

1 Transcriptome association studies of 2 neuropsychiatric traits in African 3 Americans implicate *PRMT7* in 4 schizophrenia

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14 ABSTRACT

15 In the past fifteen years, genome-wide association studies (GWAS) have provided novel insight into
16 the genetic architecture of various complex traits; however, this insight has been primarily focused on
17 populations of European descent. This emphasis on European populations has led to individuals of
18 recent African descent being grossly underrepresented in the study of genetics. With African Americans
19 making up less than two percent of participants in neuropsychiatric GWAS, this discrepancy is magnified
20 in diseases such as schizophrenia and bipolar disorder. In this study, we performed GWAS and the
21 gene-based association method PrediXcan for schizophrenia ($n=2,256$) and bipolar disorder ($n=1,019$) in
22 African American cohorts. In our PrediXcan analyses, we identified *PRMT7* ($P = 5.5 \times 10^{-6}$, local false
23 sign rate = 0.12) as significantly associated with schizophrenia following an adaptive shrinkage multiple
24 testing adjustment. This association with schizophrenia was confirmed in the much larger, predominantly
25 European, Psychiatric Genomics Consortium. In addition to the *PRMT7* association with schizophrenia,
26 we identified rs10168049 ($P = 1.0 \times 10^{-6}$) as a potential candidate locus for bipolar disorder with highly
27 divergent allele frequencies across populations, highlighting the need for diversity in genetic studies.

28 INTRODUCTION

29 Individuals of recent African ancestry have been grossly underrepresented in genomic studies. African
30 American participants make up about 2.0% of all GWAS subjects (Sirugo et al., 2019). Specifically,
31 individuals of African ancestry make up only 1.2% of all neuropsychiatric GWAS (Quansah and Mc-
32 Gregor, 2018). With the advent of polygenic risk scores, accuracy in disease prediction is critical to the
33 development of precision medicine (Khera et al., 2018); however, the lack of representative diversity in
34 the study of genomics has impacted the accuracy of genetic risk prediction across diverse populations.
35 Despite similar incidences of schizophrenia across European and African ancestry populations (De Candia
36 et al., 2013; Whiteford et al., 2013), Africans have been predicted to have significantly less disease risk
37 than their European counterparts using current GWAS summary statistics (Martin et al., 2017). Inaccuracy
38 in predicting disease risk across populations can lead to further disparities in health and treatment of
39 underrepresented populations. To prevent misclassification of genetic risk, further work in the genetics un-
40 derlying complex traits in African Americans is needed (Manrai et al., 2016). In an attempt to address this
41 discrepancy in genetic risk prediction, we performed a series of genetic association tests for schizophrenia
42 and bipolar disorder in two cohorts of African American individuals (Manolio et al., 2007; Suarez et al.,
43 2006; Smith et al., 2009).

44 Schizophrenia and bipolar disorder are two heritable neuropsychiatric disorders whose genetic com-

ponents have been attributed to the cumulative effect of thousands of loci across the genome (Ripke et al., 2014; Zhiqiang et al., 2017; Ikeda et al., 2017b). Past work shows that the genetic architectures of these two disorders significantly overlap (Bhalala et al., 2018; Allardyce et al., 2018; Consortium, 2009; Stahl et al., 2019). Up to this point, the largest GWAS of schizophrenia and bipolar disorder comprise hundreds of thousands of individuals primarily of European descent (Stahl et al., 2019; Ripke et al., 2014; Zhiqiang et al., 2017). While studies of neuropsychiatric diseases in European and Asian ancestry populations continue to grow, the scarcity of studies in African American populations persists (Ikeda et al., 2017a).

To date, one of the largest GWAS of schizophrenia in an African American population was completed by the Genetic Association Information Network (GAIN) (Manolio et al., 2007); however, this study found no SNPs to be genome-wide significant and offered little insight into the potential function of genes in schizophrenia in African Americans. The GAIN has also performed one of the largest GWAS of bipolar disorder in African Americans (Manolio et al., 2007; Smith et al., 2009). Similar to the findings of the GAIN GWAS of schizophrenia, Smith et al. found no SNPs significantly associated with bipolar disorder (Smith et al., 2009). In addition to a traditional GWAS using a logistic regression, we performed two gene-level association tests: PrediXcan and MultiXcan (Gamazon et al., 2015; Barbeira et al., 2019). PrediXcan offers a series of advantages to SNP-level analyses in detecting genetic association and functionality. First, since tests are being conducted at the gene level, PrediXcan has a lower multiple-testing burden compared to GWAS. Additionally, our understanding of functional pathways are more easily constructed for genes compared to SNPs. By using gene expression as an intermediate phenotype between genetic variation and complex phenotypes, PrediXcan results can help elucidate genetic mechanisms compared to GWAS. We completed our genetic association tests for schizophrenia in a cohort of 2,256 self-identified African American individuals from the genome-wide linkage scan of African American families (Suarez et al., 2006) and the GAIN (Manolio et al., 2007). For our study of bipolar disorder, we performed these association tests in 1,019 African American individuals from the GAIN. Using these data, we identified one gene significantly associated with schizophrenia and tested it for replication in the Psychiatric Genomics Consortium (PGC) GWAS of schizophrenia (Barbeira et al., 2018; Ripke et al., 2014; Sklar et al., 2011).

METHODS

Cohorts

The individuals in the cohorts used in this study were all self-identified African Americans. We acquired genotype and phenotype information for these individuals from the National Center for Biotechnology Information (NCBI) database of Genotypes and Phenotypes (dbGaP). The project was confirmed exempt from human subjects federal regulations under exemption number 4 by the Loyola University Chicago Institutional Review Board (project number 2014). Whole Genome genotypes and phenotypic information were acquired from three separate accessions in dbGaP (Table 1). All genotype information for these three accessions were acquired using Affymetrix Genome-Wide Human SNP Array 6.0, covering 934,940 SNPs. In total, our studies included 2,256 and 1,019 individuals for schizophrenia and bipolar disorder, respectively.

