

1 A cross-disorder MR-pheWAS of 5 major psychiatric disorders in

2 UK Biobank

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21 **Short title:** Cross-disorder MR-pheWAS of 5 major psychiatric disorders in UK
22 Biobank

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24 ABSTRACT

25 Psychiatric disorders are highly heritable and associated with a wide variety of social
26 adversity and physical health problems. Using genetic liability (rather than phenotypic
27 measures of disease) as a proxy for psychiatric disease risk can be a useful alternative
28 for research questions that would traditionally require large cohort studies with long-
29 term follow up.

30 Here we conducted a hypothesis-free phenome-wide association study in about
31 300,000 participants from the UK Biobank to examine associations of polygenic risk
32 scores (PRS) for five psychiatric disorders (major depression (MDD), bipolar disorder
33 (BP), schizophrenia (SCZ), attention-deficit/ hyperactivity disorder (ADHD) and autism
34 spectrum disorder (ASD)) with 23,004 outcomes in UK Biobank, using the open-
35 source PHEANT software package.

36 There was evidence after multiple testing ($p < 2.55 \times 10^{-6}$) for associations of PRSs with
37 226 outcomes, most of them attributed to associations of PRS_{MDD} ($n=120$) with mental
38 health factors and PRS_{ADHD} ($n=77$) with socio-demographic factors. Among others, we
39 found strong evidence of associations between a 1 standard deviation increase in
40 PRS_{ADHD} with 1.1 months younger age at first sexual intercourse [95% confidence
41 interval [CI]: -1.26, -0.94]; PRS_{ASD} with 0.01% reduced lower erythrocyte distribution
42 width [95%CI: -0.013, -0.007]; PRS_{SCZ} with 0.98 odds of playing computer games
43 [95%CI: 0.976, 0.989]; PRS_{MDD} with a 0.11 points higher neuroticism score
44 [95%CI: 0.094, 0.118] and PRS_{BP} with 1.04 higher odds of having a university degree
45 [95%CI: 1.033, 1.048].

46 We were able to show that genetic liabilities for five major psychiatric disorders
47 associate with long-term aspects of adult life, including socio-demographic factors,
48 mental and physical health. This is evident even in individuals from the general
49 population who do not necessarily present with a psychiatric disorder diagnosis.

50 **AUTHOR SUMMARY**

51 Psychiatric disorders are associated with a wide range of adverse health, social and
52 economic problems. Our study investigates the association of genetic risk for five
53 common psychiatric disorders with socio-demographics, lifestyle and health of about
54 330,000 participants in the UK Biobank using a systematic, hypothesis-free approach.
55 We found that genetic risk for attention deficit/hyperactivity disorder (ADHD) and
56 bipolar disorder were most strongly associated with lifestyle factors, such as time of
57 first sexual intercourse and educational attainment. Genetic risks for autism spectrum
58 disorder and schizophrenia were associated with altered blood cell counts and time
59 playing computer games, respectively. Increased genetic risk for depression was
60 associated with other mental health outcomes such as neuroticism and irritability. In
61 general, our results suggest that genetic risk for psychiatric disorders associates with
62 a range of health and lifestyle traits that were measured in adulthood, in individuals
63 from the general population who do not necessarily present with a psychiatric disorder
64 diagnosis. However, it is important to note that these associations aren't necessary
65 causal but can themselves be influenced by other factors, like socio-economic factors
66 and selection into the cohort. The findings inform future hypotheses to be tested using
67 causally informative designs.

68 INTRODUCTION

69 Family and twin research as well as large-scale genome-wide association studies
70 (GWAS) have shown that psychiatric disorders are highly heritable (1) and that genetic
71 risks for psychiatric disorders also are associated with socio-economic factors,
72 physical health outcomes as well as other psychiatric disorders (2-5). Using genetic
73 liability (rather than phenotypic measures of disease) as a proxy for psychiatric
74 disease risk can be a useful alternative for research questions that would traditionally
75 require long-term follow up and big datasets due to the low prevalence of some of the
76 psychiatric disorders of interest in the population (e.g. adult-onset health
77 consequences of child neurodevelopmental disorders). In addition, while high genetic
78 risk for a psychiatric disorder is not always indicative of a diagnosis of psychiatric
79 disease, it can index underlying subthreshold symptomatology that can still impact
80 later adversities and quality of life (6).

81 So far, studies have used hypothesis-driven approaches to investigate associations of
82 genetic risk for psychiatric disorders with various psychiatric and health outcomes as
83 well as lifestyle factors (7, 8). However, big data resources that are readily available,
84 such as UK Biobank with about 500,000 participants, provide rich phenotypic
85 information that can be used for hypothesis-free studies and offset the multiple testing
86 burden. Phenome scans are a type of hypothesis-free analysis where the association
87 of a trait of interest is systematically tested with a potentially large number of
88 phenotypes and can be hypothesis-generating by identifying an association when
89 there is no prior reason to expect that an association may exist. As all available
90 phenotypes are tested and the less 'significant' results published alongside those of
91 greater 'significance', phenome scans can help to reduce biases associated with

92 hypothesis-driven studies where researchers might only publish the most desirable or
93 expected results.

94 In a Mendelian randomization genome-wide association study (MR-pheWAS)
95 genetic risk is used as a proxy for lifelong liability for a disorder to explore associations
96 of this genetic liability with traits that may evolve as a consequence. Understanding
97 these associations will be essential to inform prevention or early intervention
98 strategies. However, conclusions about causality are limited due to the low predictive
99 power and high pleiotropic effects of genetic risk scores for psychiatric conditions (8).

100 The aim of this study was to investigate the associations between genetic risk for five
101 common psychiatric disorders – attention-deficit/ hyperactivity disorder (ADHD),
102 autism spectrum disorder (ASD), schizophrenia (SCZ), major depression (MDD) and
103 bipolar disorder (BP) - with a wide range of socio-demographic, lifestyle, physical and
104 mental health outcomes in UK Biobank, using the systematic hypothesis-free MR-
105 pheWAS approach.

106 **RESULTS**

107 In total 334,976 participants of white British ancestry in UK Biobank were included in
108 this study with an average age of 56 (standard deviation [SD]=8) years. A descriptive
109 overview of selected UK Biobank study sample characteristics is given in Figure 1A.
110 The UK Biobank participants are known to be more educated and healthier than the
111 average UK population which is reflected in the high percentage of people with a
112 university degree (47%) and low prevalence of current smoking (10%) in the sample,
113 which is comparable to the full UK Biobank release (9). Furthermore, 34% of
114 participants reported to have seen a general practitioner and 11% a psychiatrist for
115 nerves, anxiety, tension or depression but there are few cases of schizophrenia
116 (n=132), ADHD (n=71), ASD (n=143) or bipolar disorder (n=439). An overview of UK
117 Biobank phenotype categories is given in Figure 1B.

118 **DISORDER SPECIFIC EFFECTS**

119 The MR-pheWAS of each psychiatric disorder tested the association of the respective
120 polygenic risk score, aggregated from independent, genome-wide significant SNPs,
121 with 23,004 outcomes in UK Biobank, adjusted for age, sex and the first 10 genetic
122 principal components. There was strong evidence after multiple testing correction
123 based on the number of independent tests derived from spectral decomposition
124 ($p < 2.55 \times 10^{-6}$) for associations of either the ADHD, ASD, SCZ, MDD or BP PRS with
125 226 outcomes in 31 UK Biobank categories (Figure 2 and Table S1) as described
126 below. A less stringent 5% FDR multiple testing threshold identified 209 additional
127 outcomes also associated with at least one PRS (Figure S1). Correlations among the
128 PRS can be found in supplementary Table S2. A detailed list of all MR-pheWAS results
129 generated by the open-source PHESANT software package can be found in Table S3.

130 Unless stated as a PHEANT result, estimates for continuous outcomes are
131 generated by following up outcomes and manually curating the outcome phenotypes,
132 to compute estimates on their original scale.

