Cumulative Genetic Risk and Prefrontal Activity in Patients With Schizophrenia

Esther Walton\textsuperscript{1,2}, Jessica Turner\textsuperscript{3}, Randy L. Gollub\textsuperscript{1,4}, Dara S. Manoach\textsuperscript{1,4}, Anastasia Yendiki\textsuperscript{1}, Beng-Choon Ho\textsuperscript{5}, Scott R. Sponheim\textsuperscript{6}, Vince D. Calhoun\textsuperscript{3,7}, and Stefan Ehrlich\textsuperscript{1,2,*}

\textsuperscript{1}MGH/MIT/HMS Martinos Center for Biomedical Imaging, Massachusetts General Hospital, Charlestown, MA; \textsuperscript{2}Department of Child and Adolescent Psychiatry, University Hospital Carl Gustav Carus, Dresden University of Technology, Dresden, Germany; \textsuperscript{3}The Mind Research Network, Albuquerque, NM; \textsuperscript{4}Department of Psychiatry, University of Iowa, Iowa City, IA; \textsuperscript{5}Department of Psychiatry and the Center for Magnetic Resonance Research, University of Minnesota, Minneapolis, MN; \textsuperscript{6}Department of Electrical and Computer Engineering, University of New Mexico, Albuquerque, NM

*To whom correspondence should be addressed; Translational Developmental Neuroscience Section, Department of Child and Adolescent Psychiatry, University Hospital Carl Gustav Carus, Dresden University of Technology, Fetscherstraße 74, 01307 Dresden, Germany; tel: +49 (0)351-458-2244, fax: +49 (0)351-458-57 54, e-mail: stefan@nmr.mgh.harvard.edu

The lack of consistency of genetic associations in highly heritable mental illnesses, such as schizophrenia, remains a challenge in molecular psychiatry. Because clinical phenotypes for psychiatric disorders are often ill defined, considerable effort has been made to relate genetic polymorphisms to underlying physiological aspects of schizophrenia (so-called intermediate phenotypes), that may be more reliable. Given the polygenic etiology of schizophrenia, the aim of this work was to form a measure of cumulative genetic risk and study its effect on neural activity during working memory (WM) using functional magnetic resonance imaging. Neural activity during the Sternberg Item Recognition Paradigm was measured in 79 schizophrenia patients and 99 healthy controls. Participants were genotyped, and a genetic risk score (GRS), which combined the additive effects of 41 single-nucleotide polymorphisms (SNPs) from 34 risk genes for schizophrenia, was calculated. These risk SNPs were chosen according to the continuously updated meta-analysis of genetic studies on schizophrenia available at www.schizophreniaresearchforum.org. We found a positive relationship between GRS and left dorsolateral prefrontal cortex inefficiency during WM processing. GRS was not correlated with age, performance, intelligence, or medication effects and did not differ between acquisition sites, gender, or diagnostic groups. Our study suggests that cumulative genetic risk, combining the impact of many genes with small effects, is associated with a known brain-based intermediate phenotype for schizophrenia. The GRS approach could provide an advantage over studying single genes in studies focusing on the genetic basis of polygenic conditions such as neuropsychiatric disorders.

Key words: schizophrenia/DLPFC/working memory/intermediate phenotype/fMRI/genetic risk score

