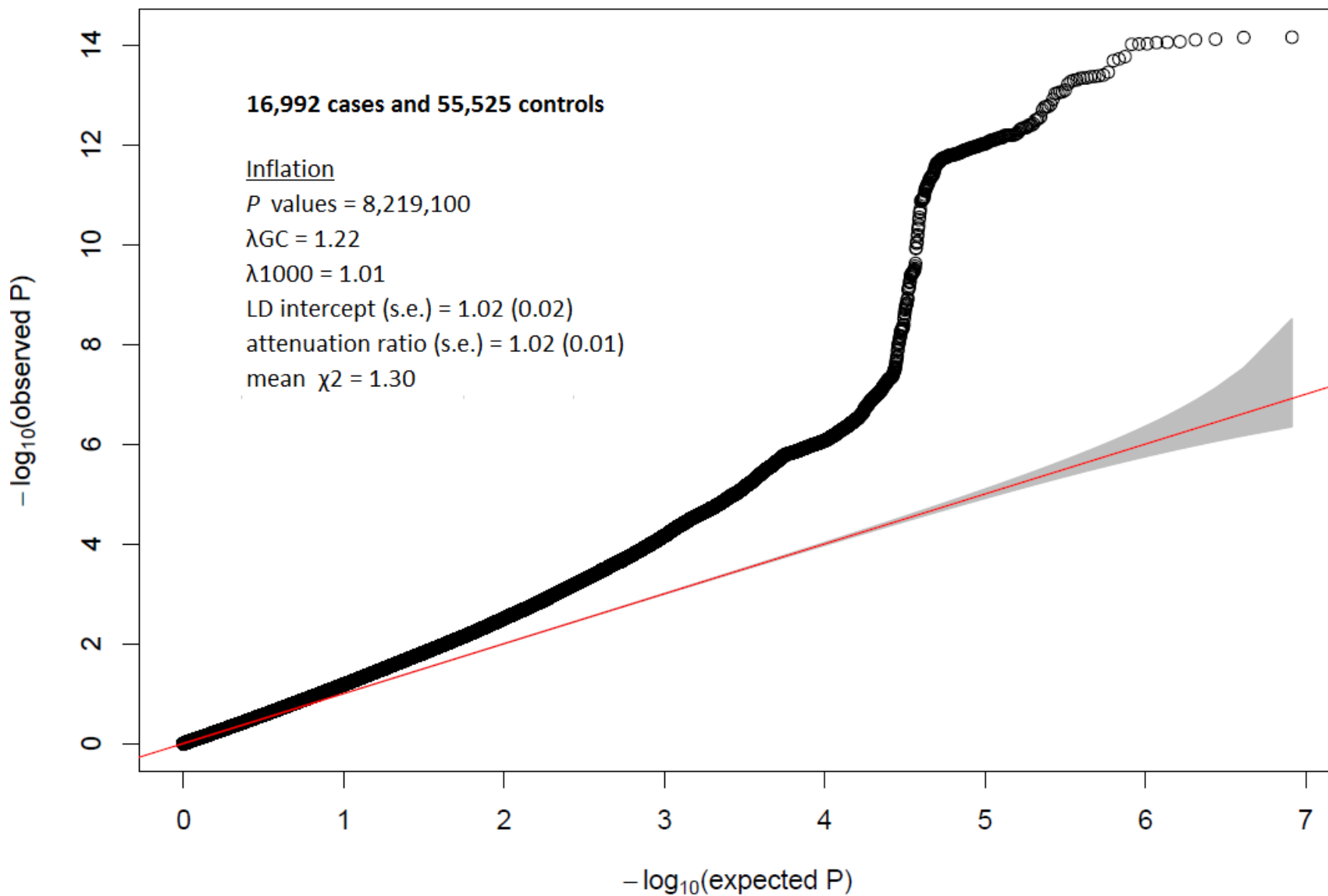


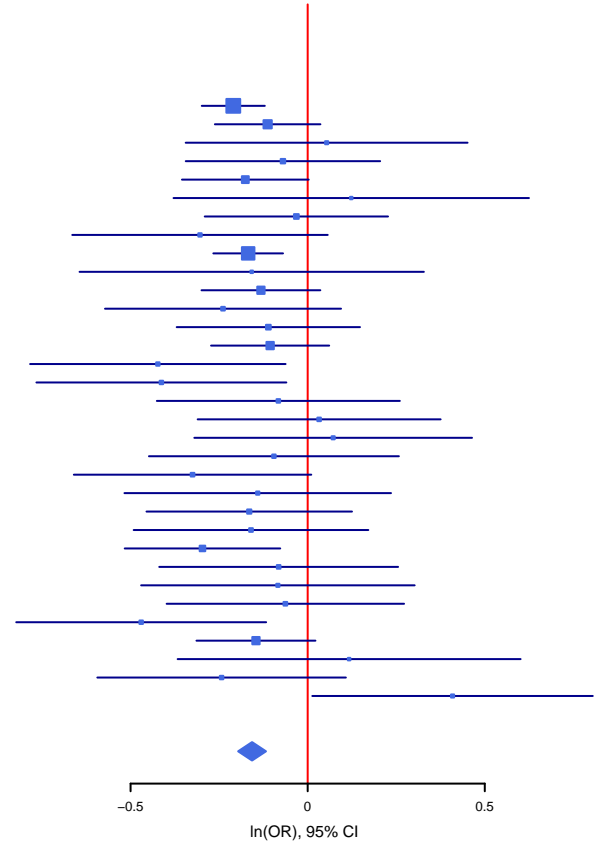
Supplementary Figure 1. Quantile-quantile (QQ) plot of association P values for the meta-analysis of anorexia nervosa (MAF ≥ 0.01 and INFO score ≥ 0.7). The red diagonal line shows the theoretical null distribution. The shading shows the 95% confidence interval bounds. MAF = minor allele frequency and INFO = imputation quality score.



Supplementary Figure 2a-h legend. Forest plots of GWAS meta-analysis results for all genome-wide significant SNPs ($P < 5 \times 10^{-8}$). The overall sample size is 16,992 cases and 55,525 controls. For specific cohort sample sizes, see Supplementary Table 1. Two SNPs were located in regions that overlapped (index SNPs of rs9821797 and rs73088112), therefore we refer to only one of these (rs9821797) in the main manuscript. The red vertical line is the reference line of no effect. The center values are $\ln(\text{OR})$ estimates from logistic regression association analyses and the error bars show the 95% confidence interval.

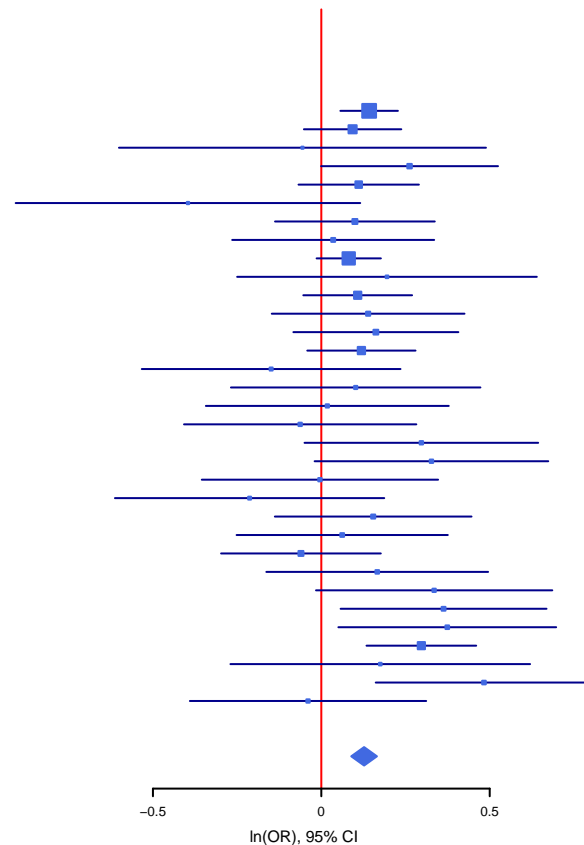
(a) chr3:rs9821797.

rs9821797		T/A		3:48718253		het_P: 0.613		het_I: 0.0	
+++++-----									
	ngt	info	p_value	f_ca(n)	f_co(n)	ln(OR)	STDerr		
aunz	0	0.99	3.79e-06	0.865(2536)	0.888(15624)	-0.21	0.0454		
chop	0	0.99	0.137	0.859(1032)	0.876(3568)	-0.113	0.076		
fin1	0	1.00	0.792	0.857(130)	0.857(503)	0.0535	0.203		
fre1	0	0.93	0.617	0.840(293)	0.859(990)	-0.0698	0.14		
gns2	0	0.97	0.0537	0.849(657)	0.870(2263)	-0.176	0.0913		
itgr	0	1.06	0.631	0.873(144)	0.868(129)	0.123	0.256		
net1	0	0.95	0.809	0.870(348)	0.873(1337)	-0.0319	0.132		
poco	0	0.97	0.0999	0.877(239)	0.912(604)	-0.304	0.184		
sedk	0	0.96	0.000821	0.869(4105)	0.886(3793)	-0.168	0.0502		
spa1	0	1.01	0.523	0.877(184)	0.886(156)	-0.158	0.248		
ukbb	0	1.02	0.122	0.872(768)	0.887(3065)	-0.132	0.0856		
ukd1	0	0.96	0.161	0.888(242)	0.908(957)	-0.239	0.17		
usa1	0	1.00	0.4	0.872(418)	0.884(750)	-0.111	0.132		
usa2	0	1.00	0.211	0.871(1304)	0.881(1307)	-0.106	0.0851		
w10	0	0.95	0.0218	0.844(156)	0.888(714)	-0.423	0.184		
w11	0	0.96	0.0215	0.847(164)	0.891(754)	-0.413	0.18		
w12	0	1.04	0.636	0.872(176)	0.879(785)	-0.0828	0.175		
w13	0	1.01	0.853	0.867(181)	0.870(789)	0.0324	0.175		
w14	0	0.99	0.719	0.898(169)	0.892(761)	0.072	0.2		
w15	0	0.97	0.596	0.874(185)	0.883(818)	-0.0953	0.18		
w16	0	1.05	0.058	0.859(174)	0.896(800)	-0.325	0.171		
w17	0	0.96	0.462	0.867(158)	0.881(761)	-0.141	0.192		
w18	0	0.94	0.263	0.840(245)	0.861(852)	-0.165	0.148		
w19	0	0.88	0.344	0.856(211)	0.873(851)	-0.16	0.169		
w1to6	0	0.90	0.0083	0.841(357)	0.874(4828)	-0.297	0.112		
w20	0	0.95	0.633	0.870(200)	0.879(802)	-0.0819	0.172		
w21	0	0.90	0.669	0.873(161)	0.887(801)	-0.0841	0.197		
w22	0	0.92	0.713	0.874(211)	0.879(845)	-0.063	0.171		
w23	0	0.87	0.00904	0.839(164)	0.885(957)	-0.47	0.18		
w24	0	0.87	0.0888	0.857(1241)	0.872(1447)	-0.146	0.0857		
w7	0	0.94	0.636	0.886(97)	0.872(831)	0.117	0.247		
w8	0	0.93	0.175	0.858(162)	0.884(1062)	-0.243	0.179		
w9	0	1.01	0.0432	0.914(180)	0.876(1021)	0.409	0.202		
Meta-analysis	0	0.97	6.99e-15	0.865(16992)	0.882(55525)	-0.157	0.0201		



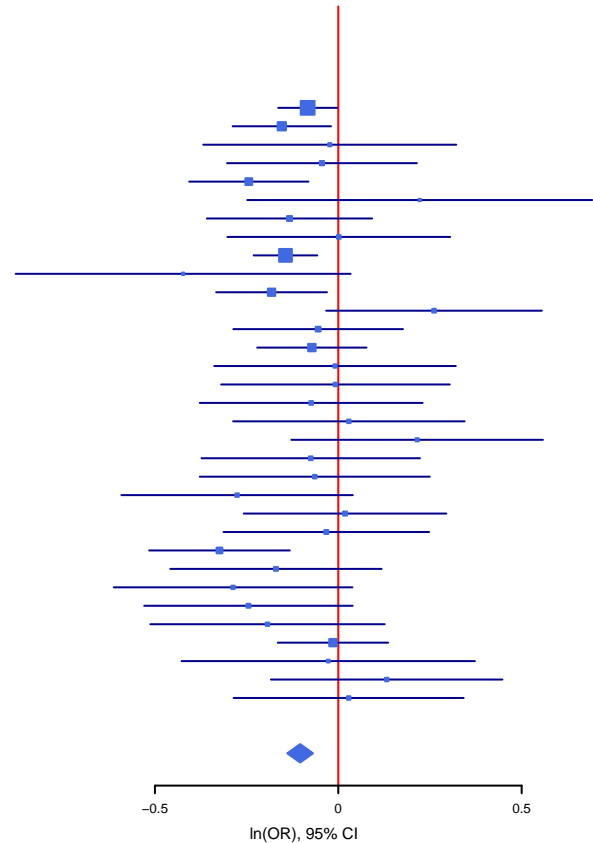
(b) chr11:6589488.

rs6589488		A/T		11:115096956			
+++++-----				het_P:	0.339	het_I:	7.9
	ngt	info	p_value	f_ca(n)	f_co(n)	ln(OR)	STDerr
aunz	0	0.96	0.00117	0.152(2536)	0.135(15624)	0.142	0.0436
chop	0	0.94	0.208	0.158(1032)	0.148(3568)	0.0929	0.0738
fin1	0	0.92	0.84	0.076(130)	0.078(503)	-0.056	0.278
fre1	0	0.94	0.0509	0.180(293)	0.145(990)	0.262	0.134
gns2	0	0.92	0.223	0.157(657)	0.146(2263)	0.111	0.0912
itgr	0	0.87	0.13	0.140(144)	0.175(129)	-0.396	0.261
net1	0	0.98	0.41	0.157(348)	0.151(1337)	0.0997	0.121
poco	0	0.92	0.818	0.179(239)	0.173(604)	0.0352	0.153
sedk	0	0.98	0.0947	0.136(4105)	0.123(3793)	0.0814	0.0487
spa1	0	0.93	0.39	0.173(184)	0.148(156)	0.195	0.227
ukbb	0	0.97	0.19	0.148(768)	0.136(3065)	0.108	0.0824
ukd1	0	0.96	0.339	0.157(242)	0.139(957)	0.139	0.146
usa1	0	0.94	0.197	0.157(418)	0.140(750)	0.162	0.125
usa2	0	0.96	0.146	0.148(1304)	0.135(1307)	0.119	0.082
w10	0	0.90	0.449	0.132(156)	0.146(714)	-0.149	0.196
w11	0	0.91	0.589	0.135(164)	0.123(754)	0.102	0.189
w12	0	0.90	0.923	0.133(176)	0.132(785)	0.0178	0.184
w13	0	0.91	0.721	0.149(181)	0.148(789)	-0.0628	0.176
w14	0	0.95	0.0925	0.150(169)	0.118(761)	0.297	0.177
w15	0	0.93	0.0641	0.147(185)	0.112(818)	0.327	0.177
w16	0	0.91	0.982	0.144(174)	0.141(800)	-0.00411	0.179
w17	0	0.90	0.297	0.114(158)	0.140(761)	-0.213	0.204
w18	0	0.96	0.302	0.150(245)	0.132(852)	0.154	0.149
w19	0	0.94	0.698	0.150(211)	0.143(851)	0.062	0.16
w1to6	0	0.93	0.619	0.126(357)	0.133(4828)	-0.0605	0.121
w20	0	0.89	0.323	0.150(200)	0.133(802)	0.166	0.168
w21	0	0.92	0.0614	0.158(161)	0.120(801)	0.335	0.179
w22	0	0.96	0.0198	0.164(211)	0.121(845)	0.363	0.156
w23	0	0.95	0.0235	0.175(164)	0.131(957)	0.374	0.165
w24	0	0.92	0.000345	0.154(1241)	0.122(1447)	0.297	0.0831
w7	0	0.91	0.439	0.148(97)	0.124(831)	0.175	0.227
w8	0	0.93	0.00317	0.189(162)	0.129(1062)	0.483	0.164
w9	0	0.95	0.825	0.123(180)	0.127(1021)	-0.0395	0.179
Meta-analysis	0	0.95	6.31e-11	0.148(16992)	0.134(55525)	0.127	0.0195



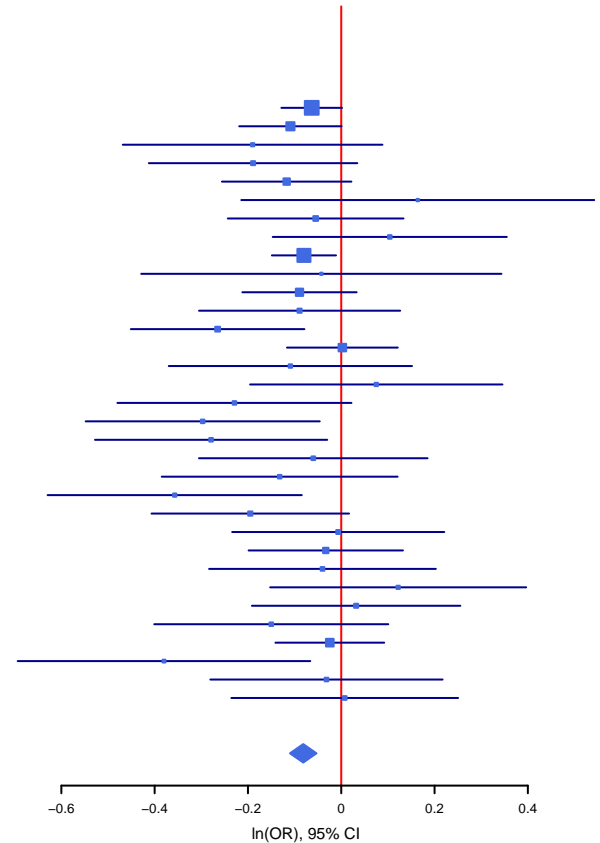
(c) chr2:2287348.

rs2287348		C/T		2:54039813		het_P:	0.309	het_I:	9.7
ngt	info	p_value	f_ca(n)	f_co(n)	ln(OR)	STDerr			
aunz	0	0.99	0.0412	0.830(2536)	0.843(15624)	-0.0836	0.041		
chop	1	1.01	0.0244	0.822(1032)	0.845(3568)	-0.154	0.0684		
fin1	1	1.01	0.894	0.796(130)	0.804(503)	-0.0233	0.176		
fre1	0	0.99	0.735	0.834(293)	0.835(990)	-0.0446	0.132		
gns2	0	0.98	0.00322	0.808(657)	0.845(2263)	-0.244	0.083		
itgr	1	0.96	0.355	0.851(144)	0.818(129)	0.222	0.24		
net1	1	0.97	0.247	0.816(348)	0.835(1337)	-0.133	0.115		
poco	0	1.04	0.993	0.849(239)	0.839(604)	0.0014	0.155		
sedk	0	0.99	0.00112	0.830(4105)	0.848(3793)	-0.144	0.0443		
spa1	0	1.02	0.0693	0.824(184)	0.884(156)	-0.423	0.233		
ukbb	1	1.00	0.0184	0.829(768)	0.853(3065)	-0.182	0.0771		
ukd1	1	1.00	0.0823	0.872(242)	0.840(957)	0.261	0.15		
usa1	1	0.99	0.64	0.830(418)	0.839(750)	-0.055	0.118		
usa2	0	1.00	0.343	0.834(1304)	0.844(1307)	-0.0721	0.076		
w10	0	1.02	0.958	0.833(156)	0.833(714)	-0.00894	0.168		
w11	0	1.01	0.961	0.821(164)	0.823(754)	-0.00773	0.159		
w12	0	1.02	0.634	0.826(176)	0.840(785)	-0.074	0.155		
w13	0	1.02	0.858	0.845(181)	0.836(789)	0.0288	0.161		
w14	0	1.05	0.219	0.873(169)	0.846(761)	0.215	0.175		
w15	0	0.98	0.622	0.819(185)	0.825(818)	-0.0752	0.152		
w16	0	1.00	0.688	0.828(174)	0.837(800)	-0.0642	0.16		
w17	0	0.99	0.0858	0.791(158)	0.838(761)	-0.276	0.161		
w18	0	0.99	0.896	0.841(245)	0.838(852)	0.0184	0.141		
w19	0	1.01	0.819	0.822(211)	0.827(851)	-0.0327	0.143		
w1to6	0	0.98	0.000925	0.792(357)	0.839(4828)	-0.324	0.0978		
w20	0	1.05	0.247	0.828(200)	0.851(802)	-0.17	0.147		
w21	0	0.99	0.0831	0.823(161)	0.856(801)	-0.287	0.166		
w22	0	1.00	0.0899	0.820(211)	0.853(845)	-0.245	0.145		
w23	0	0.95	0.237	0.821(164)	0.845(957)	-0.193	0.163		
w24	0	0.97	0.85	0.843(1241)	0.845(1447)	-0.0146	0.0768		
w7	0	1.00	0.893	0.825(97)	0.833(831)	-0.0274	0.204		
w8	0	1.04	0.413	0.845(162)	0.825(1062)	0.132	0.161		
w9	0	1.00	0.86	0.847(180)	0.844(1021)	0.0282	0.16		
Meta-analysis	7	0.99	5.62e-09	0.830(16992)	0.842(55525)	-0.104	0.0179		



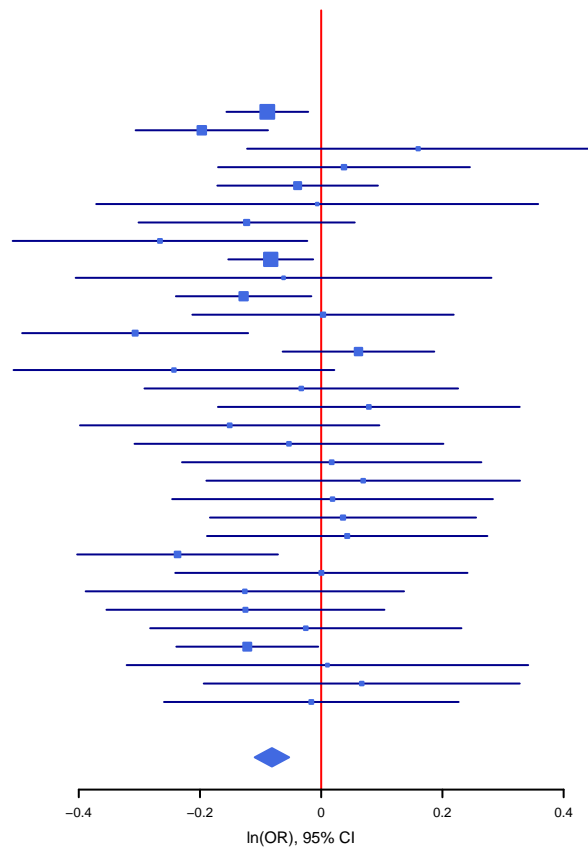
(d) chr10:rs2008387.

rs2008387		G/A		rs131448764		het_P: 0.358		het_I: 6.7		
ngt	info	p_value	f_ca(n)	f_co(n)	ln(OR)	STDerr				
aunz	0	0.93	0.0565	0.650(2536)	0.663(15624)	-0.0633	0.0332			
chop	0	0.93	0.0522	0.638(1032)	0.668(3568)	-0.109	0.056			
fin1	0	1.01	0.182	0.556(130)	0.600(503)	-0.19	0.142			
fre1	0	0.82	0.0967	0.653(293)	0.687(990)	-0.189	0.114			
gns2	0	0.89	0.0999	0.653(657)	0.677(2263)	-0.117	0.0708			
itgr	0	0.94	0.397	0.677(144)	0.642(129)	0.164	0.193			
net1	0	0.93	0.569	0.680(348)	0.698(1337)	-0.0548	0.0961			
poco	0	0.91	0.414	0.693(239)	0.661(604)	0.104	0.128			
sedk	0	0.93	0.0226	0.645(4105)	0.661(3793)	-0.08	0.0351			
spa1	0	0.77	0.829	0.677(184)	0.676(156)	-0.0427	0.197			
ukbb	0	0.93	0.153	0.652(768)	0.670(3065)	-0.0895	0.0625			
ukd1	0	0.97	0.416	0.653(242)	0.674(957)	-0.0893	0.11			
usa1	0	0.93	0.0052	0.622(418)	0.674(750)	-0.265	0.0949			
usa2	0	0.94	0.969	0.652(1304)	0.654(1307)	0.0024	0.0606			
w10	0	0.99	0.413	0.643(156)	0.664(714)	-0.109	0.133			
w11	0	0.92	0.586	0.692(164)	0.677(754)	0.0751	0.138			
w12	0	0.95	0.0747	0.645(176)	0.696(785)	-0.229	0.128			
w13	0	0.92	0.0199	0.620(181)	0.679(789)	-0.297	0.128			
w14	0	0.99	0.0276	0.617(169)	0.681(761)	-0.279	0.127			
w15	0	0.96	0.632	0.648(185)	0.665(818)	-0.06	0.125			
w16	0	0.97	0.307	0.658(174)	0.681(800)	-0.132	0.129			
w17	0	0.93	0.00996	0.635(158)	0.706(761)	-0.357	0.139			
w18	0	0.99	0.0714	0.630(245)	0.675(852)	-0.195	0.108			
w19	0	1.02	0.956	0.672(211)	0.671(851)	-0.00652	0.116			
w1to6	0	0.95	0.693	0.666(357)	0.674(4828)	-0.0332	0.0844			
w20	0	0.94	0.745	0.675(200)	0.683(802)	-0.0402	0.124			
w21	0	0.93	0.383	0.706(161)	0.676(801)	0.122	0.14			
w22	0	1.02	0.779	0.660(211)	0.653(845)	0.0319	0.114			
w23	0	1.00	0.241	0.656(164)	0.685(957)	-0.15	0.128			
w24	0	0.95	0.681	0.663(1241)	0.668(1447)	-0.0245	0.0595			
w7	0	0.99	0.0175	0.598(97)	0.678(831)	-0.38	0.16			
w8	0	1.02	0.803	0.682(162)	0.688(1062)	-0.0317	0.127			
w9	0	0.98	0.952	0.667(180)	0.661(1021)	0.00747	0.124			
Meta-analysis	0	0.94	1.72e-08	0.651(16992)	0.670(55525)	-0.0815	0.0145			



(e) chr3:rs9874207.

rs9874207		T/C		3:71019750		het_P:	0.212	het_I:	16.0
ngt	info	p_value	f_ca(n)	f_co(n)	ln(OR)	STDerr			
aunz	0	0.79	0.00971	0.483(2536)	0.500(15624)	-0.0889	0.0344		
chop	0	0.87	0.0004	0.472(1032)	0.512(3568)	-0.197	0.0557		
fin1	0	0.97	0.265	0.554(130)	0.522(503)	0.16	0.144		
fre1	0	0.84	0.723	0.488(293)	0.485(990)	0.0376	0.106		
gns2	0	0.88	0.564	0.486(657)	0.492(2263)	-0.0389	0.0676		
itgr	0	0.89	0.972	0.469(144)	0.466(129)	-0.00662	0.186		
net1	0	0.91	0.178	0.472(348)	0.500(1337)	-0.123	0.091		
poco	0	0.86	0.0313	0.452(239)	0.497(604)	-0.266	0.124		
sedk	0	0.82	0.0197	0.512(4105)	0.532(3793)	-0.0833	0.0357		
spa1	0	0.85	0.723	0.470(184)	0.493(156)	-0.0621	0.175		
ukbb	1	1.01	0.0246	0.467(768)	0.499(3065)	-0.128	0.057		
ukd1	0	0.87	0.979	0.485(242)	0.486(957)	0.003	0.11		
usa1	0	0.86	0.00122	0.450(418)	0.514(750)	-0.307	0.0951		
usa2	0	0.77	0.335	0.488(1304)	0.477(1307)	0.0615	0.0638		
w10	0	0.87	0.0727	0.460(156)	0.517(714)	-0.243	0.135		
w11	0	0.86	0.804	0.485(164)	0.494(754)	-0.0329	0.132		
w12	0	0.88	0.535	0.506(176)	0.491(785)	0.0786	0.127		
w13	0	0.87	0.233	0.483(181)	0.520(789)	-0.151	0.126		
w14	0	0.86	0.683	0.500(169)	0.512(761)	-0.0532	0.13		
w15	0	0.87	0.891	0.519(185)	0.518(818)	0.0173	0.126		
w16	0	0.83	0.599	0.542(174)	0.527(800)	0.0693	0.132		
w17	0	0.88	0.889	0.528(158)	0.524(761)	0.0187	0.135		
w18	0	0.86	0.748	0.522(245)	0.517(852)	0.036	0.112		
w19	0	0.86	0.716	0.525(211)	0.514(851)	0.0429	0.118		
w1to6	0	0.85	0.00504	0.460(357)	0.511(4828)	-0.237	0.0845		
w20	0	0.83	0.998	0.538(200)	0.536(802)	0.0003	0.123		
w21	0	0.85	0.346	0.500(161)	0.524(801)	-0.126	0.134		
w22	0	0.88	0.285	0.489(211)	0.515(845)	-0.125	0.117		
w23	0	0.85	0.846	0.513(164)	0.517(957)	-0.0254	0.131		
w24	0	0.85	0.0416	0.495(1241)	0.520(1447)	-0.122	0.0597		
w7	0	0.84	0.951	0.512(97)	0.507(831)	0.0103	0.169		
w8	0	0.81	0.616	0.520(162)	0.510(1062)	0.0669	0.133		
w9	0	0.86	0.895	0.512(180)	0.517(1021)	-0.0163	0.124		
Meta-analysis	1	0.84	2.05e-08	0.493(16992)	0.507(55525)	-0.0813	0.0145		



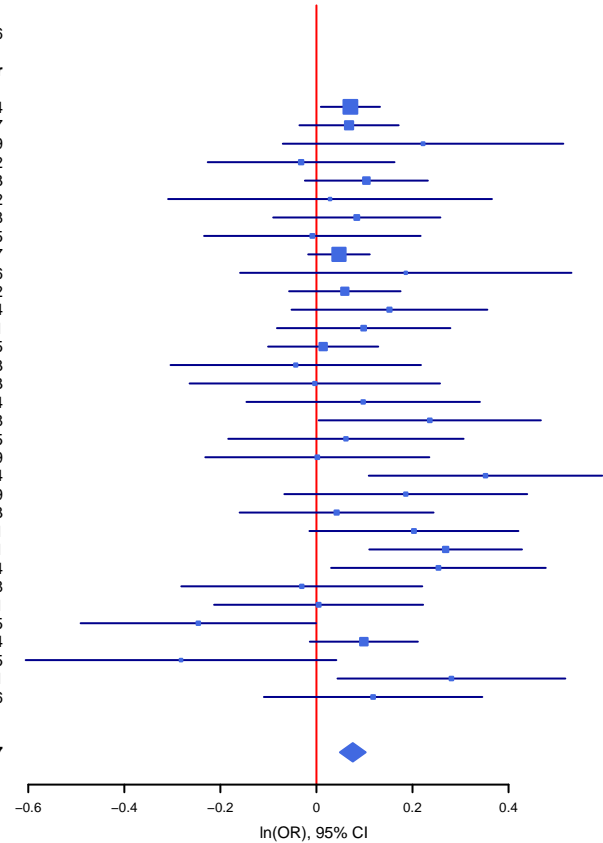
(f) chr1:rs10747478.

