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Topology of genetic associations between regional gray matter volume and intellectual ability: Evidence for a high capacity network

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ABSTRACT

Intelligence is associated with a network of distributed gray matter areas including the frontal and parietal higher 18 association cortices and primary processing areas of the temporal and occipital lobes. Efficient information trans-19 fer between gray matter regions implicated in intelligence is thought to be critical for this trait to emerge. Genetic 20 factors implicated in intelligence and gray matter may promote a high capacity for information transfer. Whether 21 these genetic factors act globally or on local gray matter areas separately is not known. Brain maps of phenotypic and genetic associations between gray matter volume and intelligence were made 23 using structural equation modeling of 3 T MRI T1-weighted scans acquired in 167 adult twins of the newly ac- 24 quired U-TWIN cohort. Subsequently, structural connectivity analyses (DTI) were performed to test the hypoth- 25 esis that gray matter regions associated with intellectual ability form a densely connected core. Gray matter regions associated with intellectual ability were situated in the right prefrontal, bilateral temporal, 27 bilateral parietal, right occipital and subcortical regions. Regions implicated in intelligence had high structural 28 connectivity density compared to 10,000 reference networks (p = 0.031). The genetic association with intelli- 29 gence was for 39% explained by a genetic source unique to these regions (independent of total brain volume), 30 this source specifically implicated the right supramarginal gyrus. 31 Using a twin design, we show that intelligence is genetically represented in a spatially distributed and densely 32 connected network of gray matter regions providing a high capacity infrastructure. Although genes for intelli- 33 gence have overlap with those for total brain volume, we present evidence that there are genes for intelligence 34 that act specifically on the subset of brain areas that form an efficient brain network. 35

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41 Introduction

36 **39** 39

Distinct regions of the cerebral cortex show consistent associations
with general intellectual ability, as shown in brain imaging studies
measuring brain structure, and activity (Haier et al., 2004; Gray and
Thompson, 2004; Jung and Haier, 2007; Narr et al., 2007; Colom et al.,
2009; Barbey et al., 2012; Schnack et al., 2014). These regions include
early information processing areas of the temporal (fusiform gyrus,
Wernicke's area) and occipital lobes (extrastriate cortex) as well as

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http://dx.doi.org/10.1016/j.neuroimage.2015.09.046 1053-8119/© 2015 Published by Elsevier Inc. higher associative processing areas of the parietal (supramarginal 49 gyrus, angular gyrus and superior parietal gyrus) and frontal lobes (dor- 50 solateral prefrontal cortex, inferior frontal gyrus and orbitofrontal cor- 51 tex). Also, subcortical areas such as the hippocampus, caudate nucleus 52 and thalamus show positive associations with intelligence (MacLullich 53 et al., 2002; Frangou et al., 2004; Haier et al., 2009; Bohlken et al., 54 2014a; Grazioplene et al., in press). The consistency of these findings 55 across the literature has led to network models of intelligence, stating 56 that cortical and subcortical gray matter regions rely on efficient com- 57 munication through white matter fiber bundles to support intelligence 58 (Jung and Haier, 2007; Deary et al., 2010; Colom et al., 2010). Indeed, 59 network analyses have revealed that individuals with a higher intelli- 60 gence have more efficient brain networks (Li et al., 2009; van den Q3 Heuvel et al., 2010; Langer et al., 2012; Fischer et al., 2014). However, 62 the extent to which brain networks for intelligence are shaped through 63 genes and environment is not known. 64

Gray matter volume and intelligence share common genetic vari- 65 ance (Thompson et al., 2001; Baaré et al., 2001; Posthuma et al., 2002; 66 Toga and Thompson, 2005). Importantly, genetic factors implicated in 67 intelligence and gray matter are found in specific regions, pertaining 68

Abbreviations: A, additive genetic; C, common environmental; CFARI, Crossing Fiber Angular Resolution of Intra-Voxel structure; DNA, deoxyribonucleic acid; DTI, diffusion tensor imaging; DWI, diffusion weighted imaging; DZ, dizygotic; E, unique environmental; EPI, echo-planar imaging; FA, fractional anisotropy; INFACT, INtravoxel Fiber Assegnment by Continuous Tractography; IQ, intelligence quotient; MZ, monozygotic; MRI, magnetic resonance imaging; ROI, region of interest; SD, standard deviation; SEM, structural equation modeling; SMG, supramarginal gyrus; STV, summed total volume; TB, total brain volume; WAIS, Wechsler adult intelligence scale.

