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- Hippocampus and amygdala volumes from magnetic resonance images
- in children: Assessing accuracy of FreeSurfer and FSL against
- manual segmentation
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ABSTRACT

The volumetric quantification of brain structures is of great interest in pediatric populations because it allows the investigation of different factors influencing neurodevelopment. FreeSurfer and FSL both provide frequently used 20 packages for automatic segmentation of brain structures. In this study, we examined the accuracy and consistency of those two automated protocols relative to manual segmentation, commonly considered as the "gold stan-22 dard" technique, for estimating hippocampus and amygdala volumes in a sample of preadolescent children aged $\frac{23}{2}$ between 6 to 11 years. The volumes obtained with FreeSurfer and FSL-FIRST were evaluated and compared with $\frac{24}{2}$ manual segmentations with respect to volume difference, spatial agreement and between- and within-method 25 correlations.

Results highlighted a tendency for both automated techniques to overestimate hippocampus and amygdala vol- 27 umes, in comparison to manual segmentation. This was more pronounced when using FreeSurfer than FSL-FIRST 28 and, for both techniques, the overestimation was more marked for the amygdala than the hippocampus. Pearson 29 correlations support moderate associations between manual tracing and FreeSurfer for hippocampus (right r = 30 0.69, p < 0.001; left r = 0.77, p < 0.001) and amygdala (right r = 0.61, p < 0.001; left r = 0.67, p < 0.001) volumes. 31 Correlation coefficients between manual segmentation and FSL-FIRST were statistically significant (right hippo- 32 campus r = 0.59, p < 0.001; left hippocampus r = 0.51, p < 0.001; right amygdala r = 0.35, p < 0.001; left amyg- 33dala r = 0.31, p < 0.001) but were significantly weaker, for all investigated structures. When computing intraclass 34 correlation coefficients between manual tracing and automatic segmentation, all comparisons, except for left hippocampus volume estimated with FreeSurfer, failed to reach 0.70. When looking at each method separately, cor- 36 relations between left and right hemispheric volumes showed strong associations between bilateral 37 hippocampus and bilateral amygdala volumes when assessed using manual segmentation or FreeSurfer. These 38 correlations were significantly weaker when volumes were assessed with FSL-FIRST. Finally, Bland-Altman 39 plots suggest that the difference between manual and automatic segmentation might be influenced by the vol- 40 ume of the structure, because smaller volumes were associated with larger volume differences between 41

These results demonstrate that, at least in a pediatric population, the agreement between amygdala and hippo- 43 campus volumes obtained with automated FSL-FIRST and FreeSurfer protocols and those obtained with manual 44 segmentation is not strong. Visual inspection by an informed individual and, if necessary, manual correction of $\,45$ automated segmentation outputs are important to ensure validity of volumetric results and interpretation of re- 46 lated findings.

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Childhood is a period of great relevance in the development of risk 61 factors for various neuropsychiatric conditions (Paus et al., 2008). 62

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1. Introduction

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Together with increased efforts in prevention, many large-scale longitudinal studies, starting in early childhood, are currently being undertaken to reveal the impact of environmental, behavioral and biological factors on subsequent developmental outcomes (Chakravarty et al., 2014; Giedd et al., 2015; Raznahan et al., 2014). Due to rapid advances of in-vivo brain imaging technologies, volumetric quantification of brain structures from structural Magnetic Resonance Imaging (MRI) is more accessible than ever. Thus, large-scale studies often acquire MRI to investigate relations between volume of specific brain structures and different aspects of behavior.

Due to their involvement in multiple neuropsychiatric and neurological conditions, the medial temporal lobe structures hippocampus and amygdala have received a considerable amount of attention. The hippocampus is one of the most commonly studied and cited brain structures in the scientific literature. Its involvement in basic cognitive functions, such as memory consolidation (Squire, 1992), psychopathologies such as PTSD (Bonne et al., 2001), major depression (Campbell and MacQueen, 2004), and neurological disorders, such as Alzheimer disease (Fox et al., 1996), is well established. The amygdala is the main structure of the limbic system associated with fear (Adolphs et al., 1994; Davis and Whalen, 2001). It has been linked to many psychopathologies including borderline personality disorder (Donegan et al., 2003; Herpertz et al., 2001), PTSD (Rauch et al., 2000) and social phobia (Stein et al., 2002). The association between negative life events during childhood, such as abuse and traumatic experiences, and the increased risk of developing psychiatric disorders later in life is well documented (Janssen et al., 2004; Johnson et al., 1999; MacMillan et al., 2001; Springer et al., 2007). It has been hypothesized that the relations between severe childhood stressors and vulnerability to psychopathologies might be mediated trough an impaired development of the hippocampus and/or amygdala (Pynoos et al., 1999; Teicher et al., 2003; Woon and Hedges, 2008). Thus, many efforts are directed at defining and clarifying the roles of the amygdala and the hippocampus in pediatric samples. From a structural neuroimaging perspective, an important challenge lies in the reliable and valid volumetric quantification of these brain regions. However, reliable volumetric estimation is methodologically limited by the anatomical complexity of these two structures.

