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Dissociation of accumulated genetic risk and disease severity in patients with schizophrenia

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Genotype–phenotype correlations of common monogenic diseases revealed that the degree of deviation of mutant genes from wild-type structure and function often predicts disease onset and severity. In complex disorders such as schizophrenia, the overall genetic risk is still often >50% but genotype–phenotype relationships are unclear. Recent genome-wide association studies (GWAS) replicated a risk for several single-nucleotide polymorphisms (SNPs) regarding the endpoint diagnosis of schizophrenia. The biological relevance of these SNPs, however, for phenotypes or severity of schizophrenia has remained obscure. We hypothesized that the GWAS 'top-10' should as single markers, but even more so upon their accumulation, display associations with lead features of schizophrenia, namely positive and negative symptoms, cognitive deficits and neurological signs (including catatonia), and/or with age of onset of the disease prodrome as developmental readout and predictor of disease severity. For testing this hypothesis, we took an approach complementary to GWAS, and performed a phenotype-based genetic association study (PGAS). We utilized the to our knowledge worldwide largest phenotypical database of schizophrenic patients (n > 1000), the GRAS (Göttingen Research Association for Schizophrenia) Data Collection. We found that the 'top-10' GWAS-identified risk SNPs neither as single markers nor when explored in the sense of a cumulative genetic risk, have any predictive value for disease onset or severity in the schizophrenic patients, as demonstrated across all core symptoms. We conclude that GWAS does not extract disease genes of general significance in schizophrenia, but may yield, on a hypothesis-free basis, candidate genes relevant for defining disease subgroups.
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Introduction

In complex disorders lacking clearly identifiable disease-causing factors, such as schizophrenia, the overall genetic risk is often >50%[1] but genotype–phenotype relationships are obscure. Recent genome-wide association studies (GWAS) on schizophrenia, building on very large cohorts of cases and controls, have uncovered and replicated a risk for several single-nucleotide polymorphisms (SNPs). Among the 10 'top hits', reaching genome-wide significance in different populations after multiple-testing correction (even though with low odds ratios), are markers in the major histocompatibility complex region and in TCF4, ZNF804A and NRGN genes.[2–6] The biological relevance of these SNPs, however, for phenotypes or severity of schizophrenia has remained unclear.

All current GWAS data on schizophrenia rely on endpoint diagnosis only and do not allow for genotype–phenotype correlations. In an approach complementary to GWAS, we employed the 'top-10' schizophrenia-associated SNPs that have emerged as genome-wide significant from several GWAS, and explored their potential contribution to the disease phenotype, including positive and negative symptoms, cognitive deficits and neurological signs (including catatonia), and age of onset of the prodrome as developmental readout and predictor of disease severity. For this phenotype-based genetic association study (PGAS), we utilized the GRAS (Göttingen Research Association for Schizophrenia) Data Collection.[7,8]

Materials and methods

Schizophrenic patients. The GRAS study was approved by the Ethics Committees of the Georg-August-University of Göttingen and of participating centers, and comprises at present 1041 patients with confirmed Diagnostic and Statistical Manual of Mental Disorders, fourth edition (DSM-IV)[9] diagnosis of schizophrenia (82.2%) or schizoaffective disorder (17.8%), examined between 2005 and 2010 in 23 collaborating centers all over Germany (Supplementary Table 1).[7,8]

Healthy subjects. Healthy subjects for the case–control study were blood donors (n = 1144), recruited according to national guidelines for blood donation.[7] Comparable to the

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patient population (Caucasian 95.5%; other ethnicities 1.8%; unknown 2.7%), almost all controls were of European–Caucasian descent (Caucasian 97.8%; other ethnicities 2%; unknown 0.2%).