Table 1. Cohort characteristics: Three separate cohorts were integrated into two main cohorts characterized by phenotype. After merging the two schizophrenia cohorts using PLINK, the number of post-QC SNPs became identical.

dbGaP Accession Number	phs000021.v3.p2	phs000167.v1.p1	phs000017.v3.p1
Phenotype	Schizophrenia	Schizophrenia	Bipolar Disorder
Total Individuals	2,220	120	1,045
Cases	1,241	15	359
Controls	979	105	686
post-QC Individuals	2,256	2,256	1,019
pre-QC SNPs	845,814	909,622	867,411
post-QC SNPs	742,015	742,015	721,050
Post-Imputation SNPs ($r^2 > 0.8$, MAF > 0.01)	12,780,487	12,780,487	12,799,548

Case-control criteria for both phenotypes was determined using the DSM-IV (Diagnostic and Statistical

84 Manual of Mental Disorders) as described previously (Smith et al., 2009). Following the update from
85 DSM-IV to DSM-V, 40 individuals, previously identified as cases under DSM-IV were removed from the
86 study because they no longer met the case criteria for DSM-V.

87 **Quality Control and Imputation**

88 Following the download of data from dbGaP, we isolated genotypes and phenotypes of African Americans
89 from those of European Americans or individuals of unidentified ethnicities for each cohort. We merged
90 the PLINK binary files from the two schizophrenia studies. At this point 906,425 SNPs were genotyped
91 in 2,256 individuals in the schizophrenia cohort and 867,411 SNPs and 1,019 individuals in the bipolar
92 disorder cohort. While the schizophrenia cohort included individuals from GAIN and the genome-wide
93 linkage scan of African American families (Suarez et al., 2006; Manolio et al., 2007), throughout the
94 rest of the paper, we will refer to the combined cohort as GAIN. In each cohort, we removed SNPs with
95 genotyping call rates less than 99% and those that significantly deviated from Hardy-Weinberg equilibrium
96 ($P < 1 \times 10^{-6}$). We then removed individuals with excess heterozygosity. Individuals greater than three
97 standard deviations from mean heterozygosity were removed from the study. We used EIGENSOFT
98 `smartpca` (Patterson et al., 2006) to generate the first ten principal components, which were used to
99 confirm self-identified ancestry (Figs. S1 and S2). After this quality control (QC), we had 742,015 SNPs
100 and 2,256 individuals total in the schizophrenia cohort and 721,050 SNPs and 1,019 individuals in the
101 bipolar disorder cohort (Purcell et al., 2007).

102 From here, the filtered data from each cohort were uploaded to the University of Michigan Imputation
103 Server for genotype imputation (Das et al., 2016). The genotypes for each cohort were imputed using
104 Eagle version 2.3 for phasing and 1000 Genomes Phase 3 version 5 (1000G) as our reference panel (Gibbs
105 et al., 2015). After this, we downloaded the imputed data from the Michigan Imputation Server and
106 converted it to PLINK binary format. We then filtered the data by removing SNPs with imputation $r^2 < 0.8$
107 and minor allele frequency (MAF) < 0.01 . At this point, we were left with 12,780,487 and 12,799,548
108 SNPs in the schizophrenia and bipolar disorder cohorts, respectively. We explored Consortium on Asthma
109 among African-ancestry Populations in the Americas (CAAPA) as an alternative reference panel for
110 imputation, but 1000G imputed more SNPs meeting our filters while simultaneously imputing SNPs with
111 MAFs identical to those imputed from CAAPA at our chosen imputation r^2 and MAF thresholds (Fig.
112 S3) (Mathias et al., 2016).

113 **Genome-Wide Association Study**

114 Using PLINK, we performed a logistic regression of the phenotype using the first ten genotypic principal
115 components as covariates to account for population structure. We used a significance threshold of P
116 $< 5 \times 10^{-8}$ to identify significantly associated SNPs. Plots were generated from PLINK results using the
117 web-based tool LocusZoom (Pruim et al., 2011).

118 **PrediXcan**

119 We performed the gene-based association test PrediXcan on both phenotypes, schizophrenia and bipolar
120 disorder, in this study. PrediXcan functions by predicting an individual's genetically regulated gene
121 expression levels using tissue-dependent prediction models trained using reference transcriptome data
122 (Gamazon et al., 2015). For our experiments we tested each phenotype across 55 prediction models. 48 of
123 these models were trained on 48 tissues in GTEx version 7 (Barbeira et al., 2018). Six of these models
124 were generated from monocyte transcriptomes of individuals in the Multi-Ethnic Study of Atheroscle-
125 rosis (MESA) cohort. The MESA models, the most diverse set of published predictors to date, were
126 built from genotypes and transcriptomes of self-identified African American, Hispanic, and European
127 individuals (Mogil et al., 2018). These models can be found at <http://predictdb.org/>. We also
128 used a model built from dorsolateral prefrontal cortex (DLPFC) data from the CommonMind Consortium
129 (Huckins et al., 2019). To impute the gene expression levels, the PLINK binary files from each cohort
130 had to be converted to PrediXcan dosage files. To do this, we used the conversion script provided at
131 <https://github.com/hakyimlab/PrediXcan/tree/master/Software>. After predict-
132 ing a genetically regulated level of expression, we tested each expression level for association with the
133 phenotype of interest. Since PrediXcan does not have a flag for performing a logistic regression with
134 covariates, we performed a logistic regression of the phenotype with the first ten principal components to
135 generate a residual phenotype in order to account for population structure. We then performed a linear
136 regression with the residual phenotype and gene expression level for each gene.