Manual segmentation is currently considered the gold standard for volumetric quantification of brain structures (Pardoe et al., 2009; Rodionov et al., 2009). However, this procedure requires sufficient anatomical and MR methodological expertise, is difficult and timeconsuming to learn, and can be associated with intra- and inter-rater variability if not performed using a consistent approach (Jack Jr. et al., 1995). In order to increase reliability and reduce potential biases associated with manual segmentation procedures, multiple protocols have been established and described in the literature for specific target regions (Jack et al., 1990; Matsuoka et al., 2003; Pruessner et al., 2000; Watson et al., 1992). Studies have demonstrated that using these protocols significantly improve intra- and inter-rater agreement (Jack et al., 1990; Matsuoka et al., 2003; Pruessner et al., 2000; Watson et al., 1992). However, these protocols require a considerable amount of training and thus further increase time demands of manual segmentation procedures. In contrast, protocols that offer the fully automated processing and segmentation of target structures from MR images are fast (speed is only limited by CPU power and availability), have excellent reproducibility and require little anatomical expertise from the end user. As a result, a number of automated protocols have recently been developed, published and received favorably by the research community. In part because they are easily and freely accessible to the research community and provide detailed documentation on usage, two of these automated procedures have gained a considerable amount of popularity. The first one is FreeSurfer (http://surfer.nmr.mgh.harvard. edu), a software developed by the Martinos Center for Biomedical Imaging (Fischl et al., 2002). FreeSurfer automatically assigns a label to each voxel from the anatomical image based on probabilistic estimations relying on Markov random fields (MRFs). The localisation and spatial relations between structures are defined according using a training set 129 of manually labeled brains. The second commonly used automated seg- 130 mentation protocol is "FIRST", provided as part of the FSL software li- 131 brary (http://fsl.fmrib.ox.ac.uk) (Patenaude, 2007; Patenaude et al., 132 2011). Using a probabilistic framework, this software estimates boundaries of brain structures based on the signal intensity of the T1 image as 134 well as the expected shape of structures to be segmented.

It is well known that neuroanatomical variations are found not only 136 in clinical populations, but also when comparing brains of normal individuals (Pruessner et al., 2002). Automated segmentation approaches 138 are based on the questionable assumption that computer algorithms 139 can reliably differentiate and delimitate anatomical regions regardless 140 of inter-individual differences in neuroanatomy, scan quality, image 141 contrast, etc. While we did not find any studies comparing the perfor- 142 mance of automated segmentation performed with FSL-FIRST and/or 143 FreeSurfer to manual segmentation in pediatric populations, the validity 144 of these protocols has previously been assessed in healthy adult controls 145 (Cherbuin et al., 2009; Morey et al., 2009; Patenaude et al., 2011) as well 146 as different clinical populations, such as Alzheimer Disease (Pipitone 147 et al., 2014; Sánchez-Benavides et al., 2010; Shen et al., 2010), mood disorders (Doring et al., 2011; Nugent et al., 2013; Tae et al., 2008), 149 temporal-lobe epilepsy (Akhondi-Asl et al., 2011; Pardoe et al., 2009) 150 and psychosis (Pipitone et al., 2014). These reports generally support 151 the ability of automated methods to detect volume difference between 152 clinical groups. However, many articles have highlighted a tendency for 153 FreeSurfer and FSL-FIRST to overestimate volume of brain structures 154 (Cherbuin et al., 2009; Doring et al., 2011; Morey et al., 2009; Nugent 155 et al., 2013; Pipitone et al., 2014; Sánchez-Benavides et al., 2010; Shen 156 et al., 2010; Tae et al., 2008). When assessing the correspondence be- 157 tween volumes derived from these two automated protocols and man- 158 ual segmentation earlier findings are variable. For the hippocampus 159 region, results usually support moderate to strong associations between 160 manual tracing and FreeSurfer, with Pearson correlation coefficients 161 ranging from 0.71 (Cherbuin et al., 2009; Sánchez-Benavides et al., 162 2010) to 0.90 (Shen et al., 2010). Studies looking at the association between hippocampus volumes derived from FSL-FIRST and manual seg- 164 mentation report Pearson correlations ranging from 0.47 (Pardoe 165 et al., 2009) to 0.67 (Nugent et al., 2013). Few studies have looked at 166 the agreement between amygdala volumes derived from automated 167 segmentation protocols and manual tracing. A study by Morey et al. 168 (2009) revealed weaker associations between manual segmentation 169 and both FSL-FIRST and FreeSurfer when estimating the amygdala vol- 170 ume than when estimating the hippocampus volume (Morey et al., 171 2009). Taken together, these results seem to indicate that the concordance between volumes derived from manual segmentation versus au- 173 tomatic protocols depend on the segmented structure as well as the 174 protocol used. Further, a report by Sánchez-Benavides suggests that 175 the accuracy of automated protocols may vary depending on neuroana- 176 tomical characteristics of studied populations (Sánchez-Benavides et al., 177 2010). More precisely, this later study highlights a larger discrepancy 178 between manually and automatically segmented volumes when used 179 on atrophic brains. Previous reports assessing the validity and accuracy 180 of FSL-FIRST and FreeSurfer were based on adult brains; it remains uncertain whether smaller brain volumes and potential changes in gray / 182 white matter contrasts in pediatric brains negatively affect the performance of these two automated segmentation software. Thus, studies investigating the validity of automated segmentation in children are 185 needed.

The goal of this article was to explore the validity of FSL-FIRST and 187 FreeSurfer in estimating hippocampus and amygdala volumes in chil- 188 dren. To do so, we compared volumes generated by these two automated techniques to volumes obtained by manual segmentation, which is 190 considered to be the "gold standard" approach. The validity of the 191 segmentation methods was investigated by means of three different ap- 192 proaches. First, we established discrepancies between volumes derived 193 from manual segmentation and automated methods. Second, to 194

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