Introduction

Family, twin, and adoption studies support a genetic basis for schizophrenia and are broadly consistent with an estimated heritability of 80\%-85\%. Genetic association studies have nonetheless yielded only weak effects for specific points of genomic variation to be related to schizophrenia. Evidence for an association is often inconsistent, ie, genetic polymorphisms identified as risk variants by some studies are often not associated with schizophrenia in other studies. For example, the catechol-O-methyltransferase (\textit{COMT}) Val108/158Met polymorphism, the most-studied common variant in schizophrenia with profound effects on \textit{COMT} enzyme activity and the dopaminergic tone, failed to be associated with the disorder on meta-analytic review.\textsuperscript{2} A more general review of 27 schizophrenia linkage studies reported that no genomic region was implicated in more than 4 of 27 studies.\textsuperscript{3} Even gene variants, which have been repeatedly linked to schizophrenia, such as zinc finger binding protein 804A (\textit{ZNF804A}), confer only a small increment in risk (OR\textless{}1.09).\textsuperscript{3} The reasons for weak and inconsistent specific genetic effects may be 2-fold. First, schizophrenia is a polygenic disease, in which not a single gene, but a combination of several genes with small effects contributes to an overall risk. For polygenic traits such as height single genes could only explain 0.3\%-0.5\% of the phenotypic variance,\textsuperscript{4} whereas combining the effects of more than 180 markers predicted more than 10\%. Second, schizophrenia is an illness with a wide range of symptoms and neuropsychological impairments, which vary within and across individuals. The practice of grouping a diverse patient population into a single diagnostic category based on clinical observation and patient reports, which may not necessarily relate to better-defined
biological constructs, may reduce the statistical power to detect associated risk variants. Furthermore, the complex clinical phenotype and associated comorbidities frequently lead to diagnostic uncertainties. This may be one of the reasons why some genetic factors identified to date confer comparable risk for both schizophrenia and bipolar disorder. On the other hand, different etiological pathways could lead to similar phenotypic expressions creating even more difficulties in mapping specific genetic effects in schizophrenia. In the current study, we aimed to combine 2 new strategies that may help address the aforementioned problems associated with polygenic diseases and phenotypic complexity: an intermediate phenotype approach and the implementation of a genetic risk score (GRS).

"Intermediate phenotypes" (also known as "endophenotypes") are heritable, disease-associated neurophysiological, cognitive, or neurobiological traits. They are thought to be more closely related to the pathophysiology of a disease than clinically observable constructs such as signs, symptoms, or diagnosis and therefore more proximal to the genetic substrate. Intermediate phenotypes may limit phenotypic complexity and reduce genetic heterogeneity thereby facilitating identification of susceptibility genes that underlie pathophysiological aspects of mental disorders.7

Dorsolateral prefrontal cortex (DLPFC) dysfunction during working memory (WM) processing is a widely acknowledged intermediate phenotype for schizophrenia.8 Compared with matched healthy controls, patients display abnormal DLPFC functioning across task difficulties.9 More specifically, DLPFC activity is assumed to follow an inverted U-shaped curve with increasing task difficulty. However, the activity pattern in schizophrenia appears to be shifted slightly so that compared with healthy controls, persons with schizophrenia show more activation for easy tasks but less activation for difficult tasks.10 Thus, patients may exhibit an "inefficiency" of the prefrontal cortex and need to recruit more neural resources than controls for the same task and may show decreased neural activity (hypofrontality) when task difficulty becomes too great and performance declines.10-12 Numerous studies have found associations between genetic risk and DLPFC inefficiency, providing evidence for the heritability of DLPFC function. Two meta-analyses on unaffected relatives of schizophrenia patients and healthy controls reported abnormalities in DLPFC functioning in the majority of reported experiments, substantiating that DLPFC inefficiency is heritable and related to elevated risk for schizophrenia.13,14 Karlsdott et al12 observed that unaffected cotwins were intermediate in performance and prefrontal activity compared with comparison subjects and patients, who showed clear signs of DLPFC inefficiency. Investigating the impact of the COMT Val108/158Met polymorphism on DLPFC activity in schizophrenia patients, unaffected siblings, and healthy controls, Egan et al15 detected an allele dosage-dependent association with DLPFC inefficiency. Prefrontal activity was greater and performance worse with increasing number of Val alleles with siblings being intermediate in performance when compared with patients and controls. Similar results were found in a study by Bertolino et al,16 who showed a COMT genotype-dependent improvement of DLPFC dysfunction after antipsychotic treatment with Val-carriers benefiting the least from the intervention, thus demonstrating a clinical relevance of assessing genetic risk.