rs10747478 T/G
 ++++++-----+++++

1:96901455

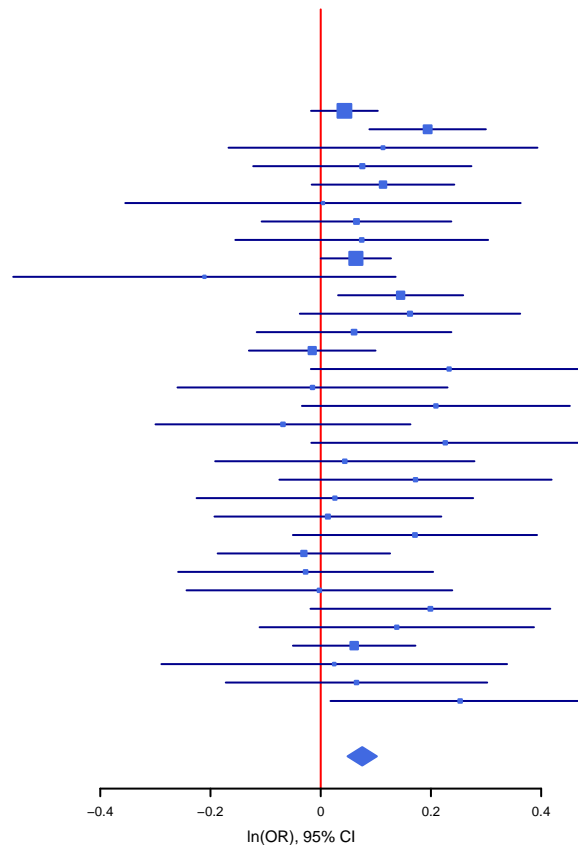
het_P: 0.149 het_I: 20.6

	ngt	info	p_value	f_ca(n)	f_co(n)	ln(OR)	STDerr
aunz	0	0.97	0.0241	0.421(2536)	0.401(15624)	0.0707	0.0314
chop	0	1.00	0.198	0.438(1032)	0.427(3568)	0.0679	0.0527
fin1	0	0.96	0.136	0.411(130)	0.366(503)	0.222	0.149
fre1	0	0.98	0.748	0.415(293)	0.427(990)	-0.0319	0.0992
gns2	0	0.96	0.111	0.455(657)	0.429(2263)	0.104	0.0653
itgr	0	1.04	0.87	0.513(144)	0.500(129)	0.0283	0.172
net1	0	0.96	0.344	0.437(348)	0.420(1337)	0.084	0.0888
poco	0	0.98	0.941	0.439(239)	0.447(604)	-0.00854	0.115
sedk	0	1.00	0.151	0.430(4105)	0.421(3793)	0.0469	0.0327
spa1	0	0.93	0.291	0.419(184)	0.387(156)	0.186	0.176
ukbb	0	0.96	0.318	0.419(768)	0.405(3065)	0.0591	0.0592
ukd1	0	1.01	0.146	0.417(242)	0.376(957)	0.152	0.104
usa1	0	0.93	0.286	0.434(418)	0.410(750)	0.0983	0.0921
usa2	0	0.94	0.811	0.421(1304)	0.419(1307)	0.014	0.0585
w10	0	0.94	0.746	0.418(156)	0.424(714)	-0.0431	0.133
w11	0	0.87	0.98	0.441(164)	0.440(754)	-0.00341	0.133
w12	0	0.94	0.435	0.438(176)	0.415(785)	0.0973	0.124
w13	0	1.02	0.0445	0.465(181)	0.408(789)	0.236	0.118
w14	0	0.96	0.624	0.440(169)	0.424(761)	0.0614	0.125
w15	0	0.99	0.988	0.439(185)	0.430(818)	0.0018	0.119
w16	0	0.96	0.00441	0.482(174)	0.404(800)	0.352	0.124
w17	0	0.98	0.149	0.443(158)	0.386(761)	0.186	0.129
w18	0	1.04	0.683	0.407(245)	0.397(852)	0.042	0.103
w19	0	0.99	0.0667	0.474(211)	0.424(851)	0.203	0.111
w1to6	0	0.93	0.000906	0.479(357)	0.420(4828)	0.269	0.0811
w20	0	0.99	0.0261	0.475(200)	0.411(802)	0.254	0.114
w21	0	0.95	0.812	0.416(161)	0.420(801)	-0.0305	0.128
w22	0	0.99	0.968	0.415(211)	0.412(845)	0.00449	0.111
w23	0	1.00	0.049	0.372(164)	0.427(957)	-0.246	0.125
w24	0	0.95	0.0856	0.421(1241)	0.399(1447)	0.0987	0.0574
w7	0	0.92	0.087	0.367(97)	0.430(831)	-0.282	0.165
w8	0	0.99	0.0203	0.473(162)	0.406(1062)	0.281	0.121
w9	0	1.00	0.308	0.449(180)	0.421(1021)	0.118	0.116
Meta-analysis	0	0.97	3.13e-08	0.431(16992)	0.412(55525)	0.0757	0.0137

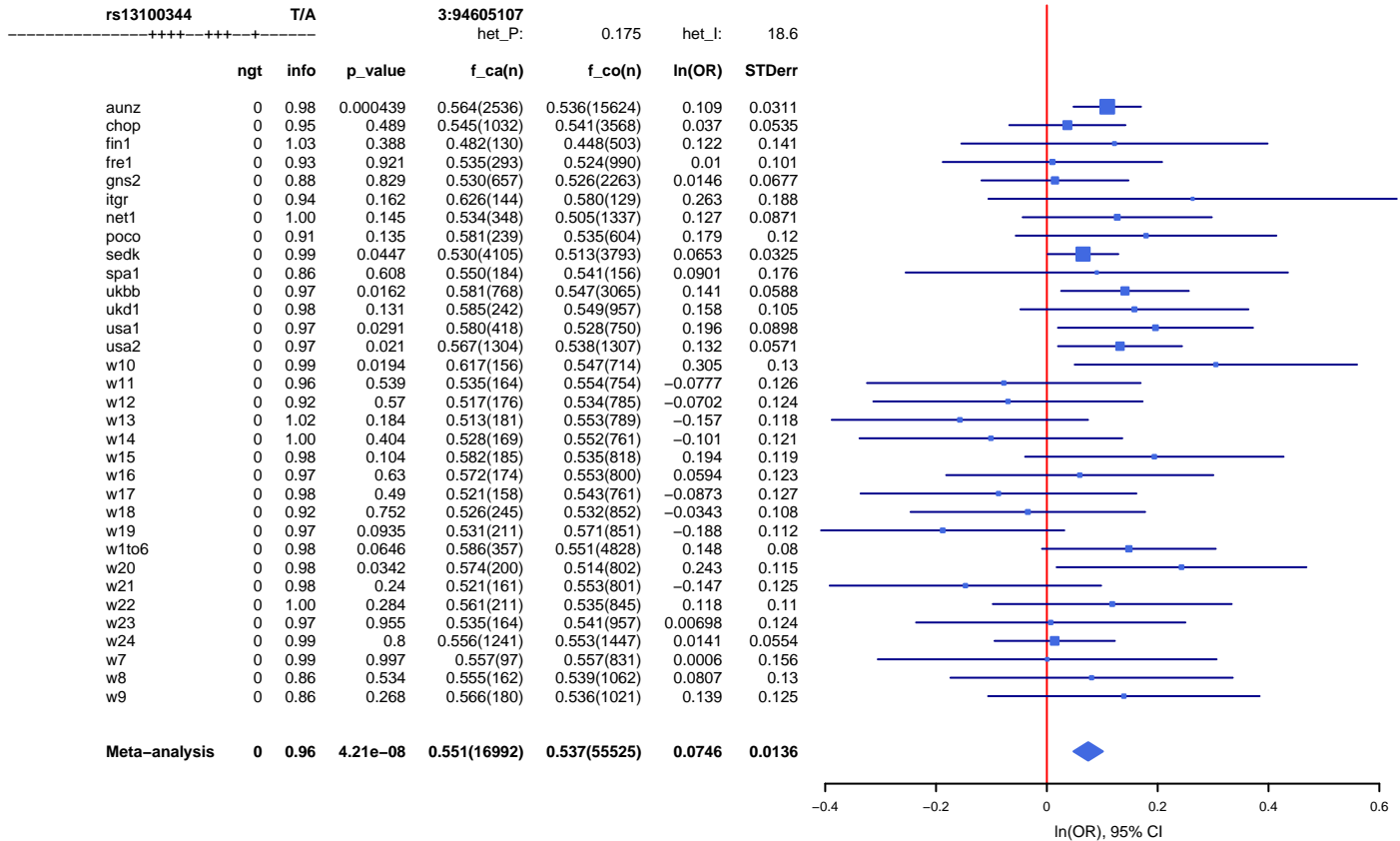


(g) chr5:rs370838138.

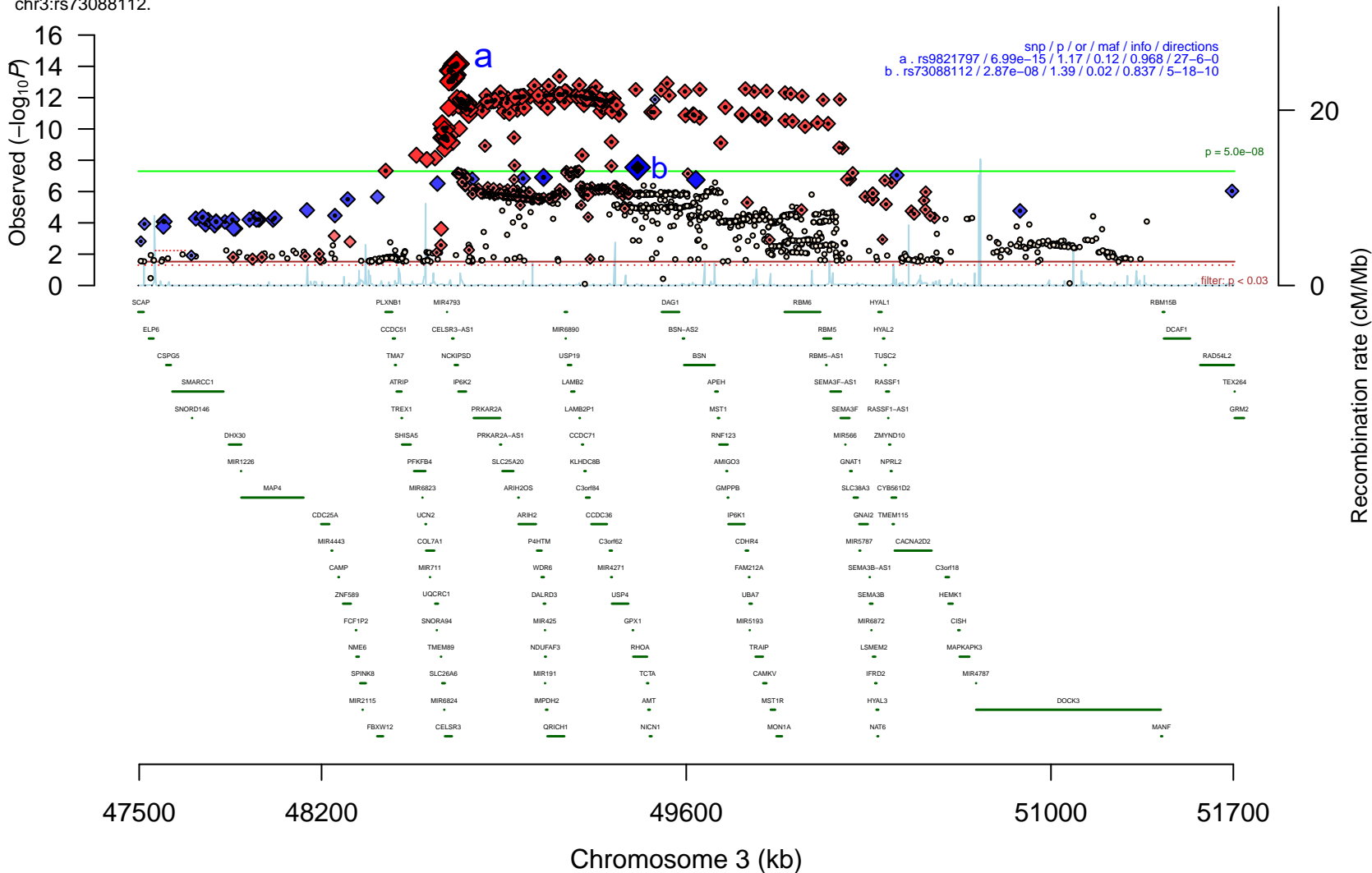
	rs370838138		G/C		5:25081845		het_P:	0.637	het_I:	0.0
	ngt	info	p_value	f_ca(n)	f_co(n)	ln(OR)	STDerr			
aunz	0	0.99	0.164	0.574(2536)	0.560(15624)	0.043	0.0309			
chop	0	0.99	0.000315	0.607(1032)	0.563(3568)	0.194	0.0539			
fin1	0	0.99	0.429	0.535(130)	0.504(503)	0.113	0.143			
fre1	0	0.98	0.455	0.617(293)	0.589(990)	0.0755	0.101			
gns2	0	0.95	0.0855	0.590(657)	0.563(2263)	0.113	0.066			
itgr	0	0.97	0.982	0.625(144)	0.629(129)	0.00409	0.183			
net1	0	1.02	0.46	0.597(348)	0.576(1337)	0.0649	0.0878			
poco	0	0.92	0.527	0.546(239)	0.520(604)	0.0744	0.117			
sedk	0	1.00	0.0499	0.563(4105)	0.548(3793)	0.0638	0.0325			
spa1	0	0.93	0.233	0.609(184)	0.664(156)	-0.211	0.177			
ukbb	0	1.01	0.0125	0.592(768)	0.555(3065)	0.145	0.0579			
ukd1	0	1.04	0.11	0.578(242)	0.535(957)	0.162	0.102			
usa1	0	0.96	0.502	0.588(418)	0.567(750)	0.0605	0.0901			
usa2	0	0.95	0.793	0.586(1304)	0.592(1307)	-0.0154	0.0586			
w10	0	1.03	0.0683	0.618(156)	0.559(714)	0.233	0.128			
w11	0	0.98	0.905	0.568(164)	0.575(754)	-0.0149	0.125			
w12	0	0.97	0.0919	0.614(176)	0.565(785)	0.209	0.124			
w13	0	1.00	0.563	0.549(181)	0.567(789)	-0.0685	0.118			
w14	0	1.02	0.0676	0.622(169)	0.570(761)	0.226	0.124			
w15	0	0.95	0.716	0.572(185)	0.562(818)	0.0437	0.12			
w16	0	0.96	0.174	0.614(174)	0.573(800)	0.172	0.126			
w17	0	0.96	0.841	0.576(158)	0.559(761)	0.0257	0.128			
w18	0	0.98	0.902	0.564(245)	0.563(852)	0.013	0.105			
w19	0	0.99	0.131	0.614(211)	0.575(851)	0.171	0.113			
w1to6	0	0.96	0.702	0.571(357)	0.579(4828)	-0.0306	0.0798			
w20	0	0.93	0.816	0.586(200)	0.591(802)	-0.0274	0.118			
w21	0	1.04	0.984	0.591(161)	0.590(801)	-0.0024	0.123			
w22	0	1.03	0.0724	0.617(211)	0.569(845)	0.199	0.111			
w23	0	0.93	0.278	0.595(164)	0.564(957)	0.138	0.127			
w24	0	0.96	0.284	0.589(1241)	0.574(1447)	0.0607	0.0567			
w7	0	0.93	0.879	0.555(97)	0.558(831)	0.0245	0.16			
w8	0	1.00	0.592	0.577(162)	0.562(1062)	0.0649	0.121			
w9	0	0.98	0.0358	0.633(180)	0.572(1021)	0.253	0.12			
Meta-analysis	0	0.98	3.17e-08	0.582(16992)	0.564(55525)	0.0753	0.0136			



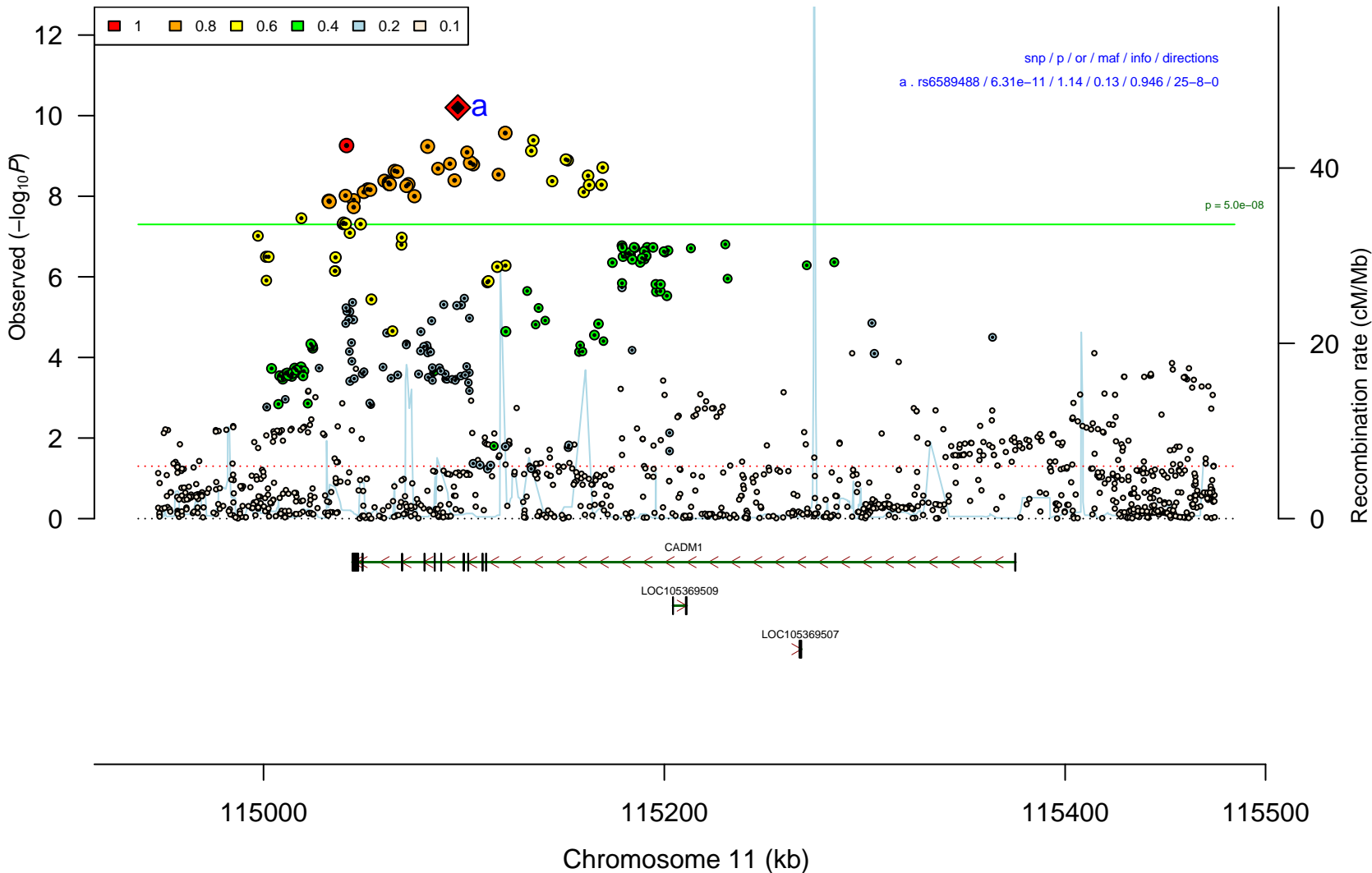
(h) chr3:rs13100344.



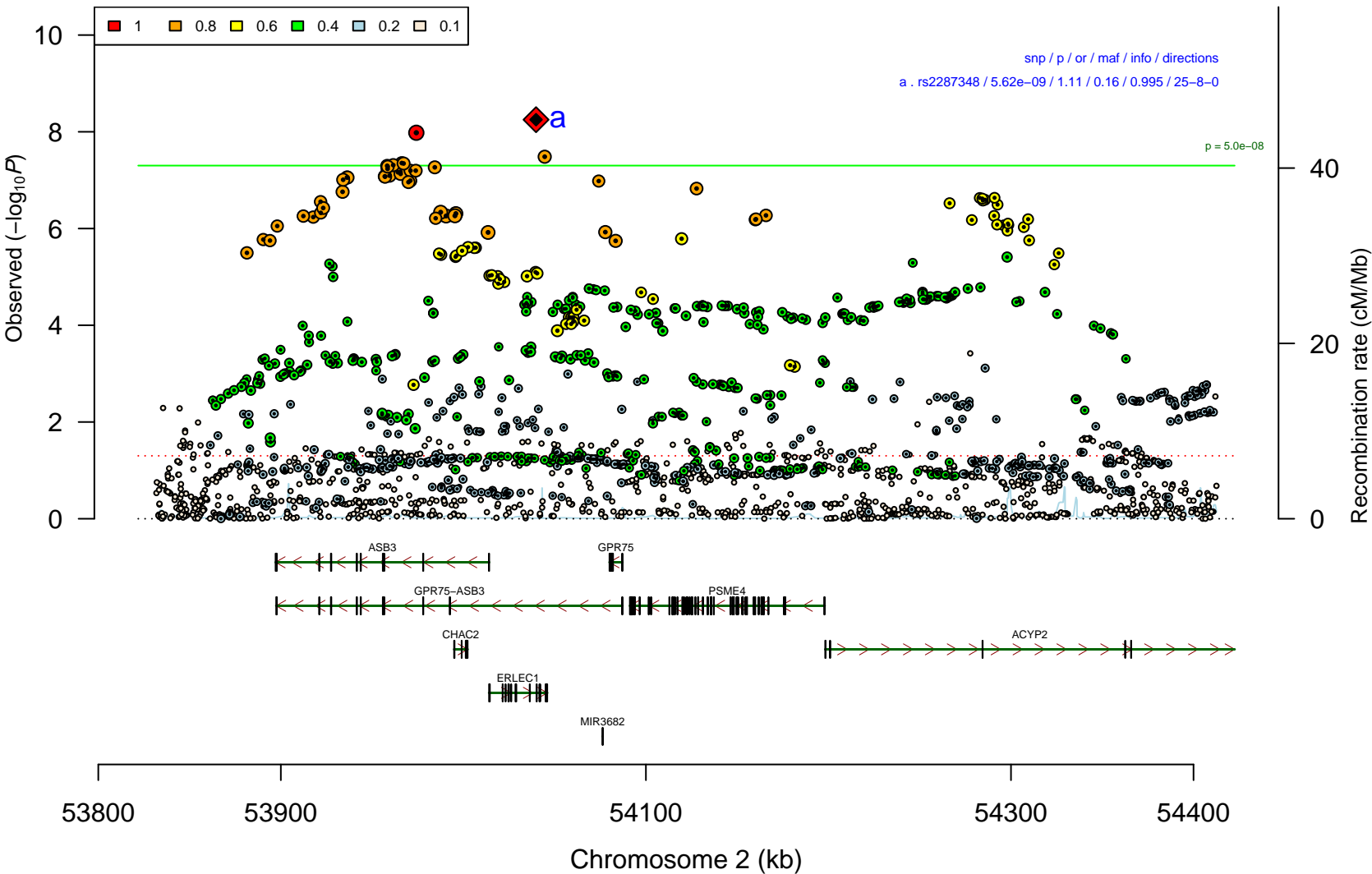
Supplementary Figure 3a-h legend. Regional plot of GWAS meta-analysis results for all genome-wide significant loci. The sample size is 16,992 cases and 55,525 controls. Two loci overlapped (index SNPs of rs9821797 and rs73088112), therefore we refer to only one of these (rs9821797) in the main manuscript. (a) chr3:rs9821797 and chr3:rs73088112.



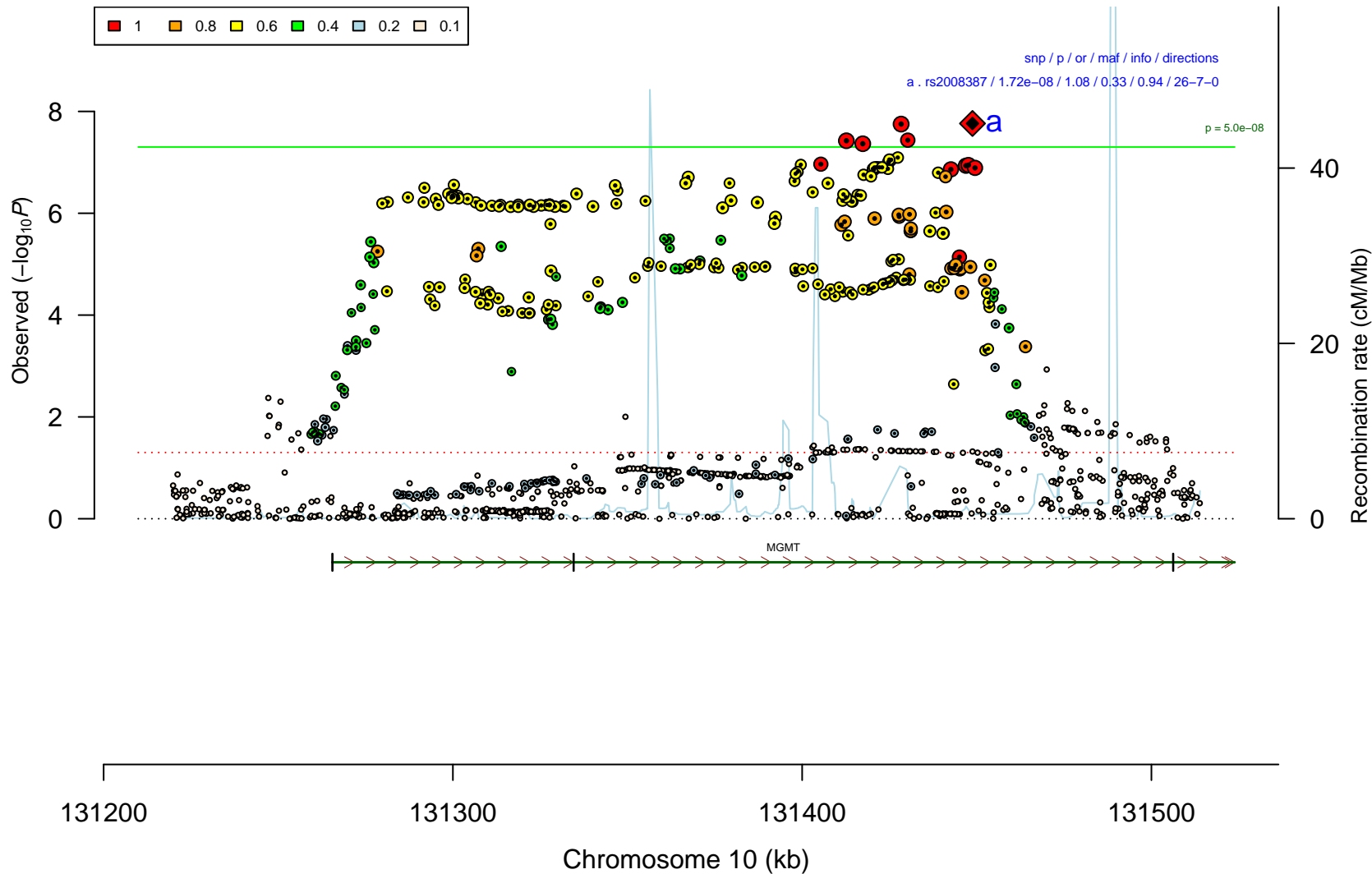
(b) chr11:6589488.



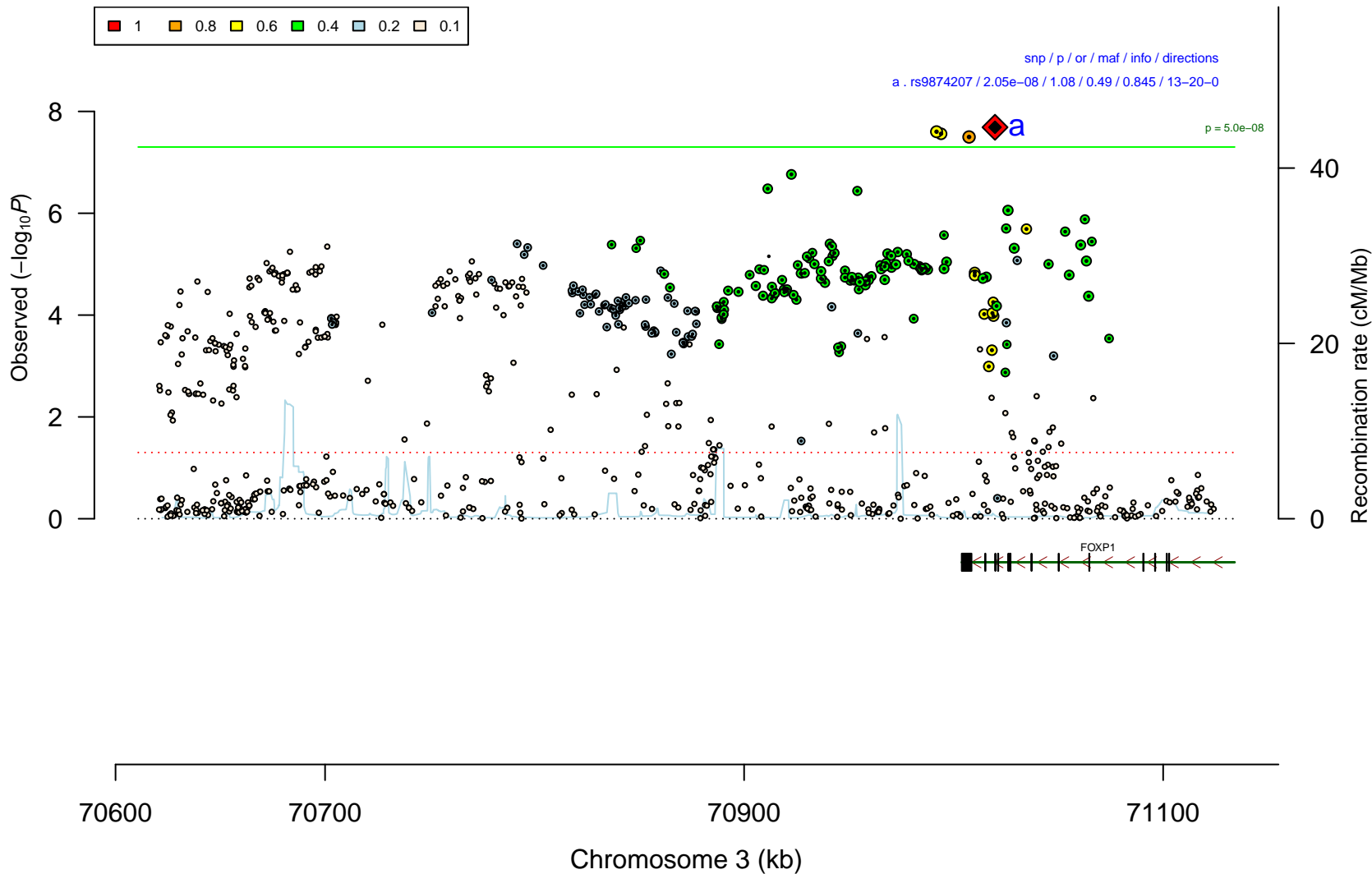
(c) chr2:2287348.