137 Following the PrediXcan association tests, we adjusted for multiple testing using the adaptive shrinkage
138 approach implemented in the R package *ashr* (Stephens, 2017). Using this package, we calculated the
139 local false sign rate (lfsr) for each test, which is similar to traditional false discovery rate approaches,
140 but takes into account both the effect sizes and standard errors of each gene-tissue pair ($n=248,605$).
141 In addition, this empirical Bayes approach uses the assumption that the distribution of actual effects is
142 unimodal with the mode at 0. We set our significance threshold for gene-tissue pairs at $\text{lfsr} < 0.2$.

143 Due to the dearth of African American neuropsychiatric cohorts, replication could not be completed
144 in an independent African American cohort. To validate our results, we compared our findings to the
145 association results of a meta-analysis of the PGC GWAS summary statistics completed using S-PrediXcan
146 (Barbeira et al., 2018; Ripke et al., 2014; Sklar et al., 2011). We performed S-PrediXcan in both
147 phenotypes using predictors from GTEx. The 2014 Ripke et al. study originally included 36,989 cases
148 and 113,075 controls composed of about 96.5% European and 3.5% Asian individuals. The 2011 Sklar et
149 al. study included 7,481 cases and 9,250 controls of primarily European and Asian descent.

150 **MultiXcan**

151 Following imputation of gene expression levels, we performed MultiXcan (Barbeira et al., 2019), a gene-
152 based association test that combines information across multiple tissues while taking their correlation into
153 account. Using the predicted expression levels in 48 tissues across GTEx, we performed MultiXcan on
154 both of the disease phenotypes.

155 **RESULTS**

156 **Schizophrenia Gene-based Association Study**

157 To better understand the genetic architecture of schizophrenia in African Americans, we performed
158 transcriptome-wide association studies using prediction models built in 55 tissues. In the GAIN cohort of
159 2,256 individuals (969 controls and 1287 cases), we predicted gene expression across 48 tissues in GTEx,
160 six models built from monocytes across MESA, and DLPFC from CommonMind (Barbeira et al., 2018;
161 Wheeler et al., 2016; Mogil et al., 2018; Huckins et al., 2019).

162 *PRMT7* was the most significantly associated gene with an lfsr of 0.119 and a p-value of 5.49×10^{-6}
163 in the atrial appendage of the heart (Table 2, Figs. 1 and 2). Increased predicted expression of *PRMT7*
164 associated with schizophrenia in 32 of 33 tissues in GTEx tested (Figs. 2 and 3). Effect sizes were
165 also positive for *PRMT7* associations with schizophrenia in 42 of 42 tissues tested in our S-PrediXcan
166 application to the PGC data (Ripke et al., 2014) (Fig. 3). Of the 42 tissues tested, 30 associations were
167 statistically significant ($P < 0.0012$) after Bonferroni adjustment for the number of tissues tested and all
168 42 tissues had $P < 0.05$ (Table S1). We found no significant gene-tissue associations using the MESA or
169 DLPFC models. While *PRMT7* in atrial appendage had the lowest lfsr across all models, *RP11-646C24.5*
170 had a lower p-value (Fig. 1), but high lfsr in both pancreas (lfsr = 0.860) and sigmoid colon (lfsr = 0.851).
171 Notably, the standard error in both of these tissues was over twice the size of that of *PRMT7*. Unlike more
172 traditional false discovery rate approaches such as Benjamini-Hochberg, both effect size and standard error
173 are used in an empirical Bayesian framework to calculate lfsr and thus the gene with the lowest p-value
174 may not be the gene with the lowest lfsr (Stephens, 2017). We found no significant associations with the
175 MultiXcan, cross-tissue model.

176 **Bipolar Disorder Gene-Based Association Study**

177 To develop a better understanding of the genetic mechanisms governing bipolar disorder in African
178 Americans, we performed PrediXcan in a cohort of 1,019 individuals (671 controls and 348 Cases).
179 Similar to our gene-based association study of schizophrenia, we performed our tests across the same 55
180 gene expression prediction models in our bipolar disorder study.

181 In the GAIN cohort of 1,019 African American individuals, no genes were identified to be significantly
182 associated with bipolar disorder. Increased predicted expression of *GREM2* in testis was the most
183 associated ($P = 2.20 \times 10^{-5}$) gene-tissue pair with bipolar disorder (Fig. 4). *KCNMB3* had the lowest lfsr
184 at 0.919. We also found no significant associations with the MultiXcan, cross-tissue model.

185 **Schizophrenia SNP-level Association Test**

186 We performed a GWAS across greater than 12 million SNPs following imputation to help elucidate the
187 role specific SNPs play in the genetics of schizophrenia in African Americans. We used the first ten

Table 2. Top PrediXcan results for schizophrenia in African Americans (GAIN cohort) by local false sign rate (lfsr). *PRMT7* makes up five of the top eight associated gene-tissue pairs by lfsr. BP are reported as transcription start site for the respective genes.