To address the polygenic nature of the disorder, we derived a cumulative summary score, which we call a GRS. This score measures risk that is based on the additive effects of several known genetic susceptibility loci for schizophrenia. The implementation of a GRS is a new and promising approach to identify people at risk and has so far been applied in other diseases such as rheumatoid arthritis,17 multiple sclerosis,18 and coronary heart disease.19 A similar approach was used in a genetic association study comparing schizophrenia patients with healthy controls.20 However, at a conventional threshold, the cumulative risk measure explained only 1% of the variance of disorder occurrence. A GRS used to predict an intermediate phenotype of schizophrenia may explain more variance than the prediction of case-control status. Given the polygenic basis and complexity of symptoms in schizophrenia, the combination of an intermediate phenotype and a GRS may improve the statistical power for detecting associations between schizophrenia-related genetic variation and pathophysiological aspects of schizophrenia. The goals of this study were (1) to derive a GRS that conveys the combined risk of several previously identified risk variants for schizophrenia and (2) to investigate the association between the risk score and whole-brain activity during a WM task. We expected an association between the risk score and DLPFC inefficiency.

Methods

Participants

Imaging genetic and behavioral data from participants of the Mind Clinical Imaging Consortium (MCIC) study of schizophrenia from 4 participating sites (the University of New Mexico [UNM], the University of Minnesota [UMN], Massachusetts General Hospital [MGH], and the University of Iowa [UI]) were used to determine genetic polymorphisms in cryo-conserved blood samples and to analyze whole-brain neural activity during a WM task. Out of a total of 285 participants, who passed imaging quality control procedures (see below), blood samples of 180 participants were available for genetic analysis, resulting in a final dataset of 79 schizophrenia patients and 99 healthy controls after genetic quality control steps (see below). Patients had a Diagnostic and Statistical Manual of Mental Disorders, Fourth Edition (DSM-IV)
diagnosis of schizophrenia or schizophreniform disorder, established using a Structured Clinical Interview for DSM disorders and a review of case files by trained clinicians. In the initial cohort, controls were matched to the patient group for age, gender, and parental education and were excluded if they had a history of a medical or axis I psychiatric diagnosis. The majority of participants were of Caucasian descent (94 healthy controls and 60 patients). For additional details about the participants and clinical measures, see online supplementary methods (SM) 1.1, SM 1.2, and references.20–22

Behavioral Task

The Sternberg Item Recognition Paradigm (SIRP) is a WM task, previously shown to consistently activate the DLPFC in healthy controls and schizophrenia patients.11 The SIRP was administered during six 46-s blocks per run for three 360-s runs. In each block, a memory set, composed of 1 (load 1), 2 (load 3), or 5 (load 5) digits, was presented (2 blocks per load condition). The Encode phase was followed by a presentation of 14 digits, one at a time (the probe phase) and participants responded to each probe to indicate whether or not the probe digit was in the memory set. Participants were instructed to respond as quickly and accurately as possible and were given a bonus of 5 cents for each correct response. This bonus was provided after completion of the scan. For additional details about the paradigm, see reference.20,21 The stimuli and responses were presented and collected using E-prime software (EPrime v1.1, Psychology Software Tools, Inc., Pittsburg, PA). Participants were excluded from further analysis, if they completed a block with less than a 78% accuracy rate and/or with more than 6 probes not answered within a block.

Image Acquisition and Preprocessing

Structural magnetic resonance imaging (MRI) data were acquired with either a 1.5T Siemens Sonata (UNM, MGH, and UI) or a 3T Siemens Trio (UMN). Functional MRI (fMRI) data were acquired with either a 1.5T Siemens Sonata (UMN) or a 3T Siemens Trio (UMN, MGH, and UI). For additional details about data acquisition, see online supplementary methods, SM 1.3. Cortical reconstruction and volumetric segmentation based on high resolution structural MRI scans were performed with the FreeSurfer surface reconstruction software http://surfer.nmr.mgh.harvard.edu Functional data were analyzed using fMRIB Software Library (FSL) (http://www.fmrib.ox.ac.uk/fsl). We fit a general linear model to the fMRI time course at each voxel in a whole-brain model to estimate the average activation during the 3 loads of the probe condition in all trials. Equal weight was given to all loads. Quality assurance steps included checks for whole-brain coverage of brain masks, motion and global mean intensity outlier timepoints, alignment of structural and functional scans, and registration problems (Echo Planar Imaging to T1 and T1 to template).