Phenotyping. Comprehensive interviews, testing and clinical ratings were conducted by an invariable team of trained examiners (psychiatrists and psychologists) using the 'GRAS Manual'.[7,8] Diagnoses of schizophrenia/schizo-affective disorders were based on the structured clinical interview for DSM-IV (SCID),[10] substantiated by information from medical records/discharge letters of all the patients. Psychopathological state, symptom severity and functional outcome were evaluated by clinical ratings (positive and negative syndrome scale (PANSS)[11] and global assessment of functioning (GAF).[9]) Neuropsychological testing (subtest 3 of Leistungsprüfsystem,[12] Trail-Making Test,[13,14] Verbal Learning and Memory test[15]) including pre-morbid intelligence (Mehrfachwahl-Wortschatz-Intelligenz test-B[16]) and neurological examination (Cambridge Neurological Inventory[17]) yielding respective composite scores, are described in detail elsewhere.[7,8]

Genotyping. SNP genotyping was performed with SimpleProbes (TIB Molbiol, Berlin, Germany) on LightCycler480 (Roche, Mannheim, Germany). All markers fulfilled the Hardy–Weinberg equilibrium.

Statistical methods. *PGAS*: Phenotype data were standardized to be normally distributed with expectation zero and variance one, and presented such that higher values always indicate better performance. Expected values of univariate and multivariate phenotypes were analyzed with linear models (Rv2.12.0), covariate adjusted and tested for additive effects of GWAS risk alleles with single-locus models (Table 1) and with a 10-loci model (Table 2). The latter simultaneously estimates for all the 10 loci regression coefficients with number of respective GWAS risk alleles. To assess statistical significance, the commonly used Bonferroni correction was employed; however, raw *P*-values are given.

Results

First, we proved in a case–control study (cases $n = 1041$; healthy controls $n = 1144$) that the GRAS population (for patient characteristics see Supplementary Table 1) provides a 'genetic data matrix' that essentially replicates the GWAS results (Table 1, upper part). In fact, screening of the 10 genome-wide hits resulted in a significant association of markers rs6913660, rs13211507 and rs3131296 in the major histocompatibility complex region (chromosome 6), and marker rs2312147 in chromosome 2 (*VRK2* gene) with an increased risk for the disease. Due to the smaller sample size compared with the GWAS studies, leading to lower power (power 70–90% for the SNPs found to reach allelic *P*-values ⩽0.05, average power of 43% over all markers), not all SNPs turned out to be significantly associated with the schizophrenia risk. Nevertheless, all of them exhibit the same direction of association as reported in GWAS.[2–6] In fact, the

GRAS sample has been included in a recent large GWAS follow-up study.[6]

We next conducted PGAS single-locus and PGAS multi-locus quantitative association analyses. In both the procedures we searched for the hypothesized association of markers with the lead symptoms of schizophrenia including developmental readouts/disease severity, both in the form of a composite construct (multivariate phenotype) and of its components separately (individual phenotypes), that is, positive and negative symptoms (PANSS), cognitive deficits (composite of executive function, reasoning, verbal learning and memory) and neurological signs (including catatonia), and/or with the age of onset of the disease prodrome. We further hypothesized that, if the markers were disease specific, they should not be associated with the schizophrenia-unrelated, general disease control variables. Hence, such variables referring to symptoms that are not in any way specific for schizophrenia were also included PANSS general psychopathology (depression, anxiety and others), global assessment of functioning and pre-morbid intelligence (basic cognitive capabilities of an individual before disease onset) (Tables 1 and 2; Supplementary Table 2).

With >1000 patients, the single-locus analysis found none of the schizophrenia phenotypes, neither individual nor multivariate, significantly influenced by any of the markers after multiple-testing correction by Bonferroni (Table 1). We thus wondered whether—instead of single genotypes—the 'genetic load' of a patient, that is, the accumulation of risk genotypes, would give us a clearer signal regarding the contribution of genetic risk to disease phenotype and severity. As illustrated in Figure 1, the composite severity score, built on the basis of the five core symptom variables (see inset), shows an essentially equal distribution of schizophrenia severity from the lowest to the highest genetic load group. As more symptoms do not necessarily reflect more severe disease and certain symptom groups may better associate with genetic load, all five variables were considered separately, too. But again, for the 10 top hits of GWAS, none of the schizophrenia symptom groups was dependent regarding its severity on the increase in the 'genetic load' (Supplementary Table 2, and Supplementary Figure 1). This becomes evident when the composite score data are presented alternatively as percentage severity over all the genetic load groups; there is no tendency of an increase in severe cases with an increase in the number of risk markers (Supplementary Figure 2). Interestingly, grouping the distribution of accumulated risk genotypes in the Icelandic GWAS sample ($n = 582$ schizophrenic individuals) yields a pattern of overall risk distribution that is comparable to that of the GRAS database (Supplementary Figure 3), further supporting the validity of our sample for the PGAS approach to the GWAS hits.

