

Assortative mating and parental genetic relatedness drive the pathogenicity of variably expressive variants

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50 **ABSTRACT**

51 We examined more than 38,000 spouse pairs from four neurodevelopmental disease cohorts and the UK
52 Biobank to identify phenotypic and genetic patterns in parents associated with neurodevelopmental
53 disease risk in children. We identified correlations between six phenotypes in parents and children,
54 including correlations of clinical diagnoses such as obsessive-compulsive disorder ($R=0.31-0.49$,
55 $p<0.001$), and two measures of sub-clinical autism features in parents affecting several autism severity
56 measures in children, such as bi-parental mean Social Responsiveness Scale (SRS) scores affecting
57 proband SRS scores (regression coefficient=0.11, $p=0.003$). We further describe patterns of phenotypic
58 and genetic similarity between spouses, where spouses show both within- and cross- disorder
59 correlations for seven neurological and psychiatric phenotypes, including a within-disorder correlation
60 for depression ($R=0.25-0.72$, $p<0.001$) and a cross-disorder correlation between schizophrenia and
61 personality disorder ($R=0.20-0.57$, $p<0.001$). Further, these spouses with similar phenotypes were
62 significantly correlated for rare variant burden ($R=0.07-0.57$, $p<0.0001$). We propose that assortative
63 mating on these features may drive the increases in genetic risk over generations and the appearance of
64 “genetic anticipation” associated with many variably expressive variants. We further identified parental
65 relatedness as a risk factor for neurodevelopmental disorders through its inverse correlations with
66 burden and pathogenicity of rare variants and propose that parental relatedness drives disease risk by
67 increasing genome-wide homozygosity in children ($R=0.09-0.30$, $p<0.001$). Our results highlight the
68 utility of assessing parent phenotypes and genotypes in predicting features in children carrying variably
69 expressive variants and counseling families carrying these variants.

70 INTRODUCTION

71 Rare genomic variants, including copy-number variants (CNVs) and single nucleotide variants (SNVs),
72 contribute to a significant proportion of individuals with neurodevelopmental and psychiatric disorders¹⁻
73 ⁷. Many of these variants are often inherited and exhibit variable expressivity and phenotypic
74 heterogeneity, in contrast to syndromic variants with invariable clinical presentation⁸. For example,
75 phenotypic heterogeneity has been observed among carriers of rare recurrent SNVs associated with
76 neurodevelopmental disorders⁹, such as the candidate schizophrenia gene *NRXN1*, which also
77 contributes to intellectual disability/developmental delay (ID/DD), autism, and congenital
78 malformations with varying degrees of penetrance¹⁰. This variable expressivity and phenotypic
79 heterogeneity can be attributed to a multi-hit model, where a primary variant sensitizes the genome for
80 neuropsychiatric features and genetic interactions of the primary variant with other (“secondary”)
81 variants in the genetic background determine the ultimate phenotypic outcome¹¹⁻¹⁴. In fact, pathogenic
82 rare CNVs on chromosome 1q21.1, 15q13.3, and 16p12.1 tend to be transmitted from a relatively mildly
83 affected parent, with the increase in phenotypic severity corresponding to an increase in the number (i.e.
84 burden) of secondary variants over generations^{11,14,15}. These secondary variants are often inherited from
85 the non-carrier parent and have been shown to act synergistically with the primary pathogenic
86 variant^{14,16}. This phenomenon of complex inheritance forms the basis of many neurodevelopmental
87 disorders and complicates predicting the phenotypic trajectory of individuals within families with these
88 disorders.

89 These complexities necessitate examining other factors, such as parental phenotype and
90 genotype, in affected probands that may elucidate their ultimate phenotypic trajectory. Previous reports
91 have shown that family history is a risk factor for neurodevelopmental disorders, with increased within
92 and cross-disorder risk observed among relatives of individuals with psychiatric disease¹⁷⁻¹⁹. This
93 increase in liability towards multiple diagnostic outcomes may be due to the overlapping genetic

94 etiologies of neurodevelopmental disorders²⁰ and is particularly relevant for understanding mechanisms
95 of variable expressivity of rare variants. Here, we sought to broaden our understanding of the effect of
96 various phenotypic and genetic factors in parents on neurodevelopmental disease outcomes in children.
97 To that end, we examined more than 38,000 spouse pairs and parent pairs of children with
98 neurodevelopmental disorders to identify risk factors that affect disease trajectories in affected children,
99 including clinical and subclinical neurological and psychiatric features and patterns of assortative mating
100 on those features, as well as genetic similarity between parents. We further investigated the implications
101 of these patterns across generations and identified potential mechanisms by which they contribute to
102 disease risk. Our findings highlight assortative mating and parental kinship as previously unexplored
103 risk factors for neurodevelopmental and neuropsychiatric disease.

104 **SUBJECTS AND METHODS**

105 **Cohort Information**

106 We assessed parental factors affecting neurodevelopmental disease in two North American autism
107 cohorts: the Simons Simplex Collection (SSC)²¹ and the Simons Foundation Powering Autism Research
108 (SPARK)²². We analyzed genetic and/or phenotypic data from 2,517 trios from SSC and 7,175 trios
109 from SPARK. We also analyzed genetic and/or phenotypic data from an additional 67 parent pairs in
110 SSC and 1,570 parent pairs in SPARK for families where the relevant proband data were not available.

111 We also collected genomic and/or phenotypic data for another neurodevelopmental disease
112 cohort of 203 samples from 43 trios, 4 larger families, and 21 spouse pairs carrying the 16p12.1
113 deletion. Probands in these families were ascertained for developmental delays or other
114 neurodevelopmental phenotypes and were identified as 16p12.1 deletion carriers through clinical
115 diagnostic tests. Families provided informed consent to give blood samples and phenotypic data under
116 Pennsylvania State University Institutional Review Board approved protocol #STUDY00000278, and
117 deidentified data from medical reports were collected under approved protocol #STUDY00017269.

118 We further assessed assortative mating and spousal relatedness in 26,610 partner pairs from the
119 UK Biobank²³, identified using Data Return 1872²⁴ (Application ID: 45023). UK Biobank is a database
120 containing genetic, lifestyle, and health information from 500,000 volunteer participants from the UK.
121 Additionally, we assessed profiles of genetic relatedness in parent pairs of children ascertained for
122 neurodevelopmental disorders carrying rare copy number variants (CNVs) from another clinical cohort
123 evaluated at GeneDx. This cohort consists of families with affected children referred by clinicians to
124 GeneDx for genetic testing and is distinct from the 16p12.1 deletion cohort. Families were subsequently
125 selected for those carrying pathogenic CNVs (see *CNV calling*), resulting in 357 parent pairs from
126 families carrying 21 distinct disease-associated CNVs (**Supp. Table 1**). The study of this cohort was
127 conducted under an IRB protocol approved by the Western Institutional Review Board (IRB #STUDY

128 1169768, WIRB Pro Number 20162523), which states that this research meets the requirements for a
129 waiver of consent. We note that the terms “spouses” and “partners” are used interchangeably, while the
130 term “parents” refers to “spouses” or “partners” in the context of affected children in this manuscript.