Genotyping

Blood samples were obtained from each participant and sent to the Harvard Partners Center for Genetics and Genomics for DNA extraction. All DNA extraction and genotyping were done blind to group assignment. Genotyping was performed at the Mind Research Network Neurogenetics Core Lab using the Illumina HumanOmni-Quad BeadChip. Further genotyping steps are described in online supplementary methods, SM 1.5. Quality control steps included common standard procedures.23 Single-nucleotide polymorphisms (SNPs) were checked for a genotyping rate of less than 90% and deviation from Hardy-Weinberg Equilibrium with a threshold of 10^{-6}. No SNPs failed these tests, and the total genotyping rate was 99.7%. We removed participants with a genotyping rate of less than 90%. In 2 other cases, where the genotyping rate was above 90% but less than 100%, missing values (one SNP per participant) were replaced with the group mean.18 Analyses were carried out with PLINK, 1.06 (http://pngu.mgh.harvard.edu/purcell/plink).

GRS Calculation

SNPs were selected based on the continuously updated meta-analysis of genetic studies on schizophrenia, available at www.schizophreniaresearchforum.org updated on February 24, 2010. We selected all significant SNPs listed under the “Top Results.” For significance value calculation, see http://www.szgene.org/methods.asp. Given a large proportion of Caucasians in our sample, we excluded SNPs, which were significant for non-Caucasian groups only (n = 5). Furthermore, we excluded common variants other than SNPs (in this case, the Variable Number Tandem Repeat in the haptoglobin gene). The apolipoprotein E (APOE) risk variants for schizophrenia represent a combination of 2 SNPs, rs429358 and rs7412. We therefore calculated APOE haplotypes in PLINK.

Based on these criteria, 55 SNPs were selected, out of which only 37 were present on the Illumina HumanOmni1-Quad BeadChip. Four more target SNPs, which were not on the Illumina chip, were replaced using tagging SNPs with a linkage disequilibrium (LD) r^2 > .8 within a 200 kbp window (based on phase 3 HapMap data). The reference allele of the tagSNP corresponding to the risk allele of the target SNP was also identified using HapMap. This resulted in a total of 41 SNPs in 34 genes, which were used for the GRS calculation (see online supplementary table 1 in SM 1.6). Furthermore, we analyzed the LD structure between these 41 SNPs in our own dataset using Haploview 4.2. Based on a threshold of r^2 > .8, only 2 SNPs (TPH1 gene on chromosome 11) were in LD suggesting a high degree of independence between all other SNPs in the GRS. An additive GRS
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Table 1. Demographic and Clinical Results

<table>
<thead>
<tr>
<th>Site</th>
<th>Sample</th>
<th>Gender (Female)</th>
<th>Race (Caucasians)</th>
<th>Age (y)</th>
<th>Cognitive Function (WRAT-III)</th>
<th>Parental SES</th>
<th>Handedness (0–12)</th>
<th>GRS</th>
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<tr>
<td></td>
<td></td>
<td>n</td>
<td>%</td>
<td>n</td>
<td>%</td>
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<td>SD</td>
<td>Mean</td>
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<td>23</td>
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<td>3</td>
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<td>84</td>
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<td>9.14</td>
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<td>10.53</td>
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<td>38.58</td>
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<tr>
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<td>34.01</td>
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<td>46.21</td>
</tr>
</tbody>
</table>

Note: Means and SDs are given, unless otherwise indicated. UI/MGH/UMN/UNM represent the 4 different study sites. Race and GRs were defined as described under “Methods” section. Parental SES was classified according to Hollingshead and handsedness, determined using the Annett Scale of Hand Preference. WRAT-III, Wide Range Achievement Test; SCZ, patient with schizophrenia; HC, healthy controls; UMN, University of Minnesota; UNM, the University of New Mexico; MGH, Massachusetts General Hospital; UI, the University of Iowa; GRs, genetic risk score; SES, socioeconomic status.

*Significantly different between SCZ patients and healthy normal controls on the basis of a Chi-square (P < .05).