We now pursued the question whether better insight into the phenotypical contribution of the GWAS hits could be obtained by performing a multilocus additive joint model analysis. This analysis weighs each marker according to its estimated relative importance as a risk genotype. Also this approach failed to uncover statistically significant associations after multiple-testing correction. Close to significance are two

Table 1 GWAS replication and PGAS single-locus quantitative association analysis

	Region/neighboring gene (Chromosome)		MHC HIST1H2BJ (Chr.6)	MHC PGBD1 (Chr.6)	MHC NOTCH4 (Chr.6)	NRGN (Chr.11)	ZNF804A (Chr.2)	MHC PRSS16 (Chr.6)	VRK2 (Chr.2)	TAOK2 (Chr.16)	TCF4 (Chr.18)	TCF4 (Chr.18)
	SNP ID		rs6913660	rs13211507	rs3131296	rs12807809	rs1344706	rs6932590	rs2312147	rs4583255	rs9960767	rs4309482
GWAS Replication	Genotype Case(%) / Control(%)		CC 73.4/65.8	TT 86.2/81.9	GG 79.3/74.4	TT 69.7/66.7	TT 38.2/35.2	TT 61.0/57.1	CC 40.8/35.3	TT 33.8/31.5	AA 88.3/89.0	AA 37.7/35.6
			CA 24.3/31.3	TC 13.2/17.5	GA 19.6/24.1	TC 27.8/30.1	TG 49.0/49.3	TC 33.7/37.4	CT 43.3/49.4	TC 47.7/47.5	AC 11.4/11.0	AG 46.3/47.2
			AA 2.3/2.9	CC 0.6/0.6	AA 1.1/1.5	CC 2.6/3.2	GG 12.9/15.4	CC 5.3/5.5	TT 15.9/15.3	CC 18.5/20.9	CC 0.3/0.1	GG 16.0/17.3
	P value		**0.0007**	**0.0195**	**0.0255**	0.3036	0.1562	0.1707	**0.0130**	0.3054	0.5058	0.5301
	Allelic Case(%) / Control(%)		C 85.5/81.4	T 92.8/90.6	G 89.1/86.5	T 83.5/81.8	T 62.6/59.9	T 77.8/75.8	C 62.5/60.0	T 57.6/55.3	A 94.0/94.4	A 60.9/59.2
			A 14.5/18.6	C 7.2/9.4	A 10.9/13.5	C 16.5/18.2	G 37.4/40.1	C 22.2/24.2	T 37.5/40.0	T 42.4/44.7	C 6.0/5.6	G 39.1/40.8
	P value		**0.0004**	**0.0094**	**0.0081**	0.1264	0.0695	0.1122	0.1042	0.1321	0.5568	0.2628
	OR (Risk Allele) (95% CI)		1.34 (C) (1.14, 1.59)	1.34 (T) (1.07, 1.68)	1.28 (G) (1.06, 1.55)	1.13 (T) (0.96, 1.33)	1.12 (T) (0.99, 1.27)	1.12 (T) (0.97, 1.30)	1.11 (T) (0.98, 1.25)	1.10 (T) (0.97, 1.24)	1.08 (C) (0.83, 1.41)	1.07 (A) (0.95, 1.21)
	OR (Risk Allele) (95% CI) in previous GWAS		1.15 (C) (1.10, 1.21)	1.24 (T) (1.16, 1.32)	1.19 (G) (1.13, 1.25)	1.15 (T) (1.10, 1.20)	1.10 (T) (1.07, 1.14)	1.16 (T) (1.11, 1.21)	1.10 (C) (1.06, 1.13)	1.08 (T) (1.05, 1.11)	1.23 (C) (1.15, 1.32)	1.09 (A) (1.06, 1.12)
PGAS	Multivariate phenotype[a]	Allele Effect	0.0054	0.0286	-0.0697	**0.0979**	-0.0174	0.0027	0.0263	0.0064	**0.1158**	-0.0066
		P value	0.8837	0.5731	0.0987	**0.0057**	0.5280	0.9322	0.3107	0.8072	**0.0346**	0.8014
	PANSS positive	Allele Effect	-0.0014	0.0736	-0.0245	0.0317	-0.0043	0.0157	0.0278	0.0522	0.1054	-0.0380
		P value	0.9811	0.3767	0.7237	0.5838	0.9233	0.7597	0.5121	0.2228	0.2396	0.3779
	PANSS negative[a]	Allele Effect	0.0004	0.1260	0.0542	**0.1572**	0.0066	0.0025	-0.0031	0.0813	0.1673	0.0587
		P value	0.9951	0.1398	0.4471	**0.0081**	0.8862	0.9624	0.9441	0.0653	0.0677	0.1870
	Cognitive score[a]	Allele Effect	-0.0521	-0.1136	0.0441	**0.1301**	0.0003	-0.0133	-0.0342	0.0433	0.0851	-0.0298
		P value	0.2802	0.0857	0.4225	**0.0048**	0.9932	0.7459	0.3169	0.2063	0.2382	0.3870
	Total CNI[a]	Allele Effect	0.0458	-0.0135	-0.1239	**0.1626**	-0.0213	0.0229	-0.0137	0.0410	0.0436	-0.0030
		P value	0.4307	0.8655	0.0639	**0.0033**	0.6197	0.6401	0.7373	0.3190	0.6070	0.9425
	Prodromal Onset	Allele Effect	0.0148	0.0462	**-0.1974**	0.0530	-0.0453	-0.0204	**0.0941**	**-0.0976**	0.1480	0.0097
		P value	0.8256	0.6259	**0.0103**	0.4174	0.3683	0.7179	**0.0428**	**0.0387**	0.1472	0.8378
	PANSS general	Allele Effect	-0.0025	0.0656	-0.0040	**0.1392**	-0.0350	-0.0026	0.0340	0.0678	0.1334	0.0244
		P value	0.9679	0.4506	0.9562	**0.0213**	0.4586	0.9604	0.4433	0.1290	0.1524	0.5885
	GAF[a]	Allele Effect	0.0035	-0.0382	-0.1233	0.1147	-0.0139	0.0378	0.0157	0.0307	0.1199	0.0196
		P value	0.9552	0.6569	0.0858	0.0558	0.7658	0.4743	0.7204	0.4891	0.1960	0.6616
	Premorbid Intelligence[b]	Allele Effect	-0.0435	-0.0434	0.0309	0.0542	0.0242	-0.0020	0.0612	-0.0824	0.0100	-0.0067
		P value	0.4770	0.6083	0.6622	0.3589	0.6008	0.9700	0.1617	0.0589	0.9134	0.8786