133 ***Attention deficit/ hyperactivity disorder***

134 PRS_{ADHD} was strongly associated with 77 outcomes (Figure 3) including 39 socio-
135 demographic factors, 33 general health and 5 mental health, brain and cognition
136 outcomes. The strongest evidence of association with PRS_{ADHD} was seen for socio-
137 demographic and lifestyle factors. A 1 SD higher PRS_{ADHD} was associated with a
138 1.09 month younger age at first sexual intercourse [95% confidence interval [CI]:
139 -1.26,-0.94] (p=2.0x10⁻¹⁶), and 0.96 lower odds of having a university degree [95% CI:
140 0.95, 0.97] (p=1x10⁻²⁹). In addition, PRS_{ADHD} was associated with younger maternal
141 and paternal age of their parents (-0.08 years [95%CI: -0.103,-0.051] p=4.4x10⁻⁹; -0.10
142 years [95% CI: -0.134,-0.067] p=3.2x10⁻⁹,respectively), 0.97 lower odds of average
143 household income [95%CI: 0.96,0.97] (p=1.3x10⁻²⁰), 1.05 higher odds of current
144 smoking [95%CI :1.03,1.06] (p=5.7x10⁻¹⁵) and 1.04 higher odds of experiencing
145 physical abuse as a child [95%CI: 1.02,1.06] (p=1.3x10⁻⁶).

146 Further, 1 SD increase in PRS_{ADHD} was associated with 17 physical health outcomes
147 related to obesity, including 0.05 kg/m² higher BMI [95%CI: 0.037,0.070] (p=7.4x10⁻¹¹),
148 leg and arm fat mass, waist circumference and trunk fat mass. Furthermore, there was
149 evidence for an association of PRS_{ADHD} with blood measures, such as 0.02 cells/L
150 higher leukocyte count [95%CI: 0.012,0.026] (p=1.4x10⁻⁷).

151 Associations seen for brain and cognition include 0.04 points lower fluid intelligence
152 score [95%CI: -0.050,-0.026] (p=1.9x10⁻⁹).

153

154 ***Autism spectrum disorder***

155 PRS_{ASD} was strongly associated with 9 outcomes (Figure 3), including 1 socio-
156 demographic, 7 general health and 1 mental health outcome.

157 The strongest association of PRS_{ASD} was found for lower erythrocyte distribution width
158 where 1 SD higher PRS_{ASD} was associated with 0.01% lower erythrocyte distribution
159 width [95% CI: -0.013, -0.007] (p=6.3x10⁻¹⁰) and 0.98 lower odds of comparative body
160 size at age 10 [95%CI:0.97,0.98] (p=6.2x10⁻¹¹). Furthermore, 1 SD higher PRS_{ASD} was
161 associated with 0.001 g/cm² lower heel bone mineral density (BMD) [95%CI:-0.002,-
162 0.001] (p=4.1x10⁻⁵).

163 The only mental health outcome that was associated with PRS_{ASD} was 1.02 higher
164 odds of being a nervous person (“suffer from nerves”) [95%CI:1.01,1.03] (p=7.7x10⁻⁷).

165 ***Schizophrenia***

166 There was strong evidence of association for PRS_{SCZ} with 11 outcomes (Figure 3),
167 including 2 socio-demographic, 4 mental health and cognition and 5 general health
168 outcomes.

169 The strongest evidence of an association with PRS_{SCZ} was detected for playing
170 computer games (OR:0.98 [95%CI:0.976,0.989] p=1.4x10⁻⁷), 0.01% lower platelet
171 distribution width [95%CI:0.003,0.006] (p=1.6x10⁻⁷) and having 0.68 lower odds of
172 glioma cancer [95%CI:0.592,0.785] (p=1.2x10⁻⁷).

173 In addition, 1 SD increased PRS_{SCZ} was associated with 0.42 sec longer duration of
174 completing an online cognitive function test (alphanumeric path) [95%CI:0.237,0.580]
175 (p=3.0x10⁻⁶) and a 19.1 mm³ reduced grey matter volume of the left putamen
176 [95%CI:-26.9,-11.3] (p=1.5x10⁻⁶).

177

178 **Major depressive disorder**

179 PRS_{MDD} was associated with 120 outcomes (Figure 3), including 18 socio-
180 demographic, 76 mental health and 26 general health outcomes.

181 Most of the associations (63%) were related to mental health, including an association
182 of higher PRS_{MDD} with higher odds of depression, anxiety, irritability, nervousness and
183 mood swings. Strongest evidence of association with PRS_{MDD} was found for
184 1.07 higher odds of “seen a doctor for nerves, anxiety, tension or depression”
185 [95%CI:1.066,1.081] (p=2.6x10⁻⁸¹), 0.11 points higher neuroticism score
186 [95%CI:0.094,0.118] (p=2.0x10⁻¹⁶) and 1.05 higher odds of having mood swings
187 [95%CI:1.045,1.060] (p=1.8x10⁻⁴⁷).

188 Furthermore, there was strong evidence of 1 SD higher PRS_{MDD} being associated with
189 socio-demographic and lifestyle traits including 1.03 higher odds of ever smoking
190 [95%CI:1.02,1.03] (p=8.4x10⁻¹⁴) and 1.04 higher odds of cannabis use
191 [95%CI:1.03,1.06] (p=2.3x10⁻⁸).

192 Associated physical health measures included 0.97 lower odds of taking medication
193 for pain relief, constipation or heartburn [95%CI:0.967,0.981] (p=4x10⁻⁶) and 1.02 odds
194 of more frequent feelings of pain, e.g. back pain [95%CI:1.015,1031] (p=6x10⁻⁹).

195 **Bipolar disorder**

196 PRS_{BP} was associated with 57 outcomes (Figure 3), including 19 socio-demographic,
197 35 general health and 3 mental health outcomes.

198 Socio-demographic and lifestyle factors included associations of higher PRS_{BP} with
199 1.04 higher odds of having a university degree [95%CI:1.033,1.048] (p=4.5x10⁻²⁶),
200 0.02 hours/day less time spent watching television [95%CI:-0.027,-0.017]

201 (p=2.8x10⁻¹⁵) and lower average alcohol intake (OR:0.98 [95%CI:0.978,0.989]
202 p=1.5x10⁻⁷).

203 General health traits included 19 traits indicating an association of 1 SD higher PRS_{BP}
204 with 0.07kg/m² lower body weight and fat mass [95%CI:-0.090,-0.058] (p=2.0x10⁻¹⁶)
205 and 5 traits related to blood measures, such as 0.004% decreased platelet distribution
206 width [95%CI:-0.006,-0.003] (p=1.8x10⁻⁶).

207 The two traits related to mental health were risk taking (OR:1.03 [95%CI:1.018,1.035]
208 p=2.9x10⁻¹⁹) and feeling fed-up (OR:0.98 [95%CI:0.975,0.989] p=2.5x10⁻⁷).

209 **CROSS DISORDER CONSIDERATIONS**

210 The highest overlap of associated outcomes of the univariable MR-pheWAS scans
211 was seen for ADHD and BP with 11 socio-economic and lifestyle and 12 general health
212 outcomes associated with both disorders (Figure 4). However, the majority of the
213 associations are directionally opposite for ADHD and BP. For example, higher
214 PRS_{ADHD} showed evidence for associations with lower educational attainment and
215 higher BMI, whereas higher PRS_{BP} was associated with higher educational attainment
216 and lower BMI. Only a higher risk of smoking initiation (“ever smoked”) was
217 directionally consistent for both PRS_{ADHD} and PRS_{BP}.

218 Furthermore, all disorder PRSs showed some evidence for association with different
219 blood cell counts, such as a decreased leukocyte count for PRS_{ADHD} and PRS_{MDD}, or
220 a decreased eosinophil count for PRS_{ADHD} and PRS_{SCZ}.

