1 Genomic dissection of bipolar disorder and schizophrenia including 28 subphenotypes

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101 Summary

102 Schizophrenia and bipolar disorder are two distinct diagnoses that share symptomology, 103 understanding the genetic factors contributing to the shared and disorder-specific symptoms will 104 be crucial for improving diagnosis and treatment. In genetic data consisting of 53,555 cases 105 (20,129 BD, 33,426 SCZ) and 54,065 controls, we identified 114 genome-wide significant loci 106 implicating synaptic and neuronal pathways shared between disorders. Comparing SCZ to BD 107 (23,585 SCZ, 15,270 BD) identified four genomic regions including one with disorder-108 independent causal variants and potassium ion response genes as contributing to differences in 109 biology between the disorders. Polygenic risk scores (PRS) identified several significant 110 correlations within case-only phenotypes including SCZ PRS with psychotic features and age of 111 onset in BD. For the first time, we discover specific loci that distinguish between BD and SCZ 112 and identify polygenic components underlying multiple symptom dimensions. These results 113 point to the utility of genetics to inform symptomology and potentially treatment.

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116 Introduction

Bipolar disorder (BD) and schizophrenia (SCZ) are severe psychiatric disorders and among the leading causes of disability worldwide(Whiteford et al., 2013). Both disorders have significant genetic components with heritability estimates ranging from 60-80%(Nöthen et al., 2010). Recent genetic and epidemiological studies have demonstrated substantial overlap between these two disorders with a genetic correlation from common variation near 0.6-0.7(Cross-Disorder Group of the Psychiatric Genomics Consortium, 2013) and high relative risks (RR) among relatives of both BD and SCZ patients (RRs for parent/offspring: BD/BD: 6.4, BD/SCZ: 2.4;

124 SCZ/BD: 5.2, SCZ/SCZ: 9.9)(Lichtenstein et al., 2009). Despite shared genetics and 125 symptomology, the current diagnostic systems("Diagnostic and Statistical Manual of Mental 126 Disorders | DSM Library," n.d.) ("WHO | International Classification of Diseases," n.d.) adhere to historical distinctions from the late 19th century and represent BD and SCZ as independent 127 128 categorical entities differentiated on the basis of their clinical presentation, with BD 129 characterized by predominant mood symptoms, mood-congruent delusions and an episodic 130 disease course and SCZ considered a prototypical psychotic disorder. Identifying genetic 131 components contributing to both disorders provides insight into the biology underlying the 132 shared symptoms of the disorders.

133 While the shared genetic component is substantial, studies to date have also implicated genetic 134 architecture differences between these two disorders(Curtis et al., 2011; Ruderfer et al., 2014). A 135 polygenic risk score created from a case only SCZ vs BD genome-wide association study 136 (GWAS) significantly correlated with SCZ or BD diagnosis in an independent sample(Ruderfer 137 et al., 2014), providing the first evidence that differences between the disorders also have a 138 genetic basis. An enrichment of rare, moderate to highly penetrant copy number variants (CNVs) 139 and *de novo* CNVs are seen in SCZ patients(CNV and Schizophrenia Working Groups of the 140 Psychiatric Genomics Consortium, 2017; Gulsuner and McClellan, 2015; Kirov et al., 2012; 141 Stone et al., 2008; Szatkiewicz et al., 2014), while, the involvement of CNVs in BD is less 142 clear(Green et al., 2016). Although the role of *de novo* single nucleotide variants in BD and SCZ 143 have been investigated in only a handful of studies, enrichment in pathways associated with the 144 postsynaptic density has been reported for SCZ, but not BD(Fromer et al., 2014; Kataoka et al., 145 2016). Identifying disorder-specific variants and quantifying the contribution of genetic variation 146 to specific symptom dimensions remain important open questions. These genetic differences will

facilitate an understanding of the dimensions of the disorders instead of the dichotomous diagnosis. For example, we have shown that SCZ patients with greater manic symptoms have higher polygenic risk for BD(Ruderfer et al., 2014). These findings demonstrate shared genetic underpinnings for symptoms across disorders and may enable us to characterize patients by genetic liability to symptom dimensions thereby informing disease course and treatment.

Here, we utilize large collections of genotyped samples for BD and SCZ along with measures identifying 28 subphenotypes to address three questions: 1) Are there specific variants, genes or pathways that are either shared by, or differentiate BD and SCZ? 2) Are the shared symptoms between these disorders driven by the same underlying genetic profiles? and 3) Can we demonstrate independent genetic signatures for subphenotypes within these disorders?

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- 158 Results
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160 Shared genetic contribution to BD and SCZ

161 We performed association analysis of BD and SCZ combined into a single phenotype, totaling 162 53,555 cases (20,129 BD, 33,426 SCZ) and 54,065 controls on 15.5 million SNP allele dosages 163 imputed from 1000 genomes phase 3(The 1000 Genomes Project Consortium, 2015). Logistic 164 regression was performed controlling for 13 principal components of ancestry, study sites and 165 genotyping platform. We identified 11,231 SNPs with p-value below our genome-wide significance (GWS) threshold of 5x10⁻⁸. After grouping SNPs in linkage disequilibrium with 166 167 each other $(r^2 > 0.2)$, 114 genomic risk loci remained. For the most significant variant in each of 168 the 114 GWS loci, we performed conditional analysis with any GWS hit within 1Mb of the 169 extent of the locus from the previously performed single disease GWAS of SCZ(Schizophrenia