131

132 **DNA Extraction and Sequencing**

133 DNA extraction and sequencing of samples within the SPARK, SSC, and UK Biobank cohorts were
134 performed by the relevant biobank. For directly recruited participants from the 16p12.1 deletion cohort,
135 DNA was extracted from peripheral blood using the QIAamp DNA Blood Maxi extraction kit (Cat. No:
136 51104, Qiagen), while clinical collaborators sent DNA samples directly to the Girirajan laboratory. SNP
137 arrays for 36 samples were generated as previously described¹⁴, while 92 additional samples were run on
138 Illumina OmniExpress 24 v 1.1 microarrays at the Northwest Genomics Center at the University of
139 Washington (Seattle, WA, USA). Illumina HiSeq X was used to generate 150-bp paired-end whole
140 genome sequence (WGS) data at $\geq 30X$ coverage for 118 individuals by Macrogen Labs (Rockville, MD,
141 USA). DNA extraction and sequencing from the GeneDx cohort were performed as previously
142 described²⁵.

143

144 **Variant calling**

145 *CNV calling*: CNVs in the GeneDx cohort were first called from exome sequencing data and confirmed
146 via microarray, as previously described²⁵. Disease-associated CNVs were identified from any overlap
147 with previously reported breakpoints for pathogenic CNVs¹⁵.

148 *SNV calling*: Single nucleotide variants (SNVs) and small indels were identified in 16p12.1 deletion
149 families using the GATK Best Practices pipeline. After removing duplicates with PicardTools²⁶, we
150 performed base-pair quality score recalibration and then used GATK HaplotypeCaller v.3.8²⁷ to call
151 variants. Calls from all samples were merged to call variants jointly using GATK GenotypeGVCFs

152 v.4.0.11²⁷, followed by variant quality score recalibration. Variants were filtered for those with an allele
153 balance between 0.25 and 0.75 or >0.9 and a read depth of ≥ 8 .

154 Whole genome (for SSC) and whole exome (for SPARK) variant call files (VCFs) were
155 downloaded from SFARI Base (<https://base.sfari.org>). Whole exome VCFs were obtained from the UK
156 Biobank using Data Field 23157. VCFs from all three cohorts were left-normalized and multi-allelic
157 records were split with BCFtools²⁸. Calls were filtered using the same criteria as above. Variants from
158 the 16p12.1 deletion cohort, SSC, and SPARK were additionally filtered for a quality ≥ 50 and a ratio of
159 quality to alternative allele depth ≥ 1.5 .

160 For all four cohorts, variant frequency in gnomAD²⁹ was annotated using vcfanno³⁰ and variants
161 were filtered for those with gnomAD exome and genome frequency $\leq 0.1\%$. Variants in the SSC,
162 SPARK, and UK Biobank cohorts were filtered for an intracohort frequency $\leq 0.1\%$. To achieve a
163 similar frequency threshold with a smaller sample size, variants in the 16p12.1 deletion cohort were
164 filtered for those that were private within related individuals. ANNOVAR³¹ was used to assign variants
165 to genes from GENCODE v19 (16p12.1 deletion cohort) or GENCODE v38 (SSC, SPARK, and UK
166 Biobank)³², and variants in all cohorts were filtered for loss-of-function (LOF), missense, or splice
167 variants in protein-coding genes. LOF variants were those annotated as 'stopgain', 'stoploss',
168 'frameshift_deletion', 'frameshift_insertion', or 'frameshift_substitution' by ANNOVAR, and missense
169 variants were those annotated as 'nonsynonymous_SNV', 'nonframeshift_deletion',
170 'nonframeshift_insertion', or 'nonframeshift_substitution'. CADD Phred-like scores³³ were annotated
171 using vcfanno as a deleteriousness metric and missense and splice variants were filtered for those with a
172 CADD Phred score ≥ 25 . For probands within SSC, variants were additionally annotated with SFARI
173 Gene Tier of the affected gene³⁴.

174 *Polygenic risk score calculations:* We obtained autism GWAS summary statistics from Grove and
175 colleagues³⁵. Duplicate and ambiguous single nucleotide polymorphisms (SNPs) were removed, and the
176 remaining SNPs were filtered for imputation INFO scores > 0.8 . We then used PLINK³⁶ to perform an

177 initial round of quality control by filtering SNPs with minor allele frequency <0.05 , Hardy-Weinberg
178 equilibrium test p-value $<1.0 \times 10^{-5}$, and genotype rate <0.05 , and removed samples missing $>1\%$ of
179 genotypes or having a Mendelian error rate $>5\%$. We then used the HRC-1000 Genomes Imputation
180 toolkit (<https://www.well.ox.ac.uk/~wrayner/tools>) to process PLINK files into individual chromosomes
181 for imputation, and used VcfCooker (<https://genome.sph.umich.edu/wiki/VcfCooker>) to convert PLINK
182 files to VCF files. SNPs were imputed using the TOPMed v.r2 imputation server using Eagle v2.4 for
183 phasing³⁷. After imputation, VCF files were converted back to PLINK format, and SNPs were again
184 filtered using the criteria from the initial quality control. SNPs were additionally filtered to remove
185 samples with $\geq \pm 3SD$ of the mean heterozygosity and strand-flipped to match the GWAS summary
186 statistics. We filtered the GWAS summary statistics for SNPs present in the HapMap3 dataset³⁸, and
187 used 1000 Genomes datasets to calculate linkage disequilibrium matrices for the SNPs. After regressing
188 betas of GWAS SNPs according to linkage disequilibrium, we used the LDpred2-auto model³⁹ to
189 calculate autism PRS for probands in the SSC cohort.