221 There was very little overlap of highly associated outcomes between the
222 neurodevelopmental domains (ADHD, ASD and SCZ).

223

224

225 **SENSITIVITY ANALYSIS**

226 We repeated our tests of association of outcomes passing the spectral decomposition
227 threshold, additionally adjusting for additional potential confounders (assessment
228 centre, genotype chip and the first 40 principal components). Estimates were highly
229 consistent with our main results, as shown in Table S4.

230 Relaxing the p-value threshold for including SNPs in the PRS resulted in some
231 changes in the results (Figure S2). For ADHD, SCZ, MDD and BP the general trend
232 was an inflation of p-values, with more associations below the significance threshold
233 (Table S5) and higher effect estimates with smaller confidence intervals. A different
234 pattern was observed for autism spectrum disorder with inconsistent results for some
235 of the outcomes, as described in detail in the supplementary Text S1. Overall the
236 strength of associations obtained for blood cell count traits across disorders varied
237 between p-value thresholds, with weaker associations found for less stringent p-value
238 thresholds.

239 **DISCUSSION**

240 In this study, we conducted a MR-pheWAS to examine the relationships between
241 genetic liability for five major psychiatric disorders and 23,004 outcomes in about
242 300,000 UK Biobank participants.

243 Our results build on a large body of literature supporting links between genetic risk for
244 psychiatric disorders with a wide variety of outcomes including psychological well-
245 being, lifestyle, socio-demographic factors and physical health (2, 4, 7, 10, 11). Our
246 findings also suggest that although different psychiatric disorders show strong genetic
247 overlap (7), genetic risk for distinct psychiatric disorders show differential associations
248 with lifestyle, socio-demographic factors and physical health as highlighted in Figure 5.
249 Genetic liability for ADHD and bipolar disorder showed the strongest associations with
250 lifestyle and social environmental factors as well as physical health. On the other hand,
251 genetic liability for major depression was most strongly associated with psychological
252 health and associations with lifestyle and socio-demographic factors were less robust.

253 We were able to replicate previously reported associations between genetic liability
254 for ADHD and lower educational attainment (12, 13), higher prevalence of smoking
255 (14), younger age at delivery (15) and higher body mass index (16). While the previous
256 findings for smoking and BMI were identified in young adults, our findings using an
257 adult population-based sample with a mean age of 56 years, suggest that associations
258 of childhood psychiatric disorder genetic liabilities with health and social outcomes
259 persist into later adulthood. Associations of genetic liability for ADHD in childhood
260 could represent effects of childhood ADHD or sub-threshold ADHD on long term social
261 and economic outcomes, or alternatively associations could be due to parental effects

262 (due to their shared genetic risk of ADHD) or horizontal pleiotropy (the same genetic
263 variants affecting multiple traits).

264 Many of the associations of genetic liability for MDD with increased mood swings,
265 irritability, feelings of loneliness and isolation are clinically known and have previously
266 been reported (5). Our results are also in line with a recent publication from the
267 Brainstorm consortium investigating genetic correlations among psychiatric disorders
268 with neurological and quantitative traits using LD score regression and GWAS
269 summary statistics, reporting high genetic correlations between most psychiatric
270 disorders and educational attainment and BMI (2, 7). However, we found little
271 evidence for associations of genetic liability for ADHD, ASD and schizophrenia with
272 mental health outcomes, such as depressive symptoms, neuroticism or anxiety; and
273 very few associations with cognitive or brain imaging outcomes, which might be
274 because of the UK Biobank being a selected sample with lower rates of psychiatric
275 disorders than the general population as discussed in the limitations section.

276 In addition to identifying previously reported associations of genetic liability for ASD,
277 our MR-pheWAS also revealed novel associations. We found a strong association of
278 genetic liability for ASD with decreased heel bone mineral density, which furthers
279 previous evidence from observational studies that children and adolescents with ASD
280 have lower bone mineral density (17, 18), higher frequency of bone fractures (19) and
281 lower vitamin D levels (20, 21), which is essential for bone metabolism. This might
282 suggest that these observed associations may be due to pleiotropic effects of genetic
283 variants associated with bone health.

284 Our MR-pheWAS of schizophrenia suggested that genetic liability for schizophrenia
285 was associated with lower grey matter volume in the left putamen and a lower risk for

286 glioblastoma cancer. Both phenotypes have been associated with schizophrenia in
287 observational studies but it is not clear whether these phenotypes are determinants or
288 consequences of schizophrenia, or due to confounding or shared genetic risk (22, 23).
289 For glioblastoma our finding could be attributed to common underlying mechanisms
290 that act in opposite directions, since it has been previously suggested that the same
291 biological pathways leading to schizophrenia may be protective for developing
292 glioblastoma (23). With respect to differences in brain volumes of schizophrenia
293 patients, two large studies by the international ENIGMA (24) and Japanese COCORO
294 (25) consortia found no notable difference in putamen volume between schizophrenia
295 patients and controls or associations of brain volumes with a PRS_{SCZ} in schizophrenia
296 patients (26). Our results suggest that genetic risk for schizophrenia could be
297 associated with putamen volume, but should be treated with caution because of
298 potential selection bias due to the highly selected subset of about 10,000 UK biobank
299 participants with brain scan data available at the time of data analysis (27). In line with
300 our results, other previous work in schizophrenia patients and their relatives identified
301 an association between schizophrenia and longer performance duration on the Trail
302 Making Test (28), which requires searching and connecting irregularly arranged
303 targets (digits and letters) in ascending order and is widely used to test for executive
304 function, cognitive ability and processing speed (29-34).

305 Altered blood cell counts were associated with genetic liability for all disorders. Many
306 psychiatric disorders previously have been associated with allergic or inflammatory
307 states (35-37), such as asthma (38) and atopic diseases (39 , 40) but it is unclear
308 whether high inflammatory states are on the causal pathway to disorder manifestation
309 or the result of comorbid and confounding behaviours associated with the disease,
310 such as restricted diet, overweight, risky behaviours or medication. Our results support

311 the possibility that altered blood cell counts could be a consequence of the disorder,
312 but we cannot rule out contributions of horizontal pleiotropic effects. Also, considering
313 the inconsistent findings from the sensitivity analyses for blood count traits, results
314 need further validation and should be treated with caution.

315 **Limitations**

316 Patients with psychiatric disorders or high genetic liability for psychiatric disorders are
317 known to be less likely to participate in studies in the first place and more likely to drop-
318 out during an ongoing study (41). Selection bias into a study as well as attrition can
319 induce collider bias (42). There is consensus that the UK Biobank sample is not
320 representative of the UK population, with participants showing, for example, lower
321 prevalence of current smoking and lower rates of mortality (9). If both having a
322 psychiatric disorder and a specific outcome (e.g. high socio-economic position) are
323 associated with participation (the collider), this can induce an association between
324 genetic risk for psychiatric disorders and the outcome, called collider bias. To reduce
325 the possibility of collider bias we limited the set of included confounders in our main
326 analysis but adjusted for assessment centre and genotype batch in sensitivity analysis.

327 A direct comparison of PHEANT estimates across the psychiatric disorders cannot
328 be done without taking the differentially powered GWASs and derived PRS into
329 account. This can also affect the number and set of outcomes associated with each
330 disorder, which allows only a relative comparison among the outcomes. Further, the
331 MDD GWAS used in the current study to calculate genetic risk scores included thirty
332 thousand participants from UK Biobank (about 10% of the GWAS sample) which might
333 have inflated our results for depression related items but is not expected to introduce
334 bias in any other traits, such as blood counts.

335 Although genetic risk scores were derived using variants associated at genome-wide
336 significance level, they can still have horizontal pleiotropic effects on different
337 disorders and traits. Hence, our reported associations cannot on its own inform about
338 causality but should be followed up with other causally informative methods. We
339 therefore encourage triangulation of results using other study designs (43, 44), such
340 as two-sample MR, negative control or twin studies.

