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Q1 Extensive learning is associated with gray matter changes in the 2 right hippocampus

Q2 Kathrin Koch ^{a,b,c,*}, Tim Jonas Reess ^{a,b,c}, Oana Georgiana Rus ^{a,b,c}, Claus Zimmer ^{a,b}

^a Department of Neuroradiology, Klinikum rechts der Isar, Technische Universität München, Ismaningerstrasse 22, 81675 Munich, Germany
 ^b TUM-Neuroimaging Center (TUM-NIC) of Klinikum rechts der Isar, Technische Universität München TUM, Ismaninger Strasse 22, 81675 Munich, Germany

6 ^c Graduate School of Systemic Neurosciences GSN, Ludwig-Maximilians-Universität, Biocenter, Großhaderner Strasse 2, 82152 Munich, Germany

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ABSTRACT

Longitudinal voxel-based morphometry studies have demonstrated increases in gray matter volume in hippocampal areas following extensive cognitive learning. Moreover, there is increasing evidence for the relevance of the subiculum in the context of learning and memory. Using longitudinal FreeSurfer analyses and hippocampus subfield segmentation the present study investigated the effects of 14 weeks of intensive learning on hippocampal and subicular gray matter volume in a sample of medical students compared to control subjects not engaged in any cognitive learning activities. We found that extensive learning resulted in a significant increase of right hippocampal volume. Volume of the left hippocampus and the subiculum remained unchanged. The current findings emphasize the role of the hippocampus in semantic learning and memory processes and provide further evidence for the neuroplastic ability of the hippocampus in the context of cognitive learning. 26

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32 Introduction

An increasing amount of evidence supports the notion that the brain 33 undergoes continuous activity-dependent neuroplastic changes across 34 the life-span. In structural imaging studies these changes have been 35 demonstrated as a consequence of various activities, such as learning 36 to juggle (Boyke et al., 2008; Draganski et al., 2004), learning of mirror 37 reading (Ilg et al., 2008), learning of new color names (Kwok et al., 38 2011), motor exercise (Niemann et al., 2014), video gaming (Kuhn 39 40 et al., 2014) or meditation (Kurth et al., 2014). Surprisingly few evidence, on the other hand, has been provided for structural changes in 41 association with learning of abstract information (Ceccarelli et al., 422009; Draganski et al., 2006). Ceccarelli et al. (2009) explored the effects 03 44 of two weeks of intensive learning and found a fronto-parietal gray matter (GM) volume increase. Draganski et al. (2006) used voxel-based 45morphometry at three different time points to investigate the effects 46 47 of intensive learning in students preparing for their medical exam. They found a significant gray matter increase in the posterior and lateral 48 parietal cortex bilaterally and a significant increase in the right hippo-4950campus during the learning period which augmented even further 51during the subsequent semester break. The primary function of the hip-52pocampus is clearly memory-related. The hippocampus plays a key

* Corresponding author at: Department of Neuroradiology, Klinikum rechts der Isar, Technische Universität München, Ismaningerstrasse 22, 81675 Munich, Germany. Fax: +49 89 41404887.

E-mail address: kathrin.koch@tum.de (K. Koch).

http://dx.doi.org/10.1016/j.neuroimage.2015.10.056 1053-8119/© 2015 Published by Elsevier Inc. role in the consolidation of information from short-term to long-term 53 memory and a number of studies have shown a correlation between 54 hippocampus volume and memory performance (Arlt et al., 2013; 55 Avery et al., 2013; Gimenez et al., 2004; Pohlack et al., 2014; Siraly 56 et al., 2015). 57

Thus, Arlt et al. (2013) and Gimenez et al. (2004) reported a correla- 58 tion between left hippocampal volume and verbal working memory, 59 Avery et al. (2013) demonstrated an association between total hippo- 60 campal volume and relational working memory and Pohlack et al. 61 (2014) reported an association between total hippocampal volume 62 and verbal working memory. 63

The hippocampus or hippocampal formation can be subdivided into 64 several subfields (dentate gyrus, areas CA3 and CA1, entorhinal cortex 65 and subiculum) (Amaral and Witter, 1995), out of which the subiculum Q4 constitutes one of the largest subfields and the major output structure. 67 There is increasing evidence indicating that the subiculum is the sub-68 field that is most strongly involved in basic memory processes and 69 semantic learning (O'Mara et al., 2009). It receives the majority of effer-70 ent information from the CA1 region of the hippocampus thus being in a 71 position to integrate, transfer and resolve information from other parts 72 of the hippocampus related to learning and memory (Amaral et al., Q5 1991; Deadwyler and Hampson, 2006).