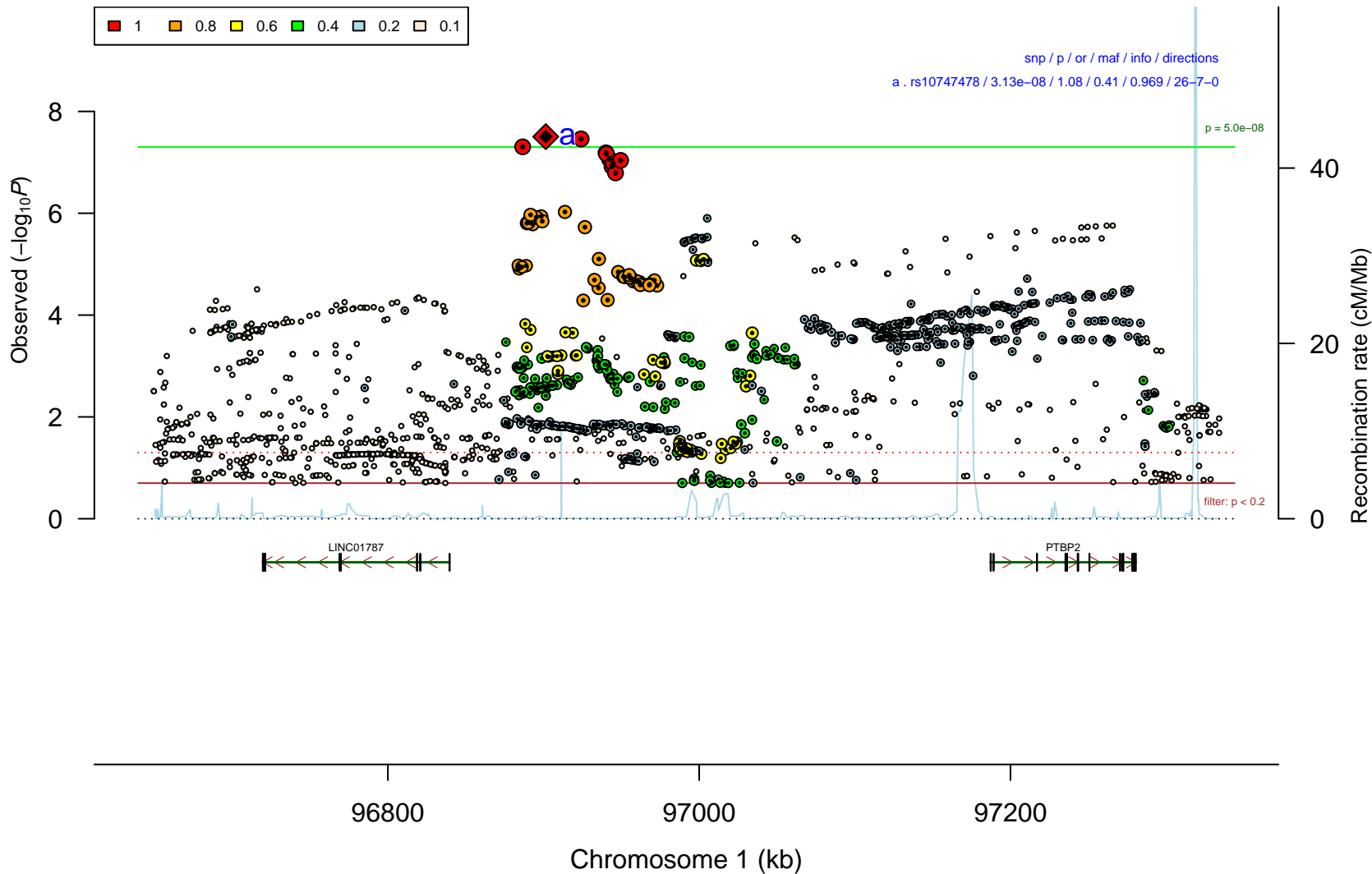
(d) chr10:rs2008387.



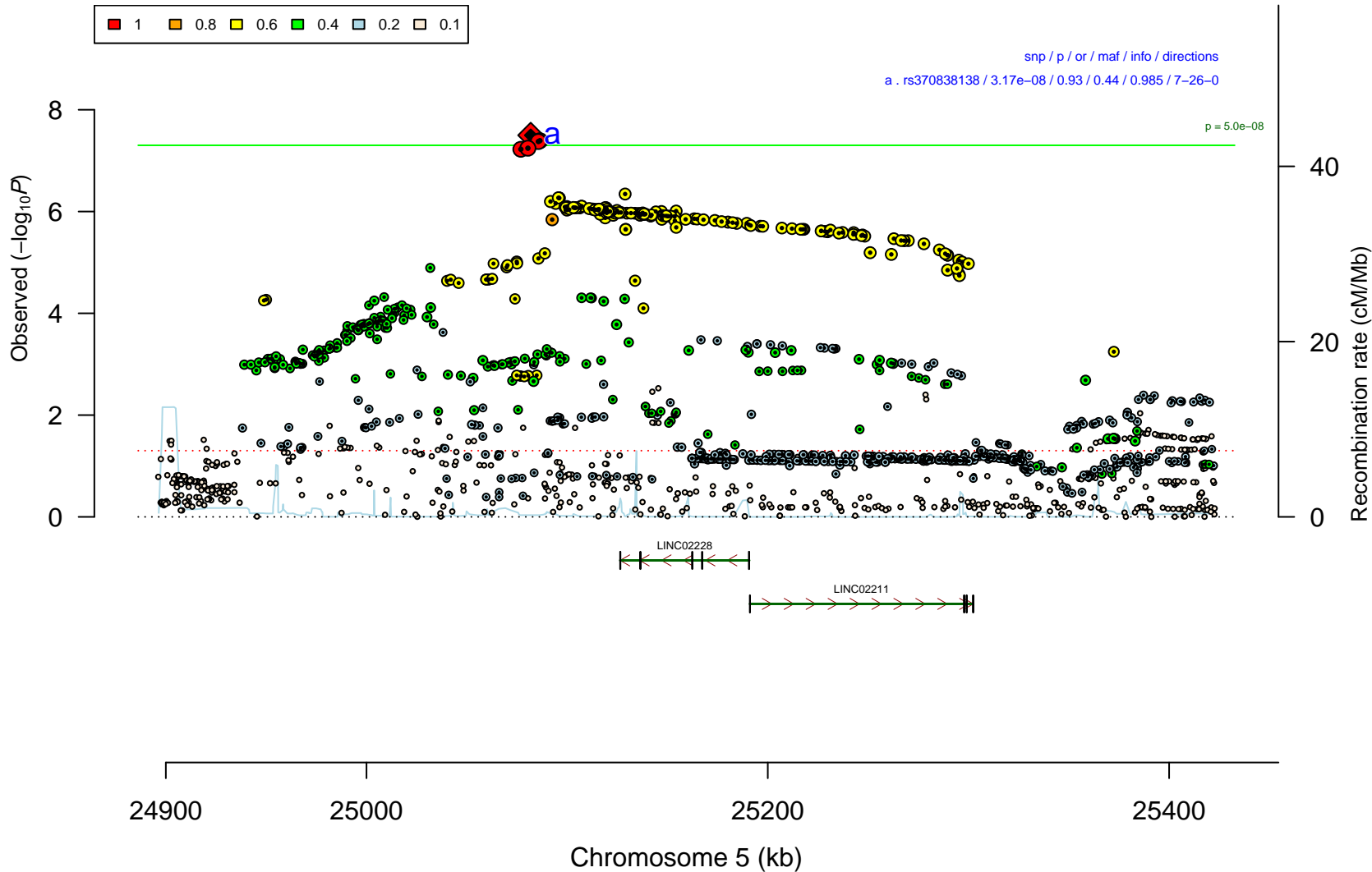
(e) chr3:rs9874207.



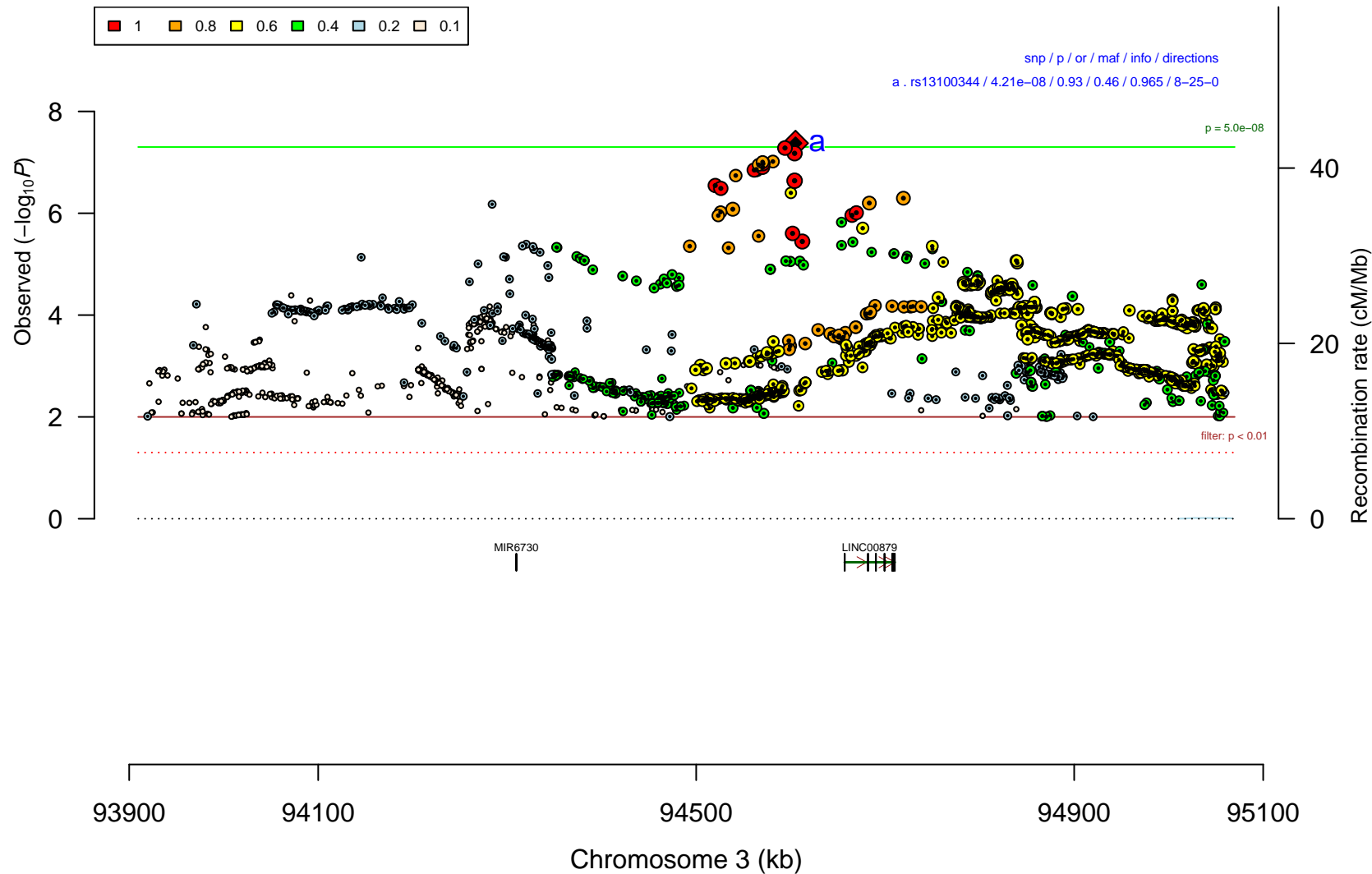
(f) chr1:rs10747478.



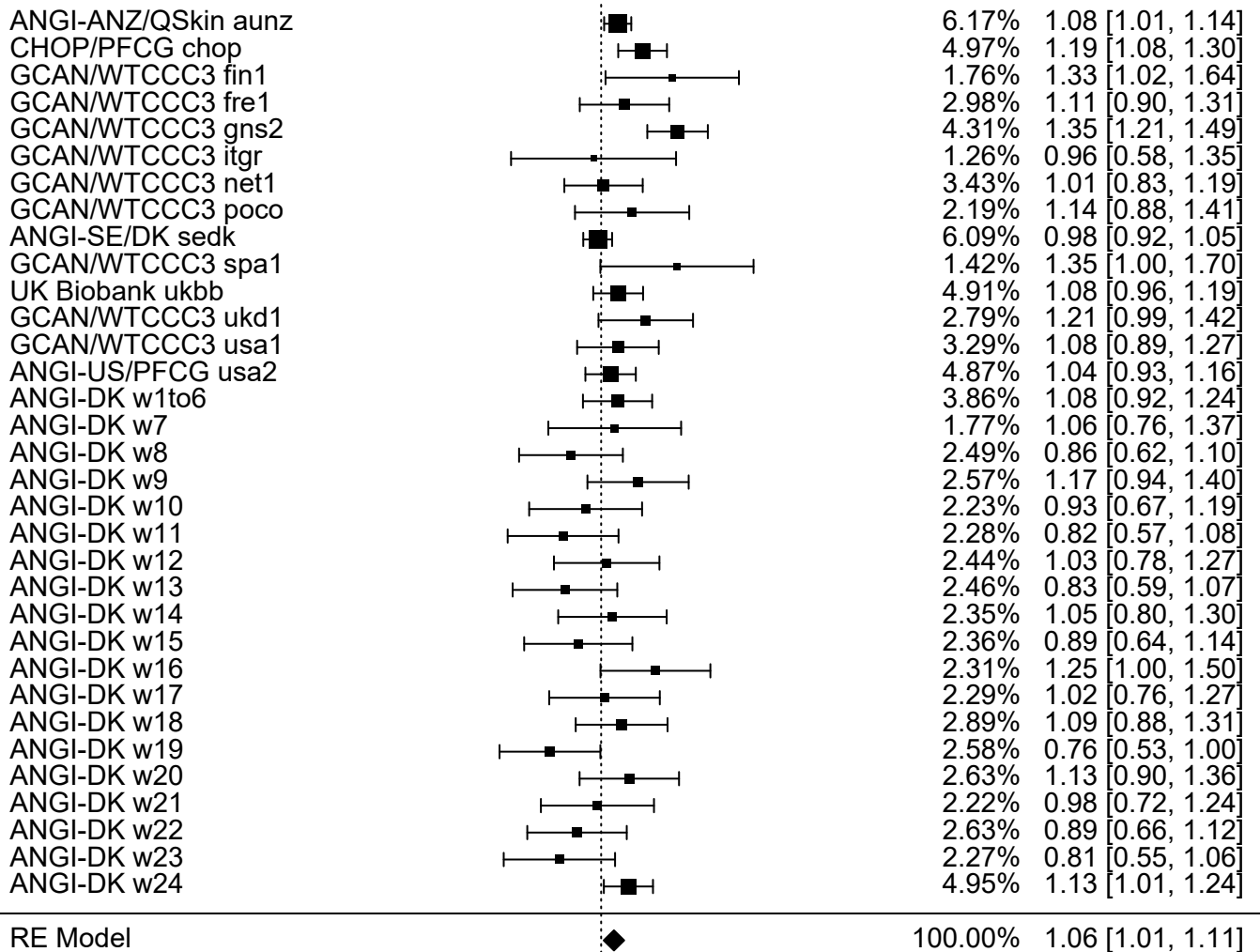
(g) chr5:rs370838138.



(h) chr3:rs13100344.



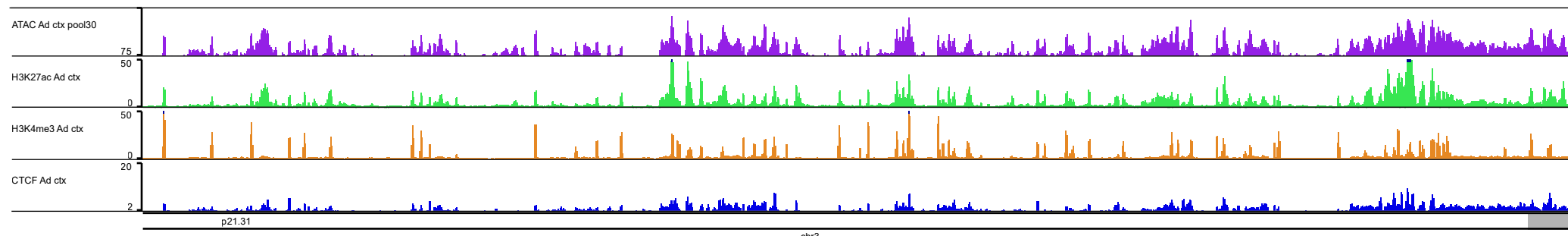
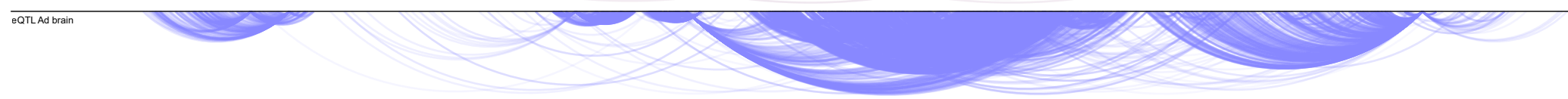
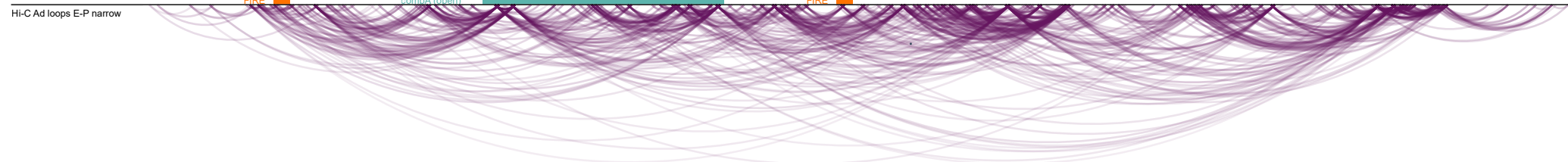
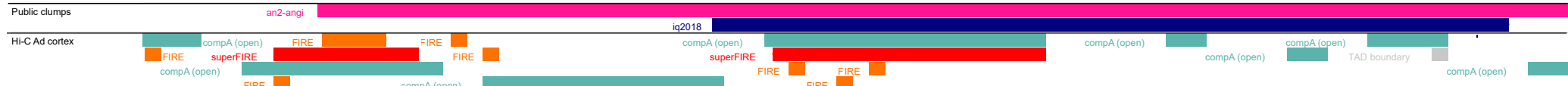
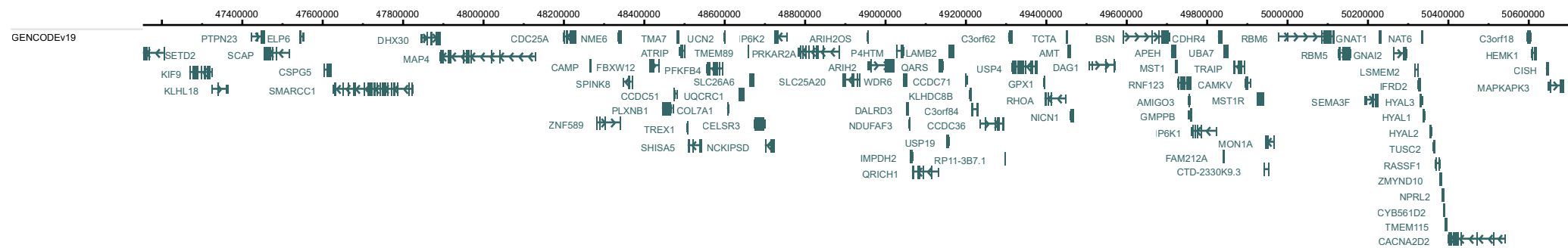
Supplementary Figure 4. A random-effects meta-analysis of the association between rs4622308 and anorexia nervosa. Previously, the first freeze of the PGC-ED revealed that this SNP was genome-wide significant (Duncan et al., 2017). In the present study, this SNP was not genome-wide significant in the primary GWAS meta-analysis using a fixed-effects model, and in a subsequent random effects meta-analysis as shown in this figure was also non-significant and showed evidence of heterogeneity (odds ratio [OR] = 1.06, 95% CI = 1.01-1.11, $P_{\text{two-tailed}} = 0.0002$, $I^2 = 53.76$). The percentages refer to the weight assigned to each cohort. The figures on the right are the percentage weight assigned to each cohort, the center values are the OR, and the error bar is the 95% confidence interval. The vertical line is the reference line of no effect. The sample size is 16,992 cases and 55,525 controls.



0.5 1 1.5 2

OR

Supplementary Figure 5a-h legend. Hi-C and eQTL analyses. These figures were generated using the WashU EpiGenome browser (<http://epigenomegateway.wustl.edu/browser>). Yellow shaded regions show the “clumps” associated with AN. The top track shows GENCODE v19 gene models. The "public clumps" track shows psychiatric GWA regions including AN GWAS. The "Hi-C Ad cortex" track shows "compartment A/B", "FIRES" (frequently interacting regions), "superFIRES" (local aggregates of FIRES), and topologically associated domain boundaries (TADs). The "Hi-C Ad loops E-P narrow" contains arcs that show the positions of high confidence chromatin interactions in adult brain (10 Kb resolution) between enhancers and/or promoters (according to ChIP-seq and brain-expressed TSS data) with a Bonferroni $P < 0.001$. The "eQTL Ad brain" track shows cis eQTL information from GTEx for all available brain tissues. The "ATAC Ad ctx pool30" track shows open chromatin data for 30 adult controls. The next three tracks show brain epigenomic marks from ChIP-seq in adult brain cortex (H3K27ac, H3K4me3, and CTCF). We selected eQTL SNP-gene pairs from CommonMind frontal cortex, GTEx in any brain region ($q < 0.05$), or in fetal cortex. Significant eQTL connections were identified by nominal $P < 0.05$ as supplied by CMC and GTEx and significant chromatin interactions were identified with a stringent Bonferroni correction for multiple testing, and only considered 10 Kb bin pairs with P value $< 2.31 \times 10^{-11}$ ($0.001/43,222,677$ tests). The chromatin interaction tests came from Fit-Hi-C with default parameters applied and FastHiC. All tests were two-tailed.



114970000 115000000 115030000 115060000 115090000 115120000 115150000 115180000 115210000 115240000 115270000 115300000 115330000 115360000 115390000 115420000 115450000 115480000

GENCODEv19 CADM1

Public clumps

Hi-C Ad cortex compA (open)