Abbreviations: CNI, Cambridge neurological inventory; CI, confidence interval; Chr, chromosome; GAF, global assessment of functioning; GWAS, genome-wide association studies; MHC, major histocompatibility complex; OR, odds ratio; PGAS, phenotype-based genetic association study; PANSS, positive and negative syndrome scale; SNP, single-nucleotide polymorphism; SNP ID, SNP identifier.
GRAS sample of schizophrenic patients, $n = 1041$; healthy control sample, $n = 1144$. Upper part shows the case–control genetic association study essentially replicating previous GWAS results. SNPs are presented from left to right in the order of OR (odds ratio). Pearson χ^2-test and Fisher's exact test (both two sided) were used for genotypic and allelic comparisons, respectively. Lower part (PGAS) gives additive effect per copy of GWAS risk allele on expected value of schizophrenia-relevant quantitative phenotypes. The multivariate phenotype combines five schizophrenia core features (positive PANSS, negative PANSS, cognitive score, total CNI and prodromal onset). All phenotypes were standardized to zero mean and variance one and presented such that larger values correspond to better performance (for this purpose, PANSS scores and total CNI were multiplied by −1). General PANSS, GAF and premorbid intelligence were included in the analyses as disease control variables. The estimate of allele effect is negative if carriers of a GWAS risk variant perform worse. It is positive, if carriers of a GWAS risk variant perform better with respect to the expected trait value for the schizophrenic sample. Allele effect size on mean trait is quantified relative to trait variability (standard deviation). P-values below 0.05 were highlighted for optical guidance but are not significant after multiple-testing adjustment.
[a]Corrected for age: PANSS negative, cognitive score and total CNI (for separate analyses and within multivariate phenotype), GAF.
[b]Corrected for language problems: pre-morbid intelligence (898 with no language problems and 108 with correction for language problems).
Exploratory exclusion of non-Caucasian subjects from the GRAS sample ($n = 48$; 4.5%) did not qualitatively alter any of the main findings in this Table.