Nevertheless, changes in gray matter volume in association with 75 learning of semantic information have only been reported for the hippo-76 campus formation as a whole whereas potential gray matter changes in 77 the subiculum have not been specifically investigated up to now. 78 Learning-related changes in gray matter volume of the hippocampus 79

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are most likely the consequences of synaptic or dendritic sprouting or 80 81 increases in synaptic strength and neuronal growth (i.e., increases of neuronal somae and nuclei) and, as such, very small (Neves et al., 82 83 2008). Hence, high image resolution is needed to detect the aggregated effects of such changes on the macroscopic level across time. Against 84 this background, in the present study we used FreeSurfer (http:// 85 surfer.nmr.mgh.harvard.edu) which is a software package capable of 86 87 detecting sub-millimeter changes in gray matter volume. The software 88 has a longitudinal image processing framework (i.e., FreeSurfer longitu-89 dinal). This framework is based on unbiased, robust, within-subject 90 template creation which has been demonstrated to successfully reduce 91variability, avoid over-regularization and increase power to detect 92structural changes across time by initializing the processing in each 93 time point with common information from the within-subject template (Reuter et al., 2012). We used this longitudinal framework to investi-94 gate the effects of intensive learning on gray matter volume and expect-95 96 ed increases in the hippocampus and more specifically the subiculum as 97 a consequence of learning. In addition, we investigated whether baseline gray matter volume predicted learning-associated increases in hip-98 pocampal gray matter volume. Finally, based on the above mentioned 99 findings of an association between memory performance and hippo-100 campus volume, we used the digit span task, an established and fre-101 102 quently employed working memory task, to explore whether memory 103 performance at the first measurement time point predicted learningassociated increases in hippocampal gray matter volume. Thus, we 104 intended to extend previous findings of a mere association between 105working memory performance and hippocampal volume and to explore 106 107a directed association by investigating whether working memory performance predicts hippocampal volume change. Identifying parameters 108 which allow the prediction of neuroplastic processes on an individual 109basis would be of high practical use, for instance in the field of cognitive 110 111 remediation and training.

112 Materials and methods

113 Subjects

We recruited 35 medical students from the Medical School of our 114 University and 24 healthy control subject. 7 medical students and 6 con-115trol subjects dropped out after the first scan resulting in a final sample 116 size of 28 right-handed healthy students (m:f = 14:14, mean age = 117 19.3 years, SD = 1.0 years) and 18 right-handed healthy control sub-118 jects (m:f = 6:12, mean age = 18.6 years, SD = 0.5 years) with no his-119 tory of neurological or psychiatric disorders or other serious medical 120 conditions. Groups were matched according to their level of education 121 122(i.e., both had the German "Abitur") and they were carefully selected regarding the amount of physical exercise and musical activities they 123were doing (i.e., professional athletes or musicians or subjects exces-124sively engaged in sports or playing an instrument were not included). 125In both groups, T₁-weighted magnetic resonance imaging (MRI) scans 126127were performed at two time points, TP1 and TP2 (i.e., 14 weeks after 128TP1). In the student group, the first scan was performed at the beginning of their first semester of medical school, the second scan was per-129formed fourteen weeks later shortly before their first semester medical 130131 exams. During this time period the students spent on average about 132190 h in class and 280 h outside of class with learning facts and information related to anatomy, chemistry, and biology. 133

The control subjects had recently started a voluntary social year. They
 were not attending any lesson or studying for any exams between TP1
 and TP2.