Hi-C Ad loops E-P narrow FIRE

eQTL Ad brain

ATAC Ad ctx pool30

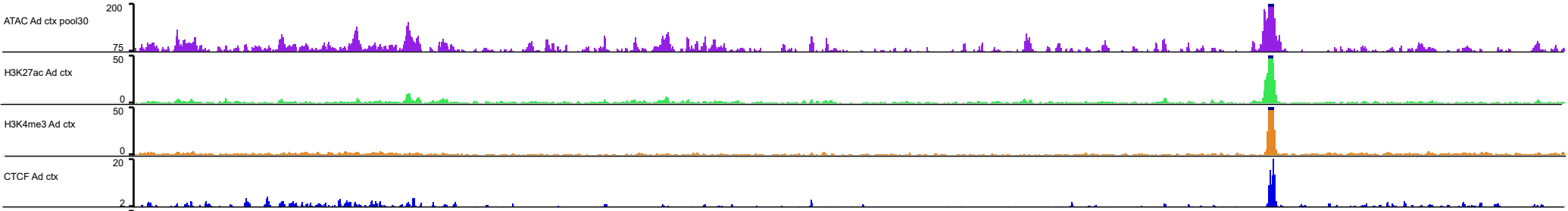
H3K27ac Ad ctx

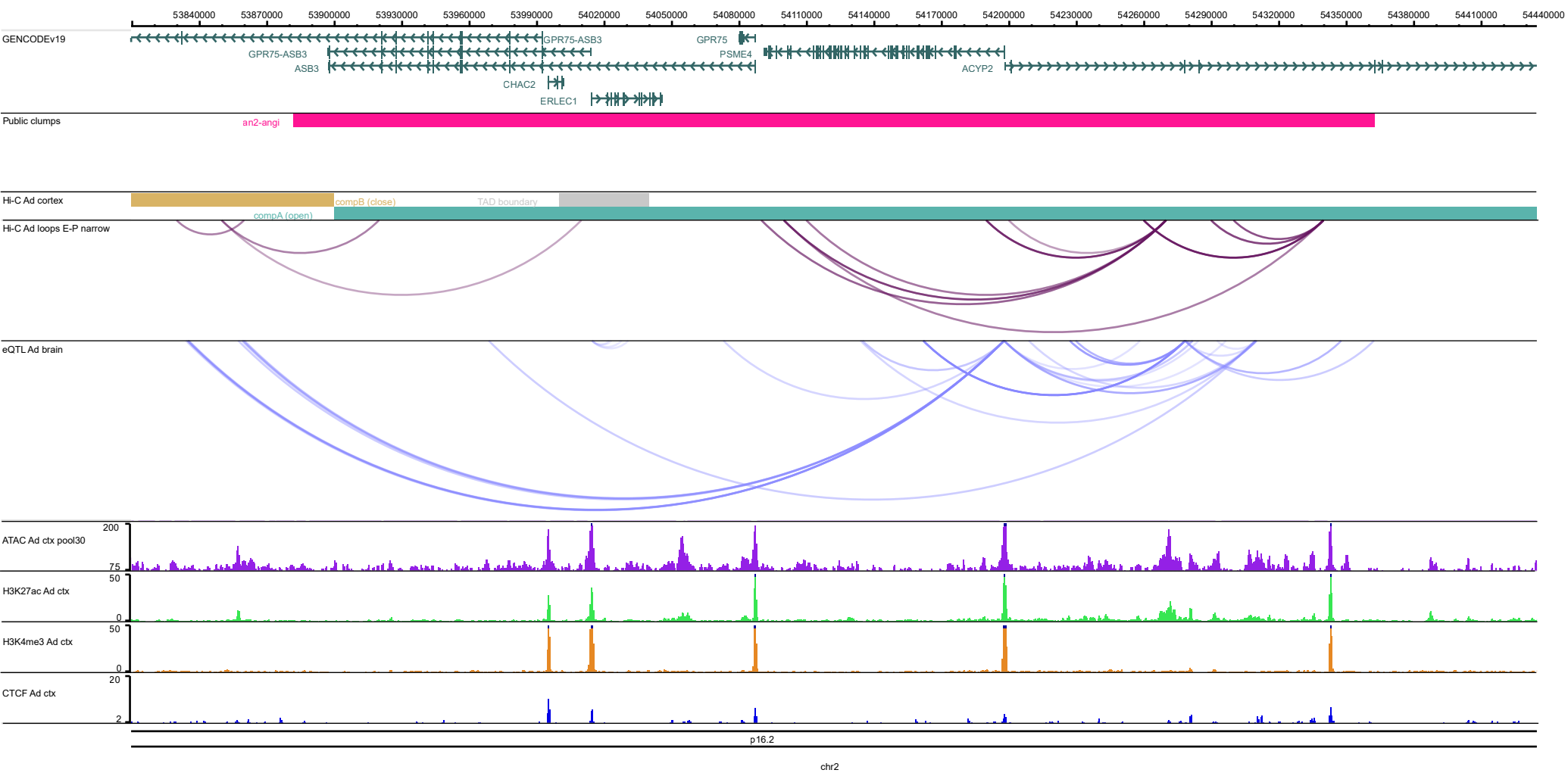
H3K4me3 Ad ctx

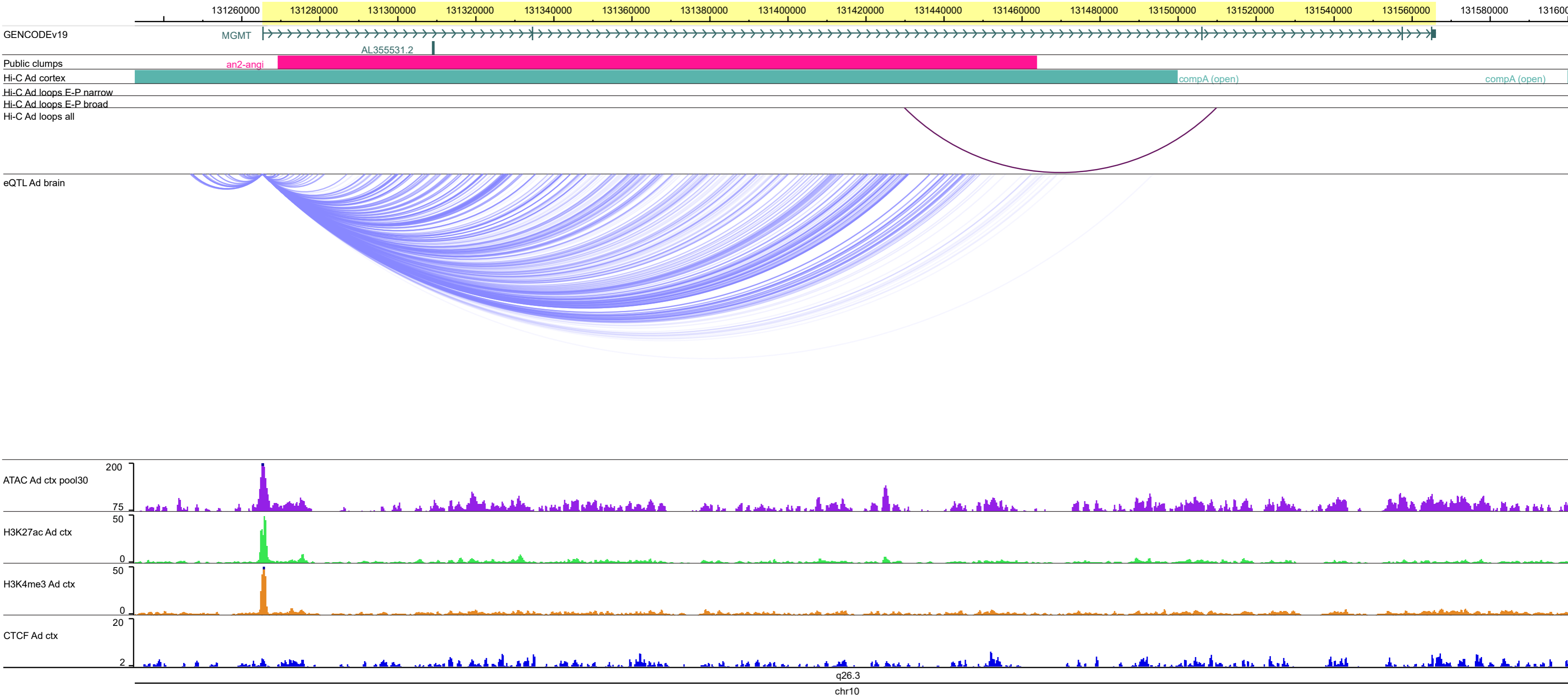
CTCF Ad ctx

chr11

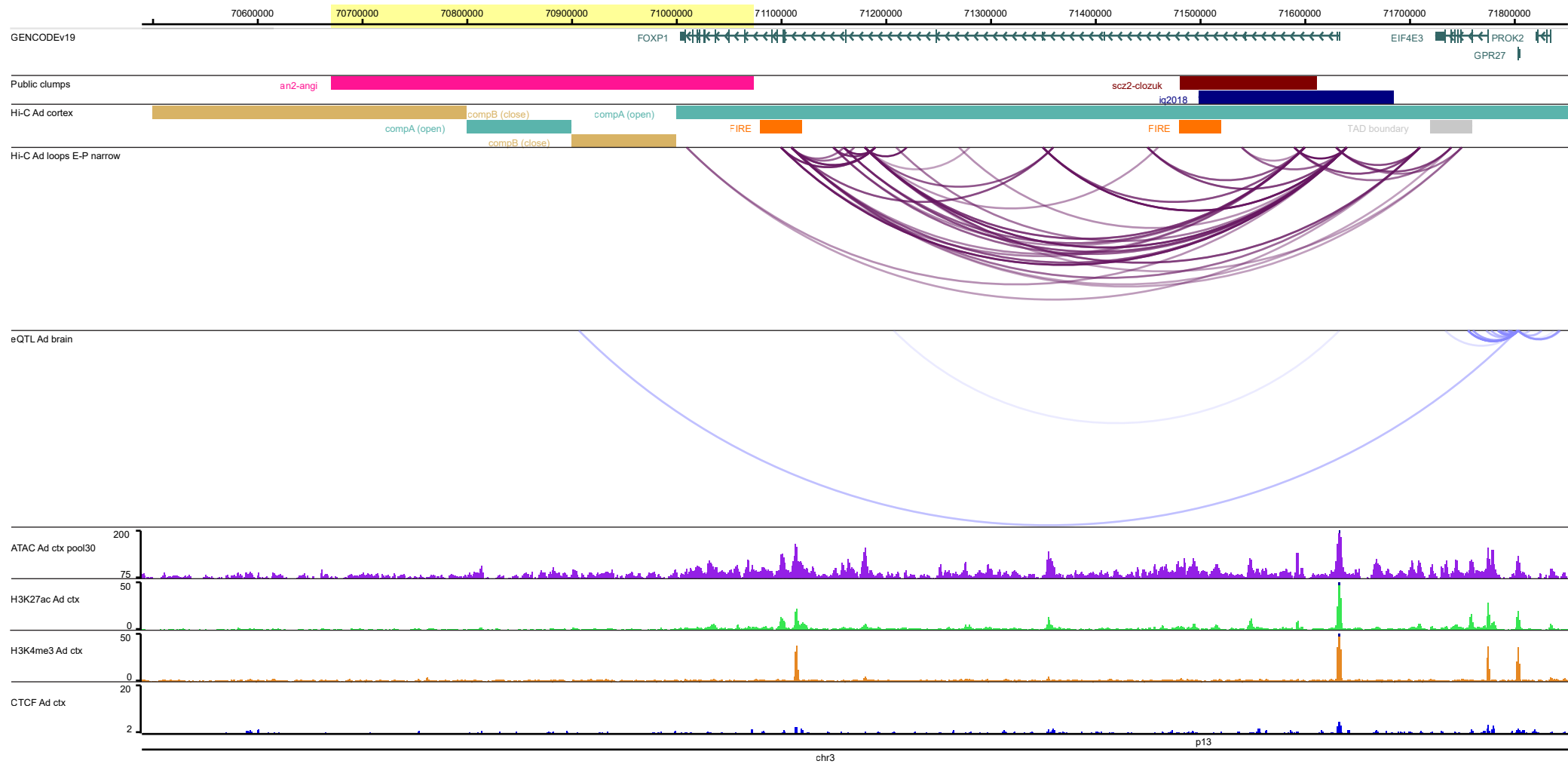
q23.3

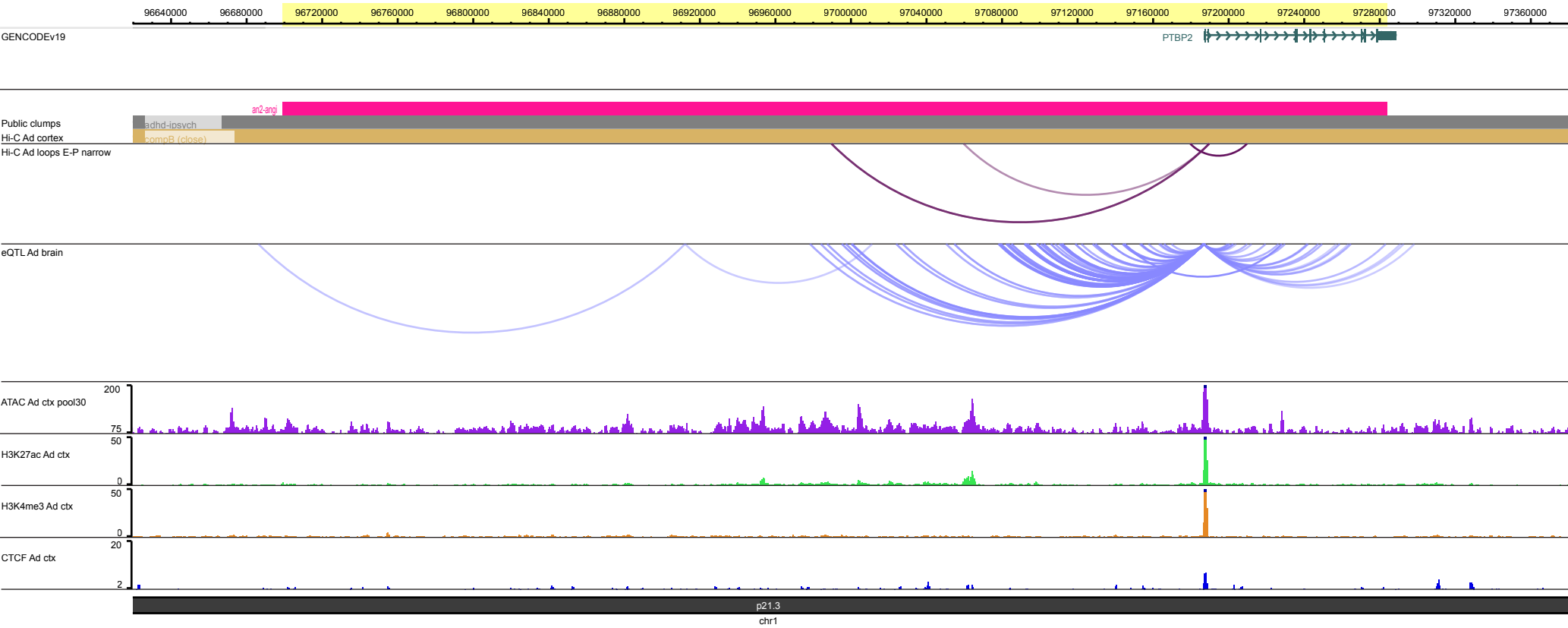






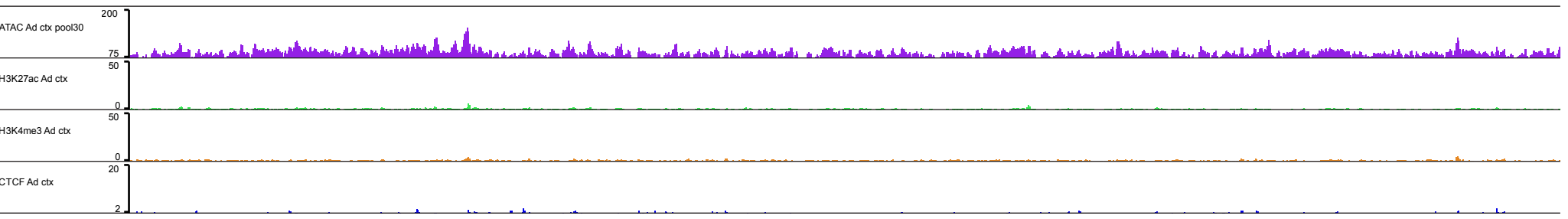
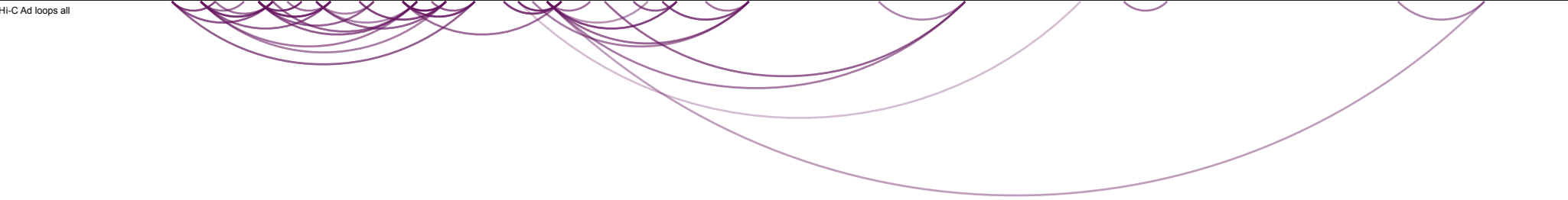
q26.3
chr10



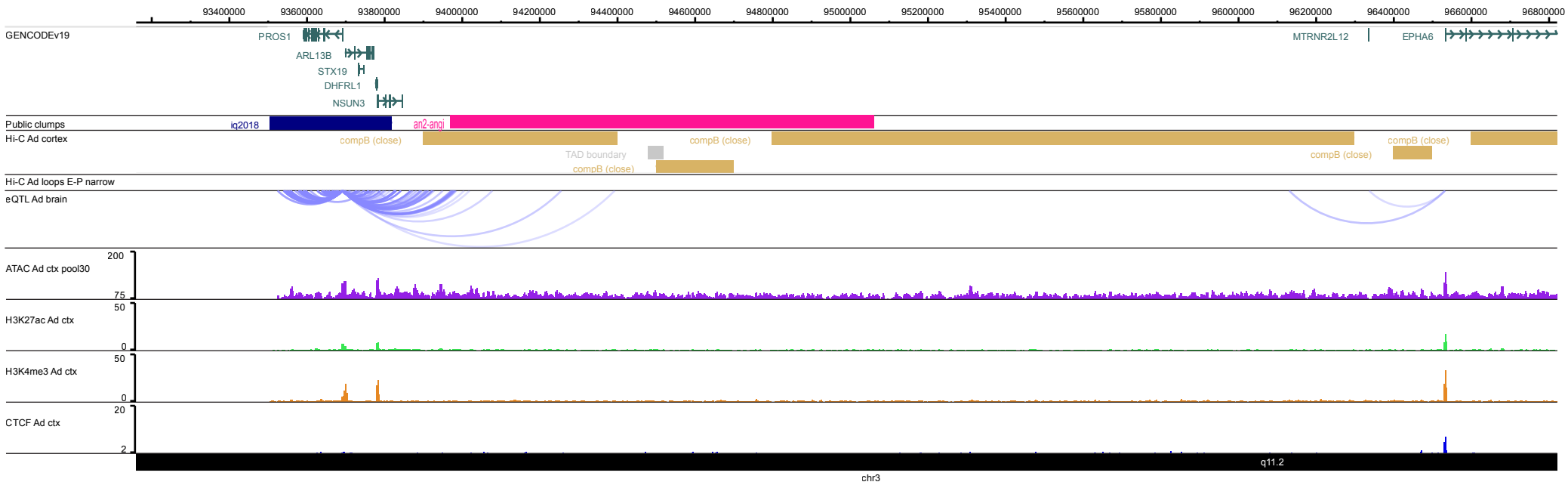


24450000 24500000 24550000 24600000 24650000 24700000 24750000 24800000 24850000 24900000 24950000 25000000 25050000 25100000 25150000 25200000 25250000 25300000 25350000 25400000

GENCODEv19 CDH10



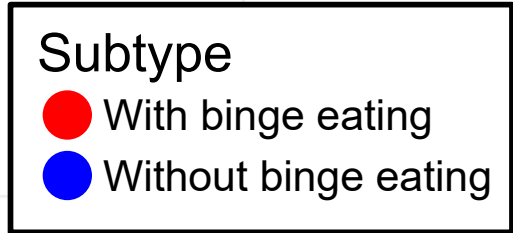
chr5



Supplementary Figure 6. Mean polygenic risk scores (PRS) according to anorexia nervosa subtype (+/- binge eating). In the datasets with available subtype data—aunz (1,417 cases with binge eating, 997 cases without binge eating), chop (358 cases with binge eating, 634 cases without binge eating), and usa2 (606 cases with binge eating, 631 cases without binge eating)— AN PRS was computed for each individual. AN PRS was derived from the primary GWAS meta-analysis summary statistics and adjusted for the principal components used in the main GWAS. Individual PRS were then aggregated into subtype group means. The center values show mean PRS and the error bars show the 95% confidence interval. Two-tailed *T* tests testing for significant differences in PRS scores by subtype were conducted for each cohort using a Bonferroni-corrected *P* threshold of < 0.017 .

Mean Polygenic Risk Score (pT=1)

4e-05
3e-05
2e-05
1e-05
0e+00



P=0.70

P=0.22

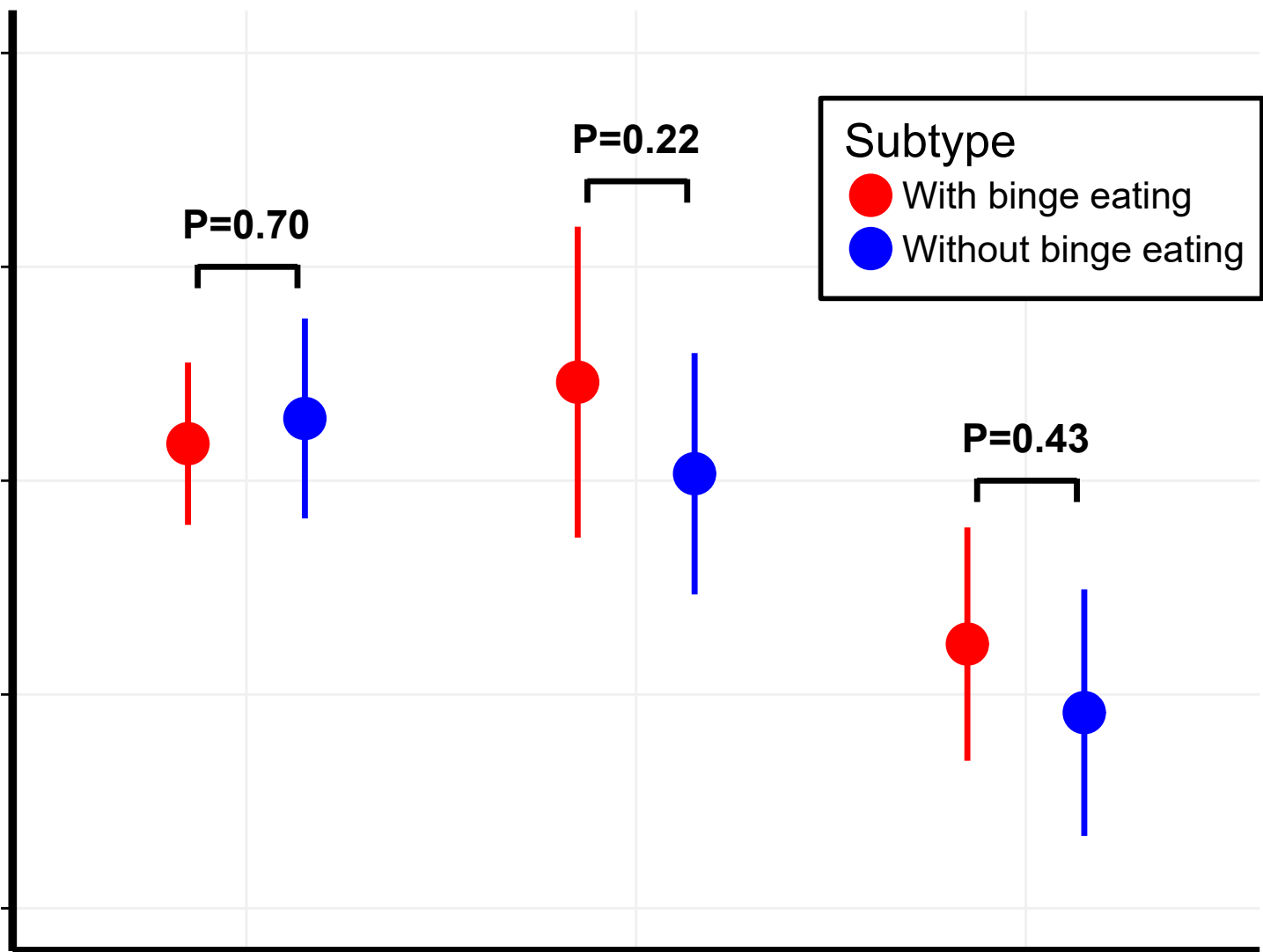
P=0.43

aunz

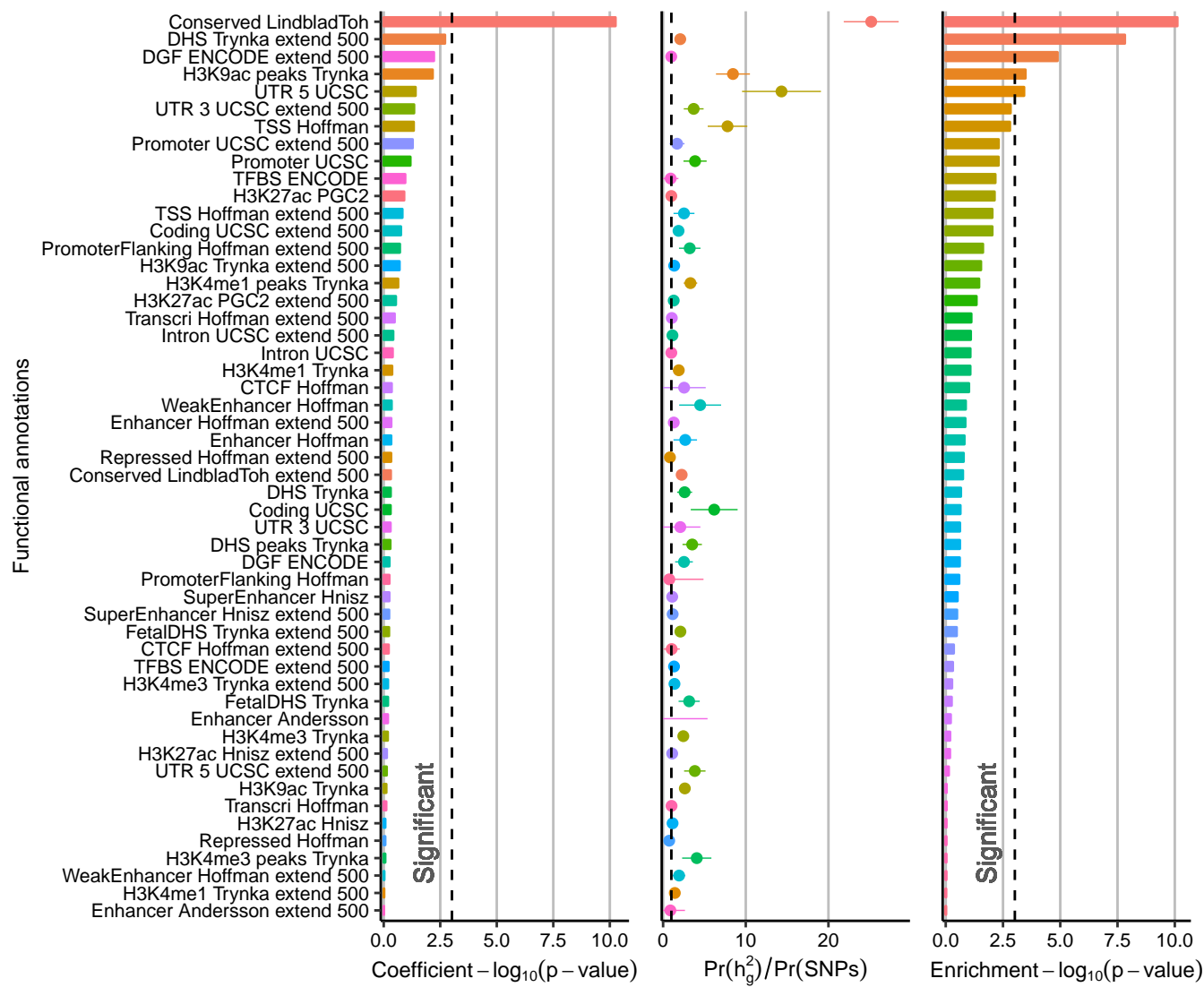
chop

usa2

Cohort

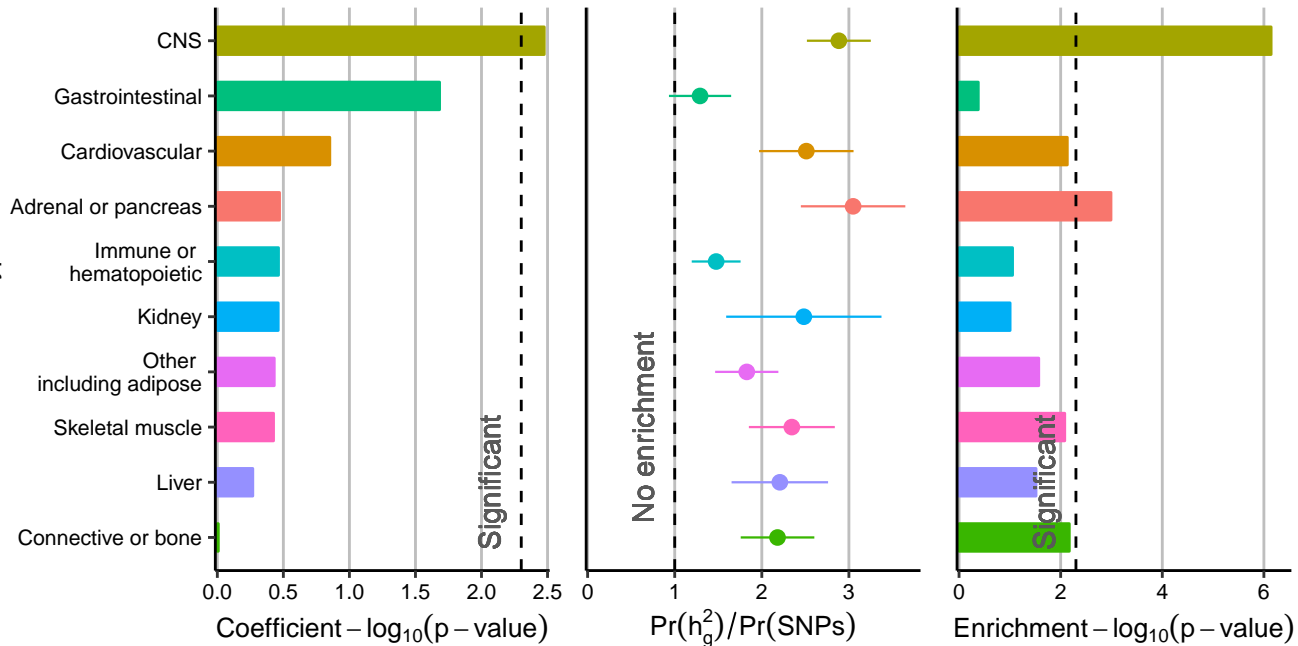


Supplementary Figure 7. Partitioned heritability analysis. The sample size is 16,992 cases and 55,525 controls. The coefficient P value (lefthand) tests for enrichment of heritability within each functional element, controlling for all other functional elements to address overlap. The enrichment P value (righthand) indicates whether this absolute enrichment is statistically significant. In each analysis, the Bonferroni-corrected threshold (vertical line) is $-\log_{10}(P) > 3.0$. The enrichment (middle) scales the heritability captured by each functional element according to the number of variants in the element (vertical line = 1, that is no enrichment). The error bar is the standard error.

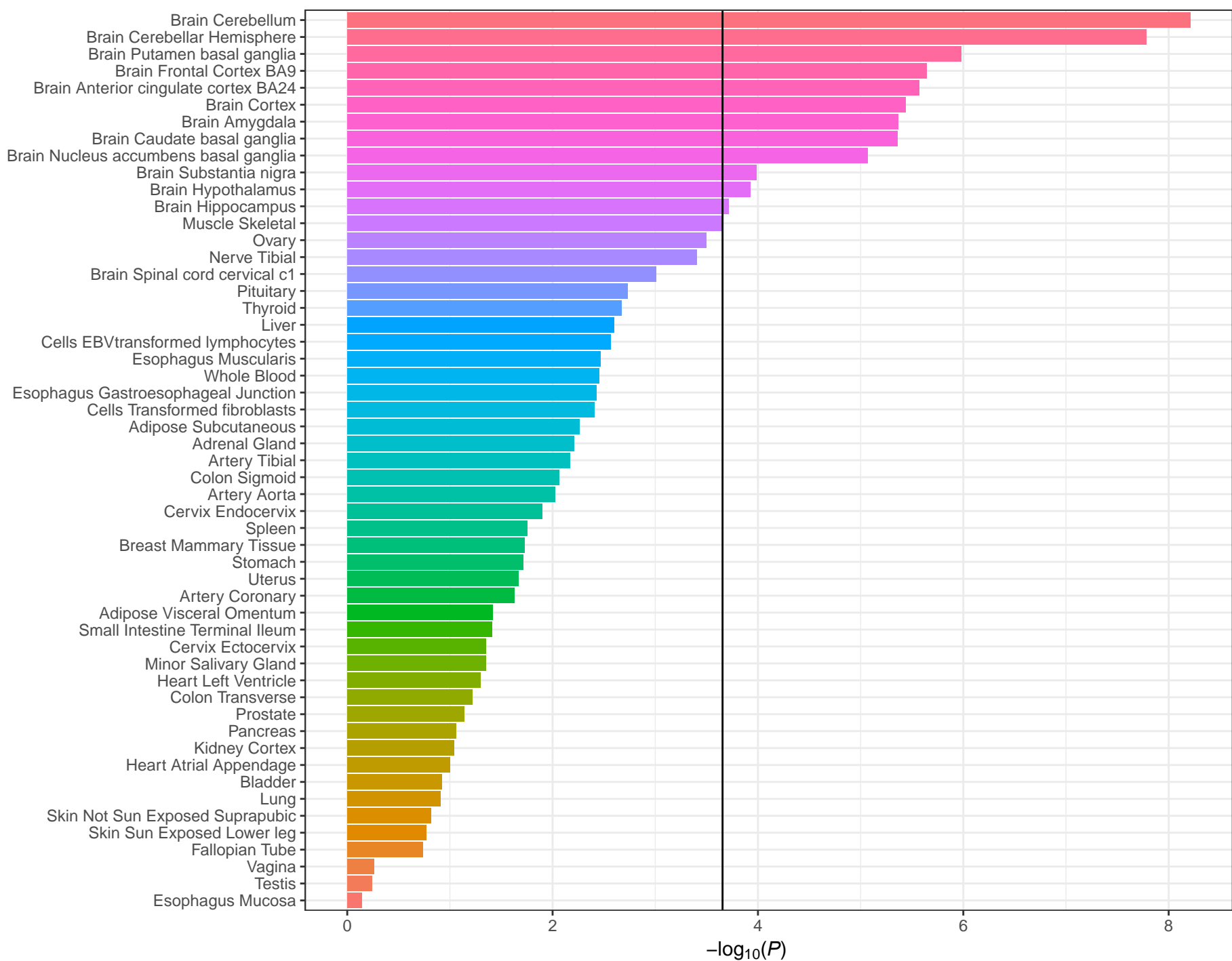


Supplementary Figure 8. Cell type group specific partitioned heritability analysis. The sample size is 16,992 cases and 55,525 controls. The coefficient P value (lefthand) tests for enrichment of heritability within each cell group, controlling for all other cell groups to address overlap. The enrichment P value (righthand) indicates whether this absolute enrichment is statistically significant. In each analysis, the Bonferroni-corrected threshold (vertical line) is $-\log_{10}(P) > 2.3$ (i.e., $0.05/10$ tests). The enrichment (middle) scales the heritability captured by each cell group according to the number of variants in the group (vertical line = 1, that is no enrichment). The error bar is the standard error.