170 Working Group of the Psychiatric Genomics Consortium, 2014) and BD(Stahl et al., 2017) and 171 identified 32 loci that were independently significant defined strictly as no single disease locus 172 within 1Mb or a GWS p-value after conditional analysis (Supplementary Table 1). We further 173 performed gene-set based tests using MAGMA(Leeuw et al., 2015) across 10,891 curated 174 pathways (Watanabe et al., 2017) and identified 8 pathways surpassing Bonferroni correction ($p < 10^{-10}$ 175 4.6x10⁻⁶) with all but one pathway implicating synaptic and neuronal biology (Supplementary 176 Table 2a). Establishing independent controls (see Methods) allowed us to perform disorder-177 specific GWAS in 20,129 BD cases vs 21,524 BD controls and 33,426 SCZ cases and 32,541 178 SCZ controls. Using these results, we compared effect sizes of these 114 loci across each 179 disorder independently (Figure 1a) showing the subsets of variants that had larger effects in SCZ 180 vs BD and vice versa.

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182 Differentiating genetic contribution to BD and SCZ

183 To identify loci with divergent effects on BD and SCZ, we performed an association analysis 184 comparing 23,585 SCZ cases with 15,270 BD cases matched for shared ancestry and genotyping 185 platform (see Methods, Figure 1b. Supplementary Figures 1-2, Table 1). Two genome-wide 186 significant loci were identified, the most significant of which was rs56355601 located on 187 chromosome 1 at position 173,811,455 within an intron of DARS2. The second most significant 188 locus was rs200005157, a four base-pair insertion/deletion, on chromosome 20 at position 189 47638976 in an intron of ARFGEF2. For both variants, the minor allele frequency was higher in 190 BD cases than SCZ cases and disease-specific GWAS showed opposite directions of effect when 191 compared to controls. We sought to identify additional disease-specific loci by comprehensively 192 incorporating expression information with association results to perform fine-mapping and

193 identify novel variants(Gamazon et al., 2015; Giambartolomei et al., 2014; Gusev et al., 2016; 194 He et al., 2013). Here, we applied the summary-data-based Mendelian randomization (SMR) 195 method(Zhu et al., 2016) (see Methods) utilizing the cis-QTLs derived from peripheral 196 blood(Westra et al., 2013), human dorsolateral prefrontal cortex (DLPFC)(Fromer et al., 2016) 197 Mind Consortium and 11 brain regions from the Common from the GTEx 198 consortium(Consortium, 2015). We identified one SNP-probe combination that surpassed the 199 threshold for genome-wide significance in blood but was also the most significant finding in 200 brain. We found that SNP rs4793172 in gene DCAKD is associated with SCZ vs BD analysis $(p_{GWAS} = 2.8 \times 10^{-6})$ and is an eQTL for probe ILMN 1811648 $(p_{eQTL} = 2.9 \times 10^{-168})$, resulting in 201 $p_{SMR} = 4.1 \times 10^{-6}$ in blood ($p_{eQTL} = 2.9 \times 10^{-25}$, $p_{SMR} = 2.0 \times 10^{-5}$ in DLFC, and $p_{eQTL} = 4.6 \times 10^{-15}$, 202 $p_{SMR} = 6.0 \times 10^{-5}$ in GTEx cerebellar hemisphere) (Supplementary Table 3, Supplementary Figure 203 204 3) and shows no evidence of heterogeneity ($p_{HET} = 0.66$) which implies only a single causal 205 variant in the locus.

206 In an effort to prioritize genes for the two GWS loci from the GWAS, we performed fine-207 mapping(Benner et al., 2016) using an LD map derived from a majority of the control samples. 208 We then performed SMR on each of the variants with causal probability greater than 1% using 209 all eQTLs from the CommonMind Consortium DLPFC reference. All the most likely causal 210 variants were shown to most significantly regulate the same gene suggesting CSE1L is the most likely relevant gene on chromosome 20 (rs200005157: causal probability=0.21, p_{GWAS}=2.4x10⁻⁸, 211 peotl 3x10⁻⁸, p_{SMR}=8.5x10⁻⁵, p_{HET}=0.34, Supplementary Table 4). For the locus on chromosome 212 213 1, SLC9C2 is the most significantly regulated gene. However, a highly significant heterogeneity 214 test indicates a complex genetic architecture making it difficult to infer a causal role for the 215 associated SNP (Supplementary Table 5). Therefore, DARS2 presents as the most likely relevant 216 gene on chromosome 1 (rs56355601: $p_{GWAS}=5.6x10^{-9}$, causal probability=0.079, p_{eQTL} 7.4x10⁻¹³, 217 $p_{SMR}=6.17x10^{-6}$, $p_{HET}=0.03$). We note however, that in both cases there are less associated 218 variants that are stronger eQTLs for these genes complicating a straightforward causal 219 interpretation. Finally, using the same gene-set test used for the combined analysis GO biological 220 process "response to potassium ion" (p=1.6x10⁻⁶) was the only pathway surpassing our 221 Bonferroni corrected significance threshold (Supplementary Table 2b).

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223 Regional joint association

224 We expanded our efforts to identify disorder-specific genomic regions by jointly analyzing 225 independent GWAS results from BD and SCZ(Pickrell et al., 2016). The genome was split into 226 1,703 previously defined approximately LD independent regions(Berisa and Pickrell, 2015). 227 Thirteen percent, or 223 regions, had a posterior probability greater than 0.5 of having a causal 228 variant for at least one disorder. Of these, 132 best fit the model of a shared causal variant 229 influencing both BD and SCZ, 88 were most likely specific to SCZ, 3 demonstrated evidence of 230 two independent variants (with one impacting each of the two disorders) and none were BD-231 specific. Of note, this approach calculates a prior probability that any given region is disease-232 specific and from these data the probability of having a BD specific region was 0.1% compared 233 to 15% for SCZ, likely a result of increased power from the larger SCZ sample size and/or a 234 difference in genetic architecture between these disorders.