Gene	Beta	t	p	se(beta)	Tissue (Predictor)	lfsr	CHR	BP
<i>PRMT7</i>	0.10	4.56	5.49×10^{-6}	0.022	Heart_Atrial_Appendage	0.119	16	68392457
<i>PRMT7</i>	0.09	4.22	2.50×10^{-5}	0.020	Cells_Transformed_fibroblasts	0.225	16	68392457
<i>PRMT7</i>	0.08	4.00	6.63×10^{-5}	0.019	Heart_Left_Ventricle	0.353	16	68392457
<i>PRMT7</i>	0.11	4.23	2.47×10^{-5}	0.025	Adrenal_Gland	0.365	16	68392457
<i>TBC1D2</i>	0.12	4.24	2.34×10^{-5}	0.028	Brain_Putamen_basal_ganglia	0.468	9	100961311
<i>NPC1</i>	0.07	3.83	1.29×10^{-4}	0.019	Thyroid	0.484	18	21086148
<i>EIF2S2P3</i>	0.08	3.81	1.41×10^{-4}	0.022	Brain_Amygdala	0.558	10	94428502
<i>PRMT7</i>	0.08	3.73	1.97×10^{-4}	0.020	Cells_EBV-transformed_lymphocytes	0.578	16	68392457

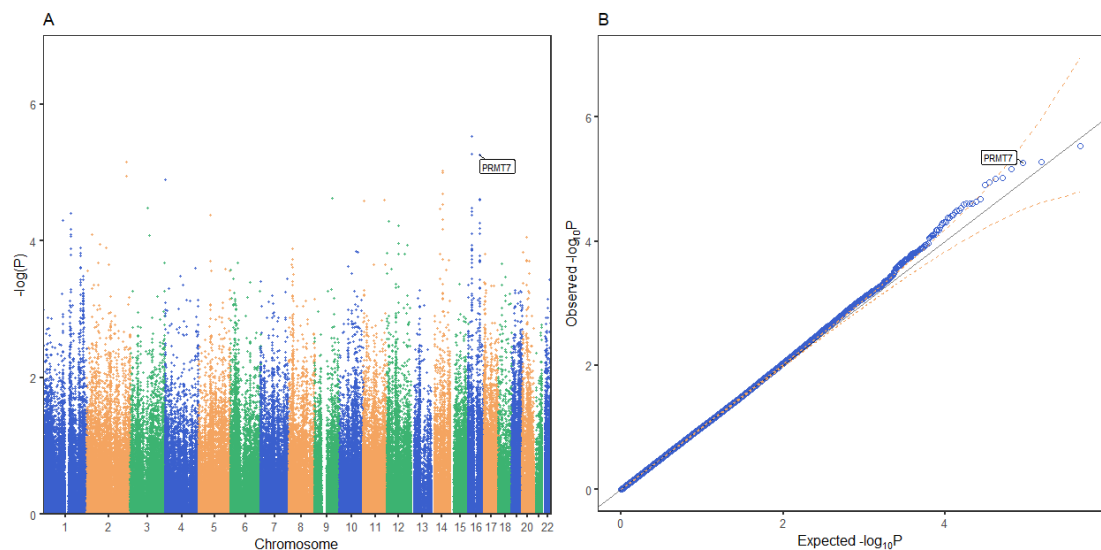


Figure 1. PrediXcan association results for schizophrenia in GAIN African Americans. Each point on the Manhattan (A) and Quantile-Quantile (B) plots represents one gene-tissue test for association with schizophrenia using GTEx version 7 gene expression prediction models. *PRMT7* expression in atrial appendage of the heart is labeled in both plots since it had the lowest lfsr (local false sign rate) of all tissues (lfsr = 0.119). Predicted *RP11-646C24.5* expression in pancreas and sigmoid colon associations are represented as the two points with lower p-values than *PRMT7*, respectively, but lfsr was greater than 0.8 for each association. Unlike more traditional false discovery rate approaches such as Benjamini-Hochberg, the gene with the lowest p-value may not be the gene with the lowest lfsr especially if the standard error of the effect size estimate is high (Stephens, 2017).

188 principal components as covariates for our logistic regression in order to adjust for population stratification
 189 in the cohort. In our SNP-level GWAS, we found no significantly associated SNPs; however, one of the
 190 most associated SNPs, rs8063446 ($P = 2.66 \times 10^{-6}$), is located at the *PRMT7* locus (Fig. 5) While not
 191 genome-wide significant, the most associated SNP in our study was rs112845369 ($P = 1.094 \times 10^{-6}$) on
 192 chromosome 15.

193 Bipolar Disorder SNP-level Association Test

194 We also performed a logistic GWAS in over 12 million SNPs in an attempt to understand the role specific
 195 SNPs play in the genetics of bipolar disorder. We similarly used the first ten principal components
 196 to adjust for population stratification in the bipolar disorder cohort. Similar to our findings in our
 197 schizophrenia GWAS, we identified no SNPs significantly associated with bipolar disorder after adjusting
 198 for multiple tests. rs10168049 on chromosome 2 was the most associated SNP ($P=1.04 \times 10^{-6}$). While