190

191 **Phenotype Assessment**

192 We assessed self-reported psychiatric phenotypes in probands and parents in the SPARK cohort and 23
193 quantitative autism and cognitive metrics in probands (**Supp. Table 2**) and two quantitative autism
194 metrics in parents (Social Responsiveness Scale (SRS)⁴⁰ and Broad Autism Phenotype Questionnaire
195 (BAPQ)⁴¹) from the SSC cohort downloaded from SFARI base (<https://base.sfari.org>). We further
196 assessed neurological and psychiatric phenotypes in 57 parent pairs from 16p12.1 deletion families.
197 Information was collected on each adult family member either through self-reports using a standardized
198 clinical questionnaire or clinician reports. Curated data was generated using a union of clinical
199 diagnoses, self-reported phenotypes, and interpreted responses to clinical questionnaires. We also
200 assessed phenotypic correlations in the UK Biobank using self-reported diagnoses of psychiatric
201 disorders from Data Field 20544.

202 Linear regression was used to assess the predictive power of parental SRS and BAPQ scores and
203 various genetic factors, including autism PRS and rare CNVs and SNVs, towards child quantitative
204 phenotypes, while tetrachoric correlations were used to assess phenotypic correlations between
205 psychiatric diagnoses in parents and children and between spouse pairs. To minimize the effects of
206 comorbidities and isolate the specific correlations of interest, pairs were removed from cross-disorder
207 tetrachoric correlations if both members of the pair had either diagnosis of interest. For example, when
208 examining the relationship between the female member of the spouse pair having anxiety and their
209 partner having depression, pairs where the male partner had anxiety or the female partner had depression
210 were excluded from the analysis.

211

212 **Kinship Calculations**

213 In the 16p12.1 deletion, SSC, SPARK, and UK Biobank cohorts, we followed the same initial SNP
214 quality control steps outlined above for PRS calculations. Additionally, SNPs within long-range linkage
215 disequilibrium regions⁴² were removed, as were samples with ± 2 standard deviations of the mean cohort
216 heterozygosity rate. For SPARK, SSC, and the 16p12.1 cohort samples, we next assessed ancestry using
217 fastStructure⁴³ to cluster samples (k=5-6) against individuals with known ancestry from the HapMap3
218 project³⁸. Kinship analysis in these cohorts was restricted to pairs with self-reported or calculated
219 European ancestry. UK Biobank samples were filtered for pairs of European ancestry using Data Field
220 22006.

221 KING⁴⁴ was used to calculate kinship coefficients for 44 parent pairs from the 16p12.1 cohort,
222 1,996 parent pairs from the SSC cohort, 4,211 parent pairs from the SPARK cohort, 6,315 partner pairs
223 from the UK Biobank cohort, and 357 parent pairs from the GeneDx cohort. In the GeneDx cohort, 11
224 parent pairs with a kinship coefficient less than -0.1 were removed from further analysis, as this highly
225 negative value may indicate that these samples come from different ancestral populations.

226

227 **Runs of Homozygosity**

228 To identify runs of homozygosity (ROH), we applied the same initial SNP quality control filters
229 described above for PRS calculations on 3,714 probands ascertained for autism in the SPARK cohort
230 and 1,981 probands from the SSC cohort. ROHs were calculated using the PLINK *homozyg* function³⁶.
231 We allowed a maximum of one heterozygous and five missing SNPs in each ROH, and runs were
232 restricted to those at least 1 Mbp in length.

233

234 **Statistical Analysis**

235 Tetrachoric correlations were performed using the *correlation* package⁴⁵ in R. Linear regression was
236 performed using *statsmodels* in Python, with proband age at the time the Autism Diagnostic Observation
237 Schedule (ADOS) was administered included as a covariate in the models. Pearson correlations and t-
238 tests were performed using the *pearsonr* and *ttest_ind* functions in the *scipy stats* (v. 1.7.1) package⁴⁶ in
239 Python (v. 3.9.7), respectively.

240 RESULTS

241 Parental phenotypes predict child phenotypes

242 To understand the relevance of parental phenotypes in neurodevelopmental disorders, we examined the
243 correlation between six parent and child psychiatric phenotypes in the Simons SPARK autism cohort
244 (**Fig. 1A, Table S4A**). We observed significant positive within-disorder correlations between mothers
245 and fathers and male and female probands for most phenotypes (Pearson correlation, $R=0.1-0.49$,
246 Bonferroni-corrected $p<0.05$) (**Fig. 1A, Table S4A**). We also observed several cross-disorder
247 correlations, including those common to both sexes, such as the correlation between depression in
248 parents and anxiety in probands ($R=0.10-0.27$, $p<0.01$), and those which are unique to one sex, such as
249 the correlation between OCD in fathers and ADHD in female probands ($R=0.13$, $p=0.003$), suggesting
250 parental and sex-specific correlations for certain pairs of features (**Fig. 1A, Table S4A**). These
251 correlations recapitulate previous patterns of within- and cross- disorder correlations observed in studies
252 of both psychiatric disease cohorts and cohorts from the general population^{17,47,48}.

253 To assess the relationship of parent and child phenotypes outside the context of a family history
254 of psychiatric disorders, we used linear models to assess the relationship between 23 quantitative
255 measures of autism and cognitive features (**Table S2**) in probands from the simplex autism cohort SSC
256 and two measures of autism features in their unaffected parents, Social Responsiveness Scale (SRS)⁴⁰
257 and Broad Autism Phenotype Questionnaire (BAPQ)⁴¹ scores (**Fig. 1B, Table S4B**). Several measures
258 of autism features in probands were significantly influenced by parental SRS scores, including
259 Repetitive Behaviors Scale (RBS)⁴⁹ ($\beta=0.14$, Bonferroni-corrected $p=1.95\times 10^{-6}$), SRS ($\beta=0.11$,
260 $p=0.002$), and Child Behavior Checklist (CBCL)⁵⁰ internalizing ($\beta=0.15$, $p=1.11\times 10^{-5}$) scores of
261 children with autism were positively associated with biparental mean SRS scores (**Fig. 1B, Table S4B**).
262 Biparental mean SRS and/or BAPQ scores were in fact stronger predictors of autism severity in affected
263 children than any individual group of rare or common genetic variants for several measures (**Fig. 1B,**