Handedness was assessed using Annett's handedness inventory. All
participants gave written informed consent to the study protocol
which is in accordance with the ethical standards of the Declaration of
Helsinki and was approved by the Ethics Committee of the Technische
Universität München, Medical School.

MRI acquisition

High-resolution anatomical T₁-weighted volume scans (MP-RAGE) 143 were collected on a 3 T whole body system equipped with a 12- 144 element receive-only head matrix coil (INGENIA, Philips). They were ob- 145 tained in sagittal orientation (TR = 9 ms, TE = 4 ms, TI = 900 ms, flip 146 angle = 8°, FOV = 240 × 240 mm², matrix = 240 mm × 240 mm, 147 number of sagittal slices = 170) with an isotropic resolution of 148 $1 \times 1 \times 1 \text{ mm}^3$.

Data analysis

We used the digit span task to assess cognitive performance and to 151 investigate a potential association between memory performance and 152 gray matter increase. The test consists of a forward and a backward ver- 153 sion. In the forward version of the test, a list of random numbers is read 154 out which the participant has to recall in the correct order. The test be- 155 gins with two numbers, increasing a number at a time until two errors 156 are committed in a row. In the backward version, participants are 157 asked to recall the digits in backward order. Thus, the forward version 158 mainly reflects working memory performance, the backward version 159 assesses predominantly manipulation of stimulus material and execu- 160 tive processing. Memory performance was investigated using a repeat- 161 ed measures one-way ANOVA with digit span forward performance as 162 the dependent variable, group as a between subject factor and measure- 163 ment time point (TP1, TP2) as a within subject factor. Digit span data of 164 one medical student and one control are missing. 165

Gray matter volume was assessed using the FreeSurfer software 166 package (version 5.3.0, http://surfer.nmr.harvard.edu). 167

The initial processing of T1 high-resolution images includes several 168 steps which have been described in previous papers (Dale et al., 1999; 169 Fischl et al., 1999). Briefly, the implemented processing stream contains 170 removal of non-brain tissue, transformation to Talairach-like space, 171 and segmentation of gray/white matter tissue. White and gray matter 172 boundary is tessellated and topological defects are automatically 173 corrected. After intensity normalization and transition of gray/white 174 matter, pial boundary is indicated by detecting the greatest shift in 175 intensity through surface deformation. Segmented data were then 176 parcellated into units based on gyral and sulcal structure, resulting 177 in values for gyrification and volume. Maps were smoothed using a 178 Gaussian kernel of 10 mm. 179

Subsequently, for the longitudinal processing, an unbiased 180 within-subject template is created using robust, inverse consistent 181 registration to estimate average subject anatomy across both mea-182 surement time points (Reuter et al., 2012). Finally, each time point 183 is processed "longitudinally", where information from the subject-184 template and from the individual runs are used (see Fig. 1, for details 185 please refer to Reuter et al. (2012)). This procedure has been demon-186 strated to significantly increase reliability and statistical power 187 in longitudinal studies (Ibarretxe-Bilbao et al., 2012; Kwan et al., 188 2012; Reuter et al., 2012).

Then, an automated subfield segmentation of the hippocampus 190 was performed using Bayesian inference and a probabilistic atlas 191 of the hippocampal formation. The hippocampal subfield volumes 192 obtained with this method have been compared to manual hippo-193 campal subfield tracings, and reliability measures were good for 194 the larger subfields (CA2/3, CA4/DG, subiculum) and only acceptable 195 for the smaller ones (CA1, presubiculum, fimbria) (Van Leemput 196 et al., 2009). Seven hippocampal subfield volumes were automatical-197 ly calculated including the fimbria (white matter), presubiculum, 198 subiculum, CA1, CA2–3, and CA4-DG fields (gray matter) as well 199 as the hippocampal fissure (cerebrospinal fluid). The procedures 200 for subfield segmentation have been described elsewhere (Van 201 Leemput et al., 2009). The present analyses focus on the subiculum. 202 To investigate the hypothesized changes in hippocampal and 203 subicular gray matter volume, for each subject and time point, 204

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