associations, one between the marker rs3131296 in *NOTCH4* (chromosome 6) and prodromal onset (developmental readout), the other between rs12807809 in chromosome 11 (near *NRGN* gene) and severity of the neurological signs (Table 2). In fact, when ignoring the multiple-testing issue, the latter marker which has a relatively low odds ratio in our case–control study (OR 1.13 here and 1.15 in the GWAS study of Stefansson et al.[5]), shows several association 'signals' that make it attractive for follow-up in future subgroup analyses. However, this marker did not associate with the cognition of schizophrenic subjects in a recent study.[18]

Discussion

How can the overall negative result regarding the phenotypical significance of basically all GWAS 'top-10' genotypes be explained? First, in clear contrast to monogenic diseases,[19] the genetic risk for schizophrenia may not simply be reflected

by phenotypical disease severity or by core symptoms of the disease. Here, an array of environmental risk factors that cannot easily be controlled for might also have a modulating role.[20] Second, the genotype –to phenotype translation may only be visible and valid for a relatively small subgroup of individuals, but still leads to significant (even though low) genetic risk odds ratios in very large GWAS samples. Third, one risk genotype may partly 'neutralize' another one, resulting in risk reduction upon combination rather than accumulation of the genetic load. This latter, seemingly paradox interaction is supported by the observation that the few effects on phenotypes found here in marker rs12807809 (chromosome 11; near *NRGN*) unexpectedly go into the opposite direction (risk genotype shows less severity). Fourth and finally, we cannot rule out that for some analyses the GRAS sample may not have enough power to detect the (certainly weak if any) phenotypical consequences of the 'top 10' GWAS hits. In this context, however, the general question

Table 2 PGAS multilocus quantitative association analysis.