The 114 GWS SNPs from the combined BD and SCZ GWAS localized into 99 independent regions (13 regions had multiple GWS SNPs), of which 78 (79%) were shared with a posterior probability of greater than 0.5. Sixty regions had at least one GWS SNP in the independent SCZ GWAS, of which 30 (50%) are shared and 8 regions contained a GWS SNP in the independent

BD GWAS, of which 6 (75%) are shared using the same definition. For the three regions 239 240 showing evidence for independent variants, two had highly non-overlapping association signals 241 in the same region stemming from independent variants. The third, on chromosome 19 presented 242 a different scenario where association signals were overlapping (Figure 2). The most significant variant in BD was rs111444407 (chr19:19358207, $p = 8.67 \times 10^{-10}$) and for SCZ was rs2315283 243 (chr19:19480575, p= 4.41×10^{-7}). After conditioning on the most significant variant in the other 244 245 disorder, the association signals of the most significant variant in BD and SCZ were largely unchanged (BD rs111444407 = 1.3×10^{-9} , SCZ rs2315283 p= 6.7×10^{-5}). We further calculated the 246 247 probability of each variant in the region being causal for both BD and SCZ(Benner et al., 2016) 248 and found no correlation (r = -0.00016). The most significant variants had the highest posterior 249 probability of being causal (SCZ: rs2315283, prob = 0.02, BD: rs111444407, prob = 0.16). Both 250 variants most significantly regulate the expression of GATAD2A in brain(Fromer et al., 2016) but in opposite directions (rs111444407 $p_{eOTL} = 6x10^{-15}$, beta = 0.105; rs2315283 $p_{eOTL} = 1.5x10^{-28}$, 251 252 beta = -0.11).

Additional work calculating heritability estimates among the same set of 1,703 regions found comparable h^2 estimates across the genome except for the major histocompatibility complex on chromosome 6 and a single region on chromosome 10 (see Supplement).

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257 **Regional SNP-heritability estimation**

Across the genome, regional SNP-heritabilities (h_{snp}^2) were estimated separately for SCZ and BD(Shi et al., 2016) and were found to be moderately correlated (r=0.25). We next defined risk regions as those containing the most associated SNP for each GWS locus. In total, there were 101 SCZ risk regions from the 105 autosomal GWS loci reported previously(Schizophrenia

262 Working Group of the Psychiatric Genomics Consortium, 2014) and 29 BD risk regions from 30 263 GWS loci reported in a companion paper(Stahl et al., 2017). Ten regions were risk regions for 264 both BD and SCZ comprising 33% of BD risk regions and 10% of SCZ risk regions. We further 265 stratified regional h_{snp}^2 by whether a region was a risk region in one disorder, none or both 266 (Supplementary Figure 4). Since the discovery data for the regions overlapped with the data used 267 for the heritability estimation, we expected within-disorder analyses to show significant results. 268 In risk regions specific to SCZ (n=91) there was a significant increase in regional h_{snp}^2 in SCZ, as 269 expected ($p = 1.1 \times 10^{-22}$), but also in BD ($p = 1.2 \times 10^{-6}$). In risk regions specific to BD (n=19), 270 significantly increased regional h_{snp}^2 was observed in BD, as expected (p = 0.0007), but not in SCZ (p = 0.89). Risk regions shared by both disorders had significantly higher h_{snp}^2 in both 271 disorders, as expected (BD p = 5.3×10^{-5} , SCZ p = 0.006), compared to non-risk regions. 272 273 However, we observed a significant increase in BD h_{snp}^2 in shared risk regions compared to BD risk regions (BD p = 0.003) but not SCZ h_{snp}^2 for shared risk regions compared to SCZ risk 274 275 regions (p = 0.62). Using a less stringent p-value threshold for defining risk regions (p $< 5 \times 10^{-6}$), 276 thereby substantially increasing the number of regions, resulted in similar results. Seven regions 277 contributed to substantially higher h²_{snp} in SCZ compared to BD but no region showed the 278 inverse pattern. Of these regions, all but one was in the major histocompatibility region (MHC), 279 the sole novel region was chr10:104380410-106695047 with regional $h_{snp}^2 = 0.0019$ in SCZ and 280 $h_{snp}^2=0.00063$ in BD.

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282 **Polygenic dissection of subphenotypes**

Subphenotypes were collected for a subset of patients with both BD and SCZ (see Methods). For
SCZ, we had clinical quantitative measurements of manic, depressive, positive and negative

285 symptoms generated from factor analysis of multiple instruments as described 286 previously(Ruderfer et al., 2014) but in larger sample sizes (n=6908, 6907, 8259, 8355 287 respectively). For BD, 24 subphenotypes were collected among nearly 13,000 cases in distinct 288 categories including comorbidities, clinical information such as rapid cycling and psychotic 289 features as well as additional disease course data such as age of onset and number of 290 hospitalizations. For each BD and SCZ patient, we calculated a polygenic risk score (PRS) using 291 all SNPs, from each of the four main GWAS analyses (BD+SCZ, BD, SCZ and SCZvsBD). We 292 then used regression analysis including principal components and site to assess the relationship 293 between each subphenotype and the 4 PRS. Specifically, we tested whether polygenic risk scores 294 of BD+SCZ, BD, SCZ or SCZvsBD were correlated with each of these subphenotypes separately 295 within BD and SCZ cases. When testing if the variance explained by the PRS was different from 296 zero, we applied a significance cutoff of p < 0.0004 based on Bonferroni correction for 112 tests. 297 In total, we identified 6 significant results after correction (Figure 3, Table 2).