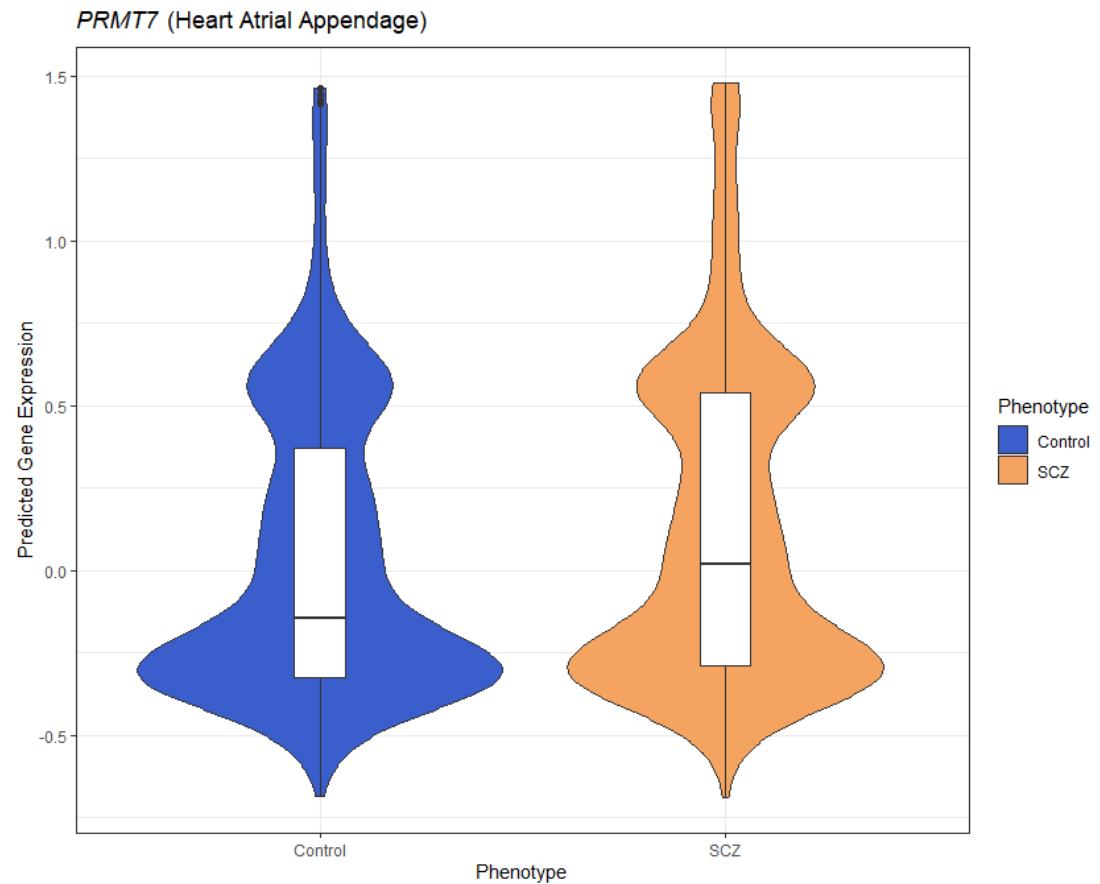


Figure 2. Predicted *PRMT7* expression is higher in schizophrenia cases than controls in GAIN. The violin plot represents the differences in density of predicted gene expression levels of *PRMT7* between cases (SCZ) and controls in heart atrial appendage from GTEx ($P = 5.49 \times 10^{-6}$).

199 not significantly associated with bipolar disorder, rs10168049 has a minor allele frequency of 0.465 in
200 African 1000G populations compared to those of European and East Asian populations with minor allele
201 frequencies of 0.042 and 0.097, respectively (Fig. 6) (Sherry, 2001; Gibbs et al., 2015). All PrediXcan and
202 GWAS summary statistics for both diseases are available at [https://github.com/WheelerLab/](https://github.com/WheelerLab/Neuropsychiatric-Phenotypes)
203 Neuropsychiatric-Phenotypes.

204 DISCUSSION

205 We performed gene-level (PrediXcan) and SNP-level association studies for schizophrenia and bipolar
206 disorder in African Americans from GAIN (Suarez et al., 2006; Manolio et al., 2007). We used summary
207 statistics from the predominantly European PGC to replicate our findings (Ripke et al., 2014; Sklar et al.,
208 2011).

209 A potential role for *PRMT7* in schizophrenia

210 *PRMT7* was significantly associated with schizophrenia in our PrediXcan analyses and one of the most
211 associated SNPs in our GWAS is just upstream of the gene (Figs. 1 and 5). Schizophrenia was associated
212 with increased expression of *PRMT7* in 32 of 33 tissues in which it was predicted in the GAIN cohort.
213 When S-PrediXcan was applied to the PGC summary stats, increased expression of *PRMT7* was associated
214 with schizophrenia in all 42 tissues in which expression was predicted (Fig. 3). *PRMT7* made up five
215 of the eight most associated gene-tissue pairs (Table 2). *PRMT7* has previously been associated with
216 SBIDD syndrome, an intellectual disability syndrome (Agolini et al., 2018). Its function in the disorder
217 remains unclear, but *PRMT7* has a functional role in neuronal differentiation, which could be a potential

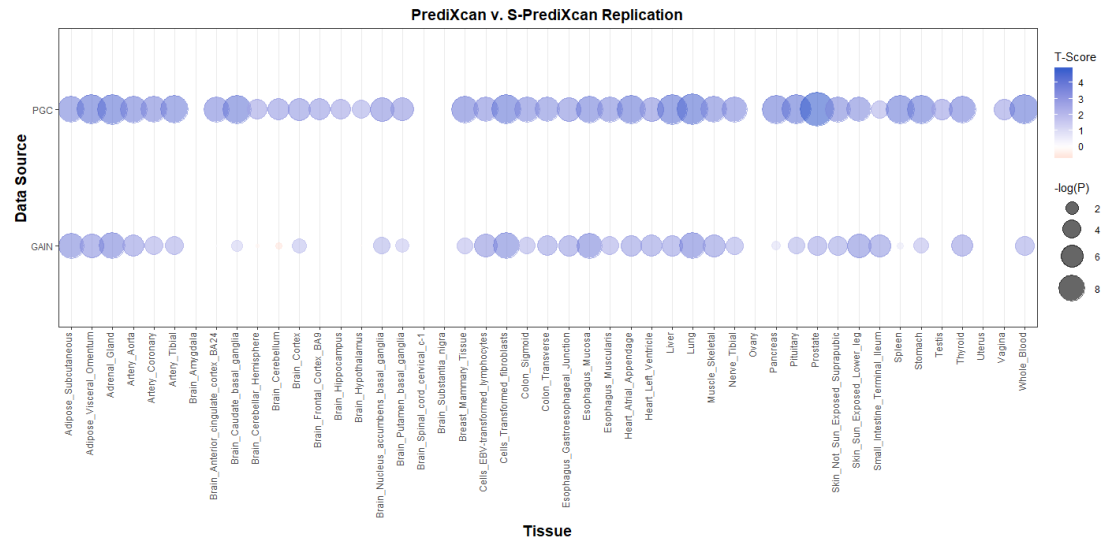


Figure 3. *PRMT7* PrediXcan discovery (GAIN) and replication (PGC) results across tissue models. In each bubble plot, the radius of the bubble is representative of the significance of *PRMT7* association with SCZ. The color of the bubble represents the test statistic with blue representing a positive direction of effect and red representing a negative direction of effect.