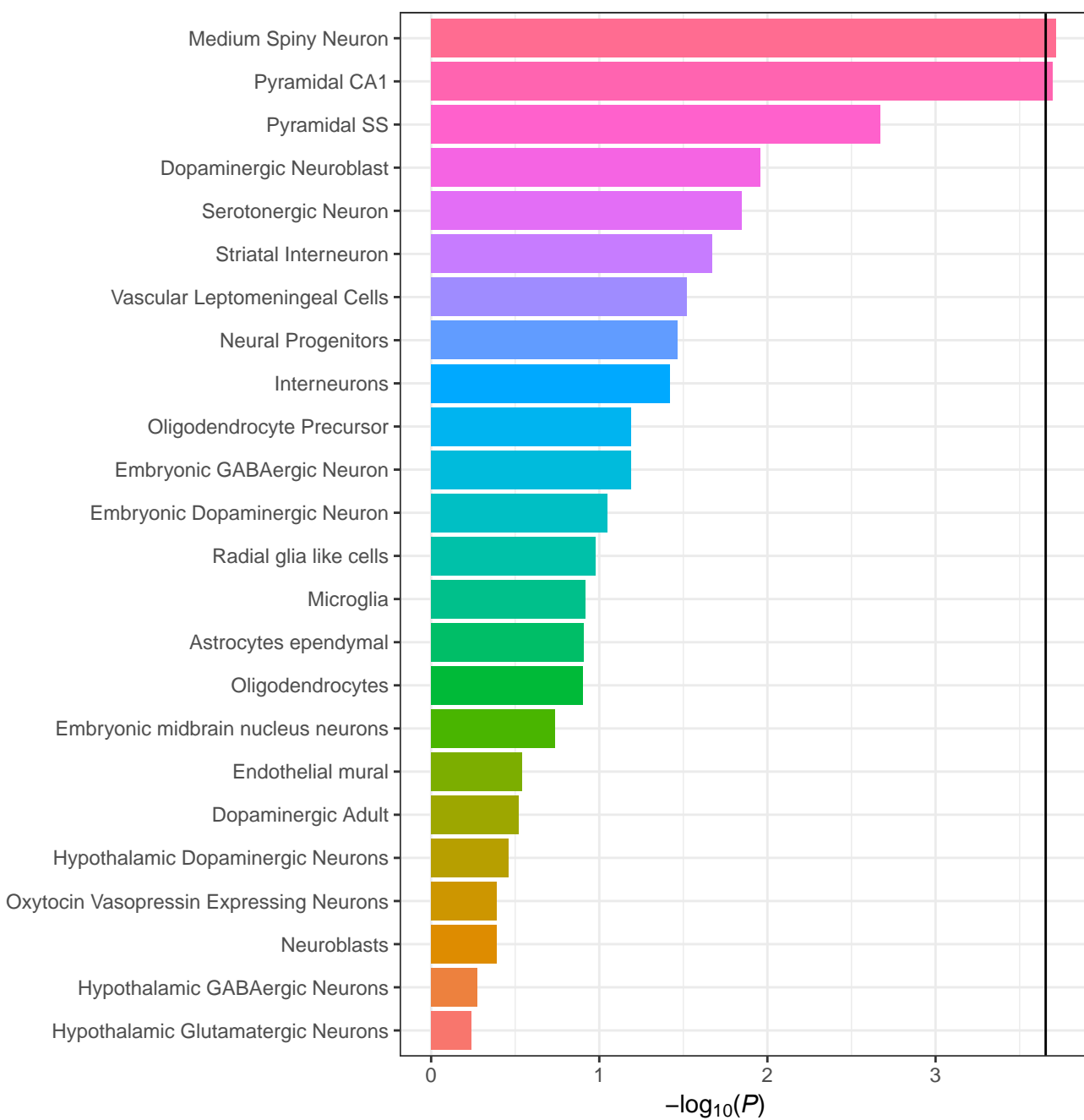
Cell type



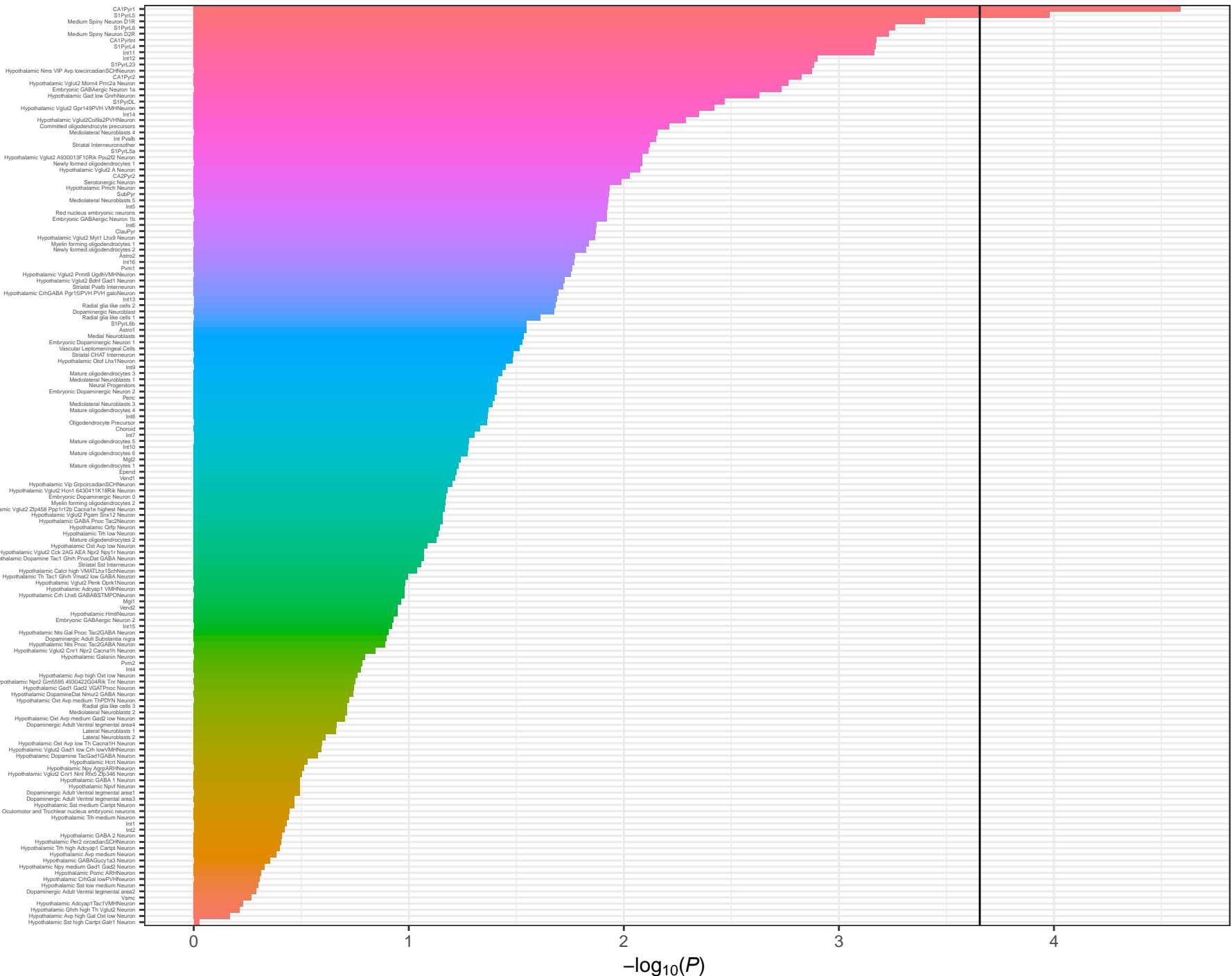
Supplementary Figure 9. P value of association between tissue specificity in GTEx and gene-level genetic association with anorexia nervosa using MAGMA. The sample size is 16,992 cases and 55,525 controls. The Bonferroni-corrected threshold (black vertical line) is $-\log_{10}(P) > 3.6$ and is based on tests across 53 tissues, 24 broad categories of cell types, and 149 KI level 2 cell types, a total of 226 tests.



Supplementary Figure 10. P value of association between tissue specificity in 24 brain cell types (level 1) and gene-level genetic association with anorexia nervosa using MAGMA. The sample size is 16,992 cases and 55,525 controls. The Bonferroni-corrected threshold (black vertical line) is $-\log_{10}(P) > 3.6$ and is based on tests across 53 tissues, 24 broad categories of cell types, and 149 KI level 2 cell types, a total of 226 tests.

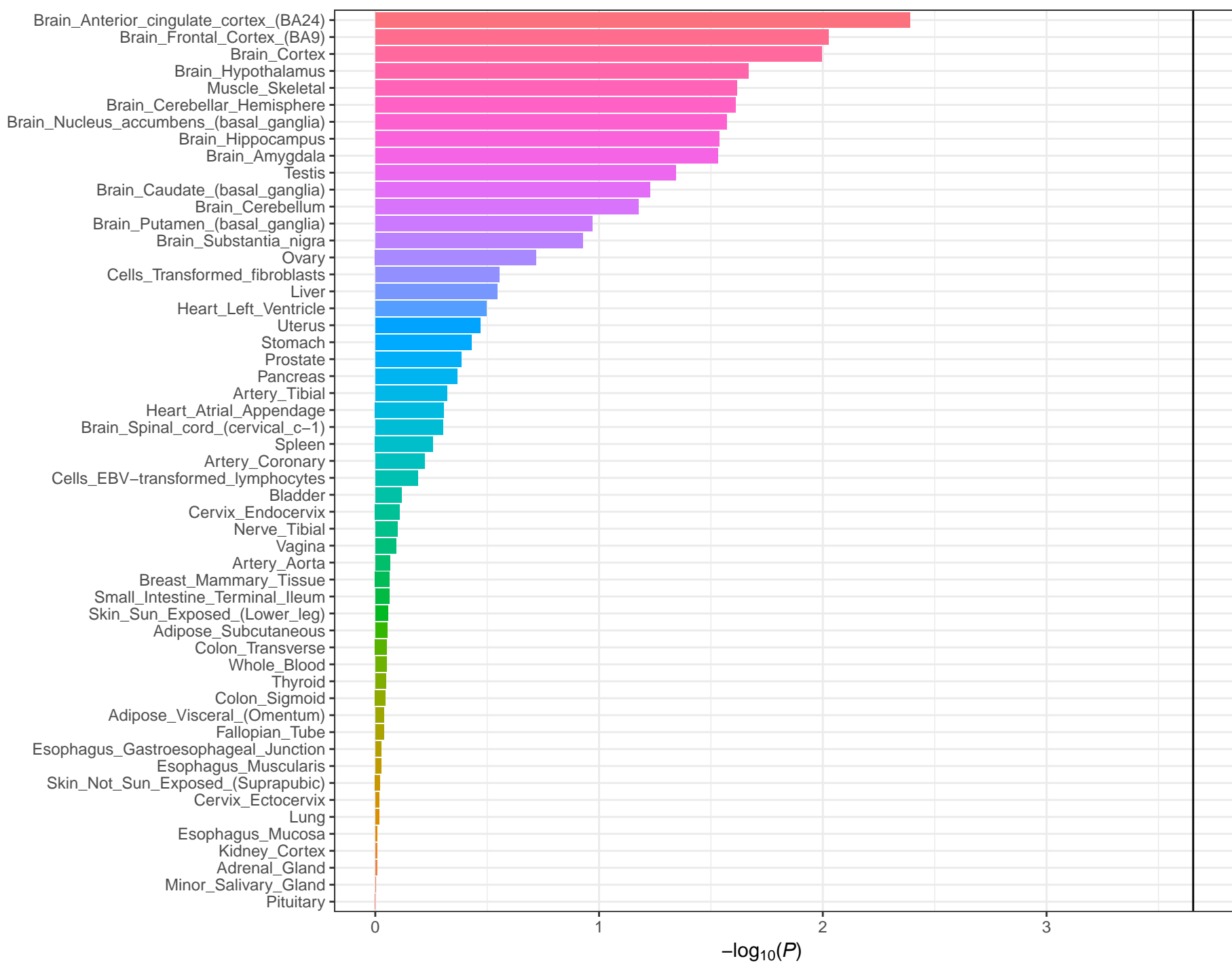


Supplementary Figure 11. P value of association between tissue specificity in 149 brain cell types (level 2) and gene-level genetic association with anorexia nervosa using MAGMA. The sample size is 16,992 cases and 55,525 controls. The Bonferroni-corrected threshold (black vertical line) is $-\log_{10}(P) > 3.6$ and is based on tests across 53 tissues, 24 broad categories of cell types, and 149 KI level 2 cell types, a total of 226 tests.

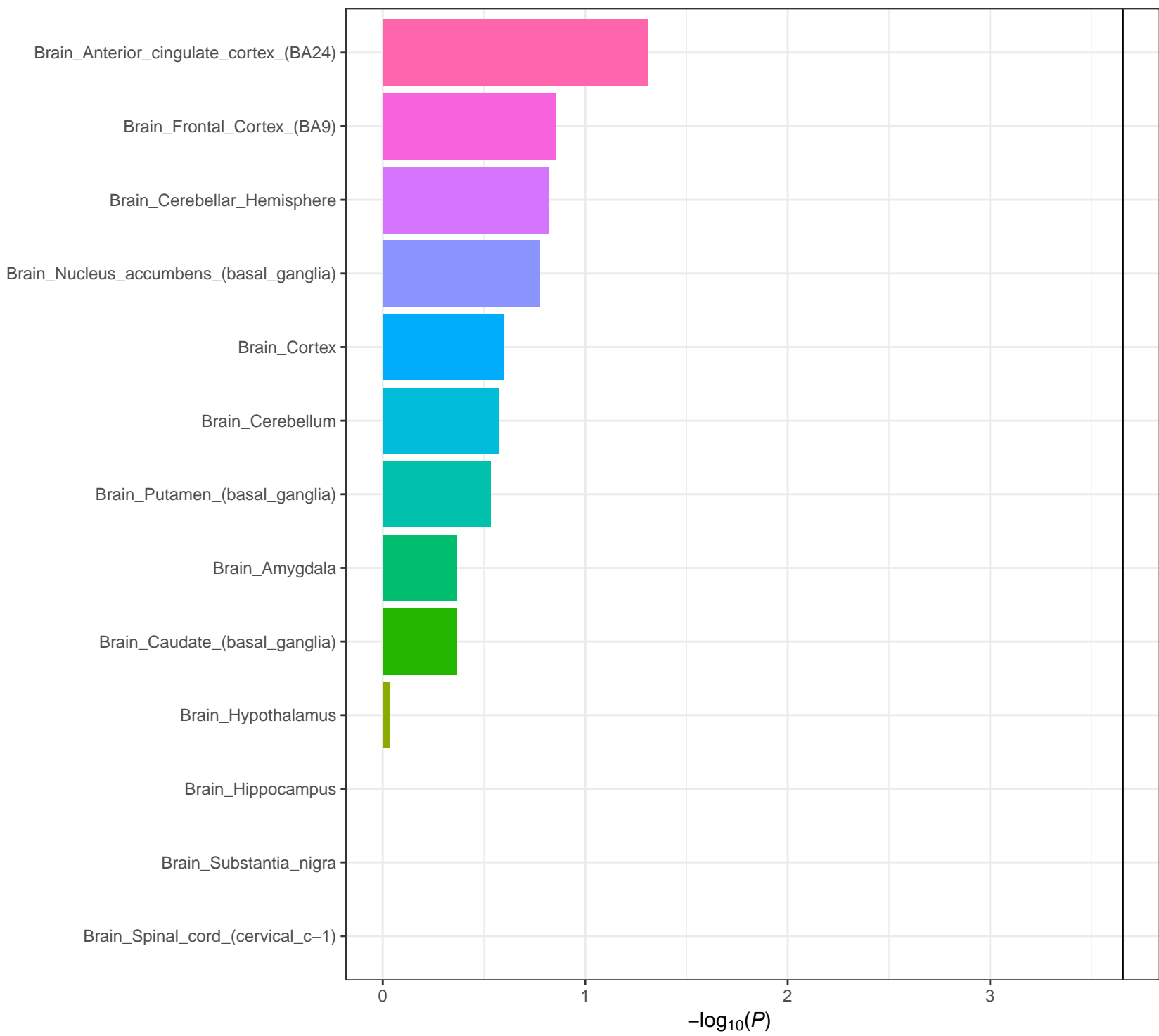


$-\log_{10}(P)$

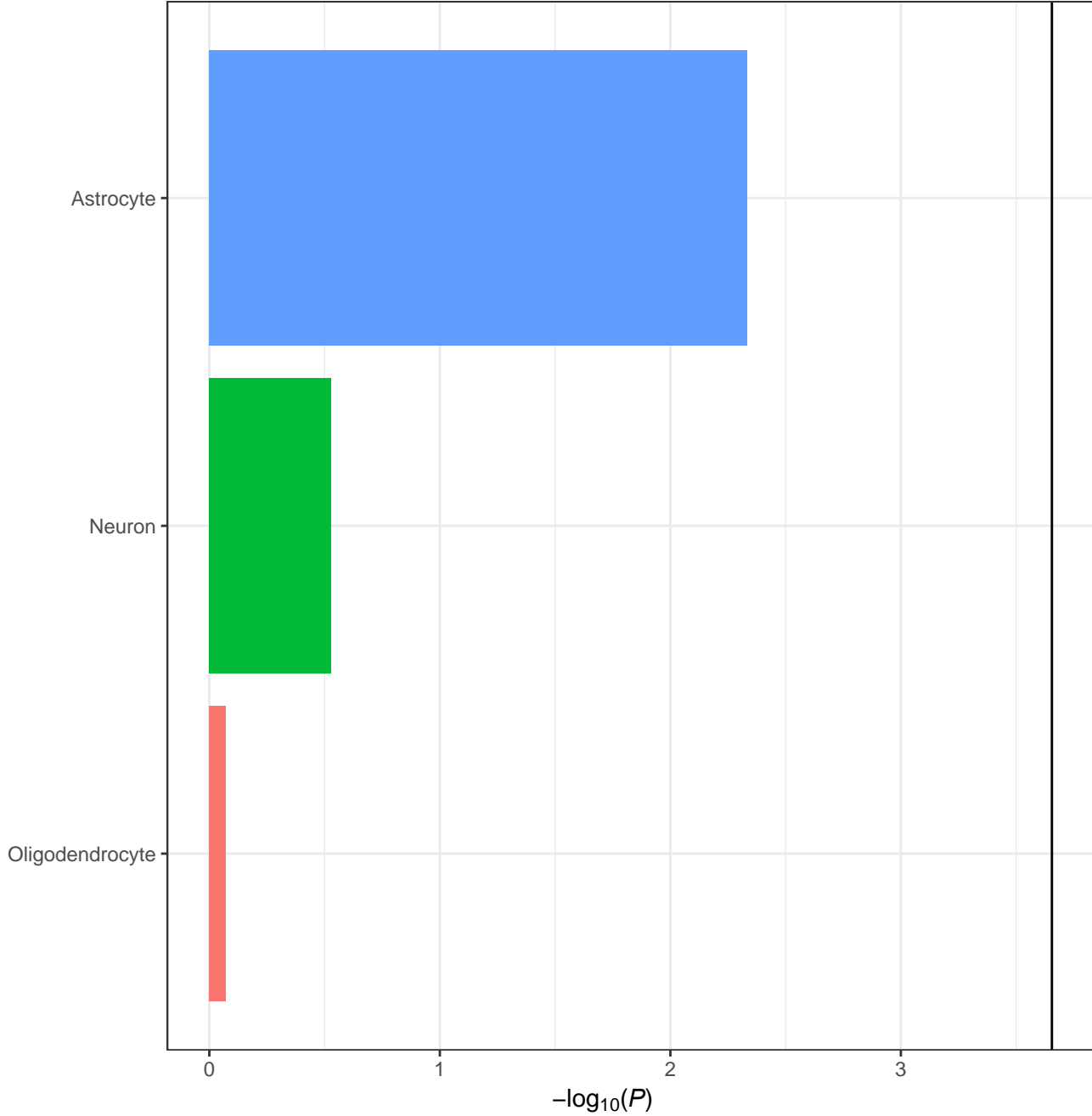
Supplementary Figure 12. P value of enrichment of heritability of anorexia nervosa in each tissue in GTEx using LD score regression applied to specifically expressed genes (LDSC-SEG). The sample size is 16,992 cases and 55,525 controls. The Bonferroni-corrected threshold (black vertical line) is $-\log_{10}(P) > 3.6$ and is based on tests across 53 tissues, 24 broad categories of cell types, and 149 KI level 2 cell types, a total of 226 tests.



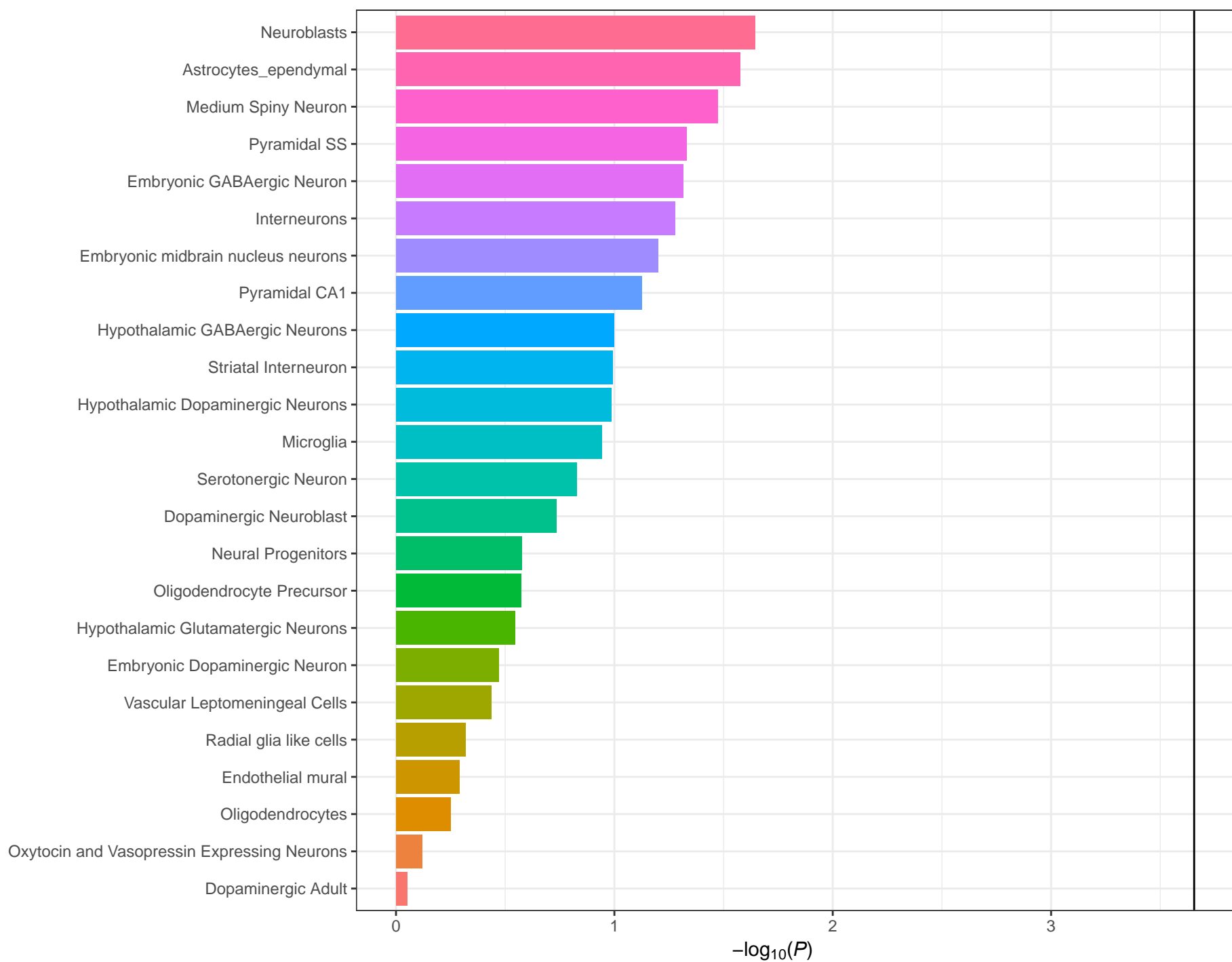
Supplementary Figure 13. P value of enrichment of heritability of anorexia nervosa in brain tissues in GTEx using LD score regression applied to specifically expressed genes (LDSC-SEG). The sample size is 16,992 cases and 55,525 controls. The Bonferroni-corrected threshold (black vertical line) is $-\log_{10}(P) > 3.6$ and is based on tests across 53 tissues, 24 broad categories of cell types, and 149 KI level 2 cell types, a total of 226 tests.



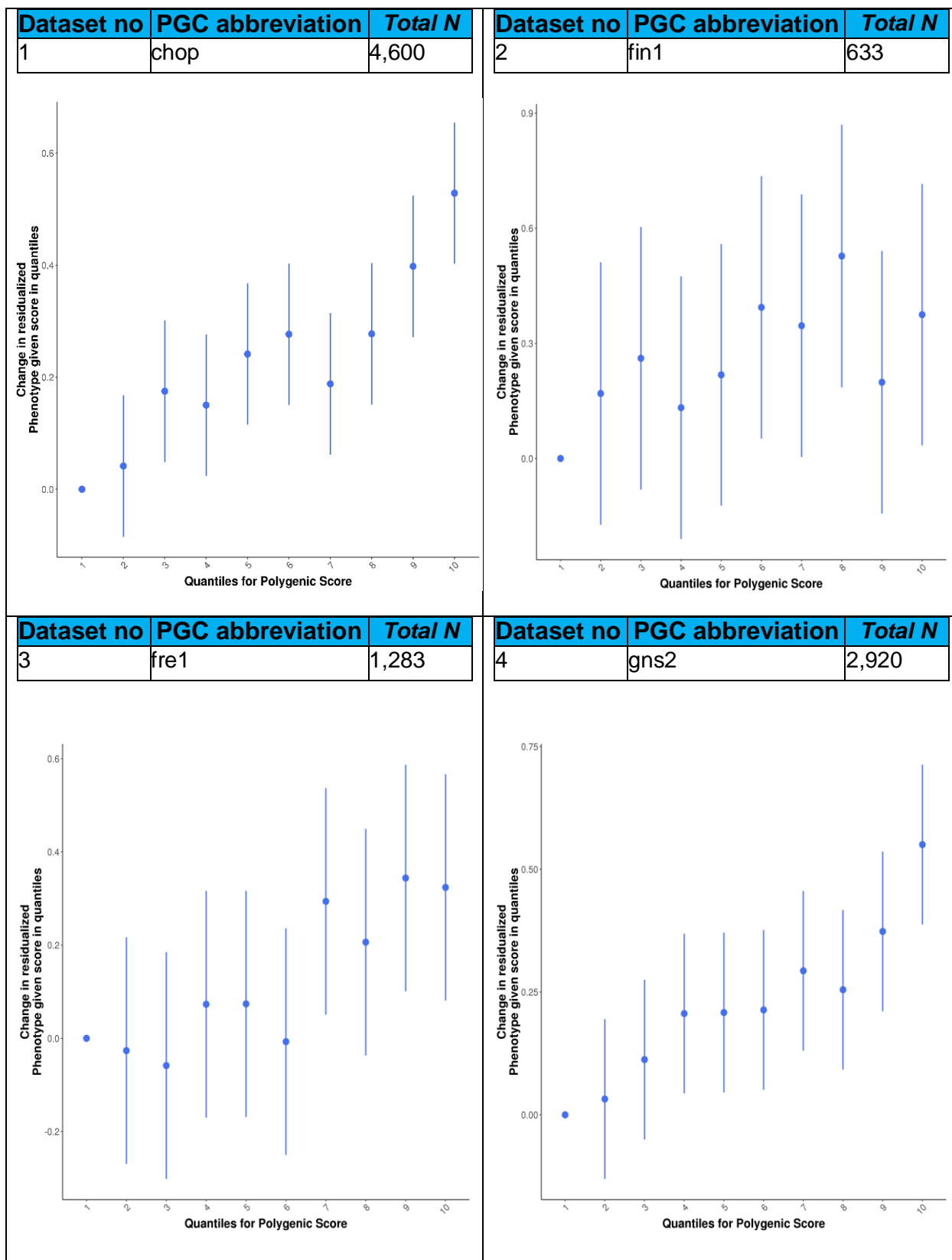
Supplementary Figure 14. *P* value of enrichment of heritability of anorexia nervosa in cell types in Cahoy database using LD score regression applied to specifically expressed genes (LDSC-SEG). The sample size is 16,992 cases and 55,525 controls. The Bonferroni-corrected threshold (black vertical line) is $-\log_{10}(P) > 3.6$ and is based on tests across 53 tissues, 24 broad categories of cell types, and 149 KI level 2 cell types, a total of 226 tests.



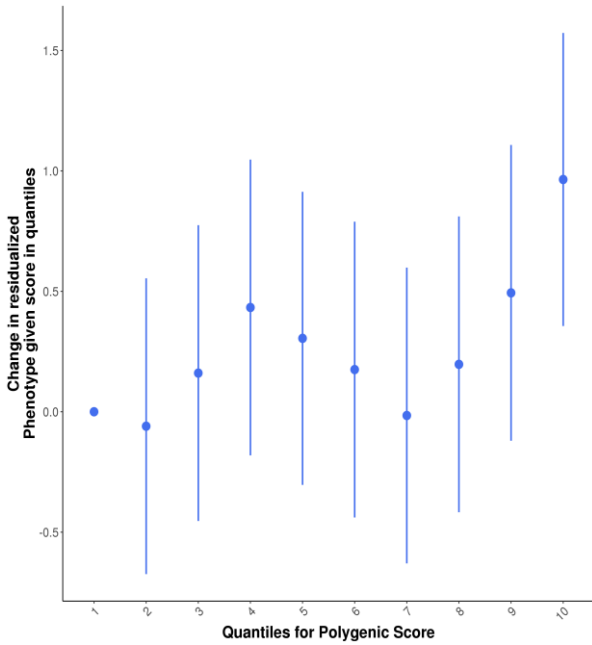
Supplementary Figure 15. P value of enrichment of heritability of anorexia nervosa in 24 brain cell types from the single-cell RNA-sequencing database (broad categories) using LD score regression applied to specifically expressed genes (LDSC-SEG). The sample size is 16,992 cases and 55,525 controls. The Bonferroni-corrected threshold (black vertical line) is $-\log_{10}(P) > 3.6$ and is based on tests across 53 tissues, 24 broad categories of cell types, and 149 KI level 2 cell types, a total of 226 tests.



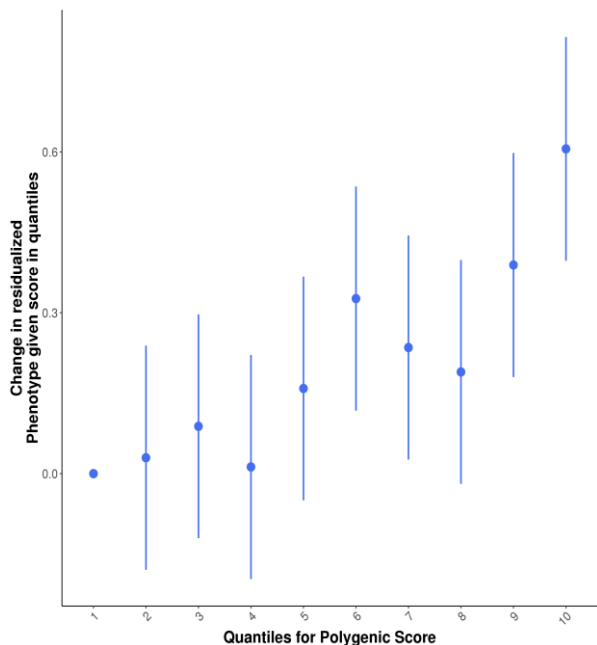
Supplementary Figure 16. Polygenic risk score (PRS) leave-one-out analysis: results for each cohort. PRS was constructed with the leave-one-out method from a GWAS with all datasets excluding the target dataset. Then, PRS was used to predict change in residualized phenotype score for anorexia nervosa (AN) risk in the target dataset (shown as the center value and an error bar which represents the 95% confidence interval of this estimate). The decile with the lowest PRS (i.e., subjects whose AN PRS is in the bottom 10%) serves as the referent. A higher residualized phenotype score indicates a higher risk of AN.