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299 A significant positive correlation existed between BD PRS and manic symptoms in SCZ cases as seen previously(Ruderfer et al., 2014) ($p=2x10^{-5}$, t=4.26) and BD PRS and psychotic features in 300 BD patients (p=5.3x10⁻⁵, t=4.04). A significant increase in SCZ PRS was seen for BD cases with 301 302 versus without psychotic features ($p=1.2x10^{-10}$, t=6.45) and patients with increased negative symptoms in SCZ patients (p=3.60x10⁻⁶, t=4.64). The BD+SCZ vs controls PRS was 303 304 significantly associated with psychotic features in BD (p=7.9x10⁻¹³, t=7.17) and negative 305 symptoms in SCZ ($p=1.5x10^{-5}$, t=4.33). The next two most significant results which did not 306 survive our conservative correction were both indicative of a more severe course in BD: 307 increased BD+SCZ PRS with increased numbers of hospitalizations in BD cases ($p=4.2 \times 10^{-4}$,

308 t=3.53) and increased SCZ PRS with earlier onset of BD ($p=7.9x10^{-4}$, t=-3.36). We assessed the 309 role of BD subtype on the correlation between SCZ PRS and psychotic features and identified a 310 significant correlation when restricted to only BD type I cases indicating the result was not likely 311 driven by BD patients with a schizoaffective subtype (BDI: 3,763 with psychosis, 2,629 without, 312 $p=1.55x10^{-5}$, Supplementary Table 6).

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We performed a GWAS for all 8 quantitative subphenotypes and 9 binary subphenotypes with at least 1,000 cases and calculated heritability and genetic correlation with BD and SCZ. Only two subphenotypes had significant h_{snp}^2 estimates using LD-score regression(Bulik-Sullivan et al., 2015) both in BD: psychotic features in BD ($h_{snp}^2=0.15$, SE=0.06) and suicide attempt ($h_{snp}^2=0.25$, SE=0.1). Only psychotic features demonstrated a significant genetic correlation with SCZ ($r_g=0.34$, SE=0.13, p=0.009).

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The significant genetic correlation demonstrates a genome-wide relationship between common variants contributing to SCZ risk and those contributing to psychotic features in BD cases. We tested whether the most significantly associated SCZ loci contributed directly to psychotic features in BD. One hundred of the 105 autosomal genome-wide significant SCZ SNPs previously published(Schizophrenia Working Group of the Psychiatric Genomics Consortium, 2014) were in our dataset after QC and 60 were in the same direction of effect for risk of psychotic features in BD (p=0.028, one-sided binomial-test).

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330 Discussion

331 Here we present a genetic dissection of bipolar disorder and schizophrenia from over 100,000 332 genotyped subjects. Consistent with earlier results(Cross-Disorder Group of the Psychiatric 333 Genomics Consortium, 2013), we found extensive genetic sharing between these two disorders, 334 identifying 114 genome-wide significant loci contributing to both disorders of which 32 are 335 novel. These findings point to the relevance of neuronal and synaptic biology for the shared 336 genetic substrate of these disorders. However, despite this degree of sharing, we identified 337 several loci that significantly differentiated between the two disorders, having opposite directions 338 of effect. We also found polygenic components that significantly correlated from one disorder to 339 symptoms of the other.

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341 Two GWS loci were identified from the case only SCZ versus BD analysis providing 342 opportunities to inform the underlying biological distinctions between BD and SCZ. The most 343 significant locus implicates DARS2 (coding for the mitochondrial Aspartate-tRNA ligase) which 344 is highly expressed in the brain and significantly regulated by the most significant SNP 345 rs56355601 ($p_{eQTL}=2.5 \times 10^{-11}$). Homozygous mutations in *DARS2* are responsible for 346 leukoencephalopathy with brainstem and spinal cord involvement and lactate elevation (LBSL), 347 which was characterized by neurological symptoms such as psychomotor developmental delay, 348 cerebellar ataxia and delayed mental development(Yamashita et al., 2013, p. 2). Based on 349 methylation analysis from the prefrontal cortex of stress models (rats and monkeys) and from 350 peripheral samples (in monkeys and human newborns), DARS2, among others, has been 351 suggested as a potential molecular marker of early-life stress and vulnerability to psychiatric 352 disorders(Luoni et al., 2016). The second most significant locus implicates CSE1L, a nuclear 353 transport factor that plays a role in cellular proliferation as well as in apoptosis(Bera et al., 2001).

354 Intronic SNPs in *CSE1L* have been associated with subjective well-being(Okbay et al., 2016) 355 and, nominally to antidepressant response(Li et al., 2016). More interestingly, CSE1L is a 356 potential target gene of miR-137, one of the well-known schizophrenia risk loci(Schizophrenia 357 Working Group of the Psychiatric Genomics Consortium, 2014), which is able to negatively 358 regulate CSE1L by interacting with complementary sequences in the 3' UTR of CSE1L(Li et al., 359 2013). Although falling short of genome-wide significance, the third most significant locus 360 implicates ARNTL (Aryl Hydrocarbon Receptor Nuclear Translocator Like), which is a core 361 component of the circadian clock. ARNTL has been previously hypothesized for relevance in 362 bipolar disorder, (Yang et al., 2008) although human genetic evidence is limited (Byrne et al., 363 2014).