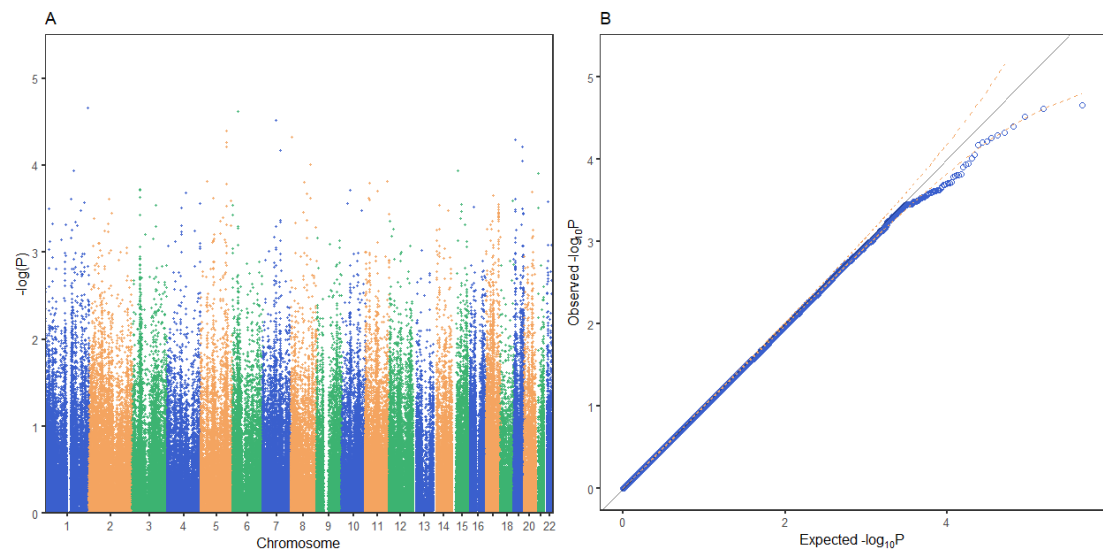


Figure 4. PrediXcan association results for bipolar disorder in GAIN. Each point on the Manhattan (A) and Quantile-Quantile (B) plots represents a gene-tissue association test for our study of bipolar disorder using GTEx models. *GREM2* on chromosome 1 was the gene most associated with bipolar disorder in our study. All of the gene associations tests across 48 tissues in GTEx are plotted in (A) and (B). 95% confidence intervals depicted by gold dotted lines (B).

218 mechanism to explore further (Dhar et al., 2012).

219 While not found to be significantly associated with schizophrenia in brain tissues, the association
220 of *PRMT7* in adrenal gland and other vascular system organs highlights the sharing of eQTLs across
221 tissues. Recently, genes in both colon and adrenal gland were identified to be significantly associated with
222 schizophrenia (Gamazon et al., 2019). Gamazon et al. highlight the opportunity to better understand the
223 genetic mechanisms of neuropsychiatric diseases outside of the context of the central nervous system.
224 A larger sample size would be needed to elucidate expression correlations and potential co-expression
225 networks underlying African American neuropsychiatric traits.