341 **Conclusion**

342 We were able to show that genetic liability for five common psychiatric disorders are
343 associated with distinct aspects of adult life, including socio-demographic factors,
344 mental and physical health. This is evident even in individuals from the general
345 population who do not necessarily present with a psychiatric disorder diagnosis.
346 However, there was surprisingly little overlap of findings for the different psychiatric
347 disorder genetic risk scores despite the high genetic and symptomatic overlap of
348 psychiatric disorders, such as schizophrenia and bipolar disorder. Furthermore, we
349 want to emphasize the benefit of using genetic instruments for hypothesis-generating
350 efforts in the field of psychiatry.

351 **METHODS**

352 **Study population**

353 Between 2006-2010 UK Biobank recruited 503,325 men and women in the UK at ages
354 40-69 years. The cohort contains a large dataset including physical measurements,
355 blood/urine/saliva samples, health and lifestyle questionnaires as well as genotype
356 (<https://www.ukbiobank.ac.uk/>).

357 For 463,010 participants genotyping was performed using the Affymetrix UK BiLEVE
358 Axiom array or Affymetrix UK Biobank Axiom® array. Participants with non-white
359 British ancestry (n=54,757) and 73,277 who have a kinship coefficient denoting a third-
360 degree relatedness were removed from an already quality checked dataset (excluding
361 participants with withdrawn consent, sex mismatch or sex aneuploidy) (45), resulting
362 in a dataset containing 334,976 participants (Figure 6).

363 UK Biobank received ethical approval from the research ethics committee (reference
364 13/NW/0382). All participants provided informed consent to participate. This work was
365 done under application number 16729 (using genetic data version 2 [500K with HRC
366 imputation] and phenotype dataset 21753).

367 **Polygenic risk scores**

368 Genetic variants were identified from the most recent GWAS summary statistics listed
369 in Table 1 with $p < 5 \times 10^{-8}$ for ADHD, ASD, SCZ, MDD and BP. This stringent p -value
370 cut-off was chosen to minimize bias introduced by horizontal pleiotropic effects of
371 genetic variants. All summary statistics were subject to standard quality control
372 including filtering for minor allele frequency ($MAF > 0.1$) and imputation quality
373 ($INFO > 0.8$) and excluding the MHC region on chromosome 6 (26-33Mb) due to its

374 complex linkage disequilibrium structure. Polygenic risk scores (PRS) were derived
375 using PRSice v2.13 by identifying independent risk alleles in approximate linkage
376 disequilibrium ($R^2 < 0.1$ within 500kb distance) and computing a weighted, standardized
377 mean score from these, as has been described previously (46).

378

379 **Table 1.** Details of GWAS used for calculating PRS

disorder	cases	controls	SNPs in PRS ¹	Source
ADHD	20,183	35,191	10	Demontis et al. (2019) (2)
ASD	18,381	27,969	2	Grove et al. (2017) (4)
Schizophrenia	36,989	113,075	113	Ripke et al. (2014) (3)
MDD	135,458	344,901	34	Wray et al. (2018) (5)
Bipolar disorder	20,129	21,524	8	Ruderfer et al. (2018) (47)

380 1- PRS derived from genome-wide significant hits ($p < 5 \times 10^{-8}$)

381 2- SNP heritability estimates reported in the corresponding discovery sample

382 ADHD – Attention deficit/hyperactivity disorder, MDD – Major depression, ASD – Autism spectrum disorder

383 **Outcomes**

384 UK Biobank provides a fully searchable data showcase
385 (<http://biobank.ctsu.ox.ac.uk/crystal/>) which at the time of data download (March
386 2018) included 23,004 outcomes (see supplementary Text S2), including lifestyle and
387 environment, socio-demographic, early life factors, anthropometry, family history and
388 depression outcomes.

389 Age, sex and the first 10 principal components derived from the genetic data were
390 included as covariates in all regression models. Age was derived from the participants
391 date of birth and the date of their first assessment centre visit. Sex was self-reported
392 and validated using genetic data.

393 **PHESANT MR-pheWAS**

394 PHESANT package (version 0.17) was used to test the association of each PRS with
395 each outcome variable in Biobank. A detailed description of PHESANT's automated
396 rule-based method is given elsewhere (48, 49). In brief, decision rules are based on
397 the variable field type and categorize each variable as one of four data types:
398 continuous, ordered categorical, unordered categorical or binary. PHESANT then
399 estimates the univariate association of the PRS (independent) with each outcome

400 variable (dependent) in a regression model, respectively. Normality of continuous data
401 is ensured by an inverse normal rank transformation prior to testing. All estimates
402 correspond to 1 SD change of the PRS. Selected continuous outcomes were followed
403 up to compute meaningful estimates on the original phenotype scale for better
404 interpretation.

405 PHESENT assigns each UK Biobank outcome to one of 91 level 3 categories based
406 on the 235 origin categories of the UK Biobank catalogue (a full list of categories is
407 provided in Table S1). Furthermore, three authors (BL, EW, ES) grouped these 91
408 categories into four prespecified higher level categories in order to aid result
409 presentation: socio-demographics and lifestyle, brain and cognition, mental health and
410 general health (Figure 1B).

411 To account for multiple testing (n=23,004 tests) we used a previously derived threshold
412 (49, 50) based on an estimate of the number of independent phenotypes calculated
413 using spectral decomposition (phenoSPD) (n=19,645). The multiple testing adjusted
414 significance threshold was $p < 2.55 \times 10^{-6}$ ($0.05/19,645$). The amount of inflation of
415 observed versus expected p-values is given as the ratio of the median chi-squared
416 statistics for observed to expected median p-values, referred to as Lambda (λ). A
417 conservative Bonferroni correction of multiple testing that assumes uncorrelated traits,
418 would yield a similar p-value threshold of $p < 2.30 \times 10^{-6}$ ($0.05/23,004$).

419 **PHESENT sensitivity analysis**

420 Analyses were re-run to assess residual confounding of assessment centre and
421 genetic batch, including them as well as all 40 principal components as additional
422 covariates for outcomes identified as strongly associated with either one of the
423 disorders PRS. These covariates were not included in the first model because this

424 could introduce collider bias if, for example, location of assessment centre is affected
425 by both genetic predisposition and outcomes, as discussed in the limitations section.

426 Furthermore, PRS were derived using various *p*-value thresholds ($p < 0.01$, $p < 0.1 \times 10^{-3}$,
427 $p < 1 \times 10^{-4}$, $p < 1 \times 10^{-5}$, $p < 1 \times 10^{-6}$) with consequently increasing numbers of SNPs (Table
428 S6) and the five MR-pheWAS were re-run with the more relaxed PRS to capture a
429 larger amount of explained variation in the disorders by accepting an increase in
430 horizontal pleiotropic effects. For MDD GWAS results were available for only 10,000
431 SNPs at these additional thresholds due to availability restrictions.

432 All analyses were performed in R version 3.2.4 ATLAS and R version 3.3.1, and the
433 code is available at [<https://github.com/MRCIEU/Psychiatric-disorder-pheWAS-UKBB>]. Git tag v0.1 corresponds to the version presented here.

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442 use of GWAS summary statistics.

443 **AUTHOR CONTRIBUTIONS**

444 BL Data curation, Formal analysis, Investigation, Methodology, Software,
445 Visualization, Writing – original draft, Writing – review & editing
446 LACM Methodology, Software, Writing – review & editing
447 LR Writing – review & editing
448 GDS Methodology, Writing – review & editing
449 AT Conceptualization, Investigation, Funding acquisition, Writing–review &
450 editing
451 KT Conceptualization, Methodology, Writing–review & editing
452 EW Conceptualization, Investigation, Methodology, Writing–review & editing
453 ES Conceptualization, Investigation, Methodology, Writing–review & editing

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460 **CONFLICT OF INTEREST**

461 The authors declare no conflict of interest.

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597

598 **Figures**

599

600 **Figure 1.** Study overview. (A) Descriptive sample overview of selected outcomes in
601 UK Biobank. (B) Categories of UK Biobank with the size of pie chart sections indicating
602 the number of included outcomes: socio-demographics (n=2,057), general health
603 (n=19,740), mental health (n=233), brain and cognition (n=974).