*Significantly different between SCZ patients and healthy normal controls on the basis of a one-way ANOVA (P < .05).

for each participant was calculated based on the following formula:

$$GRS = \sum_{i \in \{1, \ldots, 41\}} w_i X_i,$$

where $w$ is the log-transformed OR for each SNP with $w = \ln(\text{OR}_{SNP})$ and $X$ is the number of risk alleles (0, 1, or 2).18

Allelic ORs for each SNP were taken from the SZGene database. They describe the association between schizophrenia and an allele by comparing the odds of disease in an individual carrying the wild type allele to the odds of disease in an individual carrying the mutation. If ORs were significant for several populations, we used either the overall OR (all) or, if not available, the OR for Caucasians. If an OR for the protective allele was reported, the reciprocal of this reported OR was used. The GRS was weighted by multiplying the number of risk alleles with the logarithmized OR of each SNP to take different effect sizes of SNPs into account. Supplementary table 1 in section SM 1.6 shows the log-transformed OR for each SNP.

Statistical Models
We performed whole-brain analyses investigating the relationship between GRS and WM-induced brain activity for patients and controls using mixed effects models in FSL. All models were cluster-corrected according to FSL default settings with a z value of 2.3 and a P value of .05 and controlled for acquisition site. To account for nonrandom sampling of schizophrenia patients, we explicitly modeled the effects of diagnosis in our main model (model 1). In a second model, we pooled controls and patients (model 2).

To control for population stratification, we used linkage agglomerative clustering, based on pairwise identity by state (IBS) distance. In detail, we first merged the MCIC dataset with the Hapmap phase II dataset, containing 502 participants of the 4 Hapmap populations: Utah residents with ancestry from Northern and Western Europe (CEU), Han Chinese in Beijing, China (CHB), Japanese in Tokyo, Japan (JPT), and Yoruban in Ibadan, Nigeria west Africa (YRI). A symmetric $80 \times 80$ matrix of the IBS distances for all pairs of individuals was created and used for multidimensional scaling analysis. Six dimensions were extracted and included in the first 2 FSL models as covariates to correct for residual population substructure. We also ran a third model (model 3), where we included only participants of Caucasian descent, controlling for diagnosis and acquisition site. Caucasians were defined as $\pm 0.05$ units of deviation from the Hapmap CEU cluster in a plot with the first 2 principal components on the x- and y-axis. Sample characteristic analyses were carried out with SPSS 17.0.

Results

Sample Characteristics
Patients and controls did not differ in age, handedness, parental socioeconomic status (SES), or in their GRS (table 1). Patients, had a significantly lower IQ (Welch test $F = 28.86; P < .001$), were less likely to be female (Pearson Chi-square test $\chi^2 = 5.665; P = .017$) and
included fewer Caucasians (Pearson Chi-square test \( \chi^2 = 13.60; P < .001 \)). Patients and controls were distributed evenly across acquisition sites (table 1). Patients’ and controls’ accuracy on the SIRP were above 78% at all 3 WM loads but on average schizophrenia patients were less accurate (95%) than healthy controls (98%; Welch test due to unequal variances between groups with \( F = 29.54; P < .001 \)).

GRS Characteristics
GRS did not correlate with age, SIRP performance, parental SES, or IQ. Furthermore, there were no significant correlations between GRS and cumulative and current antipsychotic drug dose (for details, see online supplementary methods, SM 1.2), as well as negative and positive symptoms in the patient group (see online supplementary table 2). GRS did not differ between acquisition sites, gender, diagnostic group, or race (table 1) and followed a normal distribution both in patients and in controls (figure 1A). Forty-one SNPs from 34 different genes were included in the GRS. These genes have many different functions but can be categorized into genes involved in neurodevelopment (26%), neurotransmitter systems (32%), and other functions such as cell adhesion, cell cycle, immune response, and transcription (24%; figure 1B).

WM-Related Neural Activity
The SIRP task reliably activated WM-associated brain regions such as the DLPFC and parietal regions (figure 2A). As predicted, we could show a positive association between GRS and neural activity in the left DLPFC (model 1, \( z_{\text{max}} = 3.84, P = .0252 \), cluster-corrected; figure 2B), which was the only surviving cluster in a whole-brain analysis for a GRS main effect in a model covarying for the effects of acquisition site, diagnosis, and population stratification. This finding remained significant in a pooled analysis without controlling for diagnosis (model 2, \( z_{\text{max}} = 3.93, P = .007 \), cluster-corrected). GRS, which combined the impact of many genes with small effects (see online supplementary figure 1), accounted for 3.6% of the total variance (adjusted \( R^2 \), model 1) at the most activated DLPFC locale (\( x, y, z: -26, 50, 8 \)), corresponding to a significant \( R^2 \) change of .04 (\( P = .008 \)). There were no significant diagnosis by GRS interaction effects.