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365 The ability to generate transcriptional data on multiple tissues across many individuals using 366 RNA-sequencing has provided detailed information on the role common variants play in regulating expression of specific genes in specific tissues. These eQTLs can be integrated with 367 368 the genetic association data from GWAS to inform on the relationship between variant 369 association and variant regulation of expression for each gene. Performing this integration, we 370 identified a third genome-wide significant finding in DCAKD. The gene codes for Dephospho-371 CoA Kinase Domain Containing protein, a member of the human postsynaptic density proteome 372 from human neocortex(Bayés et al., 2011). In the mouse cortical synaptoproteome DCAKD is 373 among the proteins with the highest changes between juvenile postnatal days and adult stage, 374 suggesting a putative role in brain development(Gonzalez-Lozano et al., 2016; Moczulska et al., 375 2014). Discerning between pleiotropy (variant independently regulates expression and alters risk 376 to disease) from causality (variant regulates expression which thereby alters risk to disease) through statistical analysis alone is difficult, this analytical approach is stringent in excluding loci where colocalised SNP-phenotype and SNP-expression associations may reflect confounding driven by linkage disequilibrium (LD) (one variant regulates expression and a different variant alters risk but the variants in the region are in LD). Hence, this approach utilizes currently available data to prioritize genes, including direction of effect, for functional follow-up. These analyses will become more powered with increased sample sizes for both phenotype and eQTL data sets.

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385 Performing pathway analysis based on the full association results shows enrichment of genes 386 involved in response to potassium ions, including potassium voltage-gated channel subfamily 387 members and a number of genes regulated by cellular potassium concentration. This is in line 388 with previous genetic evidence pointing to a key etiologic role of potassium channels, in 389 particular, in BD(Judy and Zandi, 2013), which could be explained by their role in multiple 390 neurobiological mechanisms involved in the development of psychiatric disorders such as 391 regulation of the dopaminergic circuits, synaptic plasticity, and myelination(Balaraman et al., 392 2015).

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We further assessed the contribution of regions of the genome to each disorder through joint regional association and heritability estimation. These results point to an additional locus that may contribute differentially to liability to BD and SCZ. The region on chr19 shows overlapping association peaks that are driven by independent causal variants for each disorder. Both variants significantly regulate the same gene *GATAD2A* but in opposite directions. *GATAD2A* is a transcriptional repressor, which is targeted by *MBD2* and is involved in methylation-dependent

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400 gene silencing. The protein is part of the large NuRD (nucleosome remodeling and deacetylase) 401 complex, for which also HDAC1/2 are essential components. NurD complex proteins have been 402 associated with autism(Li et al., 2015). Their members, including GATAD2A, display preferential 403 expression in fetal brain development(Li et al., 2015) and in recent work has been implicated in 404 SCZ through open chromatin(Fullard et al., n.d.). Further, p66α (mouse GATAD2A) was recently 405 shown to participate in memory preservation through long-lasting histone modification in 406 hippocampal memory-activated neurons(Ding et al., 2017). SNP-heritability appears to be 407 consistently shared across regions and chromosomes between these two disorders. Regions with 408 GWS loci often explain higher proportions of heritability as expected. When looking at the effect 409 on heritability of the presence of a GWS locus in the other disorder, we identified a significant 410 increase in BD heritability for regions containing a GWS locus for SCZ but no significant 411 increase in SCZ heritability in regions having a BD one. This result suggests a directionality to 412 the genetic sharing of these disorders with a larger proportion of BD loci being specific to BD. 413 However, we cannot exclude that the asymmetry of results may reflect less power of discovery 414 for BD than SCZ. The degree to which power and subphenotypes contribute to this result 415 requires further examination.

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We note that as with nearly all GWAS findings, the calculated population-based effect sizes of the variants identified here are small and independently explain only a modest fraction to the heritability of these disorders. The identification of these variants is dependent on the ability to have highly accurate allele frequency estimates that can only be ascertained from large sample sizes. As sample sizes get larger the power to identify variants of smaller effect increases meaning that increasing sample size results in the identification of variants of smaller effect. 423 However, a small population effect size does not exclude the possibility of a substantially larger 424 effect on molecular phenotypes nor does it preclude the utility of association regions in 425 understanding biology or having a clinical impact. Efforts following up GWAS results to date 426 have demonstrated the value of these findings in pointing to genes that can aid in understanding 427 the underlying biology of the trait (Claussnitzer et al., 2015; Mohanan et al., 2018; Sekar et al., 428 2016). Further, there is a clear relationship between GWAS results of a phenotype and gene 429 targets of drugs that treat that phenotype pointing to the potential for improved therapeutic 430 understanding(Nelson et al., 2015; Ruderfer et al., 2016). A major challenge of GWAS is the 431 sheer number of findings and the substantial time/cost required for functional follow up of these 432 findings in the classical paradigms used for genes causal for monogenic disorders. In silico 433 bioinformatic analyses (such as SMR used here) that integrate GWAS results with 'omics data 434 (transcription, protein, epigenetic, etc.) have the potential to put a clearer biological focus on 435 GWAS results. Such analyses can become more complex as more reference omics data sets (with 436 genome-wide genotyping) become available. Additional analytical efforts will be required to 437 facilitate the transition from GWAS to biology but substantial data has shown there is much to be 438 learned from these variants despite their small effects(Visscher et al., 2017).