604 **Figure 2.** Overview of the distribution of disorder specific polygenic risk scores
605 ($p<5\times 10^{-8}$) associated outcomes per category of the UK Biobank variables catalogue.
606 Shown are the number of associations with polygenic risk scores for attention
607 deficit/hyperactivity disorder (ADHD), autism spectrum disorder (ASD), schizophrenia
608 (SCZ), major depression (MDD) and bipolar disorder (BP).

609 **Figure 3.** MR-PheWAS results for attention deficit/hyperactivity disorder (ADHD),
610 autism spectrum disorder (ASD), schizophrenia (SCZ), major depressive disorder
611 (MDD) and bipolar disorder (BP). Left hand panel: QQ plots of expected versus
612 observed p -values for association of PRS with all outcomes in UK Biobank. Red line
613 indicates the significance threshold (2.5×10^{-6}). Lambda (λ) indicates the degree of
614 inflation from the expected fit. Right hand panel: selected results from different
615 categories with p -values below the significance threshold and estimates generated by
616 PHESENT. Results for continuous outcomes (std. β) are the standard deviation
617 change of inverse-rank normal transformed outcome per 1 SD higher PRS.

618 **Figure 4.** Cross-disorder comparison. Shown are standardized log odds (upper
619 section in each panel) or standardized beta-values (lower section of each panel) of all
620 outcomes associated with polygenic risk scores for either attention deficit/hyperactivity
621 disorder (ADHD), autism spectrum disorder (ASD), schizophrenia (SCZ), major
622 depressive disorder (MDD) or bipolar disorder (BP) at $p<2.55\times 10^{-6}$ as indicated by

623 stars (*). For outcomes categorized as ordered-logistic, only one outcome is displayed.
624 Only associations with anthropometric measures of the right side of the body are
625 shown. Estimates were generated by PHEANT. Results for continuous outcomes
626 (std. beta) are the standard deviation change of inverse-rank normal transformed
627 outcome per 1 SD higher PRS.

628 **Figure 5.** Categories of highly associated outcomes with polygenic risk scores for
629 attention deficit/hyperactivity disorder (ADHD), autism spectrum disorder (ASD),
630 schizophrenia (SCZ), major depressive disorder (MDD) and bipolar disorder (BP). Size
631 of categories depends on the relative number of associated outcomes to the total
632 number of outcomes within each category. Only categories with more than 1 variables
633 are shown. Lifestyle and socio-demographic factors are shown in orange, physical
634 health measures are shown in green and mental health, brain and cognition traits are
635 shown in violet. Grey categories had zero hits for the corresponding disorder.

636 **Figure 6.** Overview of study sample derivation. Participants with withdrawn consent,
637 sex mismatch or sex aneuploidy were already removed from the dataset in standard
638 QC steps. (45)

639

640 **SUPPORTING INFORMATION**

641

642 **Figure S1.** Overview of the distribution of disorder specific polygenic risk scores
643 ($p < 5 \times 10^{-8}$) associated outcomes per category of the UK Biobank variables catalogue
644 after FDR adjustment for multiple testing. Shown are the number of associations with
645 polygenic risk scores for attention deficit/hyperactivity disorder (ADHD), autism
646 spectrum disorder (ASD), bipolar disorder (BP), major depressive disorder (MDD) and
647 schizophrenia (SCZ).

648 **Figure S2.** Top associated outcomes with PRS for attention deficit/hyperactivity
649 disorder (ADHD), autism spectrum disorder (ASD), schizophrenia (SCZ), bipolar
650 disorder (BP) and major depressive disorder (MDD) across different p-value
651 thresholds for SNP inclusion ($5 \times 10^{-8} - 1 \times 10^{-2}$).

652 **Table S1.** Overview of UK Biobank categories with total number of outcomes per
653 category and number of associated outcomes with polygenic risk scores passing the
654 significance threshold ($p < 2.55 \times 10^{-6}$). (ADHD- attention deficit/ hyperactivity disorder,
655 ASD- autism spectrum disorder, SCZ- schizophrenia, MDD- major depressive
656 disorder, BP- bipolar disorder)

657 **Table S2.** Correlation matrix of polygenic risk scores ($p < 5 \times 10^{-8}$). Correlation
658 coefficients are displayed on the left side, p-values on the right side of the table.
659 (ADHD- attention deficit/ hyperactivity disorder, ASD- autism spectrum disorder, SCZ-
660 schizophrenia, MDD- major depressive disorder, BP- bipolar disorder)

661 **Table S3.** MR-PheWAS results for association of generic risk of 5 common psychiatric
662 disorders with 23,004 outcomes in UK Biobank. Genetic risk is defined as weighted
663 sum of all genome-wide significant risk alleles for each disorder in 334,976 participants

664 of the UK Biobank. Estimates were generated by PHE-SANT. Results for continuous
665 outcomes are the standard deviation change of inverse-rank normal transformed
666 outcome per 1 SD higher PRS.