As an alternative way to control for possible bias due to population stratification, we ran a subsequent whole-brain model including only Caucasians (\( n = 154 \)) and covarying for acquisition site and diagnosis (model 3). The effect of GRS on left DLPFC activity remained significant (\( z_{\text{max}} = 3.84, P = .018 \)). We also found a second significant cluster in the left pars triangularis with the effect of GRS pointing in the same direction as in the DLPFC cluster (\( z_{\text{max}} = 4.58, P = .005 \); see online supplementary figure 2). As in the first model, there were no significant diagnosis by GRS interaction effects.

Discussion
Summary of Findings
In this large multicenter study, we found an overall positive correlation between additive GRS for schizophrenia and increased WM-related DLPFC activity. This finding
was highly specific for left DLPFC activity during mental scanning of WM items—a widely acknowledged intermediate phenotype for schizophrenia. The association between the GRS and DLPFC activation was not attributable to symptom severity, population stratification, or group differences in WM performance.

Schizophrenia and Prefrontal Inefficiency
As expected, the SIRP task induced reliable bilateral activations in the DLPFC and parietal lobe, regions typically found in WM tasks. The association between cumulative genetic risk for schizophrenia and WM-induced prefrontal inefficiency is in line with a large body of literature from previous neurophysiology and neuroimaging studies demonstrating DLPFC dysfunction during executive functioning in schizophrenia patients. Frontal abnormalities in schizophrenia patients during rest as well as executive functioning are well established. More than 10 years ago, Manoach et al. confirmed that schizophrenia patients showed a significantly greater magnitude of activation than controls in the DLPFC during WM processing using fMRI. These findings were interpreted as a deficit in automation, i.e., patients may fail to automate WM tasks, which in turn leads to decreased efficiency. The inefficiency hypothesis was supported by Callicott et al., who found that high-performing patients used greater resources in parts of the left DLPFC if compared with healthy controls. Similarly, Potkin et al. observed that patients had an increased DLPFC response already during moderate memory loads when compared with healthy controls. Effects were stronger on the left side. A study on medication-naive patients and controls compared DLPFC activity before and after practice of a WM task. As expected, activity decreased with practice in all groups, but this decrease was smaller in patients in the left DLPFC, even after controlling for behavioral differences. This illustrates nicely the inefficiency hypothesis—even with training patients continue to recruit more neural resources than controls to perform the same tasks. Accordingly, we observed greater prefrontal neural activity with increasing risk for schizophrenia, although task performance was high in healthy controls and schizophrenia patients.

In an additional exploratory model, where we excluded all non-Caucasian participants, we also found a second significant association between genetic risk and WM-induced activation in the left pars triangularis in the same direction as for the DLPFC cluster. This region is part of the midventrolateral prefrontal cortex, which mediates interference resolution during WM tasks. Activation in the same region has also been found to correlate with increasing task demand during a WM task and was interpreted as a compensatory response. Compensatory activation in the ventral prefrontal cortex was further closely associated with DLPFC inefficiency in schizophrenia patients.