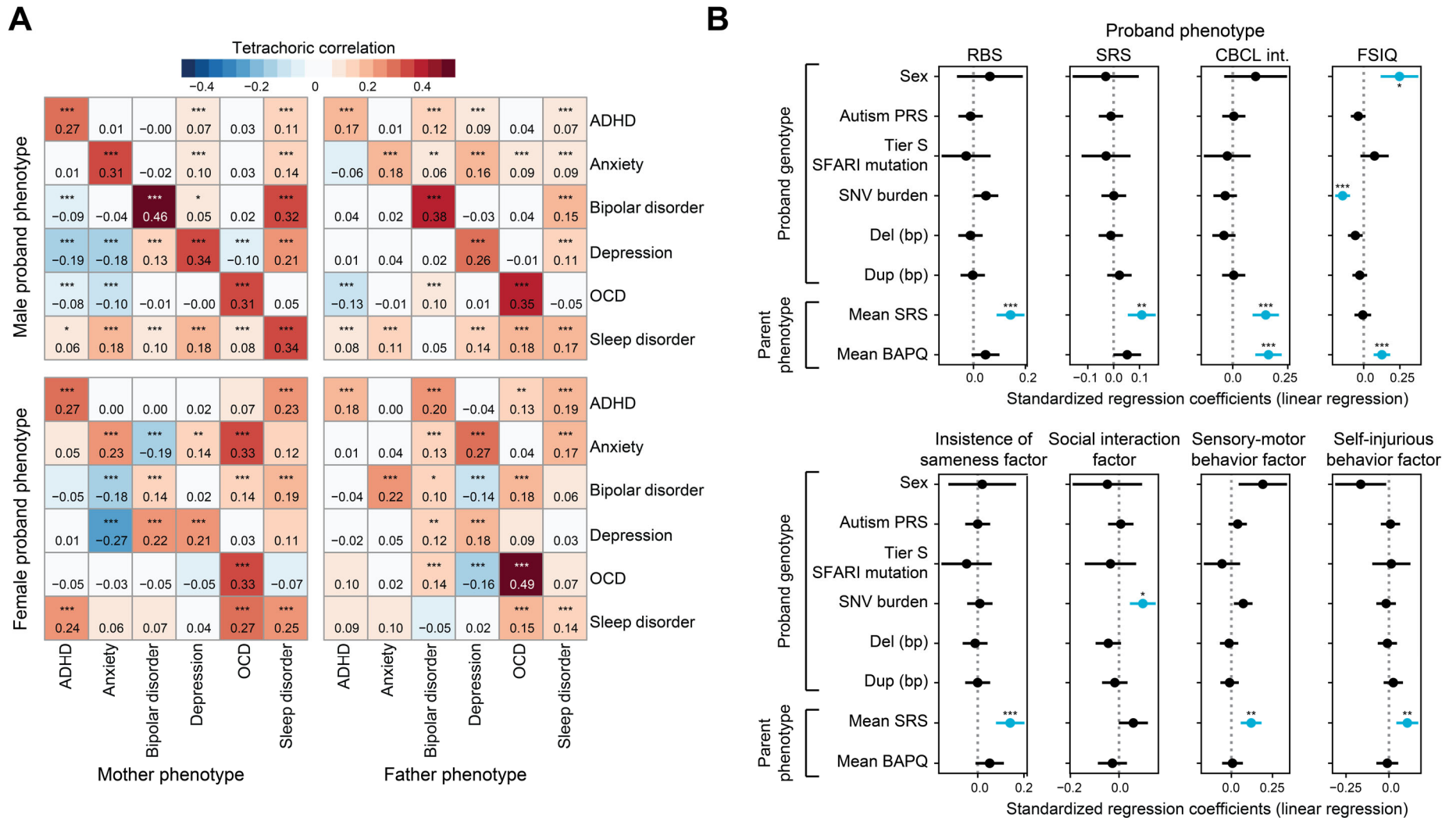


Figure 1. Parental phenotypes predict clinical outcomes in children.

(A) Tetrachoric correlations between self-reported psychiatric disorders in mothers and fathers and male and female probands within the SPARK autism cohort (n=592-4,757 for each correlation). (B) Regression coefficients from linear models incorporating effects of both genetic factors and parental phenotypes towards quantitative proband autism and cognitive phenotypes (n=1,628-2,168 for each model).

RBS: Repetitive Behaviors Scale, SRS: Social Responsiveness Scale, CBCL int: Child Behavior Checklist 6-18 internalizing, FSIQ: Full-scale IQ. Selected models shown here, full data in Table S4. Blue dots and lines indicate significance after Bonferroni-correction.

* Bonferroni-corrected $p \leq 0.05$, ** $p \leq 0.01$, *** $p \leq 0.001$.

264 **Table S4B**). These data highlight the potential utility of assessing parent phenotypes to predict child
265 phenotypes. Even in simplex families, where parents are largely unaffected, child phenotypes can be
266 predicted by parent phenotypes, suggesting some sub-clinical level of genetic liability in the parents of
267 affected children which compounds within a single generation to surpass a disease threshold. This also
268 indicates the presence of multiple genetic factors which individually are not sufficient to cause overt
269 phenotypes in parents but may additively or synergistically interact in offspring to cause more overt
270 phenotypes in subsequent generations.

271

272 **Phenotypic assortment among partner pairs**

273 Given the relevance of parental phenotypes towards predicting phenotypes in affected children, we
274 hypothesized that assortative mating may be a relevant risk factor for neurodevelopmental disease.
275 Assortative mating has been well described for neurological and psychiatric phenotypes^{24,51–58} and has
276 previously been shown to result in increasing disease liability over generations using national registrar
277 data from Sweden and Denmark^{17,47,59}. This is akin to patterns of increasing disease liability observed in
278 neurodevelopmental disease cohorts, particularly in families carrying variably expressive variants^{11,14,15}.
279 Using the 16p12.1 deletion as a paradigm for these disorders, we first examined the frequency of self-
280 reported neurological and psychiatric features in adult family members of children with the 16p12.1
281 deletion ascertained for neurodevelopmental phenotypes. Adult deletion carriers in our cohort
282 manifested several neurological and psychiatric phenotypes, including seizures (5.8%), schizophrenic
283 features (20%), depression (36%) and anxiety (39%) symptoms, and addiction phenotypes (8.0%) (**Fig.**
284 **2A**). In line with expectations of assortative mating, their non-carrier partners also manifested these
285 features at similar frequencies (seizure: 5.8%, schizophrenic features: 12%, depression: 28%, anxiety:
286 26%, addiction: 10%) (**Fig. 2A**). As these data are indicative of assortative mating, we further assessed
287 spousal assortment based on these phenotypes and were able to identify several significant within- and
288 cross-disorder correlations (**Fig. 2B, Table S4C**). For example, 16p12.1 deletion carriers with anxiety