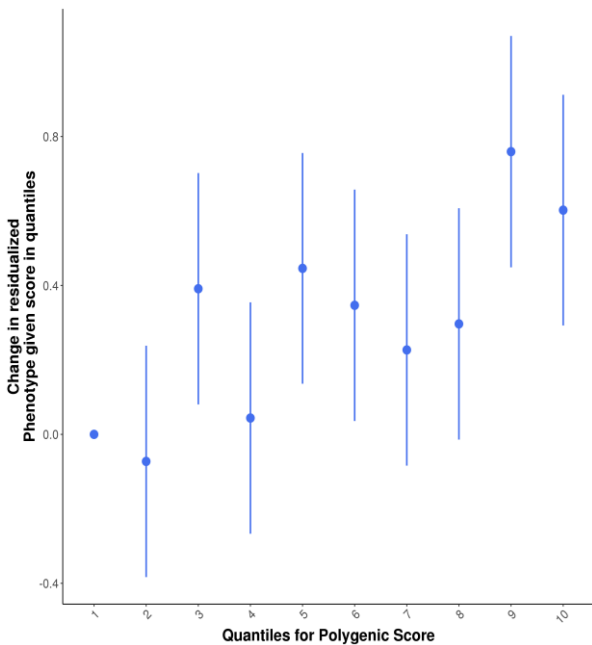
Dataset no	PGC abbreviation	Total N
5	itgr	273



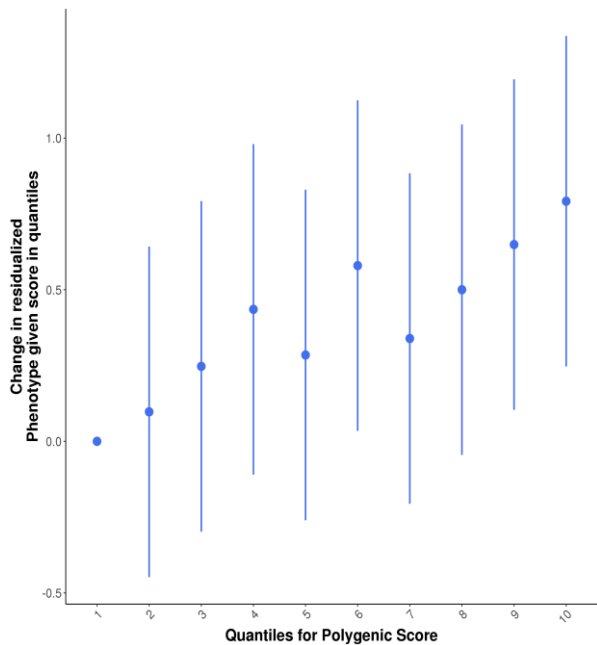
Dataset no	PGC abbreviation	Total N
6	net1	1,685



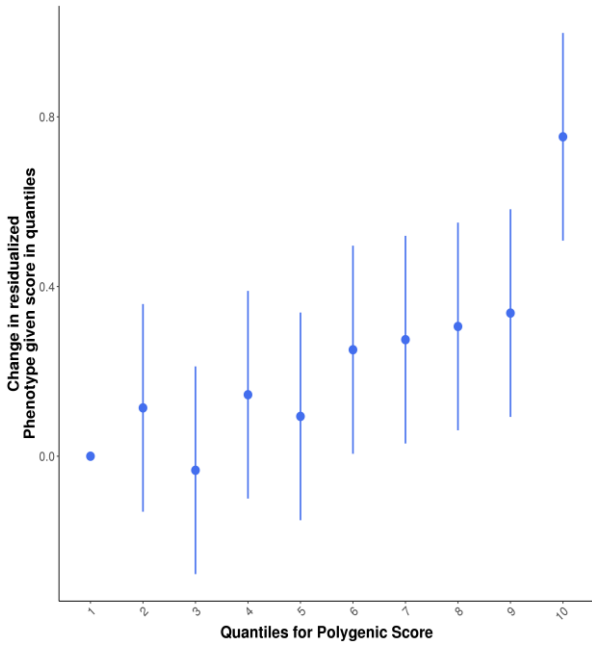
Dataset no	PGC abbreviation	Total N
7	poco	843



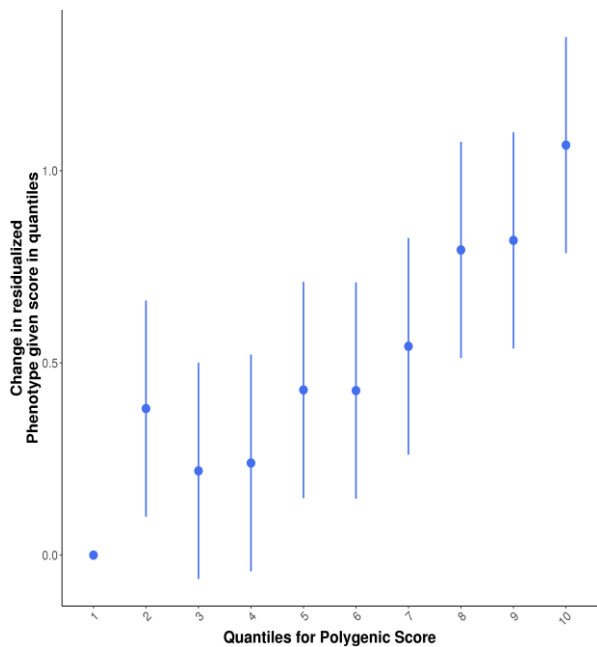
Dataset no	PGC abbreviation	Total N
8	spa1	340



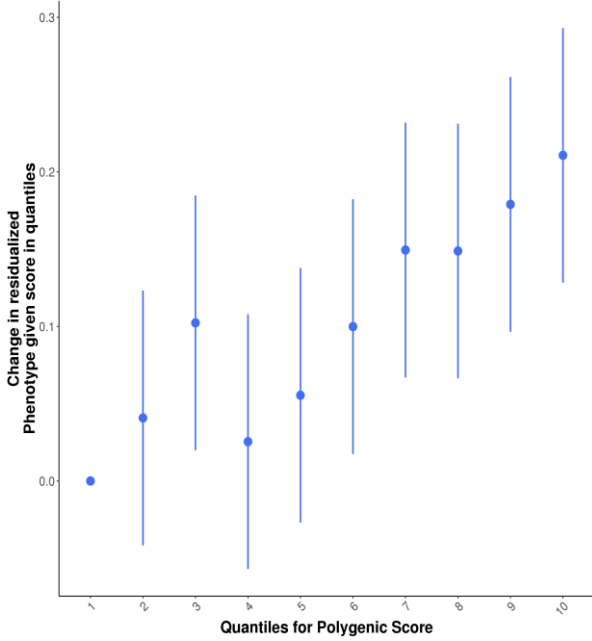
Dataset no	PGC abbreviation	Total N
9	ukd1	1,199



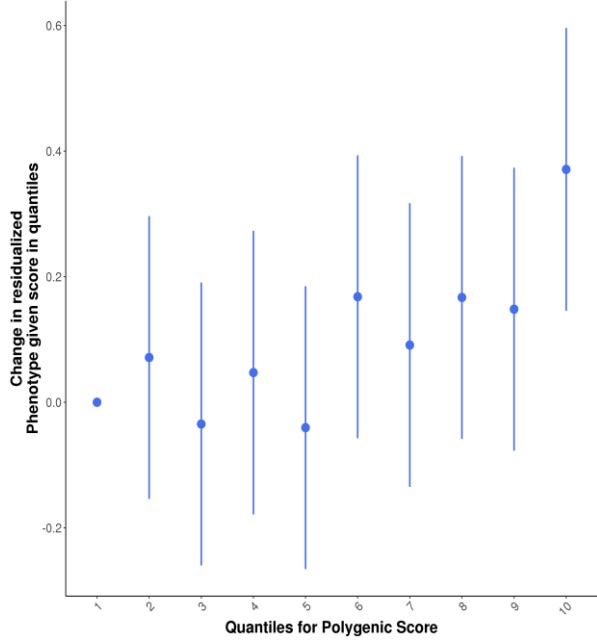
Dataset no	PGC abbreviation	Total N
10	usa1	1,168



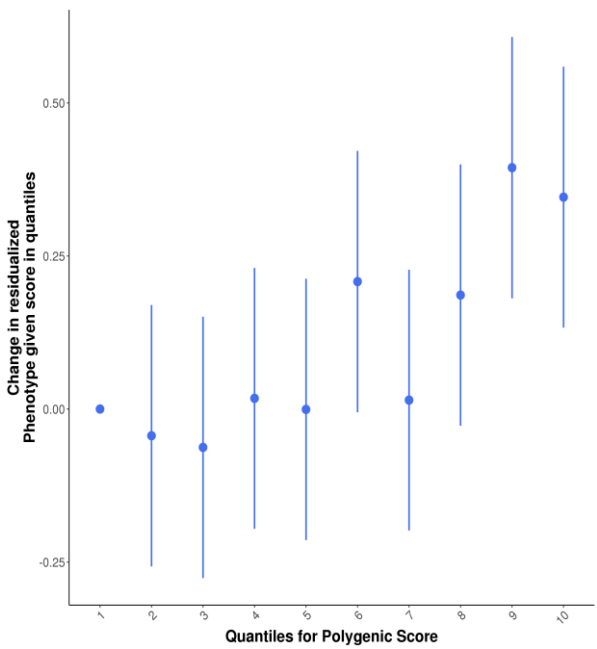
Dataset no	PGC abbreviation	Total N
11	w1to6	5,185



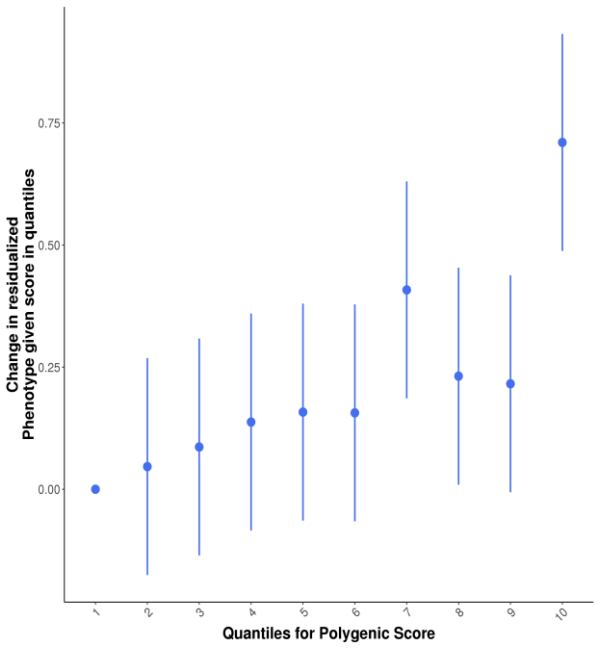
Dataset no	PGC abbreviation	Total N
12	w7	928



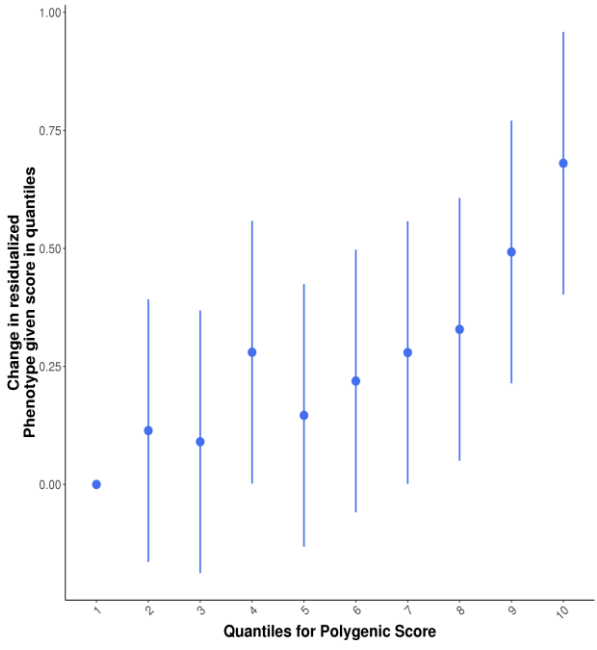
Dataset no	PGC abbreviation	Total N
13	w8	1,224



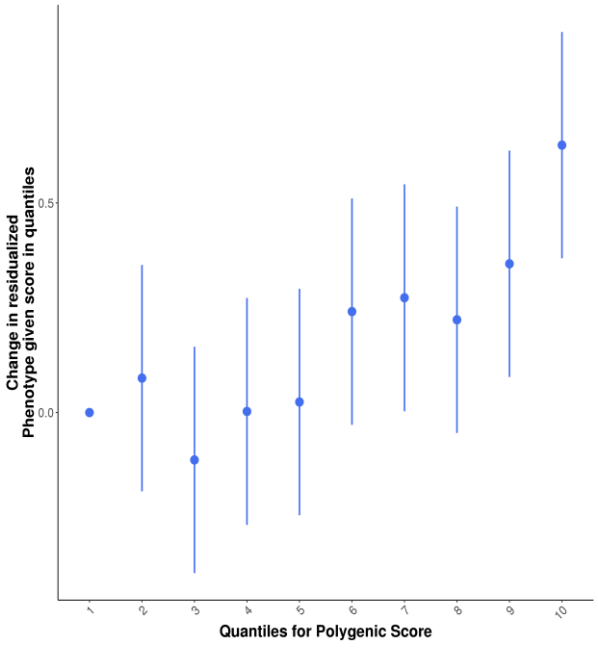
Dataset no	PGC abbreviation	Total N
14	w9	1,201



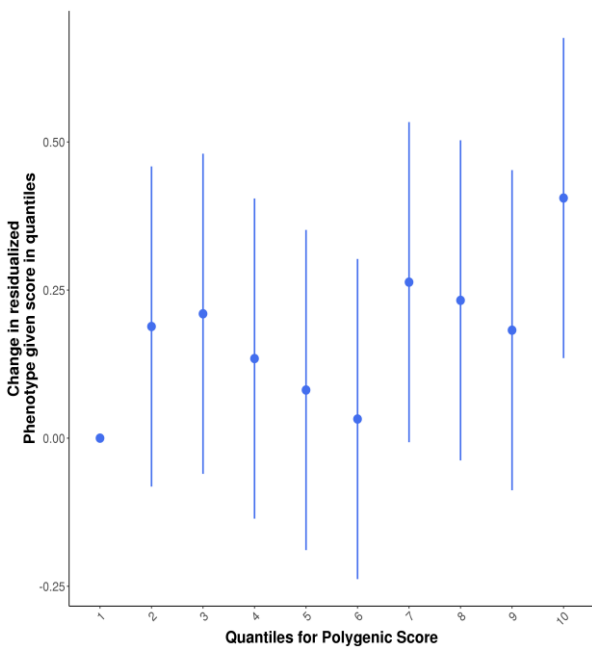
Dataset no	PGC abbreviation	Total N
15	w10	870



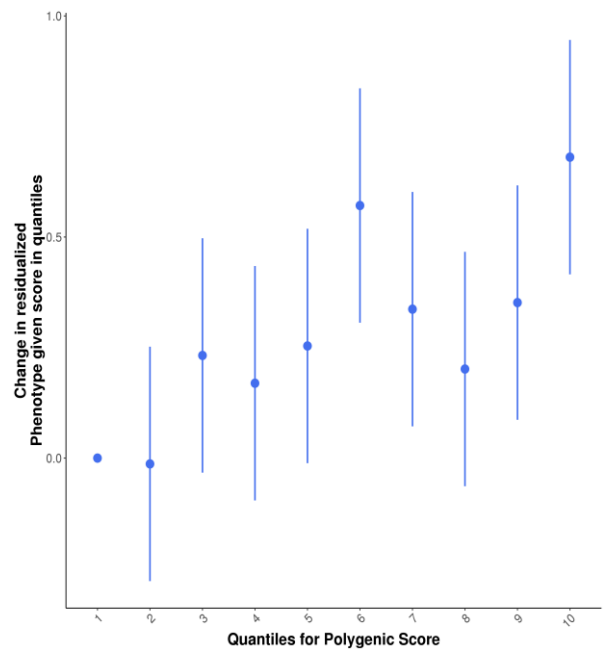
Dataset no	PGC abbreviation	Total N
16	w11	918



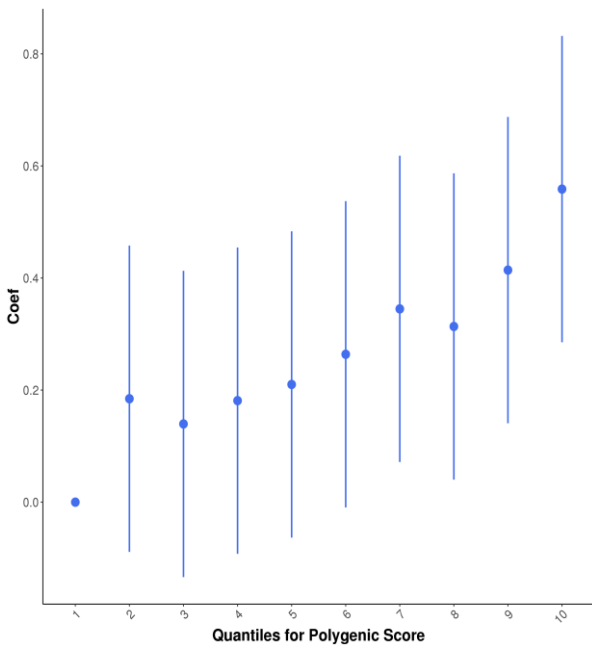
Dataset no	PGC abbreviation	Total N
17	w12	961



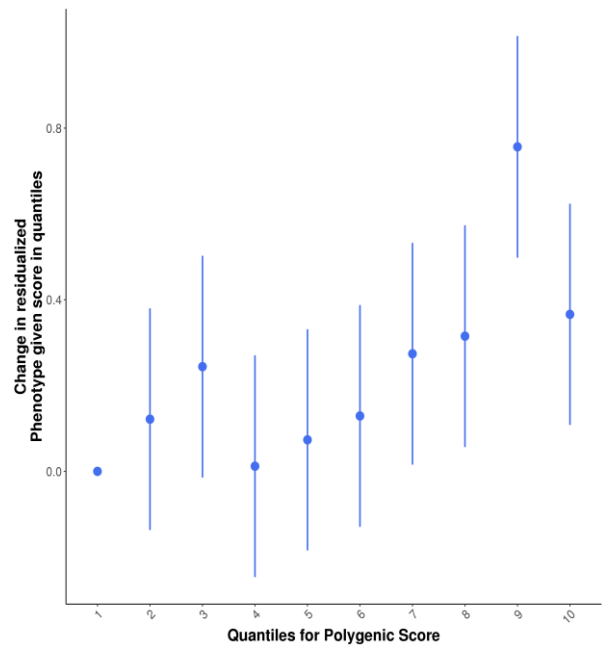
Dataset no	PGC abbreviation	Total N
18	w13	970



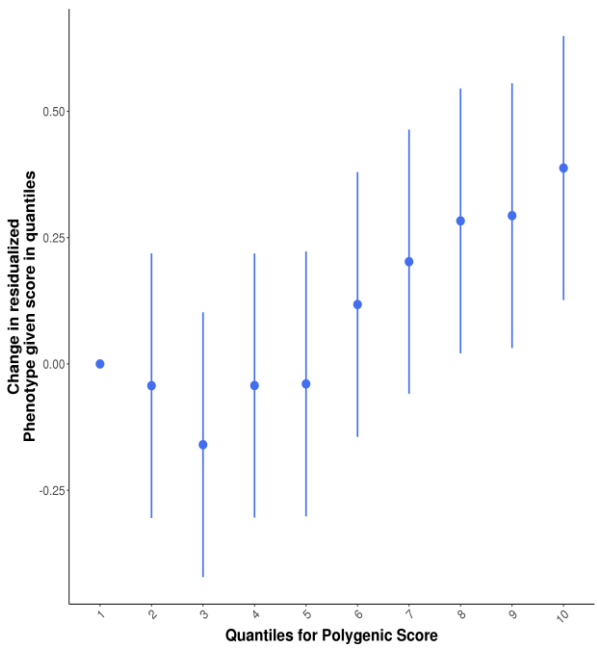
Dataset no	PGC abbreviation	Total N
19	w14	930



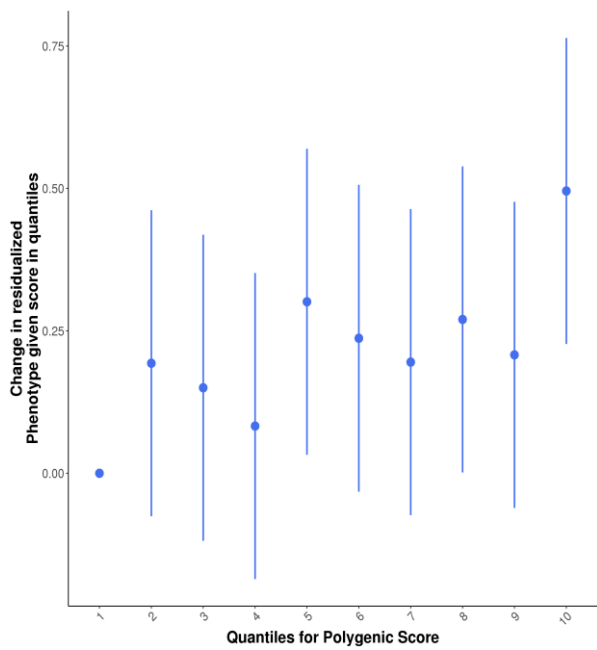
Dataset no	PGC abbreviation	Total N
20	w15	1,003



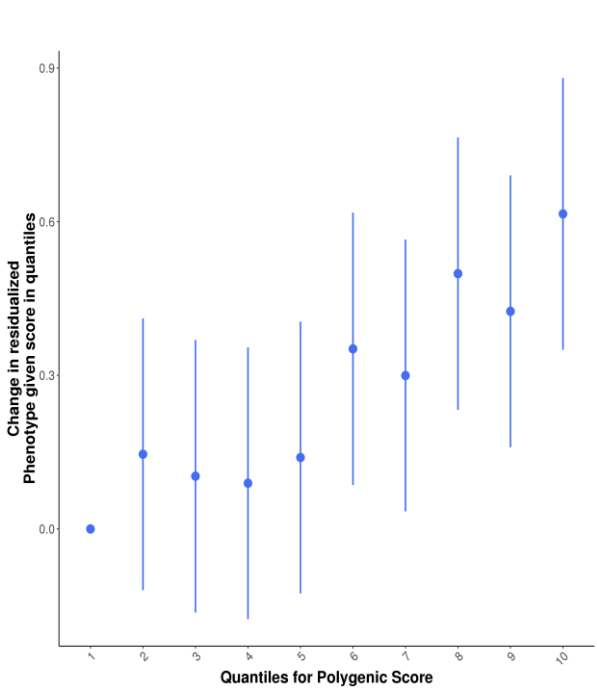
Dataset no	PGC abbreviation	Total N
21	w16	974



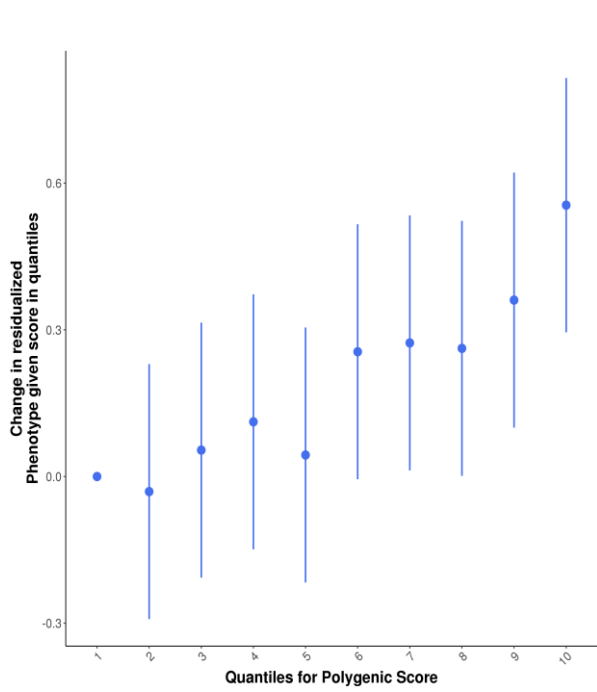
Dataset no	PGC abbreviation	Total N
22	w17	919



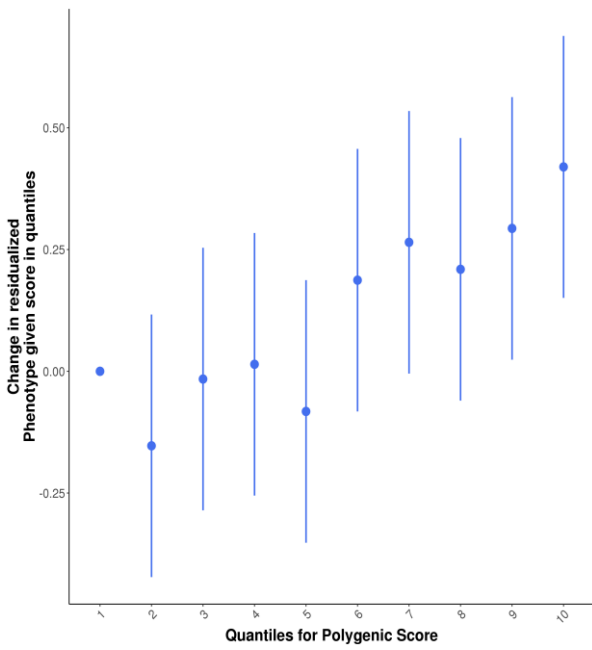
Dataset no	PGC abbreviation	Total N
23	w18	1,097



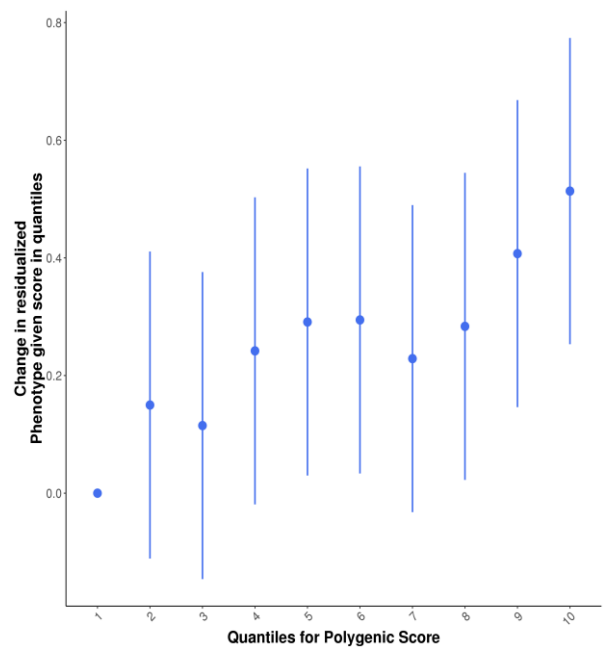
Dataset no	PGC abbreviation	Total N
24	w19	1,062



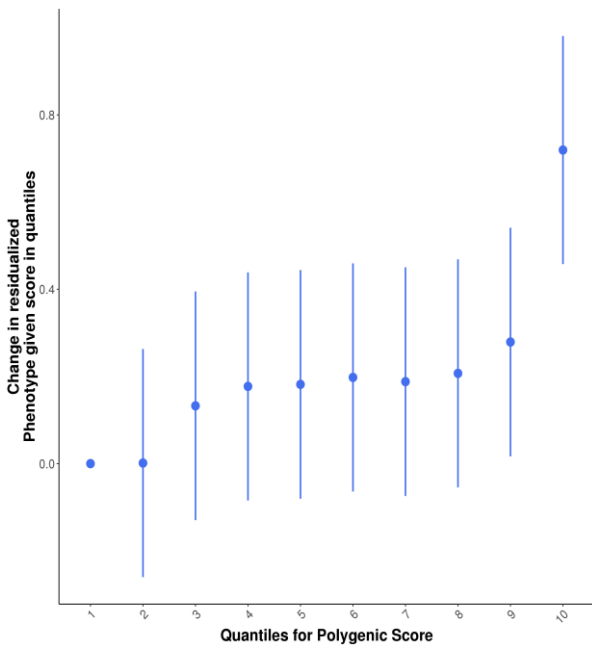
Dataset no	PGC abbreviation	Total N
25	w20	1,002



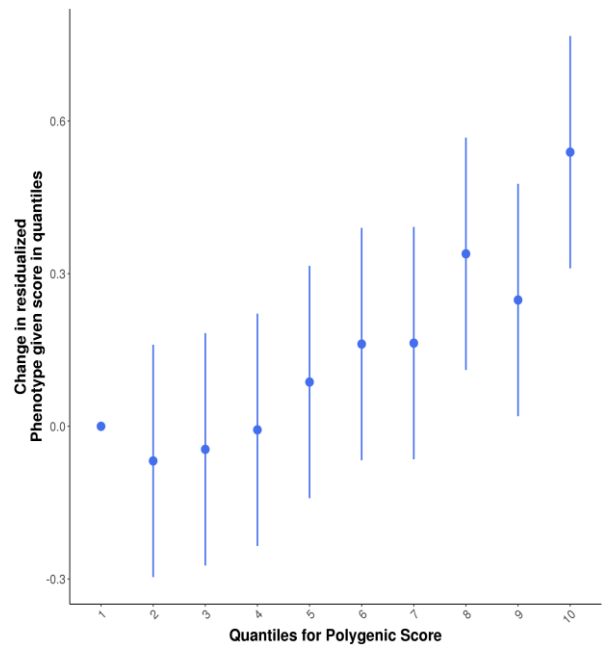
Dataset no	PGC abbreviation	Total N
26	w21	962



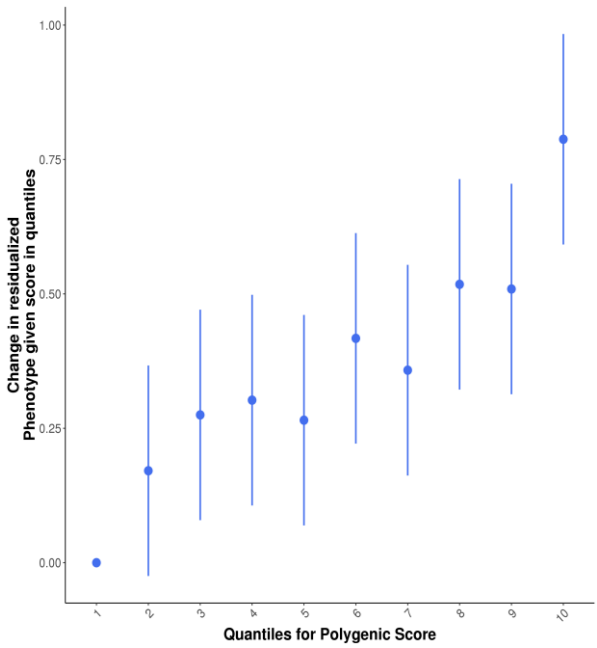
Dataset no	PGC abbreviation	Total N
27	w22	1,056



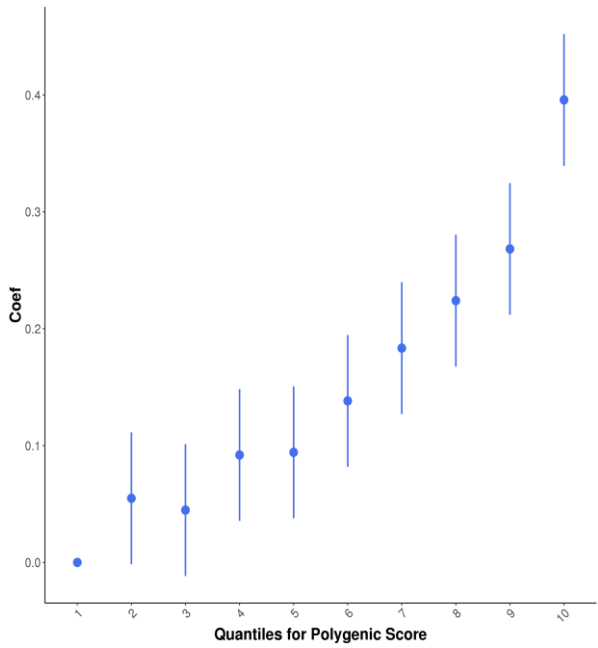
Dataset no	PGC abbreviation	Total N
28	w23	1,121



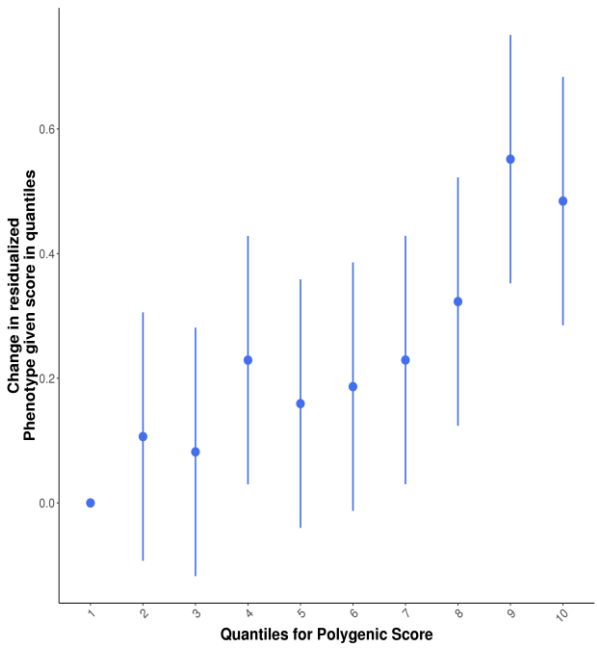
Dataset no	PGC abbreviation	Total N
29	w24	2,688



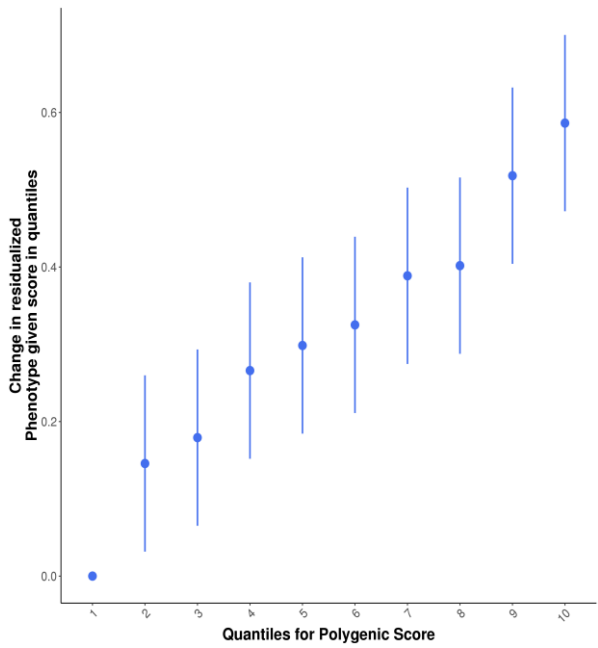
Dataset no	PGC abbreviation	Total N
30	aunz	18,160



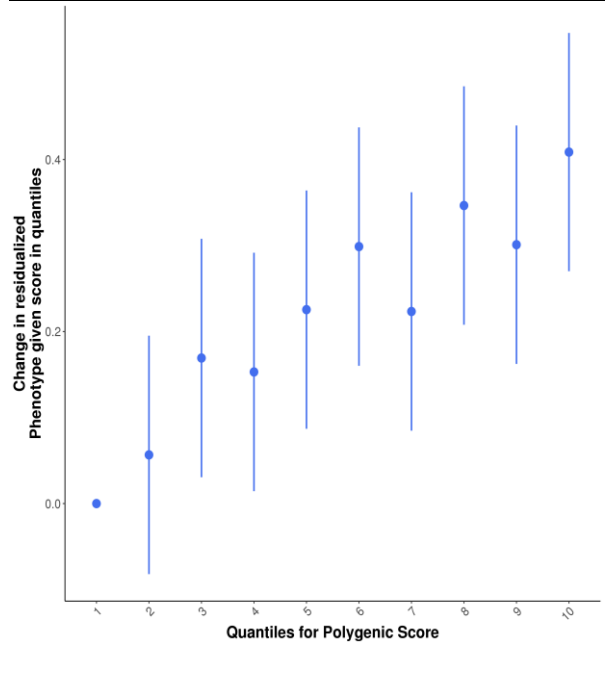
Dataset no	PGC abbreviation	Total N
31	usa2	2,611



Dataset no	PGC abbreviation	Total N
32	sedk	7,898



Dataset no	PGC abbreviation	Total N
33	ukbb	3,833



Watson, H. J., Yilmaz, Z., Thornton, L. M., Hübel, C., Coleman, J. R. I., Gaspar, H. A., . . . Bulik, C. M. (2019). Genome-wide association study identifies eight risk loci and implicates metabo-psychiatric origins for anorexia nervosa. Nature Genetics.