Genetic Risk and Prefrontal Inefficiency
The fact that GRS was linked to brain function but not to task performance or symptom severity suggests that the GRS reflects unique genetic aspects of aberrant neural responses related to schizophrenia that are perhaps not well represented in the clinical or cognitive presentations of patients or high-risk individuals. Such a finding bodes well for the validity of intermediate neural phenotypes suggesting that genetic effects are more clearly discerned through biological constructs than descriptions of complex physiological traits.
behavior. Effect sizes for genetic associations with an inter-
mediate phenotype have been found to be consistently
higher than with psychiatric diagnoses indicating that gene
effects may have a higher penetrance at the level of brain
physiology than at the level of higher order constructs such
as behavior. Associations between WM-induced abnor-
malities in prefrontal function and genetic risk variants for
schizophrenia have been repeatedly documented in studies.
The early studies from the Weinberger group estab-
lished relationships between genes with functions in the
dopaminergic system and prefrontal inefficiency during
executive functioning: COMT Val108/158Met risk allele
carriers were found to have increased WM-induced pre-
frontal activation compared with nonrisk homozygotes.31
Furthermore, risk variants of the dopamine D1 and D2
receptor and the gamma amino butyric acid (GABA)
receptor B2 were all associated with prefrontal ineffi-
ciency.32 In addition, risk variants in genes with func-
tions in brain development such as AKTI, NRG1,
DISCI, MTHFR, and DTNBPI and genes related to
other complex functions such as RGS4, APOE, and
ZNF804A were also associated with prefrontal dysfunc-
tion during executive tasks.20,35–40
Combining the effects of several risk variants for
schizophrenia predicted WM-related DLPFC activity
and explained a total amount of variance that exceeded
that of traditional case-control studies.6,41 Thus, by
studying the effects of many markers for schizophrenia
on a brain-based intermediate phenotype, we were able
to detect the additive effects of different biological path-
ways and systems that may impinge on prefrontal cortical
function but perhaps are below the level of detection in
behavior and clinical symptomatology. We did not find
significant differences in GRS, when comparing cases
and controls. Considering that the sample size is compar-
atively small and that the GRS was derived from only a
few previously studied SNPs, this could be due to a
lack of power. Given that the GRS was derived from
risk genes for schizophrenia found in previous genetic asso-
ciation studies and that DLPFC dysfunction is a well-
validated intermediate phenotype for schizophrenia, our
imaging genetics results indicate an important associa-
tion with the disorder that warrants further investigation.
Our GRS comprised genes from several major gene
genes, such as those involved in neurodevelopment
(AHII, AKTI, DISCI, IL1B, MTHFR, NOTCH4,
NRG1, PLXNA2, and RELN) and in the regulation of
neurotransmitter systems like dopamine (COMT,
SLC18A1, DRD1, and DRD2), serotonin (HTR2A
and TPH1), and GABA (GABRB2). Other genes are
involved in processes such as cell adhesion (MDGA1),
immune responses (OPCML and PRSS16), cell cycle
(RPGRIP1L), or transcription (RPP21) (http://www.ncbi.
nlm.nih.gov/gene). Our SNP selection was based on a
meta-analysis of previously published schizophrenia
case-control association studies, and conclusions about
underlying pathological pathways should be considered
with caution. The fact that these SNPs have been previ-
ously linked to schizophrenia and, in combination, predict
a widely acknowledged imaging phenotype for schizo-
phrenia lends support to the hypothesis that abnormal
neurodevelopment processes and dysfunctional neuro-
transmitter systems may have causal effects in the patho-
genesis of this burdensome disorder. However, it should
be noted that we did not explicitly select for WM-related
SNPs and that the level of evidence for each schizophrenia
risk gene varies depending on the currently available case-
control studies. For some of the SNPs included into our
GRS, the preexisting evidence might therefore be only
moderate and warrants replication in independent cohorts.
Also, lacking a complete understanding of the molecular
pathways for these traits and the relevant gene-gene inter-
action effects, our genetic model of risk was additive as-
suming a linear increase in disease susceptibility per risk
allele. With the advancements in computational modeling,
future studies should go beyond the additive model under-
lying our risk score and try to incorporate existing biolog-
ical knowledge and gene-gene interactions.

Conclusions
We derived a GRS, which combined the effects of 41 ge-
netic risk variants for schizophrenia, and demonstrated
that this score predicted DLPFC inefficiency during a
WM task, a common intermediate phenotype in schizo-
phrenia. The finding is in line with a growing number of
reports demonstrating associations between single genetic
risk variants and schizophrenia-related DLPFC dysfunc-
tion and supports an additive genetic risk model for a poly-
genic phenotype. A neural characterization of genetic risk
could help to define system neuroscience models of schizo-
phrenia. Ongoing work will show, if a GRS approach pro-
vides an advantage over studying single genes in predicting
intermediate phenotypes for neuropsychiatric disorders
and if it can be used to estimate the risk for schizophrenia
in susceptible populations.

Supplementary Material
Supplementary material is available at http://
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