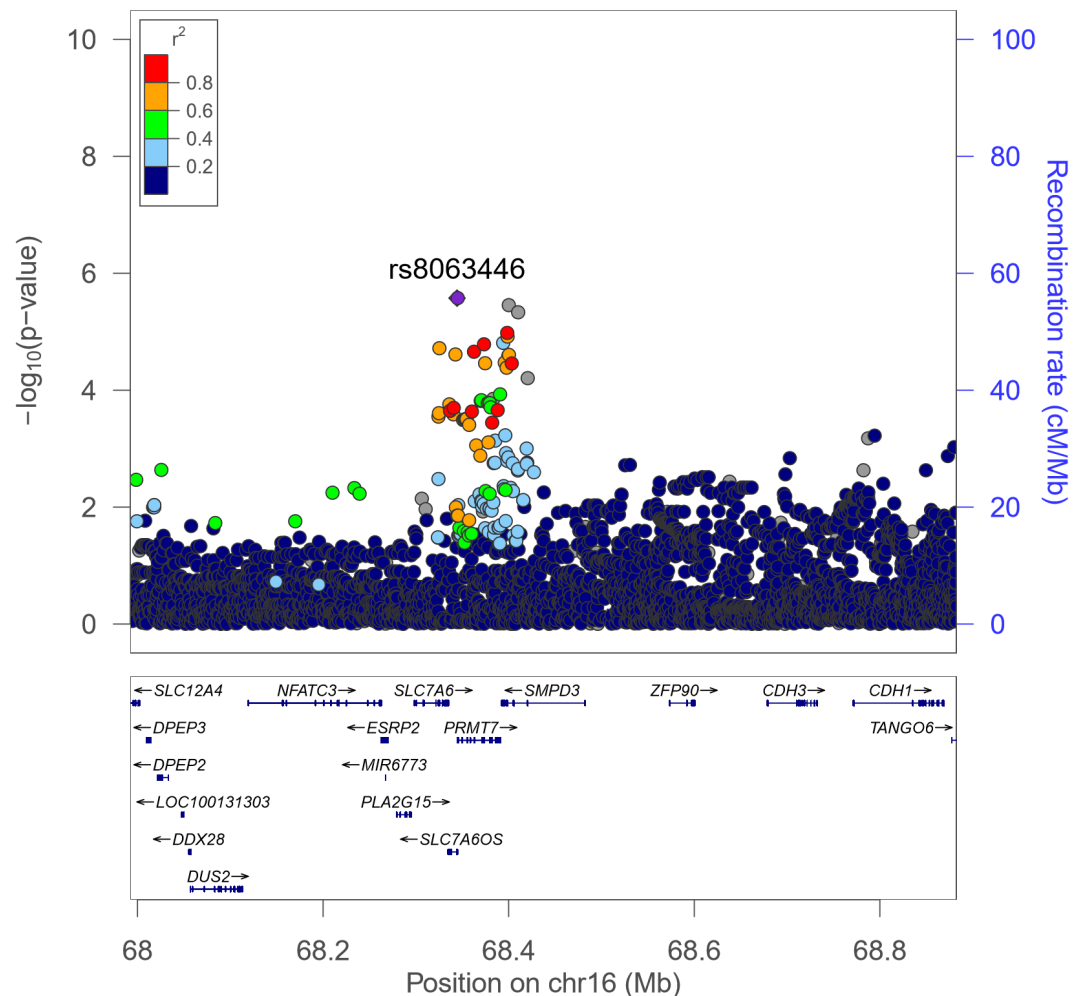


Figure 5. LocusZoom plot of the *PRMT7* locus in the GAIN GWAS for schizophrenia. rs8063446 is found in *SLC7A6OS* and 514 bp upstream of *PRMT7*. In our PrediXcan analyses, we found that increased predicted expression of *PRMT7* is associated with schizophrenia. rs8063446 is located in a linkage disequilibrium (LD) block with other SNPs associated with schizophrenia when plotted using 1000G AFR LD Population.

226 Growing need for diversity in GWAS and genetic prediction models

227 In our GWAS of bipolar disorder, rs10168049 was the most significantly associated SNP. This SNP has
 228 not been implicated in previous studies, and its low minor allele frequency in European populations
 229 suggests that it could have a larger functional impact in African populations (Fig. 6). In addition to not
 230 being identified in any published GWAS, rs10168049 is not present in any of the 55 prediction models we
 231 used to impute gene expression levels (MacArthur et al., 2017). As a result, this SNP did not contribute to
 232 predicted expression levels in our gene-based association tests. The association of this SNP with bipolar
 233 disorder needs to be replicated in larger studies; however, the lack of associations in other GWAS in the
 234 GWAS Catalog (MacArthur et al., 2017) may also be a result of ascertainment bias due to the field's focus
 235 on European populations (Lachance and Tishkoff, 2014).

236 Version 7 of the GTEx predictors we used contain data exclusively from individuals of European
 237 descent. These models are not optimal for predicting expression in African American cohorts (Mogil
 238 et al., 2018; Mikhaylova and Thornton, 2019). While they offer power driven by sample size, they do not
 239 include models with African ancestry-specific alleles that might affect susceptibility to neuropsychiatric
 240 traits in this population.

241 The use of prediction models from monocytes in MESA offers advantages with respect to similar

242 ancestry, but at the loss of nearly half of the sample size of many GTEx tissues. To ideally predict
243 expression in African American cohorts, prediction models built in more tissues from African ancestry
244 reference transcriptomes are needed. Moreover, future ancestry-specific models will not only increase
245 accuracy of expression prediction, but they will also create opportunities for different methods, such as
246 local ancestry mapping, to be applied to expression prediction by accounting for recent admixture within
247 African American cohorts (Zhong et al., 2019).

248 CONCLUSION

249 Information from this study provides promising insight into the genetic architecture of gene expression
250 underlying two neuropsychiatric disorders in African Americans. The results of our study were curbed by
251 the small sample size of GAIN. With just over 2,100 individuals in our study of schizophrenia and 1,000
252 individuals in our study bipolar disorder, our findings were limited in power compared to larger European
253 studies nearly two orders of magnitude greater in size (Ripke et al., 2014; Stahl et al., 2019).

254 The size and diversity of our prediction models further hindered our ability to identify novel genes
255 associated with these disorders. The MESA models, the most diverse of our predictors, were still limited
256 in tissue type (monocytes only) and size at 233 African American individuals, 352 Hispanic individuals,
257 and 578 European individuals (Mogil et al., 2018). The GTEx and CommonMind predictors we used,
258 while generated from a larger sample size than the MESA African American cohort for many tissues,
259 were made from exclusively European individuals (Aguet et al., 2017; Huckins et al., 2019). These
260 key limitations highlight the need to increase the number of individuals from diverse populations in the
261 study of neuropsychiatric genomics. To best characterize the molecular mechanisms that govern complex
262 traits in diverse populations, diverse models, reference panels, and study subjects need to be included in
263 genomics research.