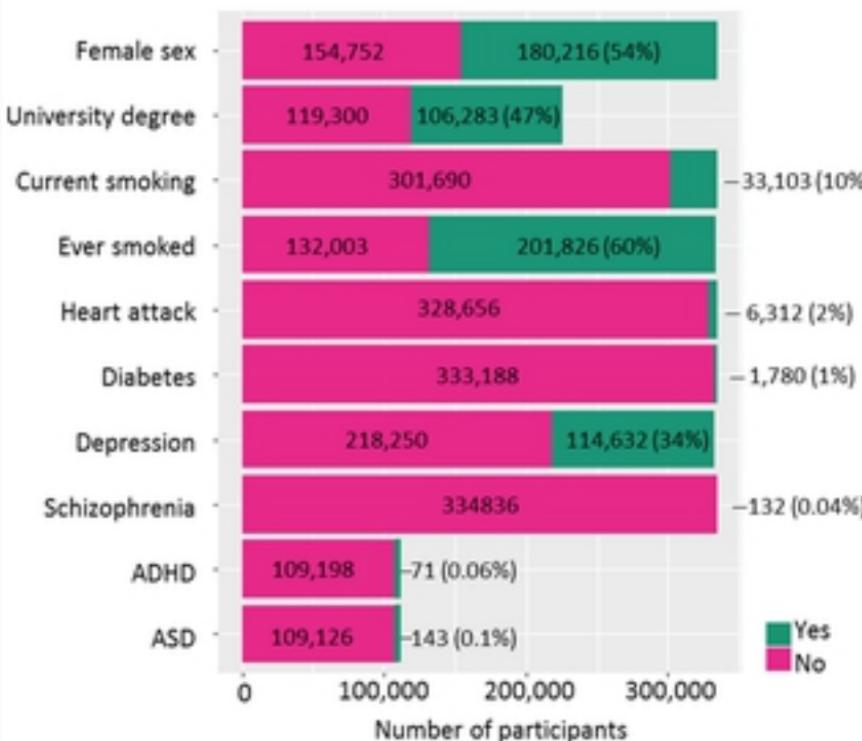
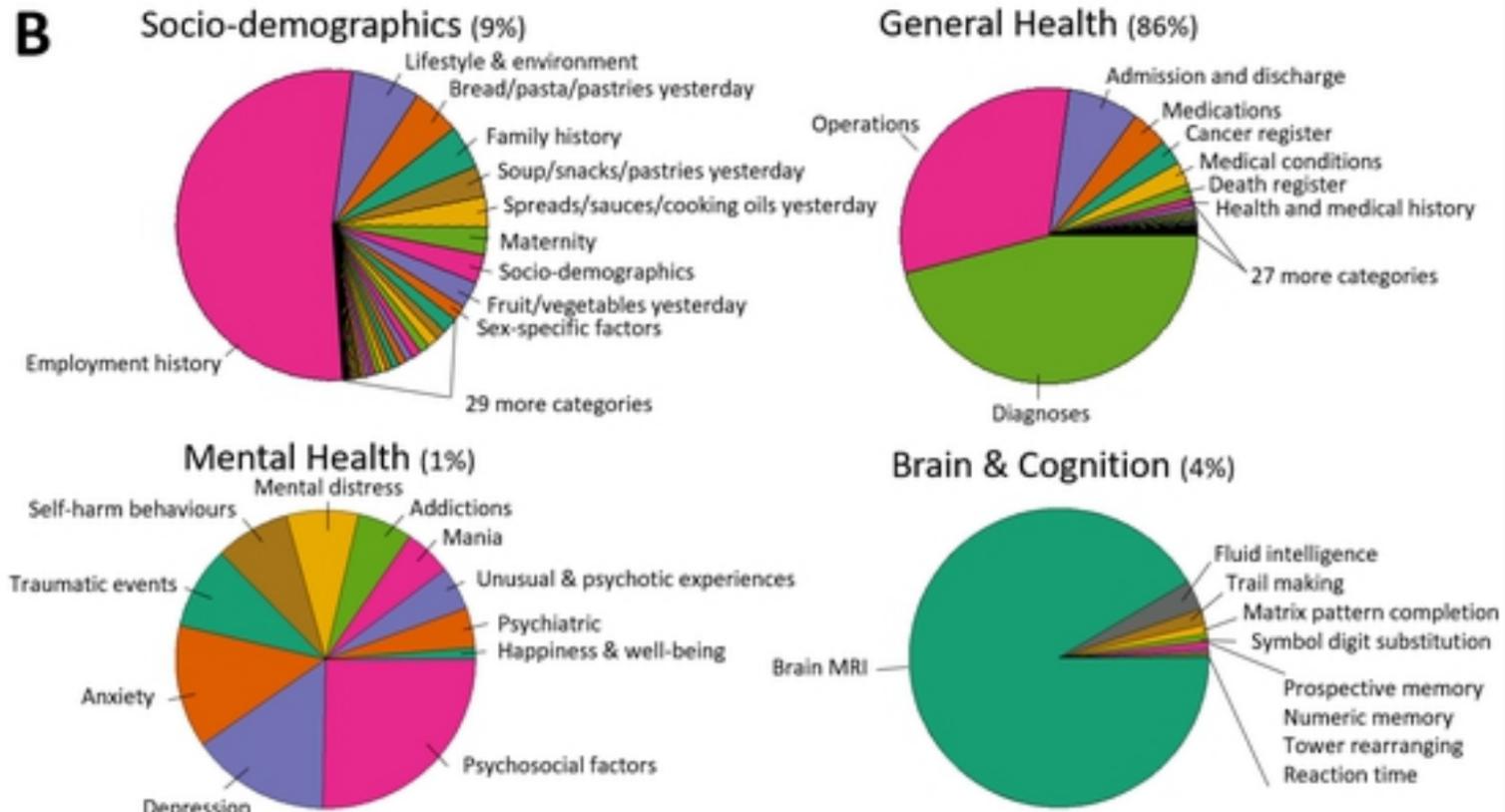
667 **Table S4.** MR-PheWAS follow-up and sensitivity results for selected continuous
668 outcomes. Genetic risk is defined as weighted sum of all genome-wide significant risk
669 alleles for each disorder in 334,976 participants of the UK Biobank. Estimates were
670 generated by linear regression on the original variable scale per 1 SD higher PRS.

671 **Table S5.** Number of strongly associated traits with PRS for attention-
672 deficit/hyperactivity disorder (ADHD), autism spectrum disorder
673 (ASD), schizophrenia (SCZ), major depressive disorder (MDD) and bipolar disorder
674 (BP) at different p-value thresholds for PRS calculation.

675 **Table S6.** Number of SNPs included in polygenic risk scores for attention-
676 deficit/hyperactivity disorder (ADHD), autism spectrum disorder
677 (ASD), schizophrenia (SCZ), major depressive disorder (MDD) and bipolar disorder
678 (BP) at different p-value thresholds.