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primarily to the medial/superior frontal, occipital and parahippocam-69 70 pal cortices and the thalamus (Hulshoff Pol et al., 2006; Brans et al., 2010; Bohlken et al., 2014b). There is evidence for considerable dif-7172ferences in the degree to which cortical and subcortical regions are shaped through genes (Wright et al., 2002; Kremen et al., 2010; Rimol 73 et al., 2010; Blokland et al., 2012; Chen et al., 2012). Although, overall 74 75shared genetic influences (i.e. through total brain volume) seem to 76out weight region-specific genetic influences (Giedd et al., 2007). 77 These findings raise the question to what extent the genetic correlation 78between regional gray matter and intellectual ability is mediated 79 through a general overlapping genetic factor (i.e. total brain volume) or through region specific genetic influences. Therefore, the first aim 80 81 of this study was to investigate the influence of environmental and ge-82 netic factors on the association between local gray matter volume and intellectual ability by use of a multivariate twin design, incorporating 83 total brain volume. Associations with intelligence subtests were also 84 explored to test for regional variation according to distinguishable cog-85 nitive abilities. 86

Interregional connectivity may be the key to the neural processes 87 that give rise to intelligence (Gray and Thompson, 2004; Jung and 88 Haier, 2007; Deary et al., 2010; Bullmore and Sporns, 2012). Although 89 genetic influences on white matter microstructure and functional con-90 nectivity have been found to overlap with cognitive ability (Chiang 04 et al., 2009; Koten et al., 2009), it remains to be investigated whether 05 genetic influences on gray matter regions implicated in intelligence 93 could be related to underlying connectivity patterns. We hypothesized 94that the gray matter areas implicated in intelligence form a densely con-9596 nected network to facilitate efficient information transfer. Therefore, 97 the second aim of this study was to characterize the connectivity of 98 gray matter regions that associate with intellectual ability using diffu-99 sion tensor imaging. This was accomplished by comparing the density 100of the structural connectivity underlying the gray matter areas implicat-101 ed in intelligence with a distribution of reference networks.

102 Methods

103 Participants

In this study, 167 adult twins participated. Participants were all part 104 of the U-Twin cohort, which was acquired between 2009 and 2013 105 106 (Bohlken et al., 2014b). The sample consists of 45 complete monozygotic (MZ), 35 complete dizygotic (DZ) twin pairs and 7 individuals from 107 incomplete (1 MZ, 6 DZ) pairs of whom a magnetic resonance imaging 108 109 (MRI) brain scan and intelligence quotient (IQ) were obtained. Of the DZ twins, five pairs were of opposite sex (DOS). All participants were 110 111 between the age of 18 and 67 years (mean = 32.7, SD = 13.6 years). Zygosity was determined by resemblance for polymorphic DNA markers. 112 Upon participation, all subjects gave their written informed consent. 113 This study was approved by the Medical Ethical Committee of the 114 UMC Utrecht and the experiments were in accordance with the Declara-115116 tion of Helsinki. Two DZ twins and one MZ twin did not complete 117 the WAIS III intelligence test; therefore an evaluation of intellectual ability was obtained in 164 participants. Two DZ twins did not com-118plete the MRI part of the study. Therefore a complete structural 119(T1-weighted + DWI (diffusion weighted imaging)) MRI dataset was 120121acquired in 165 participants.

122 Assessment of intellectual ability

An evaluation of intellectual ability was obtained by means of a
 shortened version of the WAIS III general intelligence test, consisting
 of five subtests: Digit Symbol Substitution, Block Design, Arithmetics,
 Digit Span and Information. All five subtests were used to calculate a
 proxy for the full-scale IQ, which will be referred to as IQ from now on.