| Region/neighboring gene (chromosome)		MHC HIST1H2BJ (Chr.6)	MHC PGBD1 (Chr.6)	MHC NOTCH4 (Chr.6)	NRGN (Chr.11)	ZNF804A (Chr.2)	MHC PRSS16 (Chr.6)	VRK2 (Chr.2)	TAOK2 (Chr.16)	TCF4 (Chr.18)	TCF4 (Chr.18)
SNP ID		rs6913660	rs13211507	rs3131296	rs12807809	rs1344706	rs6932590	rs2312147	rs4583255	rs9960767	rs4309482
PGAS											
Multivariate phenotype[a]	Allele effect	−0.0024	0.0996	**−0.1081**	**0.0966**	−0.0209	−0.0024	0.0241	−0.0038	**0.1233**	−0.0147
	P-value	0.9669	0.1277	**0.0204**	**0.0062**	0.4466	0.9582	0.3498	0.8846	**0.0254**	0.5749
PANSS positive	Allele effect	−0.0475	0.1224	−0.0710	0.0280	0.0026	0.0144	0.0295	0.0464	0.1124	−0.0428
	P-value	0.6211	0.2615	0.3593	0.6302	0.9542	0.8508	0.4902	0.2844	0.2160	0.3268
PANSS negative[a]	Allele effect	−0.0424	0.1669	0.0004	**0.1455**	−0.0016	−0.0253	0.0023	0.0671	0.1462	0.0462
	P-value	0.6627	0.1285	0.9955	**0.0145**	0.9717	0.7449	0.9573	0.1288	0.1130	0.3024
Cognitive score[a]	Allele effect	−0.0468	−0.1442	0.0972	**0.1262**	−0.0012	0.0449	−0.0274	0.0434	0.0834	−0.0398
	P-value	0.5390	0.0927	0.1108	**0.0066**	0.9738	0.4590	0.4246	0.2077	0.2524	0.2530
Total CNI[a]	Allele effect	0.0617	0.0484	**−0.1535**	**0.1663**	−0.0306	0.0041	−0.0141	0.0247	0.0498	−0.0091
	P-value	0.5025	0.6402	**0.0392**	**0.0027**	0.4770	0.9552	0.7292	0.5490	0.5595	0.8263
Prodromal onset	Allele effect	0.0587	0.1900	**−0.2631**	0.0545	−0.0527	−0.0536	0.0757	**−0.1099**	0.1715	−0.0006
	P-value	0.5733	0.1125	**0.0020**	0.4038	0.2967	0.5183	0.1057	**0.0212**	0.0966	0.9903
PANSS general	Allele effect	−0.0049	0.0916	−0.0425	**0.1312**	−0.0391	−0.0239	0.0368	0.0583	0.1184	0.0150
	P-value	0.9609	0.4169	0.5992	**0.0311**	0.4110	0.7644	0.4089	0.1962	0.2089	0.7418
GAF[a]	Allele effect	−0.0260	0.0129	−0.1548	0.1096	−0.0157	0.0685	0.0192	0.0229	0.1166	0.0141
	P-value	0.7905	0.9073	0.0531	0.0686	0.7394	0.3825	0.6637	0.6096	0.2136	0.7547
Pre-morbid intelligence[b]	Allele effect	−0.0744	−0.0148	0.0496	0.0533	0.0335	0.0449	0.0576	−0.0739	0.0097	−0.0086
	P-value	0.4439	0.8928	0.5299	0.3714	0.4741	0.5648	0.1920	0.0946	0.9171	0.8469

Abbreviations: CNI, Cambridge neurological inventory; Chr, chromosome; GAF, global assessment of functioning; MHC, major histocompatibility complex; PGAS, phenotype-based genetic association study; PANSS, positive and negative syndrome scale; SNP, single-nucleotide polymorphism; SNP ID, SNP identifier. GRAS sample of schizophrenic patients, $n = 1010$ individuals with complete genotype information for 10 GWAS loci. SNPs are presented from left to right in the order of OR (odds ratio; compare Table 1). Instead of considering the SNPs separately as in Table 1, their joint additive effect is estimated simultaneously, allowing for different relative weights of the loci. Allele effect coefficients (and their associated *P*-values) for each additive variant in a multiple regression model are displayed. The multivariate phenotype combines five schizophrenia core features (positive PANSS, negative PANSS, cognitive score, total CNI and prodromal onset). All phenotypes were standardized to zero mean and variance one and presented such that larger values correspond to better performance (for this purpose, PANSS scores and total CNI were multiplied by −1). General PANSS, GAF and pre-morbid intelligence were included in the analyses as disease control variables. The estimate of allele effect is negative if carriers of a GWAS risk variant perform worse. It is positive, if carriers of a GWAS risk variant perform better with respect to the expected trait value for the schizophrenic sample. Allele effect size on mean trait is quantified relative to trait variability (standard deviation). Weight estimates with *P*-values below 0.05 were highlighted for optical guidance but are not significant after multiple-testing adjustment.
[a]Corrected for age: PANSS negative, cognitive score and total CNI (for separate analyses and within multivariate phenotype), GAF.
[b]Corrected for language problems: pre-morbid intelligence (886 with no language problems and 107 with correction for language problems).
Exploratory exclusion of non-Caucasian subjects from the GRAS sample ($n = 48$; 4.5%) did not qualitatively alter any of the main findings in this Table.
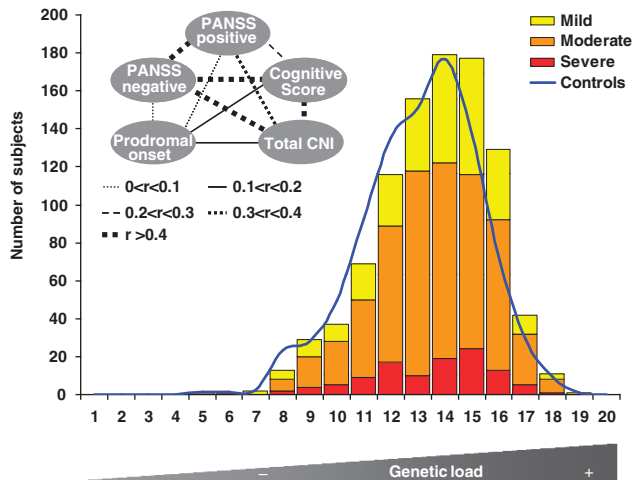
Figure 1 Distribution of phenotype severity and cumulative genetic load with respect to the number of GWAS-identified 'top-10' risk SNP alleles in the GRAS population (bar graph). Phenotype severity is based on a composite score of the five core features of schizophrenia, displayed as inset (including intercorrelations between items). Score range in the GRAS sample is divided into three equal parts and ranked as mild, medium and severe disease phenotype. The blue line denotes the comparison of the risk SNP allele distribution in the healthy control sample.