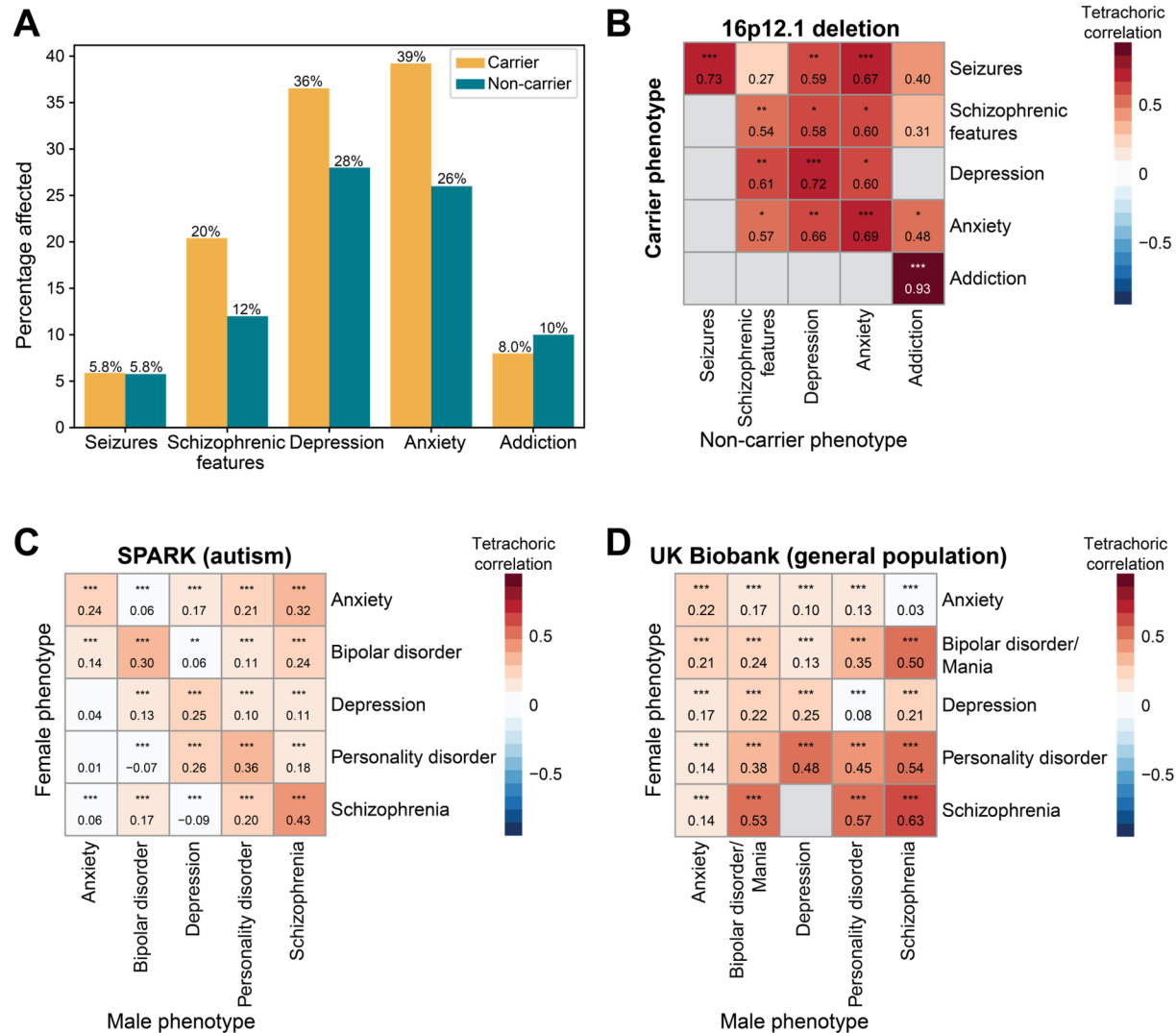


Figure 2. Patterns of assortative mating in multiple cohorts.

(A) Percentage of carrier and non-carrier spouses in a cohort of 16p12.1 deletion carrier families with five neurological and psychiatric disorders (n=49-52). (B-D) Tetrachoric correlations of neurological and psychiatric phenotypes between (B) carrier and non-carrier spouses in the 16p12.1 deletion cohort (n=28-50), (C) mothers and fathers in the SPARK autism cohort (n=3,956-6,142), and (D) female and male spousal partners in the UK Biobank (n=19,468-22,953). Grey boxes indicate insufficient sample size for correlation. * Bonferroni-corrected $p \leq 0.05$, ** $p \leq 0.01$, *** $p \leq 0.001$.

289 were likely to have partners with anxiety ($R=0.69$, Bonferroni-corrected $p=8.14\times 10^{-7}$) or depression
290 ($R=0.66$, $p=0.001$) (**Fig. 2B, Table S4C**). These data confirm that the well-documented patterns of
291 assortative mating for psychiatric phenotypes are typically prevalent in families carrying variably
292 expressive variants with a high load of genetic liability for neurodevelopmental disorders. We also
293 compared patterns of phenotypic assortment in parents of children in an affected cohort and the general
294 population by analyzing five self-reported psychiatric phenotypes in the SPARK autism cohort and the
295 UK Biobank (**Fig. 2C-D, Table S4D**). Although these phenotypes occur much more frequently in the
296 SPARK cohort (**Supp. Table 3**), positive within- and cross- disorder correlations were observed in both
297 of these cohorts (**Fig. 2C-D, Table S4D**). For example, women in both cohorts with bipolar disorder or
298 mania were more likely to have partners with bipolar disorder (SPARK: $R=0.30$, Bonferroni-corrected
299 $p=7.69\times 10^{-129}$; UKB: $R=0.24$, $p=3.53\times 10^{-304}$) or schizophrenia (SPARK: $R=0.24$, $p=3.90\times 10^{-80}$; UKB:
300 $R=0.50$, $p=0$) (**Fig 2C-D**). This confirms that spousal assortment on psychiatric phenotypes occurs both
301 in individuals ascertained for disease and those from the general population.