679 **Text S1.** Sensitivity analysis for autism spectrum disorder

680 **Text S2.** UK Biobank outcomes description

A**B****Figure 1**

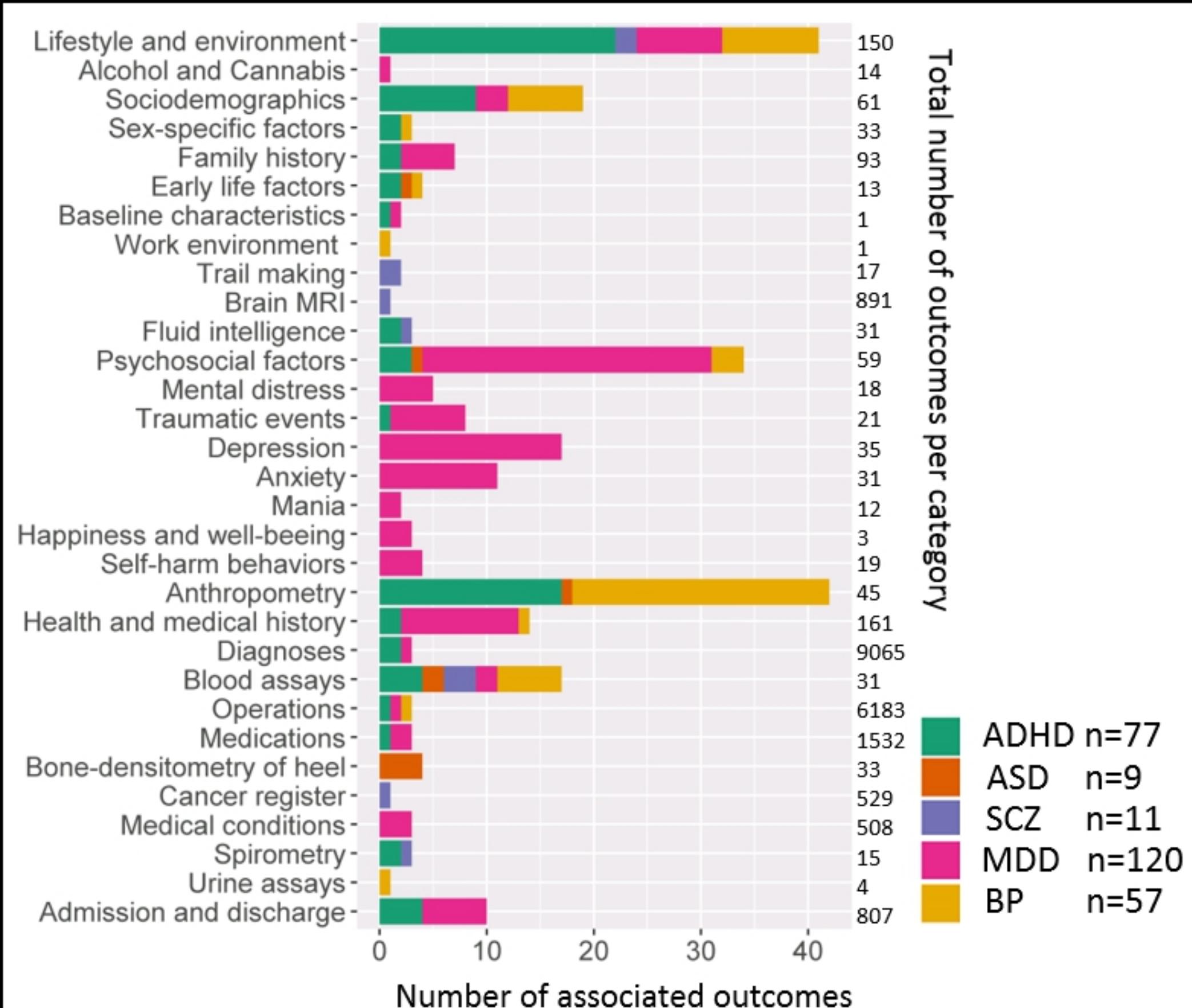


Figure 2

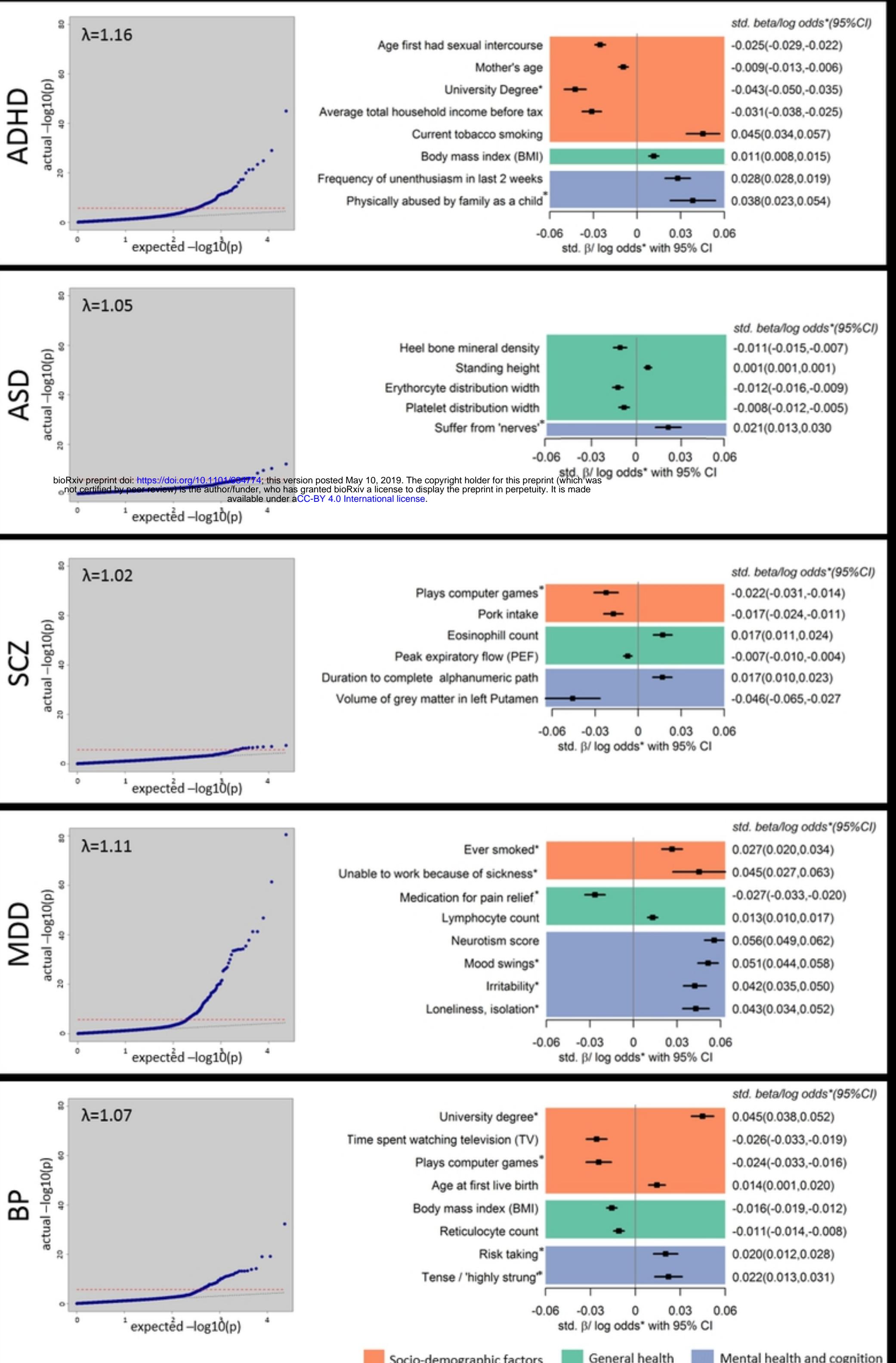
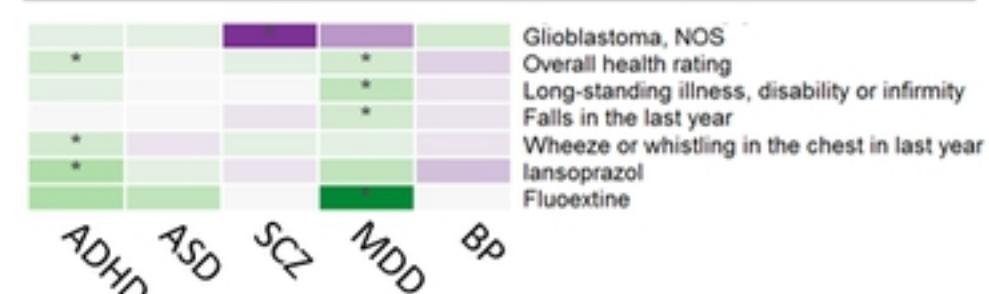
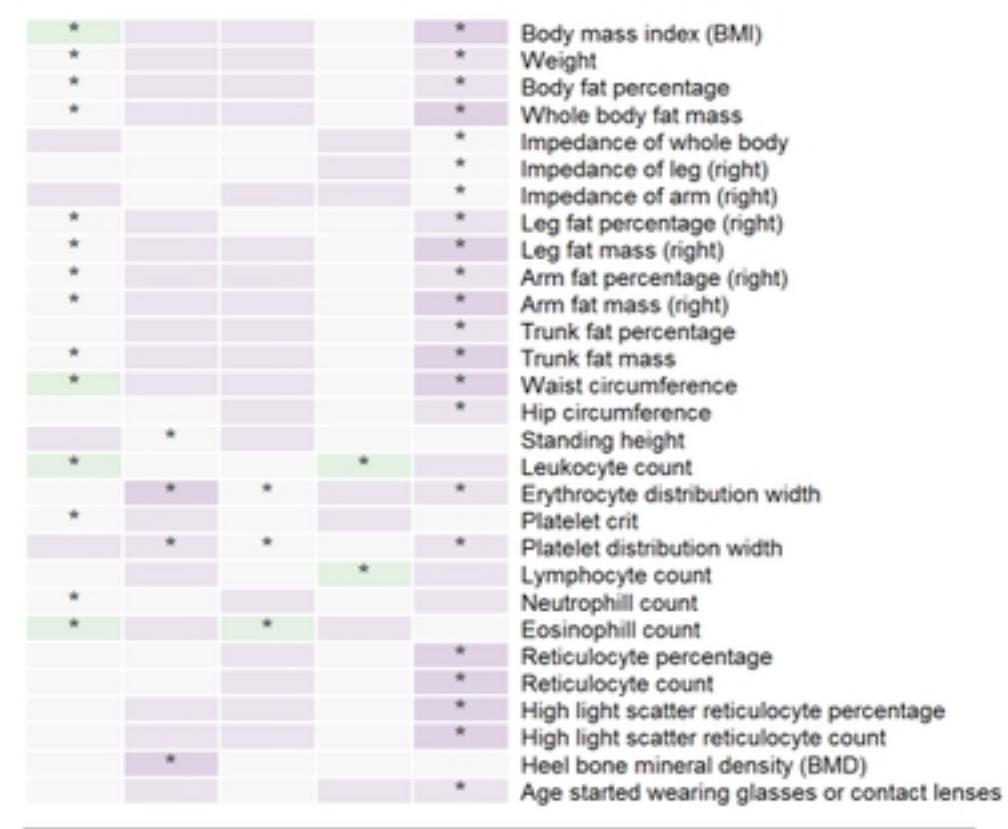