Supplementary Information

TABLE OF CONTENTS

Supplementary Note	3
Anorexia Nervosa Genetics Initiative (ANGI)	3
Eating Disorders Working Group of the Psychiatric Genomics Consortium	5
Additional Methods	11
Samples and study design	11
Eating Disorders Working Group of the Psychiatric Genomics Consortium (PGC-ED) Freeze 1	11
Genetic Consortium for Anorexia Nervosa/Wellcome Trust Case Control Consortium-3 (GCAN/WTCCC-3)	12
Anorexia Nervosa Genetics Initiative (ANGI)	12
UK Biobank	15
Merging of case and control data	15
Statistical power	16
Quality control and covariates	16
Anorexia nervosa subtype phenotypes	17
GWAS of related traits in UK Biobank	17
Genotyping, imputation and QC	17
GWAS of BMI, body fat percentage, fat mass, and fat free mass	18
GWAS of physical activity	18
GWAS of anxiety	19
GWAS of neuroticism	19
Gene expression	19
Predicted tissue-specific gene expression	21
A general note on multiple testing correction	21
Additional Results	22
Primary GWAS meta-analysis	22
Genomic inflation and residual confounding	22
Previous hit	23
Chromosome X	23
Female-only secondary GWAS	23
eQTL and Hi-C interactions	23
Multi-trait conditional and joint analysis	24
Clinical investigations	24

Anorexia nervosa subtype	24
Males with anorexia nervosa	25
Within-trait prediction: polygenic risk scoring	25
Gene-wise analysis	25
Pathway analysis	26
Tissue and cell type analyses	26
Predicted tissue-specific gene expression	27
Cross-trait analysis	27
Genetic correlations	27
Generalized summary data-based Mendelian randomization	27
References	29

Supplementary Note

Anorexia Nervosa Genetics Initiative

Jessica H Baker ¹	Anders Juréus ⁸	Claes Norring ^{4,5}
Andrew W Bergen ^{2,3}	Allan S Kaplan ^{21,22,23}	Catherine M Olsen ¹¹
Andreas Birgegård ^{4,5}	Walter Kaye ²⁴	Richard Parker ¹¹
Joseph M Boden ⁶	Martin Kennedy ²⁵	John F Pearson ³²
Harry Brandt ⁷	Katherine M Kirk ¹¹	Nancy L Pedersen ⁸
Cynthia M Bulik ^{1,8,9}	Mikael Landén ^{8,26}	Liselotte Petersen ^{27,28}
Steven Crawford ⁷	Janne T Larsen ^{13,27,28}	Michael Strober ^{33,34}
Laramie E Duncan ¹⁰	Virpi M Leppä ⁸	Patrick F Sullivan ^{1,8,35}
Scott Gordon ¹¹	Paul Lichtenstein ⁸	Laura M Thornton ¹
Jakob Grove ^{12,13,14,15}	Nicholas G Martin ¹¹	Tracey D Wade ³⁶
Katherine A Halmi ¹⁶	Manuel Mattheisen ^{4,5,12,29}	Hunna J Watson ^{1,37,38}
Anjali K Henders ¹⁷	James Mitchell ³⁰	Thomas Werge ³⁹
L. John Horwood ⁶	Grant W Montgomery ^{11,17,31}	David C Whiteman ¹¹
Craig Johnson ¹⁸	Preben Bo Mortensen ^{13,27,28}	D. Blake Woodside ^{22,23,40,41}
Jennifer Jordan ^{19,20}	Melissa A Munn-Chernoff ¹	Zeynep Yilmaz ^{1,35}

- 1 Department of Psychiatry, University of North Carolina at Chapel Hill, Chapel Hill, North Carolina, US
- 2 BioRealm, LLC, Walnut, California, US
- 3 Oregon Research Institute, Eugene, Oregon, US
- 4 Department of Clinical Neuroscience, Karolinska Institutet, Stockholm, Sweden
- 5 Center for Psychiatry Research, Stockholm Health Care Services, Stockholm City Council, Stockholm, Sweden
- 6 Christchurch Health and Development Study, University of Otago, Christchurch, New Zealand
- 7 The Center for Eating Disorders at Sheppard Pratt, Baltimore, Maryland, US
- 8 Department of Medical Epidemiology and Biostatistics, Karolinska Institutet, Stockholm, Sweden
- 9 Department of Nutrition, University of North Carolina at Chapel Hill, Chapel Hill, North Carolina, US
- 10 Department of Psychiatry and Behavioral Sciences, Stanford University, Stanford, California, US
- 11 QIMR Berghofer Medical Research Institute, Brisbane, Australia
- 12 Department of Biomedicine, Aarhus University, Aarhus, Denmark
- 13 The Lundbeck Foundation Initiative for Integrative Psychiatric Research (iPSYCH), Aarhus, Denmark
- 14 Centre for Integrative Sequencing, iSEQ, Aarhus University, Aarhus, Denmark
- 15 Bioinformatics Research Centre, Aarhus University, Aarhus, Denmark
- 16 Department of Psychiatry, Weill Cornell Medical College, New York, New York, US
- 17 Institute for Molecular Bioscience, University of Queensland, Brisbane, Australia
- 18 Eating Recovery Center, Denver, Colorado, US
- 19 Department of Psychological Medicine, University of Otago, Christchurch, New Zealand
- 20 Canterbury District Health Board, Christchurch, New Zealand
- 21 Centre for Addiction and Mental Health, Toronto, Canada
- 22 Institute of Medical Science, University of Toronto, Toronto, Canada
- 23 Department of Psychiatry, University of Toronto, Toronto, Canada
- 24 Department of Psychiatry, University of California San Diego, San Diego, California, US
- 25 Department of Pathology and Biomedical Science, University of Otago, Christchurch, New Zealand
- 26 Department of Psychiatry and Neurochemistry, Institute of Neuroscience and Physiology, Sahlgrenska Academy at the University of Gothenburg, Gothenburg, Sweden
- 27 National Centre for Register-Based Research, Aarhus BSS, Aarhus University, Aarhus, Denmark
- 28 Centre for Integrated Register-based Research (CIRRAU), Aarhus University, Aarhus, Denmark
- 29 Department of Psychiatry, Psychosomatics and Psychotherapy, University of Würzburg, Würzburg, Germany
- 30 Department of Psychiatry and Behavioral Science, University of North Dakota School of Medicine and Health Sciences, Fargo, North Dakota, US
- 31 Queensland Brain Institute, University of Queensland, Brisbane, Australia
- 32 Biostatistics and Computational Biology Unit, University of Otago, Christchurch, New Zealand

- 33 Department of Psychiatry and Biobehavioral Science, Semel Institute for Neuroscience and Human Behavior, University of California Los Angeles, Los Angeles, California, US
- 34 David Geffen School of Medicine, University of California Los Angeles, Los Angeles, California, US
- 35 Department of Genetics, University of North Carolina at Chapel Hill, Chapel Hill, North Carolina, US
- 36 School of Psychology, Flinders University, Adelaide, Australia
- 37 School of Psychology, Curtin University, Perth, Australia
- 38 School of Paediatrics and Child Health, University of Western Australia, Perth, Australia
- 39 Department of Clinical Medicine, University of Copenhagen, Copenhagen, Denmark
- 40 Centre for Mental Health, University Health Network, Toronto, Canada
- 41 Program for Eating Disorders, University Health Network, Toronto, Canada

Eating Disorders Working Group of the Psychiatric Genomics Consortium

Roger AH Adan^{1,2,3}
Lars Alfredsson⁴
Tetsuya Ando⁵
Ole A Andreassen⁶
Jessica H Baker⁷
Andrew W Bergen^{8,9}
Wade H Berrettini¹⁰
Andreas Birgegård^{11,12}
Joseph M Boden¹³
Ilka Boehm¹⁴
Claudette Boni¹⁵
Vesna Boraska Perica^{16,17}
Harry Brandt¹⁸
Gerome Breen^{19,20}
Julien Bryois²¹
Katharina Buehren²²
Cynthia M Bulik^{7,21,23}
Roland Burghardt²⁴
Matteo Cassina²⁵
Sven Cichon²⁶
Maurizio Clementi²⁵
Jonathan RI Coleman^{19,20}
Roger D Cone²⁷
Philippe Courtet²⁸
Steven Crawford¹⁸
Scott Crow²⁹
James J Crowley^{11,30}
Unna N Danner²
Oliver SP Davis^{31,32}
Martina de Zwaan³³
George Dedoussis³⁴
Daniela Degortes³⁵
Janiece E DeSocio³⁶
Danielle M Dick³⁷
Dimitris Dikeos³⁸
Christian Dina³⁹
Monika Dmitrzak-Weglarz⁴⁰
Elisa Docampo^{41,42,43}
Laramie E Duncan⁴⁴
Karin Egberts⁴⁵
Stefan Ehrlich¹⁴
Geòrgia Escaramís^{41,42,43}
Tõnu Esko^{46,47}
Xavier Estivill^{41,42,43,48}
Anne Farmer¹⁹
Angela Favaro³⁵
Fernando Fernández-Aranda^{49,50}
Manfred M Fichter^{51,52}
Krista Fischer⁴⁶
Manuel Föcker⁵³
Lenka Foretova⁵⁴
Andreas J Forstner^{26,55,56,57,58}
Monica Forzan²⁵
Christopher S Franklin¹⁶
Steven Gallinger⁵⁹
Hélène A Gaspar^{19,20}
Ina Giegling⁶⁰
Johanna Giuranna⁵³
Paola Giusti-Rodríguez³⁰
Fragiskos Gonidakis⁶¹
Scott Gordon⁶²
Philip Gorwood^{15,63}
Monica Gratacos Mayora^{41,42,43}
Jakob Grove^{64,65,66,67}
Sébastien Guillaume²⁸
Yiran Guo⁶⁸
Hakon Hakonarson^{68,69}
Katherine A Halmi⁷⁰
Ken B Hanscombe⁷¹
Konstantinos Hatzikotoulas^{16,72}
Joanna Hauser⁷³
Johannes Hebebrand⁵³
Sietske G Helder^{19,74}
Stefan Herms^{26,56,58}
Beate Herpertz-Dahlmann²²
Wolfgang Herzog⁷⁵
Anke Hinney⁵³
L. John Horwood¹³
Christopher Hübel^{19,21}
Laura M Huckins^{16,76}
James I Hudson⁷⁷
Hartmut Imgart⁷⁸
Hidetoshi Inoko⁷⁹
Vladimir Janout⁸⁰
Susana Jiménez-Murcia^{49,50}
Craig Johnson⁸¹
Jennifer Jordan^{82,83}
Antonio Julià⁸⁴
Gursharan Kalsi¹⁹
Deborah Kaminska⁸⁵
Allan S Kaplan^{86,87,88}
Jaakko Kaprio^{89,90}
Leila Karhunen⁹¹
Andreas Karwautz⁹²
Martien JH Kas^{1,93}
Walter H Kaye⁹⁴
James L Kennedy^{86,87,88}
Martin Kennedy⁹⁵
Anna Keski-Rahkonen⁸⁹
Kirsty Kiezebrink⁹⁶
Youl-Ri Kim⁹⁷
Lars Klareskog⁹⁸
Kelly L Klump⁹⁹
Gun Peggy S Knudsen¹⁰⁰
Maria C La Via⁷
Mikael Landén^{21,101}
Janne T Larsen^{65,102,103}
Stephanie Le Hellard^{104,105,106}
Virpi M Leppä²¹
Robert D Levitan^{86,87,88}
Dong Li⁶⁸
Paul Lichtenstein²¹
Lisa Lilienfeld¹⁰⁷
Bochao Danae Lin¹
Jolanta Lissowska¹⁰⁸
Jurjen Luyckx¹
Pierre J Magistretti^{109,110}
Mario Maj¹¹¹
Katrin Mannik^{46,112}
Sara Marsal⁸⁴
Christian R Marshall¹¹³
Nicholas G Martin⁶²
Manuel Mattheisen^{11,12,64,114}
Morten Mattingsdal⁶
Sara McDevitt^{115,116}
Peter McGuffin¹⁹
Sarah E Medland⁶²
Andres Metspalu^{46,117}
Ingrid Meulenbelt¹¹⁸
Nadia Micali^{119,120,121}
James Mitchell¹²²
Karen Mitchell¹²³
Alessio Maria Monteleone¹¹¹
Palmiero Monteleone¹²⁴
Preben Bo Mortensen^{65,102,103}
Melissa A Munn-Chernoff⁷
Benedetta Nacmias¹²⁵
Marie Navratilova⁵⁴
Claes Norring^{11,12}
Ioanna Ntalla³⁴
Catherine M Olsen⁶²
Roel A Ophoff^{1,126}
Julie K O'Toole¹²⁷
Leonid Padyukov⁹⁸
Aarno Palotie^{47,90,128}
Jacques Pantel¹⁵
Hana Papezova⁸⁵
John F Pearson¹²⁹
Nancy L Pedersen²¹
Liselotte Petersen^{65,102,103}
Dalila Pinto⁷⁶
Kirstin L Purves¹⁹
Raquel Rabionet^{130,131,132}
Anu Raevuori⁸⁹
Nicolas Ramoz¹⁵
Ted Reichborn-Kjennerud^{100,133}
Valdo Ricca^{125,134}
Samuli Ripatti^{47,89,135}

Stephan Ripke^{136,137,138}
Franziska Ritschel^{14,139}
Marion Roberts^{19,82,140}
Alessandro Rotondo¹⁴¹
Dan Rujescu^{51,60}
Filip Rybakowski¹⁴²
Paolo Santonastaso¹⁴³
André Scherag¹⁴⁴
Stephen W Scherer¹⁴⁵
Ulrike Schmidt^{20,146}
Nicholas J Schork¹⁴⁷
Alexandra Schosser¹⁴⁸
Jochen Seitz²²
Lenka Slachtova¹⁴⁹
P. Eline Slagboom¹¹⁸
Margarita CT Slof-Op ‘t
Landt^{150,151}

Agnieszka Slopień¹⁵²
Sandro Sorbi^{125,153}
Michael Strober^{154,155}
Garret D Stuber^{7,156}
Patrick F Sullivan^{7,30,21}
Beata Świątkowska¹⁵⁷
Jin P Szatkiewicz³⁰
Ioanna Tachmazidou¹⁶
Elena Tenconi³⁵
Laura M Thornton⁷
Alfonso Tortorella^{158,159}
Federica Tozzi¹⁶⁰
Janet Treasure^{20,146}
Artemis Tsitsika¹⁶¹
Marta Tyszkiewicz-Nwafor¹⁵²
Konstantinos Tziouvas¹⁶²
Annemarie A van Elburg^{2,163}

Eric F van Furth^{150,151}
Gudrun Wagner⁹²
Esther Walton¹⁴
Hunna J Watson^{7,164,165}
Thomas Werge¹⁶⁶
David C Whiteman⁶²
Elisabeth Widen⁹⁰
D. Blake Woodside^{87,88,167,168}
Shuyang Yao²¹
Zeynep Yilmaz^{7,30}
Eleftheria Zeggini^{16,72}
Stephanie Zerwas⁷
Stephan Zipfel¹⁶⁹

1 Brain Center Rudolf Magnus, Department of Translational Neuroscience, University Medical Center Utrecht, Utrecht, The Netherlands

2 Center for Eating Disorders Rintveld, Altrecht Mental Health Institute, Zeist, The Netherlands

3 Institute of Neuroscience and Physiology, Sahlgrenska Academy at the University of Gothenburg, Gothenburg, Sweden

4 Institute of Environmental Medicine, Karolinska Institutet, Stockholm, Sweden

5 Department of Behavioral Medicine, National Institute of Mental Health, National Center of Neurology and Psychiatry, Tokyo, Japan

6 NORMENT KG Jebsen Centre, Division of Mental Health and Addiction, University of Oslo, Oslo University Hospital, Oslo, Norway

7 Department of Psychiatry, University of North Carolina at Chapel Hill, Chapel Hill, North Carolina, US

8 BioRealm, LLC, Walnut, California, US

9 Oregon Research Institute, Eugene, Oregon, US

10 Department of Psychiatry, Center for Neurobiology and Behavior, Perelman School of Medicine at the University of Pennsylvania, Philadelphia, Pennsylvania, US

11 Department of Clinical Neuroscience, Karolinska Institutet, Stockholm, Sweden

12 Center for Psychiatry Research, Stockholm Health Care Services, Stockholm City Council, Stockholm, Sweden

13 Christchurch Health and Development Study, University of Otago, Christchurch, New Zealand

14 Division of Psychological and Social Medicine and Developmental Neurosciences, Faculty of Medicine, Technische Universität Dresden, Dresden, Germany

15 INSERM 1266, Institute of Psychiatry and Neuroscience of Paris, Paris, France

16 Wellcome Sanger Institute, Wellcome Genome Campus, Hinxton, Cambridge, UK

17 Department of Medical Biology, School of Medicine, University of Split, Split, Croatia

18 The Center for Eating Disorders at Sheppard Pratt, Baltimore, Maryland, US

19 Institute of Psychiatry, Psychology and Neuroscience, Social, Genetic and Developmental Psychiatry (SGDP) Centre, King’s College London, London, UK

20 National Institute for Health Research Biomedical Research Centre, King’s College London and South London and Maudsley National Health Service Foundation Trust, London, UK

21 Department of Medical Epidemiology and Biostatistics, Karolinska Institutet, Stockholm, Sweden

22 Department of Child and Adolescent Psychiatry, Psychosomatics and Psychotherapy, RWTH Aachen University, Aachen, Germany

23 Department of Nutrition, University of North Carolina at Chapel Hill, Chapel Hill, North Carolina, US

24 Department of Child and Adolescent Psychiatry, Klinikum Frankfurt/Oder, Frankfurt, Germany

25 Clinical Genetics Unit, Department of Woman and Child Health, University of Padova, Padova, Italy

26 Institute of Medical Genetics and Pathology, University Hospital Basel, Basel, Switzerland

27 Life Sciences Institute and Department of Molecular and Integrative Physiology, University of Michigan, Ann Arbor, Michigan, US

- 28 Department of Emergency Psychiatry and Post-Acute Care, CHRU Montpellier, University of Montpellier, Montpellier, France
- 29 Department of Psychiatry, University of Minnesota, Minneapolis, Minnesota, US
- 30 Department of Genetics, University of North Carolina at Chapel Hill, Chapel Hill, North Carolina, US
- 31 MRC Integrative Epidemiology Unit, University of Bristol, Bristol, UK
- 32 School of Social and Community Medicine, University of Bristol, Bristol, UK
- 33 Department of Psychosomatic Medicine and Psychotherapy, Hannover Medical School, Hannover, Germany
- 34 Department of Nutrition and Dietetics, Harokopio University, Athens, Greece
- 35 Department of Neurosciences, University of Padova, Padova, Italy
- 36 College of Nursing, Seattle University, Seattle, Washington, US
- 37 Department of Psychology, Virginia Commonwealth University, Richmond, Virginia, US
- 38 Department of Psychiatry, Athens University Medical School, Athens University, Athens, Greece
- 39 l'institut du thorax, INSERM, CNRS, UNIV Nantes, CHU Nantes, Nantes, France
- 40 Department of Psychiatric Genetics, Poznan University of Medical Sciences, Poznan, Poland
- 41 Barcelona Institute of Science and Technology, Barcelona, Spain
- 42 Universitat Pompeu Fabra, Barcelona, Spain
- 43 Centro de Investigación Biomédica en Red en Epidemiología y Salud Pública (CIBERESP), Barcelona, Spain
- 44 Department of Psychiatry and Behavioral Sciences, Stanford University, Stanford, California, US
- 45 Department of Child and Adolescent Psychiatry, Psychosomatics and Psychotherapy, University Hospital of Würzburg, Centre for Mental Health, Würzburg, Germany
- 46 Estonian Genome Center, University of Tartu, Tartu, Estonia
- 47 Program in Medical and Population Genetics, Broad Institute of the Massachusetts Institute of Technology and Harvard University, Cambridge, Massachusetts, US
- 48 Genomics and Disease, Bioinformatics and Genomics Programme, Centre for Genomic Regulation, Barcelona, Spain
- 49 Department of Psychiatry, University Hospital of Bellvitge –IDIBELL and CIBERobn, Barcelona, Spain
- 50 Department of Clinical Sciences, School of Medicine, University of Barcelona, Barcelona, Spain
- 51 Department of Psychiatry and Psychotherapy, Ludwig-Maximilians-University (LMU), Munich, Germany
- 52 Schön Klinik Roseneck affiliated with the Medical Faculty of the University of Munich (LMU), Munich, Germany
- 53 Department of Child and Adolescent Psychiatry, University Hospital Essen, University of Duisburg-Essen, Essen, Germany
- 54 Department of Cancer, Epidemiology and Genetics, Masaryk Memorial Cancer Institute, Brno, Czech Republic
- 55 Institute of Human Genetics, University of Bonn, School of Medicine & University Hospital Bonn, Bonn, Germany
- 56 Department of Genomics, Life and Brain Center, University of Bonn, Bonn, Germany
- 57 Department of Psychiatry (UPK), University of Basel, Basel, Switzerland
- 58 Department of Biomedicine, University of Basel, Basel, Switzerland
- 59 Department of Surgery, Faculty of Medicine, University of Toronto, Toronto, Canada
- 60 Department of Psychiatry, Psychotherapy and Psychosomatics, Martin Luther University of Halle-Wittenberg, Halle, Germany
- 61 1st Psychiatric Department, National and Kapodistrian University of Athens, Medical School, Eginition Hospital, Athens, Greece
- 62 QIMR Berghofer Medical Research Institute, Brisbane, Australia
- 63 CMME, hôpital Sainte-Anne (GHU Paris Psychiatrie et Neurosciences), Paris Descartes University, Paris, France
- 64 Department of Biomedicine, Aarhus University, Aarhus, Denmark
- 65 The Lundbeck Foundation Initiative for Integrative Psychiatric Research (iPSYCH), Aarhus, Denmark
- 66 Centre for Integrative Sequencing, iSEQ, Aarhus University, Aarhus, Denmark
- 67 Bioinformatics Research Centre, Aarhus University, Aarhus, Denmark
- 68 Center for Applied Genomics, Children's Hospital of Philadelphia, Philadelphia, Pennsylvania, US
- 69 Department of Pediatrics, Perelman School of Medicine, University of Pennsylvania, Philadelphia, Pennsylvania, US
- 70 Department of Psychiatry, Weill Cornell Medical College, New York, New York, US
- 71 Department of Medical and Molecular Genetics, King's College London, Guy's Hospital, London, UK
- 72 Institute of Translational Genomics, Helmholtz Zentrum München, Neuherberg, Germany

- 73 Department of Adult Psychiatry, Poznan University of Medical Sciences, Poznan, Poland
- 74 Zorg op Orde, Leidschendam, The Netherlands
- 75 Department of General Internal Medicine and Psychosomatics, Heidelberg University Hospital, Heidelberg University, Heidelberg, Germany
- 76 Department of Psychiatry, and Genetics and Genomics Sciences, Division of Psychiatric Genomics, Icahn School of Medicine at Mount Sinai, New York, New York, US
- 77 Biological Psychiatry Laboratory, McLean Hospital/Harvard Medical School, Boston, Massachusetts, US
- 78 Eating Disorders Unit, Parklandklinik, Bad Wildungen, Germany
- 79 Department of Molecular Life Science, Division of Basic Medical Science and Molecular Medicine, School of Medicine, Tokai University, Isehara, Japan
- 80 Faculty of Health Sciences, Palacky University, Olomouc, Czech Republic
- 81 Eating Recovery Center, Denver, Colorado, US
- 82 Department of Psychological Medicine, University of Otago, Christchurch, New Zealand
- 83 Canterbury District Health Board, Christchurch, New Zealand
- 84 Rheumatology Research Group, Vall d'Hebron Research Institute, Barcelona, Spain
- 85 Department of Psychiatry, First Faculty of Medicine, Charles University, Prague, Czech Republic
- 86 Centre for Addiction and Mental Health, Toronto, Canada
- 87 Institute of Medical Science, University of Toronto, Toronto, Canada
- 88 Department of Psychiatry, University of Toronto, Toronto, Canada
- 89 Department of Public Health, University of Helsinki, Helsinki, Finland
- 90 Institute for Molecular Medicine Finland, Helsinki Institute of Life Science, University of Helsinki, Helsinki, Finland
- 91 Institute of Public Health and Clinical Nutrition, Department of Clinical Nutrition, University of Eastern Finland, Kuopio, Finland
- 92 Eating Disorders Unit, Department of Child and Adolescent Psychiatry, Medical University of Vienna, Vienna, Austria
- 93 Groningen Institute for Evolutionary Life Sciences, University of Groningen, Groningen, The Netherlands
- 94 Department of Psychiatry, University of California San Diego, San Diego, California, US
- 95 Department of Pathology and Biomedical Science, University of Otago, Christchurch, New Zealand
- 96 Institute of Applied Health Sciences, University of Aberdeen, Aberdeen, UK
- 97 Department of Psychiatry, Seoul Paik Hospital, Inje University, Seoul, Korea
- 98 Rheumatology Unit, Department of Medicine, Center for Molecular Medicine, Karolinska Institutet and Karolinska University Hospital, Stockholm, Sweden
- 99 Department of Psychology, Michigan State University, East Lansing, Michigan, US
- 100 Department of Mental Disorders, Norwegian Institute of Public Health, Oslo, Norway
- 101 Department of Psychiatry and Neurochemistry, Institute of Neuroscience and Physiology, Sahlgrenska Academy at the University of Gothenburg, Gothenburg, Sweden
- 102 National Centre for Register-Based Research, Aarhus BSS, Aarhus University, Aarhus, Denmark
- 103 Centre for Integrated Register-based Research (CIRRAU), Aarhus University, Aarhus, Denmark
- 104 Department of Clinical Science, K.G. Jebsen Centre for Psychosis Research, Norwegian Centre for Mental Disorders Research (NORMENT), University of Bergen, Bergen, Norway
- 105 Dr. Einar Martens Research Group for Biological Psychiatry, Center for Medical Genetics and Molecular Medicine, Haukeland University Hospital, Bergen, Norway
- 106 Department of Clinical Medicine, Laboratory Building, Haukeland University Hospital, Bergen, Norway
- 107 American School of Professional Psychology, Argosy University, Northern Virginia, Arlington, Virginia, US
- 108 Department of Cancer Epidemiology and Prevention, M Skłodowska-Curie Cancer Center - Oncology Center, Warsaw, Poland
- 109 BESE Division, King Abdullah University of Science and Technology, Thuwal, Saudi Arabia
- 110 Department of Psychiatry, University of Lausanne-University Hospital of Lausanne (UNIL-CHUV), Lausanne, Switzerland
- 111 Department of Psychiatry, University of Campania "Luigi Vanvitelli", Naples, Italy
- 112 Center for Integrative Genomics, University of Lausanne, Lausanne, Switzerland
- 113 Department of Paediatric Laboratory Medicine, The Hospital for Sick Children, Toronto, Canada
- 114 Department of Psychiatry, Psychosomatics and Psychotherapy, University of Würzburg, Würzburg, Germany
- 115 Department of Psychiatry, University College Cork, Cork, Ireland