302

303 **Genetic consequences of assortative mating**

304 Partner assortment will result in positive correlations between their risk alleles, which underly the
305 phenotypes driving assortment⁶⁰. While genetic risk of assortative mating is often measured using
306 association studies of common variants^{24,60-62}, we hypothesized that rare deleterious variants also
307 correlate among partners, given the links between rare variants and psychiatric disorders on which
308 partners tend to assort^{7,63,64}. To confirm this pattern in the assessed cohorts, we correlated the burden of
309 rare deleterious coding single nucleotide variants (SNVs) between partners in each cohort (**Fig. 3A**). We
310 observed that the burden of rare variants positively correlated between partners in the 16p12.1 deletion
311 cohort (Pearson's $R=0.46$, $p=2.79\times 10^{-4}$), SPARK ($R=0.52$, $p<10^{-12}$), Simons Simplex Collection (SSC,
312 $R=0.57$, $p<10^{-12}$), and the UK Biobank ($R=0.07$, $p=6.80\times 10^{-9}$) cohorts (**Fig. 3A**), confirming that partner

313 assortment based on related phenotypes drives correlations in genetic risk. Interestingly, the correlation
314 was much weaker in the UK Biobank ($R=0.07$) than the disease cohorts ($R\geq 0.46$), possibly due to the
315 lower frequency of psychiatric phenotypes (**Supp. Table 3**) and lower burden of genetic risk in this
316 cohort. The ultimate outcome of this correlation in burden is an increase in the variance of genetic
317 liability over generations⁶⁵, where affected individuals will have more affected children and unaffected
318 or mildly affected individuals will have more mildly affected children (**Fig. 3B**), mimicking patterns
319 observed in families carrying variably expressive variants. For example, 16p12.1 deletion carrier
320 families typically show increases in clinical severity over generations, from unaffected or mildly
321 affected grandparents to parents with psychiatric features and then to children with neurodevelopmental
322 disorders (**Fig. 3C**) and this increase in phenotypic severity corresponds to an increase in genetic
323 burden^{11,14,16} (**Supp. Figure 1**). Taken together, this suggests that assortative mating drives the increases
324 in genetic liability that result in the severe phenotypes observed in neurodevelopmental disease cohorts,
325 particularly in carriers of variably expressive variants.

326

327 **Parental relatedness as a risk factor for neurodevelopmental disease**

328 Another parental factor which can increase disease risk is the level of consanguinity or genetic
329 relatedness of parents, through increasing both the frequency of autosomal recessive conditions and
330 genome-wide homozygosity^{66,67}. Long stretches of homozygosity have been associated with increased
331 risk for several neurodevelopmental disorders^{68,69}. Although the parents assessed in our cohorts are not
332 consanguineous, we hypothesized that increases in parental relatedness lead to increases in genome-
333 wide homozygosity and increased risk of neurodevelopmental disorders. To assess this, we examined
334 large (≥ 1 Mbp) runs of homozygosity (ROH) in 3,714 and 1,981 children of the parent pairs analyzed in
335 the SPARK and SSC cohorts, respectively. We found a significant correlation between parental
336 relatedness (as measured by kinship coefficients) and the number of large ROH in both male and female

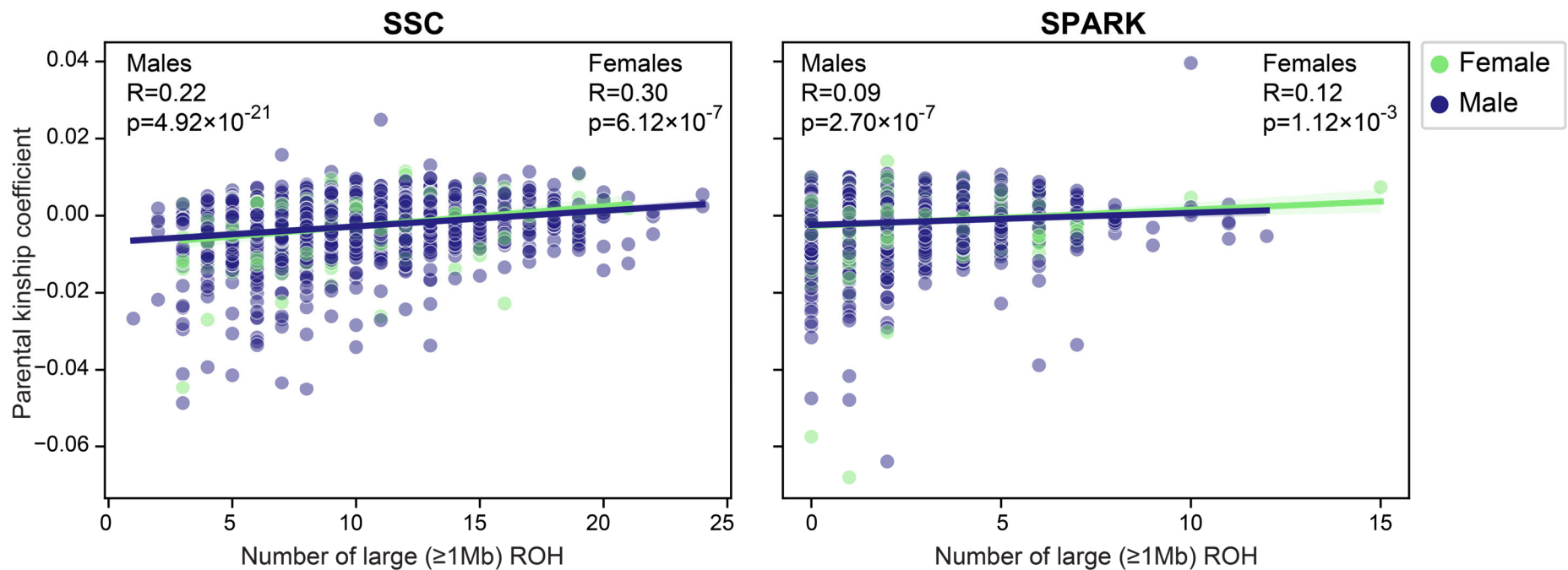


Figure 4. Parental relatedness correlates with homozygosity in children.

Pearson correlations of parental relatedness (measured as kinship coefficients) and the number of large (≥ 1 Mb) runs of homozygosity in male and female probands in the SSC (male $n=1,711$, female $n=270$) and SPARK (male $n=2,956$, female $n=758$) autism cohorts.

337 probands from both SSC (male $R=0.22$, $p=4.92\times 10^{-21}$, female $R=0.30$, $p=6.12\times 10^{-7}$) and SPARK (male
338 $R=0.09$, $p=2.70\times 10^{-7}$, female $R=0.12$, $p=1.12\times 10^{-3}$) (**Fig. 4**), indicating that increased spousal
339 relatedness could lead to increased disease risk by unmasking recessive alleles within those intervals.

