennett, *Assessment of Coma and Impaired Consciousness - Practical*
308 *Scale*. *Lancet*, 1974. **2**(7872): p. 81-84.
- 309 15. Tournier, J.D., F. Calamante, and A. Connelly, *MRtrix: Diffusion tractography in crossing fiber*
310 *regions*. *International Journal of Imaging Systems and Technology*, 2012. **22**(1): p. 53-66.
- 311 16. Bigler, E.D., et al., *Heterogeneity of brain lesions in pediatric traumatic brain injury*.
312 *Neuropsychology*, 2013. **27**(4): p. 438-51.
- 313 17. Adams, H.H., et al., *Rating method for dilated Virchow-Robin spaces on magnetic resonance*
314 *imaging*. *Stroke*, 2013. **44**(6): p. 1732-5.
- 315 18. Dubost, F., et al., *Enlarged perivascular spaces in brain MRI: Automated quantification in*
316 *four regions*. *NeuroImage*, 2019. **185**: p. 534-544.
- 317 19. Wardlaw, J.M., M.C. Valdés Hernández, and S. Muñoz-Maniega, *What are white matter*
318 *hyperintensities made of? Relevance to vascular cognitive impairment*. *Journal of the*
319 *American Heart Association*, 2015. **4**(6): p. e001140.
- 320 20. Nachev, P., et al., *Enantiomorphic normalization of focally lesioned brains*. *Neuroimage*,
321 2008. **39**(3): p. 1215-1226.
- 322 21. Foulon, C., et al., *Advanced lesion symptom mapping analyses and implementation as*
323 *BCBtoolkit*. *Gigascience*, 2018. **7**(3): p. 1-17.

- 324 22. Guo, C., et al., *Repeatability and reproducibility of FreeSurfer, FSL-SIENAX and SPM brain*
325 *volumetric measurements and the effect of lesion filling in multiple sclerosis*. European
326 Radiology, 2019. **29**(3): p. 1355-1364.
- 327 23. Fischl, B., *FreeSurfer*. Neuroimage, 2012. **62**(2): p. 774-81.
- 328 24. Adler, S., et al., *MELD Protocol 4 - Lesion Masking*. 2018: protocols.io.
329 doi:10.17504/protocols.io.n9udh6w.
- 330 25. Adler, S., et al., *Novel surface features for automated detection of focal cortical dysplasias in*
331 *paediatric epilepsy*. Neuroimage Clin, 2017. **14**: p. 18-27.
- 332 26. Desikan, R.S., et al., *An automated labeling system for subdividing the human cerebral cortex*
333 *on MRI scans into gyral based regions of interest*. Neuroimage, 2006. **31**(3): p. 968-80.
- 334 27. Pardoe, H.R., R. Kucharsky Hiess, and R. Kuzniecky, *Motion and morphometry in clinical and*
335 *nonclinical populations*. Neuroimage, 2016. **135**: p. 177-85.
- 336 28. King, D.J., et al., *Structural-covariance networks identify topology-based cortical-thickness*
337 *changes in children with persistent executive function impairments after traumatic brain*
338 *injury*. NeuroImage, 2021. **244**.
- 339 29. Radwan, A.M., et al., *Virtual brain grafting: Enabling whole brain parcellation in the presence*
340 *of large lesions*. NeuroImage, 2021. **229**: p. 117731.
- 341 30. Imms, P., et al., *Exploring personalized structural connectomics for moderate to severe*
342 *traumatic brain injury*. Network Neuroscience, 2022: p. 1-24.

343