arises of how much clinical significance a genetic association may have if several thousand patients are needed to reveal a tiny effect on disease severity or phenotype. Thus, building on >1000 patients, one would at least have expected some more signals to pop up (that is, more nominal P-values around 0.05 in the PGAS part of Table 1). In fact, the estimated power assessed according to Lettre et al.,[21] who performed simulations in a similar context (sample size, normally distributed phenotype), amounts to overall around 80% at the $\alpha = 0.05$ level for the PGAS approach. The presented additive model, chosen to match our hypothesis of cumulated risk, performed similarly to a co-dominant one but appeared to be slightly more powerful.

After all, we note that entirely different genotypes that were never found to be significantly associated with any schizophrenia risk in GWAS, still may profoundly modulate the schizophrenic phenotype, for example, of genes encoding neuregulin-1, complexin 2 or COMT.[7,22,23] On the other hand, GWAS finds may be of general rather than disease-specific significance. Several studies have for instance suggested that schizophrenia and affective disorders are on a continuum of liability. Genetic linkage and association studies have proposed common disease loci for both the disorders.[2,24] Likewise, family studies show that first-degree relatives of bipolar patients have a higher risk for schizophrenia compared with first-degree relatives of healthy controls.[25,26] Also other psychiatric diseases like alcoholism or major depression have been found to be associated with certain schizophrenia risk genes, for example, DISC1.[27] Thus, exploration of many other phenotypes available in the GRAS database, including candidate intermediate phenotypes[28,29] or those reflecting a more dimensional approach to the disease,[30] might potentially be interesting. Purpose of the present study, however, was to cover the lead symptoms of schizophrenia in the first place.

To conclude, GWAS approaches in diseases as complex as schizophrenia do not lead to the reconstruction of a 'common

disease mechanism' or to the discovery of 'classical disease genes', as such genes obviously do not exist. What makes our study important for the clinician is that we can show, for the first time, that the combination of a whole battery of genetic pre-disposing factors (the 'top 10' GWAS finds in schizophrenia) in individual patients will not make their schizophrenic phenotype any different or worse than that of those patients who do not carry these genetic factors. Importantly, however, GWAS results may guide, on a hypothesis-free basis, to the identification of totally unexpected candidate genes involved in certain disease aspects in subgroups of patients, as they can be defined by PGAS. In order to get closer to understanding the disorders as complex and heterogeneous as schizophrenia, GWAS and PGAS will have to go hand in hand.

Conflict of interest

The authors declare no conflict of interest.

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Supplementary Information accompanies the paper on the Translational Psychiatry website (http://www.nature.com/tp)