340 We then asked how this parental relatedness could interact with other neurodevelopmental
341 disease factors. We calculated kinship coefficients as a measure of relatedness between partner pairs
342 from all four cohorts and found negative correlations between spousal kinship and the average spousal
343 rare variant burden in all three neurodevelopmental disease cohorts (16p12.1 deletion: $R=-0.30$, $p=0.04$;
344 SPARK: $R=-0.13$, $p=1.05\times 10^{-17}$, SSC: $R=-0.23$, $p=1.34\times 10^{-23}$) (**Fig. 5A**). However, no significant
345 correlation was observed for partners in the UK Biobank ($R=-0.02$, $p=0.07$) (**Fig. 5A**). We then
346 examined parental relatedness in the context of severe damaging mutations in children ascertained for
347 autism in SSC. We found that kinship coefficients of 102 parent pairs of children with loss-of-function
348 mutations in canonical autism-associated genes (Tier S SFARI genes)³⁴ were significantly lower than
349 kinship coefficients in 1,251 parent pairs of affected children without damaging mutations in autism-
350 associated genes (one-tailed t-test $p=0.041$) (**Fig. 5B**). We also examined kinship coefficients of 346
351 parent pairs from an independent cohort of children ascertained for neurodevelopmental disorders
352 carrying 21 different pathogenic CNVs (**Supp. Table 2**). We found that parental kinship had an inverse
353 relationship with severity as well as penetrance of the primary variant in children. Kinship coefficients
354 were significantly higher among parents of children with inherited CNVs compared to parents of
355 children with *de novo* CNVs (one-tailed t-test, $p=0.03$) (**Fig. 5C**). Similarly, parents of children with
356 duplications, which are typically associated with incomplete penetrance⁷⁰ and less severe features⁷¹ than
357 deletions, had higher kinship coefficients compared to parents of children with deletions (one-tailed t-
358 test, $p=0.02$) (**Fig. 5D**). Although not statistically significant, we also identified a trend of higher kinship
359 values in families with variably expressive CNVs compared to syndromic CNVs (one-tailed t-test,
360 $p=0.14$) (**Fig. 5E**).

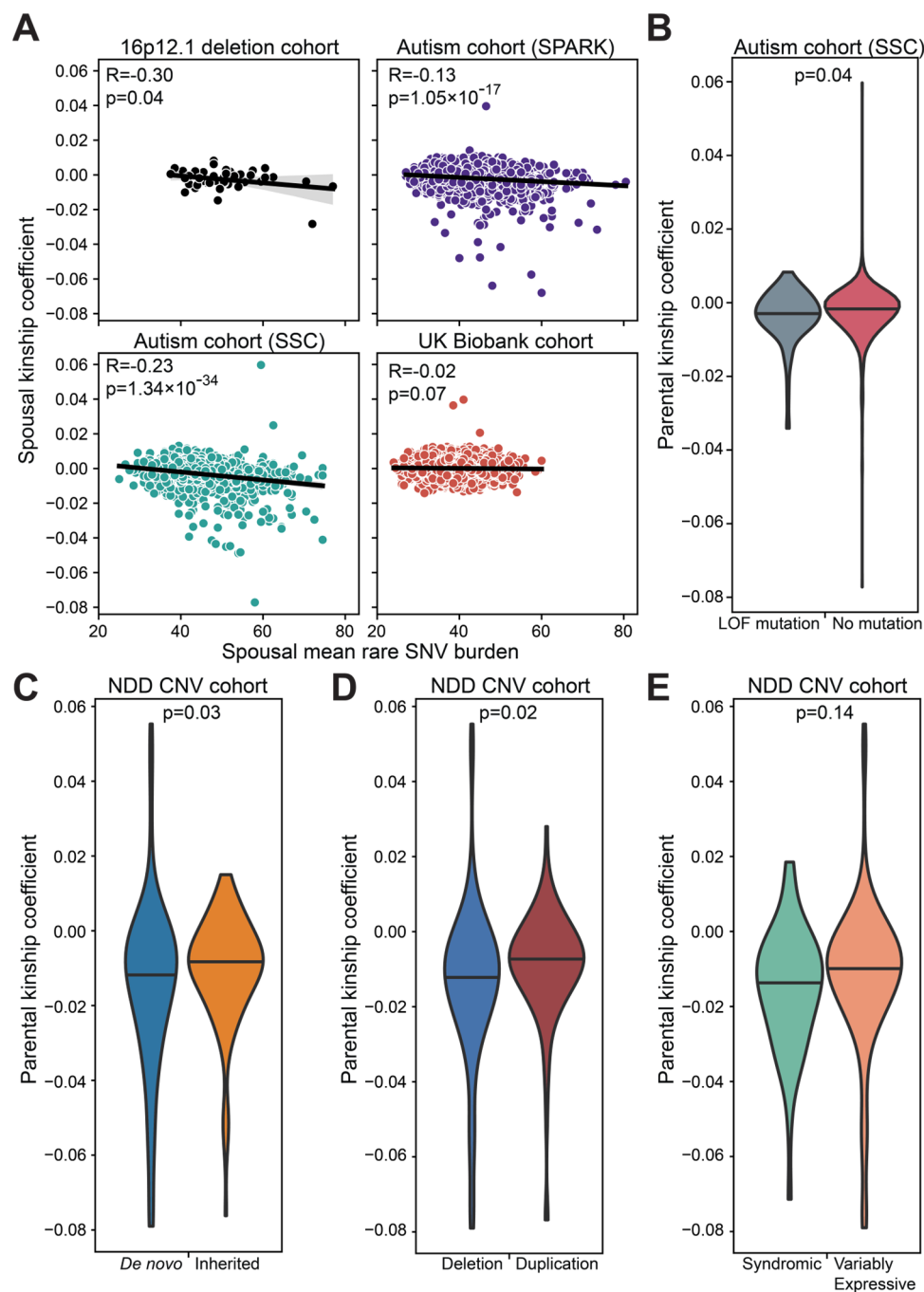


Figure 5. Parental relatedness is a risk factor for neurodevelopmental disorders.

(A) Pearson correlations of the average burden of rare, likely deleterious variants and relatedness (measured as kinship coefficients) in spouse pairs from the 16p12.1 deletion ($n=44$), SPARK ($n=4,095$) and SSC ($n=1,813$) autism, and UK Biobank ($n=6,315$) cohorts. (B) Comparison of parental relatedness among SSC probands with loss-of-function mutations in Tier S SFARI genes ($n=102$) and without mutations in SFARI genes ($n=1,243$). Lines indicate median. One tailed t-test. (C-E) Comparisons of parental relatedness among probands ascertained for neurodevelopmental disease with (C) *de novo* ($n=129$) and inherited ($n=116$) CNVs, (D) deletions ($n=249$) and duplications ($n=97$), and (E) syndromic ($n=47$) and variably expressive ($n=299$) CNVs. Lines indicate median. One tailed t-test.