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We have now identified multiple genomic signatures that correlate between one disorder and a clinical symptom in the other disorder, illustrating genetic components underlying particular symptom dimensions within these disorders. Medical symptoms, including those seen in psychiatric disorders, can manifest through a multitude of causes. The classic example often used is headache for which many different paths lead to the same symptom. Psychiatric symptoms also have many potential causes. For example, symptoms of psychosis can be the result of highly

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446 heritable diseases such as BD and SCZ but also infectious and neurodegenerative diseases, 447 sleep/sensory deprivation or psychedelic drugs. Demonstrating a shared biological underpinning 448 to these symptoms suggests they could be treated through modulating the same pathway. As 449 previously shown, we find a significant positive correlation between the PRS of BD and manic 450 symptoms in SCZ. We also demonstrate that BD cases with psychotic features carry a 451 significantly higher SCZ PRS than BD cases without psychotic features and this result is not 452 driven by the schizoaffective BD subtype. Further, we show that increased PRS is associated 453 with more severe illness. This is true for BD with psychotic features having increased SCZ PRS, 454 earlier onset BD having higher SCZ PRS and cases with higher BD+SCZ PRS having a larger 455 number of hospitalizations. We demonstrated that psychotic features within BD is a heritable 456 trait and GWS loci for SCZ have a consistent direction of effect in psychotic features in BD, 457 demonstrating the potential to study psychosis more directly to identify variants contributing to 458 that symptom dimension.

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460 This work illustrates the utility of genetic data, in aggregate, at dissecting symptom 461 heterogeneity among related disorders and suggests that further work could aid in characterizing 462 patients for more personalized treatment. Genetic risk scores have demonstrated their ability to 463 inform and predict pathology(Cleynen et al., 2016) and more recently have been shown to be 464 able to identify patients with risk equivalent to monogenic variants (Khera et al., 2017). In 465 psychiatry, we lack objective biological measurements (biomarkers) with which to assess the 466 ability of a genetic signature to predict or inform. Lacking diagnostic pathology for psychiatric 467 disorders leaves a genuine opportunity for the genetics to drive diagnosis and treatment to a 468 much larger degree than in other domains. One potential model assumes that each individual has

469 a quantitative loading of a series of symptom dimensions (i.e. manic, psychotic, cognitive, etc.) 470 and that these symptoms can be assessed at the genetic level to characterize a patient's 471 dysfunction and used to inform disease course and optimal treatment. Making this a reality will 472 require more detailed information on disease course and outcomes. For example, if treatment 473 response data existed for these samples one could ask whether a genetic loading for psychosis 474 was correlated with response to treatment. Initial work has already shown the potential of this 475 approach using a SCZ PRS to inform lithium response in BD(Amare et al., 2018). Ultimately, 476 the goal will be to quantify multiple genetic loadings of each individual's illness and use those 477 measures to inform treatment based on the outcomes of previous individuals with similar 478 profiles.

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In conclusion, we present a detailed genetic dissection of BD and SCZ pointing to substantial shared genetic risk but also demonstrating that specific loci contribute to the phenotypic differences of these disorders. We show that genetic risk scores can correspond to symptoms within and across disorders. Finally, we present data that points to these disorders being neither independent nor the same but sharing particular symptom dimensions that can be captured from the genetics and used to characterize patients to ultimately inform diagnosis and treatment.

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487 Author Contributions:

DMR, PS and KSK managed and organized the group. DMR, SR, JB, EAS, JMWP, NM, AWC,
APSO, LMOL and VT contributed to analyses. Subphenotype collection and organization was
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- 592 The authors declare no competing interests.
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917	Figure Legends
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919	Figure 1. Associated Genomics Loci Shared and Divergent Between BD and SCZ
920	a) Odds ratios (OR) from independent data sets of BD (blue) and SCZ (red) for each of the 114
921	genome-wide significant variants in the BD and SCZ vs controls GWAS. b) Manhattan plot for
922	SCZ vs BD GWAS.
923	
924	Figure 2. Genomic Region with Disorder Independent Causal Variants in BD and SCZ
925	Regional association plot of chr19 18.4 - 20 Mb. Independent GWAS of SCZ (top) and BD
926	(bottom) are displayed along with genes in the region. Each point is a SNP with the most
927	significant SNP labeled and colored purple, other points are colored by LD with most significant
928	SNP. Despite overlapping regional association these loci are independent across disorders and
929	conditional analysis in both directions yields equivalent significance.
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931	Figure 3. Polygenic Risk Score Dissection of Clinical Symptom Dimensions
932	Effect size (calculated by dividing regression estimate by standard error) from regression
933	analysis including ancestry covariates for each subphenotype and PRS for BD (x-axis) and SCZ

934 (y-axis). Point size represents -log10(p-value) with SCZ (red) and BD (blue). Numbered 935 subphenotypes are 1) comorbid migraine, 2) panic attacks 3) suicide attempt 4) mixed states 5) 936 rapid cycling 6) comorbid eating disorder 7) comorbid OCD 8) year of birth 9) suicide ideation 937 10) panic disorder 11) number of suicide attempts 12) depressive symptoms (SCZ) 13) episodes 938 depressive 14) episodes total 15) positive symptoms (SCZ) 16) irritable mania 17) age of onset 939 depression 18) family history 19) episodes mixed mania 20) unipolar mania 21) alcohol 940 substance dependence 22) age of onset mania 23) age at interview 24) number of 941 hospitalizations. All subphenotypes are in BD except those labeled (SCZ).

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943 **Table Legends**

944

945 Table 1. Most Significant Associated Loci from SCZ vs BD GWAS

Association results for the five most significant variants in the SCZ vs BD GWAS with the top two being genome-wide significant. Each variant includes results from the independent BD vs controls and SCZ vs controls GWAS and the comparable p-value from a heterogeneity test when performing a two cohort meta-analysis of SCZ and BD.