Brain image acquisition and processing

MRI scans were acquired on a Philips Achieva scanner operating 129 at 3 T, using an eight-channel SENSE head-coil. A T1-weighted 3D 130 fast-field echo scan was acquired from each participant. Scan acquisition 131 was performed using the following parameters: 220 0.8 mm contiguous 132 slices; echo time (TE) 4.6 ms; repetition time (TR) 10 ms; flip angle 8°; 133 in-plane voxel size 0.75×0.75 mm². 134

Cortical volume of 34 anatomically delineated cortical ROIs and 135 the volume of 7 subcortical ROIs was extracted in each hemisphere 136 using the FreeSurfer 5.1.0 structural imaging pipeline (Fischl et al., 137 2004; Kremen et al., 2010). Also, a measure of total brain volume (TB) 138 was calculated by summing the total gray matter volume, the cortical 139 white matter volume and the cerebellar volume. 140

The DWI scan consisted of a single shot EPI-DTI with 30 diffusion 141 weighted volumes (b = 1000 s/mm²) with non-colinear gradient direc-142 tions and five diffusion-unweighted volumes (b = 0 s/mm^2), TR/TE = 143 7035/68 ms, FOV 240 mm, matrix 128/128, 75 slices at 2 mm thickness, 144 no gap, SENSE factor 3, no cardiac gating. Two DWI datasets were ac-145 quired in the transverse plane per subject using the same parameters 146 but with reverse k-space readout, allowing for correction of susceptibil-147 ity artifacts and increasing signal to noise ratio. 148

Preprocessing of the DWI scans was performed with the diffu- 149 sion toolbox of Andersson et al. (Andersson and Skare, 2002; 150 Andersson et al., 2003) and in-house developed software (Mandl Q6 et al., 2010). First, susceptibility artifacts were corrected by calculating 152 a distortion map based on the two b = 0 images acquired with reversed 153 k-space readout. Subsequently it was applied to the two sets of 154 30 direction-weighted images. This resulted in a corrected DWI set 155 consisting of a single b = 0 image and 30 corrected weighted images, 156 thereby avoiding the need for non-linear registration approaches to 157 the T1-weighted images (Andersson et al., 2003). The DWI set was 158 corrected for Eddy-current distortions and small head movements by 159 realigning all scans to the diffusion-unweighted image (Andersson 160 and Skare, 2002).

Tensor fitting was performed using a constrained compressed 162 sensing algorithm called Crossing Fiber Angular Resolution of Intra-163 Voxel structure (CFARI) (Landman et al., 2012). CFARI models the diffusion profile in each voxel as a finite mixture of discrete and independent compartments, defining the diffusivity within each compartment separately. This has the advantage that it provides a robust 167 framework for identifying intra-voxel structure and is able to estimate fiber tracts in areas of high fiber complexity (e.g. crossing fibers) despite the limited number of 30 orientations in which the DWI data were acquired. CFARI is implemented in the Java Image Science Toolkit and is 171 publicly available (http://www.nitrc.org/projects/jist) (Lucas et al., 172 2010).

Tractography was performed using an approach called INtravoxel 174 Fiber Assegnment by Continuous Tractography (INFACT) (Landman 175 et al., 2012), which is a continuous tracking method based on the 176 FACT algorithm (Mori and van Zijl, 2002). All voxels with fractional Q7 anisotropy (FA) > 0.3 were used as starting seeds for tractography. 178 Tracing was ended when a voxel with FA < 0.1 was encountered or 179 when the turning angle exceeded 45°. A Runge-Kutta solver was 180 used for determining tract continuation. After tractography was 181 completed, all fibers shorter than 10 mm were discarded, as they 182 were deemed spurious. Finally, all remaining fibers were linearly ex- 183 tended by 5 mm in the orientation prior to termination to maximize 184 the probability of penetration into the gray matter. The presence of a $\ _{185}$ white matter connection between two gray matter regions was de- 186 termined by labeling each streamline with the gray matter areas it 187 connects based on the anatomical segmentation mask. The anatom- 188 ical segmentation mask was registered to the DWI scan using a six- 189 parameter rigid body transformation. A total of six scans needed to 190 be excluded from further DTI/network analysis due to unreliable 191 measurements. 192

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