361 These patterns are in accordance with a liability threshold model (LTM), which posits that an
362 individual will manifest disease once the sum of various independent risk factors exceeds a certain
363 threshold⁷². An inverse relationship between all independent risk factors contributing to disease is
364 implied by the LTM, and has been observed previously when examining additive effects of rare and
365 common variants towards neurodevelopmental disease^{1,73,74}. Thus, the negative relationship between
366 parental kinship and rare SNV burden or presence of highly pathogenic mutations supports parental
367 relatedness as a risk factor for variably expressive pathogenic variants, likely through increased levels of
368 genome-wide homozygosity. This is a particularly compelling hypothesis considering the lack of
369 correlation between spousal kinship and mean spousal rare variant burden seen in the UK Biobank pairs
370 (**Fig. 5A**). Unlike the other cohorts, the pairs in the UK Biobank were not ascertained for children with
371 neurodevelopmental disease and thus may not exist within the same range of liability threshold as the
372 other cohorts.

373 **DISCUSSION**

374 Assortative mating has previously been hypothesized to explain the high degree of heritability for
375 complex neurodevelopmental and psychiatric disorders^{54,75}. However, the implications of assortative
376 mating have not been explored in the context of rare variably expressive pathogenic variants. Here, we
377 confirm that the well-documented patterns of spousal assortment on neuropsychiatric features^{24,51–58}
378 hold true for families carrying rare disease-associated variants. It is possible that the observed
379 phenotypic similarity in partners may be caused by mechanisms outside of assortative mating, including
380 partner convergence⁷⁶ or, in disease cohorts, caregiver burden on parents of affected children⁷⁷.
381 However, we find it unlikely that these mechanisms could account for all observed partner similarity for
382 three key reasons. Firstly, previous studies have indicated that convergence has only a modest effect on
383 spousal similarity^{76,78}. Further, of the 27 participants in the 16p12.1 deletion cohort who indicated the
384 age at depression symptom onset, 70% (19/27) began having symptoms before the birth of their first
385 child. Thus, caregiver burden is an unlikely cause of depression in these individuals and, although these
386 data are limited, we expect similar trends for other disorders. Finally, if spousal phenotypic similarity
387 were the result of convergence or caregiver burden, we would not expect to see a correlation in genetic
388 risk between partners. However, as we do see a strong correlation in rare variant burden, we can
389 conclude that active assortment accounts for much of the observed phenotypic similarity between
390 partners.

391 Partner assortment on these phenotypes drives increases in genetic risk alleles across generations
392 in affected individuals⁷⁵, similar to the observed patterns of increasing secondary variant burden that are
393 well documented in cohorts of variably expressive CNVs^{11,14} and are concordant with a “two-hit”
394 mechanism of pathogenicity. Under assortative mating, individuals with similar psychiatric phenotypes
395 but distinct underlying genetic etiologies may preferentially mate and have a child who carries genetic
396 risks from both parents, leading to a compounding of genetic burden over generations. Increased disease

397 liability in children resulting from assortative mating has been previously shown^{17,59}, and the
398 implications for this finding in the context of variably expressive rare variants has been previously
399 hypothesized⁷⁹. Here, our analysis provides more evidence to support the relationship between parental
400 assortment on psychiatric phenotypes and compounding genetic burden towards disease severity.
401 Unique combinations of rare variants arising from recent mutations have been identified as a major
402 contributor to disease^{80,81} and assortative mating may drive accumulation of those new combinations.
403 This observation has important implications on the trajectory of individuals within a family carrying a
404 variably expressive variant and provides insights into the genetic “anticipation” observed in these
405 families.

406 We further highlight the role of spousal relatedness towards complex disorders, showing that
407 spousal kinship acts as a risk factor for variably expressive variants. Long stretches of homozygosity
408 have previously been implicated in neurodevelopmental disorders^{68,69} and we show here that parental
409 relatedness increases genome-wide homozygosity and, likely through this mechanism, acts as a risk
410 factor for neurodevelopmental disease. Increases in spousal kinship are often due to population
411 stratification⁷⁶, which has heretofore been an underexplored risk factor for neurodevelopmental
412 disorders and may represent an exciting new avenue for further research.

413 While these results have interesting implications for our understanding of disease pathogenicity,
414 our study has some limitations. We restricted our analysis of phenotypic assortment to only a subset of
415 distinct phenotypes, and while we attempted to control for comorbidities that could inflate correlation
416 estimates, it is likely that some cross-disorder contamination remains. Further, all cohorts assessed here
417 were primarily of European descent and further research is needed to understand how assortment and
418 partner similarity may vary across racial and ethnic groups, particularly for disease-associated rare
419 variants. This may be particularly relevant in the context of isolated populations or those practicing
420 endogamy, who, due to their unique recent genetic history, may have a higher load of both parental

421 relatedness and rare variant burden^{82,83}. Further research is required to examine the genetic architecture
422 in these populations and how these changes in architecture may impact disease risk.

423 In summary, we provide evidence for assortative mating on neuropsychiatric disorders,
424 particularly in parents of children with neurodevelopmental phenotypes. We show patterns of rare
425 variant burden similarities between spouses and highlight compounding of rare variant burden over
426 generations resulting from phenotypic assortment. We also highlight increased spousal kinship as a risk
427 factor for neurodevelopmental disorders through the unmasking of recessive alleles. Overall, our results
428 highlight the importance of assessing parental history of neuropsychiatric disease and parental
429 relatedness for more accurate assessment of neuropsychiatric disease risk in children and counseling of
430 families carrying variably expressive variants.

431

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437

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445 Approved researchers can obtain the SSC and SPARK datasets described in this study by applying at
446 <https://base.sfari.org>. This research has been conducted using data from UK Biobank, a major
447 biomedical database. More information about UK Biobank is available at <https://www.ukbiobank.ac.uk/>.

448

449 **DECLARATIONS OF INTERESTS**

450 Dr. Santhosh Girirajan reports grants from the National Institutes of Health outside the submitted work.
451 Drs. Lisa Dyer and Jane Juusola are employees of GeneDx, LLC. All other authors report no financial
452 relationships with commercial interests.

453

454 **DATA AND CODE AVAILABILITY**

455 Microarray and genome sequencing data from participants in the 16p12.1 deletion cohort who consented
456 to having their information shared anonymously will be made available at NCBI dbGaP study accession
457 phs00245. Phenotype and genetic data from the SPARK and SSC autism and UK Biobank cohorts
458 through the cohorts' data portals: SFARI base (<https://base.sfari.org>) and UK Biobank database
459 (<https://www.ukbiobank.ac.uk>), respectively. All code for analysis will be available at GitHub
460 (https://github.com/girirajanlab/Assortative_mating_project).

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