950

951 Table 2. Complete Results of Polygenic Risk Score Dissection Analysis

952 Polygenic scoring results of all four GWAS phenotypes (BD+SCZ vs controls, BD vs controls, 953 SCZ vs controls and SCZ vs BD) and 24 subphenotypes from BD and 4 subphenotypes from 954 SCZ, rows without case/control counts are quantitative measures. Significance and effects are 955 from regression analysis of subphenotype on PRS including principal components of ancestry 956 and site as covariates. Effect is the regression estimate divided by the standard error. 957

958 Supplementary Figure Legends

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- Figure S1. Related to Figure 1b. Regional Association Plot and Forest Plot for the First
 Genome-wide Significant Hit in the SCZ vs BD GWAS.
- Figure S2. Related to Figure 1b. Regional Association Plot and Forest Plot for the Second
 Genome-wide Significant Hit in the SCZ vs BD GWAS.
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966 Figure S3. Related to Summary-data-based Mendelian Randomization. Detailed 967 Association of DCAKD from SMR.

968 Results at the *DCAKD* locus from SMR analysis of SCZ vs BD. Top plot, brown dots represent 969 the *P* values for SNPs from SCZ vs BD GWAS, diamonds represent the *P* values for probes from 970 the SMR test. Bottom plot, the eQTL *P* values of SNPs from the Westra study for the 971 ILMN_1811648 probe tagging *DCAKD*. The top and bottom plots include all the SNPs available 972 in the region in the GWAS and eQTL summary data, respectively, rather than only the SNPs 973 common to both data sets. Highlighted in red is the gene (*DCAKD*) that passed the SMR and 974 HEIDI tests.

975

Figure S4. Related to Regional SNP-heritability estimation. Heritability Estimates for BD and SCZ in Genome-wide Significant Regions of BD and SCZ.

978 Regional SNP-heritability estimates for SCZ and BD stratified by whether the region contains

979 the most significant variant in a genome-wide significant locus in BD, SCZ, neither or both.

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982 STAR Methods

983 CONTACT FOR REAGENT AND RESOURCE SHARING

- 984 Summary statistics from GWAS are publically available at
- 985 https://www.med.unc.edu/pgc/results-and-downloads/downloads. Genotype and phenotype data
- 986 use is restricted and governed by the Psychiatric Genetics Consortium. Further information and
- 987 requests for analytical results or additional information should be directed to and will be fulfilled
- 988 by the Lead Contact, Douglas Ruderfer (<u>douglas.ruderfer@vanderbilt.edu</u>).

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990 SUBJECT DETAILS

991 Genotyped Sample Description

992 SCZ samples are a substantial subset of those analyzed previously(Schizophrenia Working 993 Group of the Psychiatric Genomics Consortium, 2014). BD samples are the newest collection 994 from Psychiatric Genomics Consortium Bipolar Disorder Working Group(Stahl et al., 2017). To 995 ensure independence of the data sets, individuals were excluded until no individual showed a 996 relatedness (pihat) value greater than 0.2 to any other individual in the collection, while 997 preferentially keeping the case over the control for case-control related pairs. In total 1,795 BD 998 cases, 1,165 SCZ cases and 27,274 controls were removed (most of which were previously 999 known), leaving 20,129 BD cases 33,426 SCZ cases and 54,065 controls for the final meta-1000 analysis.

For analyses directly comparing BD and SCZ, we matched cases from both phenotypes on genotyping platform and ancestry, resulting in 15,270 BD cases versus 23,585 SCZ cases. Hence, we were able to match 76% of BD cases and 71% of SCZ cases for this case vs case analysis. All samples were collected with approval from respective IRBs and informed consent was obtained.

1006

1007 Sub-phenotype Description

1008 BD sub-phenotypes were collected by each study site using a combination of diagnostic 1009 instruments, case records and participant interviews. Ascertainment details for each study site are 1010 described in the supplementary data of the PGC Bipolar Working Group paper(Stahl et al., 1011 2017). The selection of phenotypes for collection by this group was determined by literature 1012 searches in order to determine phenotypes with prior evidence for heritability. It was further 1013 refined dependent on the availability of phenotype data across a range of study sites and the 1014 consistency by which the phenotypes were defined. Schizophrenia subphenotypes are the same 1015 as described previously(Ruderfer et al., 2014) but in a larger proportion of patients.

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1017 METHOD DETAILS

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1019 QUANTIFICATION AND STATISTICAL ANALYSIS

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1021 Quality Control, Imputation, Association Analysis and Polygenic Risk Score Testing

1022 Quality control and imputation were performed on each of the study cohort datasets (n=81), 1023 according to standards established by the Psychiatric Genomics Consortium (PGC). The quality 1024 control parameters for retaining SNPs and subjects were: SNP missingness < 0.05 (before 1025 sample removal); subject missingness (p < 0.02); autosomal heterozygosity deviation (| F_{het} | < 1026 (0.2); SNP missingness < 0.02 (after sample removal); difference in SNP missingness between cases and controls < 0.02; and SNP Hardy-Weinberg equilibrium (p > 10⁻⁶ in controls or p > 1027 10^{-10} in cases). Genotype imputation was performed using the pre-phasing/imputation stepwise 1028 1029 approach implemented in IMPUTE2(Howie et al., 2011) / SHAPEIT(Delaneau et al., 2013) 1030 (chunk size of 3 Mb and default parameters). The imputation reference set consisted of 2,186 1031 phased haplotypes from the full 1000 Genomes Project dataset (August 2012, 30,069,288 1032 variants, release "v3.macGT1"), all variants align to human genome build 19 (hg19). After 1033 imputation, we used the best guess genotypes (genotype probability > 0.8), for further robust 1034 relatedness testing and population structure analysis. Here we required very high imputation 1035 quality (INFO > 0.8) and low missingness (<1%) for further quality control. After linkage disequilibrium (LD) pruning ($r^2 < 0.02$) and frequency filtering (MAF > 0.05), there were 14,473 1036 1037 autosomal SNPs in the data set. Principal component estimation was done with the same 1038 collection of autosomal SNPs. We tested the first 20 principal components for phenotype association (using logistic regression with study indicator variables included as covariates) and 1039 1040 evaluated their impact on the genome-wide test statistics using λ . Thirteen principal components 1041 namely in all association 1,2,3,4,5,6,7,8,10,12,15,18,20 were included analyses 1042 $(\lambda=1.45)$. Analytical steps were repeated for SCZ vs BD analysis.