chr2:122078406 G/A

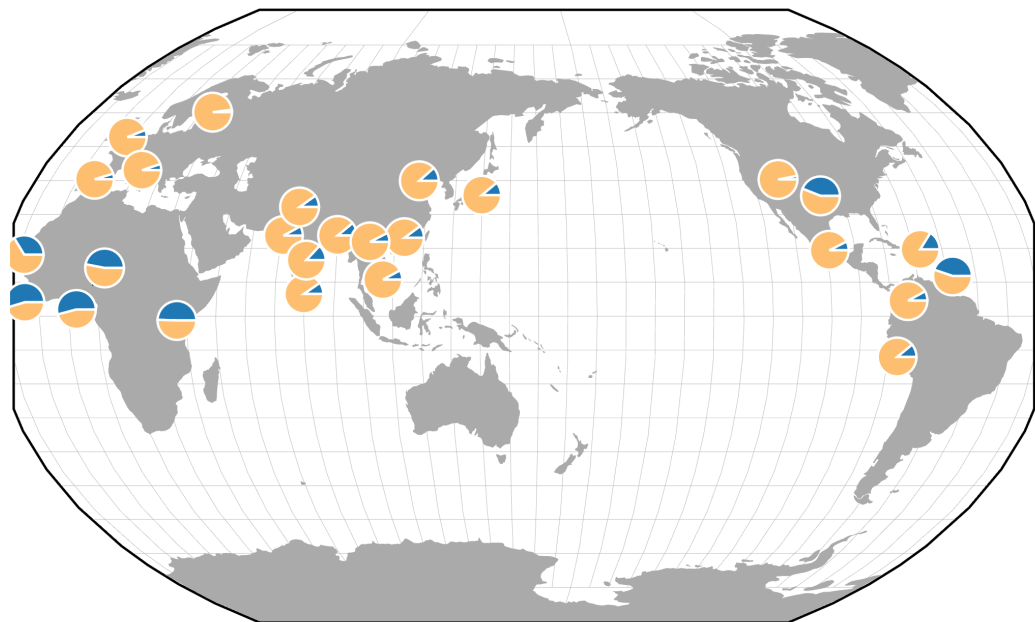


Figure 6. rs10168049 frequency across 1000G populations. Representative minor allele frequencies (MAFs) of rs10168049, which associated with bipolar disorder in GAIN ($P=1.04 \times 10^{-6}$) in different populations from 1000G. The global MAF of this SNP in 1000G is 0.187; however, the MAF reaches up to 0.541 in the YRI (Yoruba people in Ibadan, Nigeria) population of 1000G. This figure was generated using the Geography of Genetic Variants Browser (Marcus and Novembre, 2017).

264 **DECLARATIONS**

265 **Conflicts of Interests**

266 The authors declare no conflicts of interest.

267 **Author Contributions**

268 Peter N. Fiorica conceived and designed the experiments, performed the experiments, analyzed results,
269 contributed analysis tools, drafted the manuscript, prepared figures and/or tables, reviewed drafts of the
270 manuscript.

271 Heather E. Wheeler conceived and designed the experiments, analyzed results, contributed analysis
272 tools, drafted the manuscript, reviewed drafts of the manuscript.

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277 Fiorica). The funders had no role in study design, data collection and analysis, decision to publish, or
278 preparation of the manuscript.

279 **Data Availability**

280 All datasets analyzed were downloaded from the NCBI dbGaP at <https://www.ncbi.nlm.nih.gov/gap/> using the accession numbers phs000021.v3.p2, phs000167.v1.p1, and phs000017.v3.p1. Gene
281 prediction models were downloaded from PredictDB at <http://predictdb.org/>. All scripts and
282 summary statistics are available at
283 <https://github.com/WheelerLab/Neuropsychiatric-Phenotypes>.

285 **ACKNOWLEDGEMENTS**

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287 and Molecular Genetics of Schizophrenia - nonGAIN Sample (MGS_nonGAIN), was provided by
288 Genomics Research Branch at NIMH (see below) and the genotyping and analysis of samples was provided
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336 SUPPLEMENTAL TABLES & FIGURES

337 **Supplemental Figure 1: Principal Component Analysis of Schizophrenia Genotype Data** We per-
338 formed principal component analysis on the GAIN cohort merged with three populations from version
339 three of the HapMap Project (Belmont et al., 2003). Each point on the plot represents one individual in the
340 study plotted across axes for their first and second principal components. The three HapMap populations
341 plotted are Chinese in Beijing and Japanese in Tokyo (ASN), European ancestry in Utah (CEU), and
342 Yoruba people in Ibadan, Nigeria (YRI).

343 **Supplemental Figure 2: Principal Component Analysis of Bipolar Disorder Genotype Data** We
344 performed principal component analysis on the GAIN cohort merged with three populations from version
345 three of the HapMap Project (Belmont et al., 2003). Each point on the plot represents one individual in the
346 study plotted across axes for their first and second principal components. The three HapMap populations
347 plotted are Chinese in Beijing and Japanese in Tokyo (ASN), European ancestry in Utah (CEU), and
348 Yoruba people in Ibadan, Nigeria (YRI).

349 **Supplemental Figure 3: Comparison of MAFs across Imputation Reference Panels** We imputed
350 genotypes using the University of Michigan Imputation Server using either 1000G or CAAPA as the
351 reference panel. (A-D) depict the MAFs of SNPs from the GAIN schizophrenia study. We saw a similar
352 pattern of MAFs in the GAIN data of the bipolar disorder study. (A) depicts the MAF of SNPs at the
353 intersection of each reference panel before filtering by $r^2 > 0.8$ and $MAF > 0.01$. (B) Depicts MAFs
354 of SNPs in 1000G and CAAPA from (A) that passed the filters of $r^2 > 0.8$ and $MAF > 0.01$ and were
355 included in the GTEx prediction models across 44 tissues. (C) shows a plot of the MAFs of filtered SNPs
356 from 1000G and CAAPA found in the MESA predictors. (D) shows a plot of the MAFs of filtered SNPs
357 from 1000G and CAAPA that were included in our GWAS.

358 **Supplemental Table 1: S-PrediXcan Results of PGC SCZ Data for PRMT7** The table includes
359 the results for PRMT7 in our S-PrediXcan application to the PGC GWAS summary statistics across 42
360 tissues in which gene expression was predicted.

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