Figure 3

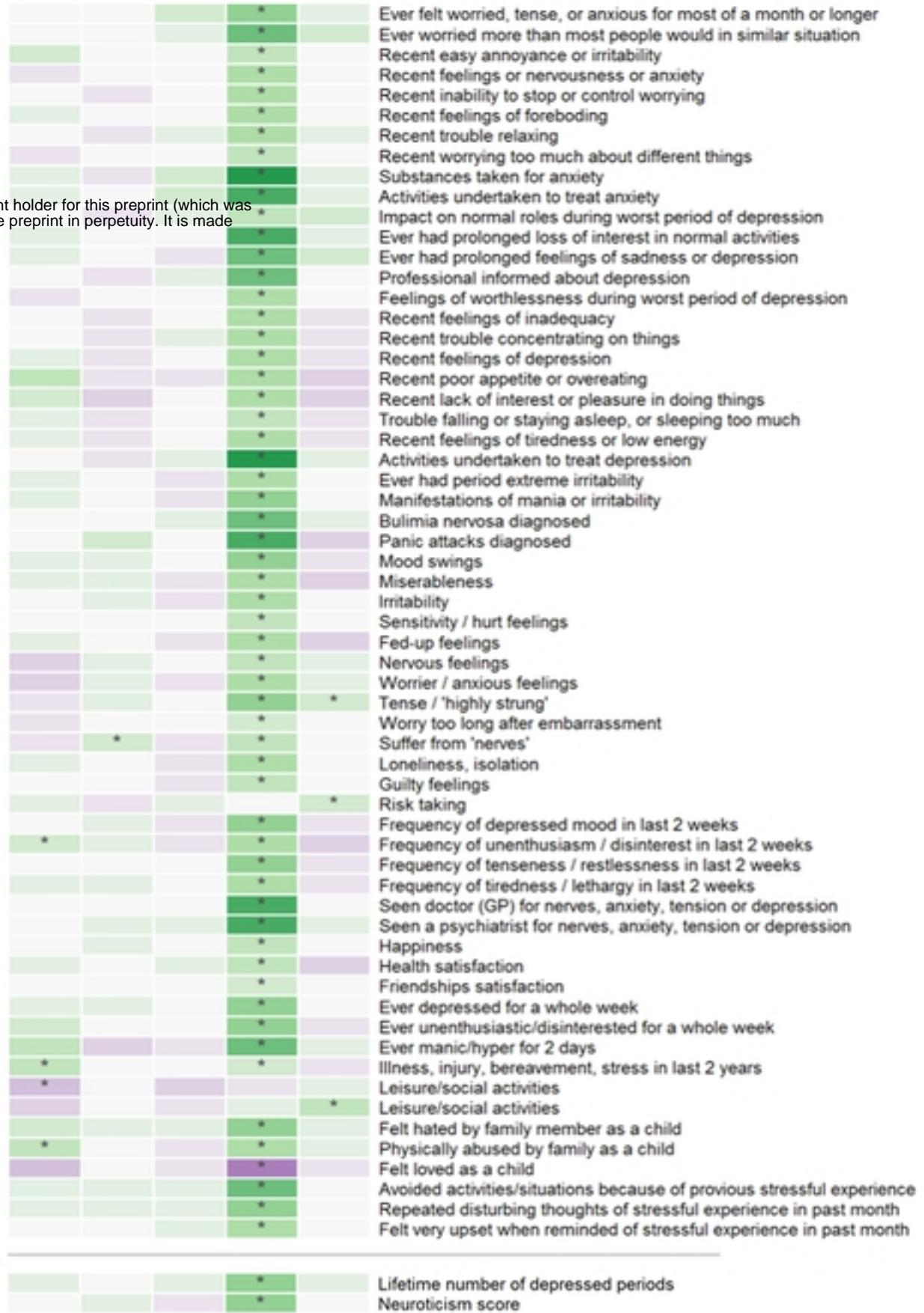
Socio-demographics



General health



Mental health



Brain & cognition

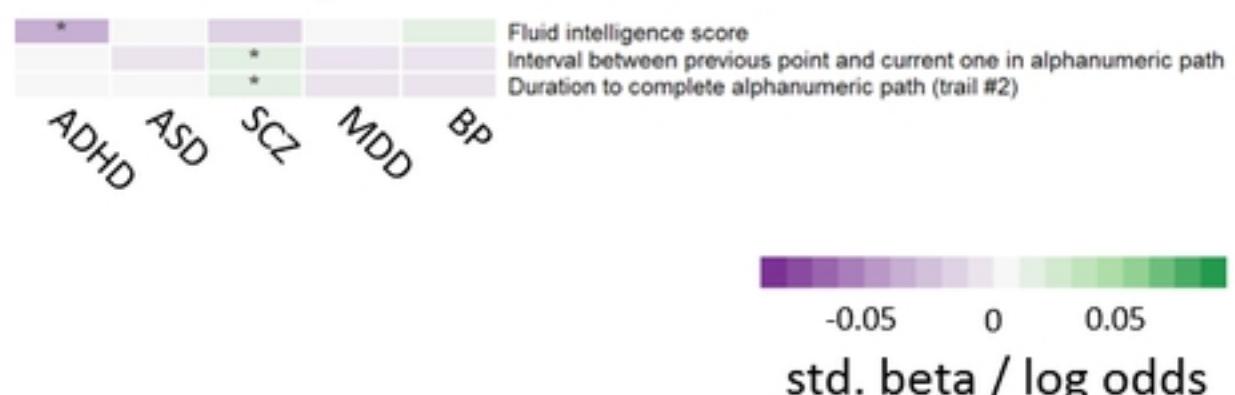


Figure 4

ADHD

ASD

SCZ

Anthropometry

Early-life factors

- Health & medical history
 - Family history
 - Traumatic events
 - Medications
- Fluid Intelligence
 - Memory
 - Attention
- Sex-specific factors
 - Androgen levels
 - Estrogen levels
- MDD
 - Depression
 - Anxiety
 - Sleep
 - Cognition

MDD

The diagram illustrates the complex relationships between mental health outcomes and various social, environmental, and individual factors. At the top, 'MDD' is shown in large, bold, black letters. Below it, 'Depression' is written in large blue letters. To the left of 'Depression', 'Happiness & well-being' is written in large blue letters, with 'Psychosocial factors' and 'Traumatic events' listed below it. A large blue box labeled 'Mental distress' is positioned at the bottom right. Various risk and protective factors are arranged around these central concepts, including:

- Top Left (Risk Factors):** Lifestyle & environments, Health & medical history, Family history, Blood assays, Mania, Alcohol & Cannabis, Sociodemographics (Demographics, Education), and Traumatic events.
- Top Right (Protective Factors):** Self-harm behaviors, Anxiety, Stress resilience, and Self-esteem.
- Bottom Left (Risk Factors):** Psychosocial factors, Traumatic events, Mental distress, and Mental health (Anxiety, Depression, Psychosis, Self-harm).
- Bottom Right (Protective Factors):** Self-harm behaviors, Anxiety, Stress resilience, and Self-esteem.

Bone densitometry

Early-life factors

Psychosocial factors	Medications
Admission discharge	Medications
Decision making	Medications
Fluid intake/excretion	Medications
Happiness/Well-being	Medications

Blood assays

BP

Anthropometry

```

graph TD
    A[Urine assays] --- B[Lifestyle & Environment  
Genetics]
    B --- C[Sex-specific factors]
    C --- D[Early-life factors  
Sociodemographics]
    B --- E[Medical history  
Administrative factors/charge]
    E --- F[Medications]
    C --- D
    E --- F
  
```

Blood assays Trail making

Figure 5

463,010 UK Biobank participants with
genetic data after standard QC

54,757 removed due to non-white
British ancestry

73,277 removed due to relatedness

334,976 participants