We performed four main association analyses (Figure 1), i.e. (i) GWAS of BD and SCZ as a single combined case phenotype, as well as disorder-specific GWAS using independent control sets in (ii) BD cases vs BD controls and (iii) SCZ cases vs SCZ controls, and (iv) association analysis of SCZ cases vs BD cases. For all GWS loci from the GWAS of BD and SCZ vs controls we identified any GWS loci within 1Mb from the extent of the locus in the previously published PGC SCZ vs controls(Schizophrenia Working Group of the Psychiatric Genomics

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1049 Consortium, 2014) and the most recent PGC GWAS of BD vs controls(Stahl et al., 2017) and 1050 performed conditional analysis. Specifically, we transformed the genotype probabilities of the 1051 disease variant into dosages and used it as an additional covariate for the association analysis for 1052 the BD+SCZ vs controls index SNP. This was done within each cohort and an OR based inverse 1053 SE weighted meta-analysis was performed for the final result. All datasets were included except 1054 for those with trios.

1055

1056 Summary-data-based Mendelian Randomization (SMR)(Zhu et al., 2016)

1057 SMR is a method that integrates summary level GWAS data with gene expression quantitative 1058 trait loci (eQTL) identified in independent data sets. This integration aims to identify variants 1059 that have pleotropic effects on expression of a given gene and the phenotype. While significant 1060 findings may indeed reflect a causal path from variant to phenotype through expression, it is 1061 impossible to discern statistically between pleiotropy and causality. However, the method can 1062 remove linkage as driving the result, and uses the data available to prioritise amongst genes in 1063 genomic regions that show association with disease. We used SMR as a statistical fine-mapping 1064 tool applied to the SCZ vs BD GWAS results to identify loci with strong evidence of causality 1065 via gene expression. SMR analysis is limited to significant (FDR < 0.05) cis SNP-expression 1066 quantitative trait loci (eQTLs) with MAF > 0.01. eQTLs passing these thresholds were combined 1067 with GWAS results in the SMR test, with significance (p_{SMR}) reported at a Bonferroni-corrected 1068 threshold for each eQTL data set. The eQTL architecture may differ between genes. For 1069 example, through LD, many SNPs can generate significant associations with the same gene, but 1070 in some instances multiple SNPs may be independently associated with the expression of a gene. 1071 After identification of significant SNP-expression-trait association through the SMR test, a 1072 follow-up heterogeneity test aims to prioritize variants by excluding regions for which there is 1073 conservative evidence for multiple causal loci ($p_{HET} < 0.05$). SMR analyses were conducted 1074 using eQTL data from whole peripheral blood(Westra et al., 2013), dorsolateral prefrontal cortex 1075 generated by the CommonMind Consortium⁸, and 11 brain sub-regions from the GTEx 1076 consortium(Consortium, 2015).

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1078 Regional joint GWAS

1079 Summary statistic Z-scores were calculated for each marker in each of the four main GWAS 1080 results, using the logistic regression coefficient and its standard error. Rare SNPs (MAF < 0.01), 1081 and SNPs with a low INFO score (< 0.3) in either dataset were removed. The causal variant 1082 relationships between SCZ and BD were investigated using the Bayesian method software pw-1083 gwas (v0.2.1), with quasi-independent regions determined by estimate LD blocks in an analysis 1084 of European individuals (n=1,703)(Berisa and Pickrell, 2015; Pickrell et al., 2016). Briefly, pw-1085 gwas takes a Bayesian approach to determine the probability of five independent models of 1086 association. (1) There is no causal variant in BD or SCZ; (2) a causal variant in BD, but not SCZ 1087 (3); a causal variant in SCZ, but not BD; (4) a shared causal variant influencing both BD and 1088 SCZ; (5) two causal variants where one influences BD, and one influences SCZ (Figure 2). The 1089 posterior probability of each model is calculated using model priors, estimated empirically 1090 within pw-gwas. Regions were considered to support a particular model when the posterior 1091 probability of the model was greater than 0.5.

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1093 Regional SNP-heritability estimation

1094 We calculated local SNP-heritability independently for SCZ and BD using the Heritability 1095 Estimator from Summary Statistics (HESS) software(Shi et al., 2016) for each of the 1096 independent regions defined above. The sum of these regional estimates is the total SNP-1097 heritability of the trait. To calculate local SNP-heritability HESS requires reference LD matrices 1098 representative of the population from which the GWAS samples were drawn. We utilized the 1099 1000 genomes European individuals as the reference panel(The 1000 Genomes Project 1100 Consortium, 2015). Unlike pw-gwas(Pickrell et al., 2016), HESS does not assume that only one 1101 causal variant can be present in each region.

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1103 DATA AND SOFTWARE AVAILABILITY

1104 Summary statistics from GWAS are publically available at
1105 https://www.med.unc.edu/pgc/results-and-downloads/downloads.
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