- 116 HSE National Clinical Programme for Eating Disorders, Cork, Ireland
- 117 Institute of Molecular and Cell Biology, University of Tartu, Tartu, Estonia
- 118 Department of Biomedical Data Science, Leiden University Medical Centre, Leiden, The Netherlands
- 119 Department of Psychiatry, Faculty of Medicine, University of Geneva, Geneva, Switzerland
- 120 Division of Child and Adolescent Psychiatry, Geneva University Hospital, Geneva, Switzerland
- 121 Great Ormond Street Institute of Child Health, University College London, London, UK
- 122 Department of Psychiatry and Behavioral Science, University of North Dakota School of Medicine and Health Sciences, Fargo, North Dakota, US
- 123 National Center for PTSD, VA Boston Healthcare System, Department of Psychiatry, Boston University School of Medicine, Boston, Massachusetts, US
- 124 Department of Medicine, Surgery and Dentistry "Scuola Medica Salernitana", University of Salerno, Salerno, Italy
- 125 Department of Neuroscience, Psychology, Drug Research and Child Health (NEUROFARBA), University of Florence, Florence, Italy
- 126 Center for Neurobehavioral Genetics, Semel Institute for Neuroscience and Human Behavior, University of California Los Angeles, Los Angeles, California, US
- 127 Kartini Clinic, Portland, Oregon, US
- 128 Center for Human Genome Research at the Massachusetts General Hospital, Boston, Massachusetts, US
- 129 Biostatistics and Computational Biology Unit, University of Otago, Christchurch, New Zealand
- 130 Saint Joan de Déu Research Institute, Saint Joan de Déu Barcelona Children's Hospital, Barcelona, Spain
- 131 Institute of Biomedicine (IBUB), University of Barcelona, Barcelona, Spain
- 132 Department of Genetics, Microbiology and Statistics, University of Barcelona, Barcelona, Spain
- 133 Institute of Clinical Medicine, University of Oslo, Oslo, Norway
- 134 Department of Health Science, University of Florence, Florence, Italy
- 135 Institute for Molecular Medicine Finland (FIMM), HiLIFE Unit, University of Helsinki, Helsinki, Finland
- 136 Analytic and Translational Genetics Unit, Massachusetts General Hospital, Boston, Massachusetts, US
- 137 Stanley Center for Psychiatric Research, Broad Institute of the Massachusetts Institute of Technology and Harvard University, Cambridge, Massachusetts, US
- 138 Department of Psychiatry and Psychotherapy, Charité - Universitätsmedizin, Berlin, Germany
- 139 Eating Disorders Research and Treatment Center, Department of Child and Adolescent Psychiatry, Faculty of Medicine, Technische Universität Dresden, Dresden, Germany
- 140 Faculty of Medicine & Health Sciences, University of Auckland, Auckland, New Zealand
- 141 Department of Psychiatry, Neurobiology, Pharmacology, and Biotechnologies, University of Pisa, Pisa, Italy
- 142 Department of Psychiatry, Poznan University of Medical Sciences, Poznan, Poland
- 143 Department of Neurosciences, Padua Neuroscience Center, University of Padova, Padova, Italy
- 144 Institute of Medical Statistics, Computer and Data Sciences, Jena University Hospital, Jena, Germany
- 145 Department of Genetics and Genomic Biology, The Hospital for Sick Children, Toronto, Canada
- 146 Institute of Psychiatry, Psychology and Neuroscience, Department of Psychological Medicine, King's College London, London, UK
- 147 J. Craig Venter Institute (JCVI), La Jolla, California, US
- 148 Department of Psychiatry and Psychotherapy, Medical University of Vienna, Vienna, Austria
- 149 Department of Pediatrics and Center of Applied Genomics, First Faculty of Medicine, Charles University, Prague, Czech Republic
- 150 Center for Eating Disorders Ursula, Rivierduinen, Leiden, The Netherlands
- 151 Department of Psychiatry, Leiden University Medical Centre, Leiden, The Netherlands
- 152 Department of Child and Adolescent Psychiatry, Poznan University of Medical Sciences, Poznan, Poland
- 153 IRCSS Fondazione Don Carlo Gnocchi, Florence, Italy
- 154 Department of Psychiatry and Biobehavioral Science, Semel Institute for Neuroscience and Human Behavior, University of California Los Angeles, Los Angeles, California, US
- 155 David Geffen School of Medicine, University of California Los Angeles, Los Angeles, California, US
- 156 Department of Cell Biology and Physiology, University of North Carolina at Chapel Hill, Chapel Hill, North Carolina, US
- 157 Department of Environmental Epidemiology, Nofer Institute of Occupational Medicine, Lodz, Poland
- 158 Department of Psychiatry, University of Naples SUN, Naples, Italy
- 159 Department of Psychiatry, University of Perugia, Perugia, Italy

- 160 Brain Sciences Department, Stremble Ventures, Limassol, Cyprus
- 161 Adolescent Health Unit, Second Department of Pediatrics, "P. & A. Kyriakou" Children's Hospital, University of Athens, Athens, Greece
- 162 Pediatric Intensive Care Unit, "P. & A. Kyriakou" Children's Hospital, University of Athens, Athens, Greece
- 163 Faculty of Social and Behavioral Sciences, Utrecht University, Utrecht, The Netherlands
- 164 School of Psychology, Curtin University, Perth, Australia
- 165 School of Paediatrics and Child Health, University of Western Australia, Perth, Australia
- 166 Department of Clinical Medicine, University of Copenhagen, Copenhagen, Denmark
- 167 Centre for Mental Health, University Health Network, Toronto, Canada
- 168 Program for Eating Disorders, University Health Network, Toronto, Canada
- 169 Department of Internal Medicine VI, Psychosomatic Medicine and Psychotherapy, University Medical Hospital Tuebingen, Tuebingen, Germany

Additional Methods

Samples and study design

Psychiatric Genomics Consortium (PGC) abbreviations for the 33 datasets meta-analyzed in this study are shown in **Supplementary Table 1**. **Supplementary Table 1** shows cohort case and control numbers, SNP numbers, and lambda at ascertainment [pre-quality control (QC)] and post-QC, **Supplementary Table 2** describes ascertainment including how cases were evaluated in the primary studies, and case and control sources and inclusion and exclusion criteria. **Supplementary Table 3** gives the genotyping platforms used. Details of the contributing studies and cohorts are provided below. The study is a secondary analysis of data collected from the studies described and did not involve direct recruitment and contact with participants or collection of identifiers linked to participants. The IRB of the University of North Carolina at Chapel Hill gave approval for this project (reference number 15-3307).

Eating Disorders Working Group of the Psychiatric Genomics Consortium (PGC-ED) Freeze 1

Full details of these data are given in the Freeze 1 paper of the PGC-ED (<http://www.med.unc.edu/pgc>) authored by Duncan et al.¹

To summarize, Duncan et al. datasets included the Children’s Hospital of Philadelphia/Price Foundation Collaborative Group (CHOP/PFCG) case-control data from the anorexia nervosa GWAS of Wang et al.⁶, case-only data from the Genetic Consortium for Anorexia Nervosa/Wellcome Trust Case Control Consortium-3 (GCAN/WTCCC-3; <https://www.wtccc.org.uk>) included in Boraska et al.², and since many of the controls used in Boraska et al. were not permitted to be re-used, control cohorts were sourced as described in Duncan et al.¹. Control cohorts from a similar geographic location and genotyped on an Illumina platform were preferentially sought. Briefly, ethical approval for each site in GCAN/WTCCC3 was obtained from the local ethics committee. All participants provided written informed consent in accordance with the Declaration of Helsinki^{1,2}. Ethical approval for each site in the PFCG was obtained separately from their own institution’s human subjects committee. Informed consent was obtained from all study participants³⁻⁵. Controls for the PFCG cases were obtained from CHOP. The Research Ethics Board of CHOP approved the study, and written informed consent was obtained from all subjects or their parents⁶. All cases and controls in PGC-ED Freeze 1 were also included in the current analysis¹. As per Freeze 1, the Italy-North cases from Boraska et al. were not included due to a lack of ancestrally-matched controls accessible for our study. Pre-QC datasets with < 100 cases were identified (Germany, Greece, Italy-South, Norway) and merged with other data to form larger datasets. The combinations were an excellent ancestral match according to principal components analysis (PCA). Italy-South and Greece were merged to form the cohort *itgr*. Czech and Poland data were merged and formed the cohort *poco*, and Norway, Germany, and Sweden data were merged and formed the cohort *gns2* (see the section below called “Genetic Consortium for Anorexia Nervosa/Wellcome Trust Case Control Consortium-3” for information about Poland and Sweden data). These samples are included in *chop*, *fin1*, *fre1*, *itgr*, *gns2*, *net1*, *poco*, *spa1*, *ukdl*, and *usa1* (see the section below for other data in *poco* and *gns2*).

Genetic Consortium for Anorexia Nervosa/Wellcome Trust Case Control Consortium-3 (GCAN/WTCCC-3)

Additional GCAN/WTCCC-3 cohorts that were not part of Duncan et al. due to a lack of controls (Poland) or small $N < 100$ of cases (Sweden) were included in our analysis as we were able to identify ancestrally matched cases and/or controls. We sourced Poland controls from the BoMa/MooDS-PGC study (Bonn/Mannheim Bipolar study; Systematic Investigation of the Molecular Causes of Major Mood Disorders and Schizophrenia study)^{7,8}. These controls were also used in the PGC bipolar disorder GWAS⁹. These controls were produced by the International Agency for Research on Cancer (IARC; <https://www.iarc.fr>) and the Centre National de Génotypage (CNG; <https://www.cng.fr>) GWAS Initiative for a study of upper aerodigestive tract cancers¹⁰ and genotyped on the Illumina HH317k. They were drawn from a hospital-based case-control sample recruited by the Nofer Institute of Occupational Medicine in Lodz and a population-based case-control sample recruited by the Cancer-Center and Institute of Oncology in Warsaw. Controls were not screened for neuropsychiatric phenotypes. These combinations above proved an excellent ancestral match according to PCA. These samples are included in *poco* and *gns2*.

Anorexia Nervosa Genetics Initiative (ANGI)

ANGI (<https://www.med.unc.edu/psych/eatingdisorders/our-research/angi>) is a multi-country effort to identify the genetic causes of anorexia nervosa and involves international research teams in the US, Sweden, Denmark, and Australia with assistance from New Zealand. Details on recruitment strategies, case definitions, and methods for ANGI have been reported previously and are outlined briefly below^{11,12}. All ANGI controls were screened for eating disorder phenotypes and some for additional psychiatric phenotypes.

Australia and New Zealand (ANGI-ANZ).

Cases from ANGI-ANZ were recruited the following way. Individuals who resided in Australia (age ≥ 13 years) or New Zealand (age ≥ 14 years) self-identified or were referred to the study. Those interested in study participation completed the consent process and online diagnostic questionnaire (ED100K-V1). Cases met anorexia nervosa criteria based on DSM-IV-TR. Once the questionnaire was completed, the participant provided a blood sample. In New Zealand, witnessed informed consent was obtained prior to sample collection. Genotyping was conducted at the Broad Institute on the Illumina Global Screening Array. Controls were obtained from the QSkin Sun and Health Study (<https://qskin.qimrberghofer.edu.au>)¹³. Briefly, ~40,000 men and women aged 40-69 years were randomly sampled from Queensland, Australia. Those who indicated no eating disorders history from a checklist were included as controls. QSkin was established to study the etiology of cutaneous melanoma and other cancers of the skin; the cohort is followed up passively through linkage with health registers. QSkin participants provided Oragene saliva samples, and DNA was genotyped at Erasmus University Rotterdam in the Netherlands on the Illumina Global Screening Array. Ethical approval for the Australian component of the study was provided by the QIMR Berghofer Human Research Ethics Committee (QIMR-HREC approval P1339). Those interested in study participation completed the informed consent process via online submission prior to taking an online diagnostic questionnaire (ED100K-V1). Participants younger than 18 years then completed a paper questionnaire which required parent/guardian co-signature. Ethical approval for the New Zealand component was provided by the Health and Disability Ethics Committee of the New Zealand

Ministry of Health. In New Zealand, witnessed informed consent was obtained prior to sample collection. Participants under 18 years required parent/guardian co-signature. Ethical approval for the QSkin study was provided by the Human Research Ethics Committee of the QIMR Berghofer Medical Research Institute (QIMR-HREC approval P1309). QSkin participants were given the option to provide written informed consent or to consent online in order to take part in the QSkin study. The samples described formed cohort *aunz*.

Denmark (ANGI-DK). National register and biobank. Cases and controls were identified primarily using the national register and biobank system. Genotypes came from Guthrie cards held by the Danish Neonatal Screening Biobank at Statens Serum Institute. Samples from this biobank are linked to the Danish register system via the unique Danish personal identification number. The individuals were born between 1981 and 2005 and had to be alive and a resident of Denmark on their first birthday and have a known mother. Cases had International Classification of Diseases (ICD-10) anorexia nervosa (F50.0)¹⁴ diagnoses assigned by psychiatrists at inpatient and outpatient psychiatric services, and were identified using the Danish Psychiatric Central Research Register¹⁵. Controls were randomly selected from the same nationwide birth cohort. Cases and controls were specifically ascertained and genotyped as a part of ANGI, and additional control genotypes came from the Lundbeck Foundation Initiative for Integrative Psychiatric Research (iPSYCH; <http://ipsych.au.dk>)¹⁶. DNA from the dried blood spots was extracted, whole-genome amplified in triplicates, and genotyped in 23 batches based on birth year. The first wave (batch) contains the youngest participants (born in 2004) and wave 23 consists of the oldest participants (born in 1981). Wave 24, a supplementary batch, comprised all participants with an ICD-10 atypical anorexia nervosa lifetime diagnosis (F50.1, most commonly diagnosed when all anorexia nervosa criteria except amenorrhea were met) up to and including 2013 and F50.0 cases diagnosed during 2013. Comparisons showed that F50.0 and F50.1 samples matched on register information tested, i.e., age of diagnosis, frequency of lifetime psychiatric diagnoses and intellectual disability, urbanicity, and maternal and paternal age at childbirth. Genotyping was performed on the Illumina PsychChip array at the Broad Institute of Harvard and MIT. GenCall¹⁷ and Birdseed¹⁸ were used to call variants with MAF > 0.01. Call sets were merged after pre-QC on individual call sets. Data processing and GWAS analyses were performed on secure servers at the GenomeDK high-performance computing cluster (<http://genome.au.dk>).

DNA preparation, genotyping, genotype calling, QC methods, and imputation were performed on the broader iPSYCH cohort waves, which also included psychiatric cases with non-anorexia nervosa diagnoses (schizophrenia, depression, bipolar disorder, autism spectrum disorder, and attention-deficit/hyperactivity disorder). The full iPSYCH cohort described contains ~86,000 individuals, including about 57,000 cases with at least one of the noted psychiatric disorders and around 30,000 controls. Each wave that was processed had ~3,500 participants.

Case-control data were filtered for the primary GWAS analysis such that cases required an anorexia nervosa or atypical anorexia nervosa lifetime diagnosis and controls required no lifetime anorexia nervosa or iPSYCH psychiatric diagnoses listed above. To address batch effects, a GWAS was conducted on waves separately, except waves 1 to 6 which were combined in a block due to smaller case sample sizes. Ancestry principal components (PCs) and PCs that

captured batch effects and other possible sources of variation were included as covariates (**Supplementary Table 18**). ANGI-DK samples formed cohorts $w1$ to $w24$.

Individuals in ANGI-DK did not provide written informed consent. The register data collection is pseudo-anonymized (<http://ipsych.au.dk/about-ipsych/data-processing-and-data-security-atipsych>) and the re-identification key for linking the Danish civil registration number is stored separately from phenotype and genotype data. An exemption from consent is legally possible in Denmark if approved by The Danish Scientific Ethics Committee (Videnskabetisk Komité). This exemption was given for all samples and was provided in 2012 to the iPSYCH study, with the most recent approval granted in 2018.

Denmark clinic samples. A Danish clinical cohort was obtained. Cases were defined as patients with at least one recorded hospital admission during which an ICD-10¹⁴ diagnosis of F50.0 or F50.1 was given. Clinical cases consisted of women born 1947-1980 (age range: 35 to 68 years). Samples were genotyped at the Broad Institute on the Illumina Global Screening Array. Ethical approval with the protocol no. H-KF-01-024/01 was obtained from the competent Danish authority, De Videnskabetiske Komiteer for Region Hovedstaden (The Capital Region of Denmark's Committees on Health Research Ethics). All participants provided written informed consent prior to being included into the study. The samples described are included in cohort *sedk*.

Sweden (ANGI-SE). The primary recruitment strategy involved Riksät-National Quality Register for Eating Disorders Treatment¹⁹, which includes eating disorder-specific information from individuals seeking eating disorder treatment in Sweden since 1999. Potential cases identified through Riksät were sent a letter asking them to complete a follow-up questionnaire, which included the ED100K-V1 questionnaire. In the second recruitment strategy, study nurses at the Stockholm Centre for Eating Disorders (<http://stockholmatstorningar.se>) (SCÄ) recruited cases for ANGI. When patients with anorexia nervosa came into this center, a research nurse discussed the study with them and reviewed the consent. When participants consented, a blood sample was taken, and the participant was directed to complete the online diagnostic questionnaire. The third strategy was community outreach, specifically using traditional media (i.e., TV, radio, and newspapers) and social media including the Swedish ANGI website (<http://www.angi.se>), directly linking to the questionnaire. The final recruitment strategy for cases and controls involved LifeGene (<https://www.lifegene.se>)²⁰, an ongoing study initiated in 2010 to evaluate how genes, environment, and lifestyle affect health. Individuals enrolled in LifeGene completed an eating disorder assessment similar to the online diagnostic questionnaire and provided a blood sample. An anorexia nervosa algorithm for LifeGene was harmonized with the ED100K-V1 questionnaire for case and control identification. All cases met anorexia nervosa criteria based on DSM-IV-TR²¹, and controls screened negative for a history of eating disorders. Genotyping was performed at the Broad Institute on the Illumina Global Screening Array. The Swedish component of ANGI (Riksät, SCÄ, and Community) was approved by the Regional Ethical Review Board in Stockholm (dnr: 2013/112-31/2). Individuals who wished to participate were mailed consent forms along with the vials for blood samples. Signed consent forms were returned with the samples. The Regional Ethical Review Board of Stockholm provided initial ethical approval of LifeGene. All participants provided consent online²⁰. The Swedish component of ANGI obtained approval, as stated above, for use of LifeGene samples and data as

part of ANGI-SE. These samples described are included in cohort *sedk* (see the section above called “Denmark Clinic Samples” for other samples included in *sedk*).

United States (ANGI-US). Individuals who resided in the US (ages ≥ 12 years) self-identified or were referred to the study. Individuals completed a brief online screener to determine eligibility as a case or control for ANGI. Those deemed eligible completed the consent process and online diagnostic questionnaire (ED100K-V1). All cases met anorexia nervosa criteria based on DSM-IV-TR, and controls screened negative for a history of eating disorders. Once the questionnaire was completed, the participant provided a blood sample. Approximately 1,000 controls were recruited from the community. Additional control samples were obtained from The Price Foundation Anorexia Nervosa Trios Study^{5,22}. These additional controls reported no history of eating disorders, had no first degree relative with an eating disorder, and screened negative for other Axis I psychopathology. Genotyping was performed at the Broad Institute on the Illumina Global Screening Array. Ethical approval for the US component of ANGI was granted by the University of North Carolina at Chapel Hill’s Institutional Review Board (IRB# 13-0081). Individuals who were interested in study participation (and deemed eligible by the brief screen in the US) contacted the study team and completed the consent process. Although some provided written consent, most participants provided consent over the phone after a complete review of the consent forms. These samples formed cohort *usa2*.

UK Biobank

The UK Biobank (<http://www.ukbiobank.ac.uk/>) is a large prospective study of ~500,000 residents of the United Kingdom aged from 40 to 69 years old²³. UK Biobank aims to provide insights into the causes, prevention, and treatment of various illnesses. Recruitment occurred between 2006-2010. The present study uses data from the July 2017 release including the second wave of genetic data. Cases were identified by primary and secondary ICD-10¹⁴ diagnosis from linked health care records and self-report diagnosis of anorexia nervosa by a clinical professional in the UK Biobank mental health questionnaire. Controls were screened for any psychiatric disorder (Chapter V: Mental and behavioral disorders). UK Biobank participants provided electronic signed consent at their baseline assessment visit. UK Biobank was approved by the NHS Health Research Authority North West-Haydock Research Ethics Committee (reference 16/NW/0274). The current study was completed as part of approved UK Biobank application 27456. The samples described formed cohort *ukbb*.

Merging of case and control data

When sourcing control data, we prioritized controls that were ancestrally matched and genotyped on a similar platform to cases. There are two instances whereby cases were matched with controls from another country [i.e., *aunz* New Zealand cases ($n = 430$) were merged with Australian cases ($n = 2,261$) and controls ($n = 17,158$), and *sedk* Denmark cases ($n = 129$) were merged with Sweden cases ($n = 4,118$) and controls ($n = 4,035$)]. Country is unlikely to be confounded with case-control status in these cohorts. Ancestral matching was undertaken by visual inspection of a principal components analysis PC1 v PC2 plot for the merged data, and the matches were excellent with cases and controls randomly interspersed. Further, the first five ancestry PCs and any PCs that significantly differed ($P < 0.05$) between case and control cohort were included as covariates in GWAS to capture nuanced ancestry or batch effect differences.

Statistical power

Identified susceptibility variants in psychiatric genetics typically have an OR of ~ 1.1 ²⁴. Our study was acceptably powered to detect susceptibility variants in this range (80% power to detect an OR of 1.09-1.19 with an additive model, 0.9% lifetime risk²⁵, $\alpha = 5 \times 10^{-8}$, MAF 0.05–0.5). Prior experience in other PGC disorders gives us reason to believe that when sample sizes reach an inflection point for power to detect multiple GWAS hits, the number of significant loci begins to increase linearly as samples were added after this inflection point. It is not surprising to see several borderline significant hits when first exceeding the inflection point, as has been observed in schizophrenia²⁶⁻²⁸. As more cases are added, it is probable that we will get more lead hits above this borderline range.

Quality control and covariates

For the GWAS analysis, the default QC parameters in Ricopili were used and are described as follows. Ricopili QC begins with a pre-filter SNP call rate of > 0.95 , which is useful for cases and controls genotyped on different platforms. Next, QC involves sample filters, then SNP filters. Default sample filters are a call rate (cases/controls) ≥ 0.98 , heterozygosity inbreeding coefficient ≤ 0.2 (cases/controls), and sex violations. Default SNP filters are a call rate ≥ 0.98 , case-control missingness difference ≤ 0.02 , no valid association P value (invariant), and violations of Hardy-Weinberg equilibrium (in controls $P > 10^{-6}$, in cases $P > 10^{-10}$). Some cohorts required the application of more stringent thresholds to reduce bias (**Supplementary Table 18**). Ancestry outliers were removed based on plotting the first two principal components (PCs) in a principal components analysis (PCA) containing each cohort and five reference cohorts (1000 Genomes Phase 3 EUR, AFR, EAS, SAS, AMR)²⁹ to retain European samples. Samples showing familial structure and/or cryptic relatedness, or duplicates were removed ($\hat{\pi} > 0.2$) during PCA. For the Danish waves, we conducted additional relatedness testing across all the waves combined, and then removed one individual from each related pair ($\hat{\pi} > 0.2$). PCs significantly associated with the phenotype were identified for inclusion as covariates. For the *aunz* cohort, QC and PCA was done externally. For QC, see **Supplementary Table 18**. PCA (20 principal components) were computed using smartpca (EigenSoft 6.0.1) on the cleaned data from all individuals used in the current paper in conjunction with the genotypes of $\sim 50,000$ additional individuals available at QIMR Berghofer, and the population reference data from the Genome-EUTWIN populations (<https://ega-archive.org/datasets/EGAD00000000043>) and HapMap Phase 3 populations³⁰. Analyses were run using a filtered set of genotypes available across all genotyping projects (N SNPs $\sim 40,000$). Individuals beyond six standard deviations (SDs) of the European PC1 and PC2 centroid were excluded from analysis. The data were then put through Ricopili PCA module; the first five ancestry components and PCs significantly associated with the phenotype were always included as covariates (**Supplementary Table 18**). To the extent that national laws and regulations permitted, we examined sample overlap across cohorts by performing LD score bivariate regressions and estimating genetic covariance intercepts to assess sample overlap^{31,32} (**Supplementary Table 19**).

Some of the cohorts required additional QC beyond the default process and parameters used in PGCs GWAS pipeline, Ricopili. **Supplementary Table 18** shows the additional QC steps applied, if any, and the PCs used as covariates, by cohort. The first five PCs were automatically

included to adjust for ancestry effects. Tests were done on post-QC data to investigate whether any of the PCs differ significantly between cases and controls, and if so, PCs nominally associated with the phenotype ($P < 0.05$) were also included in the GWAS as covariates.

Anorexia nervosa subtype phenotypes

We defined subtypes based on the presence or absence of binge eating. The rationale for this choice was that the current DSM-5³³ subtyping (i.e., restricting versus binge/purge) is a clinical, rather than a biological distinction. “Purging” behavior is a heterogeneous category and includes several behaviors either alone or in combination (e.g., self-induced vomiting, laxative abuse, diuretic abuse, other inappropriate compensatory behaviors), and our sample size is insufficiently powered—and phenotyping in several samples inadequately detailed—to identify individuals with various purging behavior constellations. In contrast, binge eating is more uniformly defined, represents a clear departure from restrictive eating behavior, and lies on the appetite continuum. Although the twin-based genetic correlation between binge eating and self-induced vomiting is high ($r_g = 0.74$)³⁴, less is known about other purging methods. Larger samples sizes will enable additional group distinctions and allow us to comment on the biological appropriateness of the current DSM-5 subtypes.

We defined anorexia nervosa with and without binge eating using available phenotypic data. We were able to use *chop*, *aunz*, and *usa2* datasets for this analysis. For *aunz* (ANGI-ANZ) and *usa2* (ANGI-US), anorexia nervosa with binge eating was defined as reporting ever 1) "having eating binges when you ate what most people would regard as an unusually large amount of food in a short period of time" and 2) "having a sense of loss of control during those eating binges". The absence of binge eating was determined by a “no” answer to either item. For *chop* (CHOP/PFCG), the presence of binge eating was defined as reporting a history of binge eating by structured or semi-structured interview. The no binge eating group had to report no lifetime history of bulimia nervosa and no history of binge eating. Purging was not used in these definitions and binge eating did not need to occur within episodes of anorexia nervosa. The percentage of available subtype data within these cohorts was 95% overall. Future analyses with larger samples sizes will increase confidence in results from this analysis.

GWAS of related traits in UK Biobank

Seven UK Biobank GWAS were performed to facilitate genetic correlation investigations. The phenotypes were BMI (Hübel, Gaspar, Coleman, Hanscombe, Purves...Breen, unpublished report), body fat percentage³⁵, fat mass (Hübel, Gaspar, Coleman, Hanscombe, Purves...Breen, unpublished report), fat-free mass³⁵, physical activity³⁵, anxiety³⁶, and neuroticism³⁵.

Genotyping, imputation and QC

Briefly, blood samples were genotyped on two arrays, which share nearly all of their content: the UK BiLEVE array ($N = 49,949$) or the UK Biobank Axiom array ($N = 438,414$). Genotyping was conducted by Affymetrix across 33 different batches of approximately 4,700 samples each. UK Biobank provides extensive information on sample processing on its website (biobank.ctsu.ox.ac.uk/crystal/refer.cgi?id=155583). UK Biobank performed stringent QC on the genotyping data at the Wellcome Trust Centre for Human Genetics (WTCHG). For further

details, see: biobank.uct.ac.za/crystal/refer.cgi?id=155580. Prior to imputation, all variant sites with a call rate below 90% were filtered out. Imputation was carried out by UK Biobank using a merged UK10K-1000 Genomes Phase 3 reference panel and the Haplotype Reference Consortium (HRC) panel³⁷ (for further information, please see³⁸). UK Biobank preferentially retained SNPs imputed to HRC for SNPs present in both imputation panels. Imputation was conducted using the IMPUTE4 program³⁸.

We excluded non-European participants identified by *k*-means clustering ($k = 4$) on the first two PCs derived from the genotype data, and related individuals (KING relatedness metric > 0.088 , equivalent to an identical-by-descent coefficient of 0.25). SNPs were excluded if they had a MAF $< 1\%$, if no call was made in $> 2\%$ of samples following imputation, if they were imputed with low confidence (INFO < 0.8), deviated substantially from Hardy-Weinberg equilibrium (HWE test, $P < 10^{-7}$), if they were not genotyped and were not part of the HRC panel. Additional QC and other information about these GWAS is given in **Supplementary Table 20**.

GWAS of BMI, body fat percentage, fat mass, and fat free mass

The GWASs on BMI, body fat percentage, fat mass, and fat free mass were a cross-sectional analysis of 155,961 healthy European participants from the UK Biobank. To identify genetic variation uniquely associated with body composition not confounded by illnesses and their downstream effects or metabolism-altering medication, we applied stringent exclusion criteria (e.g., psychiatric, gastrointestinal and endocrine illnesses, hormonal and antidiabetic medication). Body composition was assessed using Tanita BC-418 MA scale (Tanita Corporation, Arlington Height, IL). We included 7,794,483 genotyped and imputed SNPs and insertion-deletion variants with a MAF $\geq 1\%$. We covaried for assessment center, genotyping batch, smoking status, alcohol consumption, menopause, age, and socioeconomic status (measured by the Townsend Deprivation Index)³⁹. We accounted for underlying population stratification by including the first six ancestry PCs, calculated on the European subsample GWAS cohorts. We used BGENIE v1.2 (<https://jmarshini.org/bgenie>) for sex-specific analyses and meta-analyzed these sex-specific GWAS using METAL⁴⁰ (<http://csg.sph.umich.edu/abecasis/metal>).

GWAS of physical activity

We calculated sex-specific GWAS of physical activity with imputed genotype data in 29,496 male and 36,758 female ($N = 66,254$) individuals in the UK Biobank, including age (at recruitment), genotyping array, and genetic PCs 1–20 as covariates. Physical activity in the UK Biobank was measured continuously over a period of seven days with a wrist-worn accelerometer. General physical activity quality control of raw data is described in detail elsewhere⁴¹. They used a wear-time adjusted 7-day average measure of activity, including only individuals meeting UK Biobank QC criterion: good wear-time, good calibration, calibration on own data, and no problem indicators. Analyses were performed on the intersection of this UK Biobank subset with those passing general genotyping QC: in European ancestry subset; used in the calculation of ancestry PCs; without excess relatives in the UK Biobank sample; no putative sex chromosome aneuploidy; and were not outliers for heterozygosity and genotype missingness. General genotyping considerations, raw data QC, and imputation procedure in the UK Biobank are described in detail elsewhere³⁸. 20 PCs provided by the UK Biobank were used.

GWAS of anxiety

We conducted sex-specific GWAS of anxiety disorders with imputed genotype data on 25,443 cases and 58,113 controls³⁶. Cases met criteria for probable lifetime anxiety disorder diagnosis if they either self-reported a professional diagnosis of any of the five core anxiety disorders (generalized anxiety disorder, panic disorder, specific phobia, agoraphobia, or social anxiety disorder) or met criteria for probable lifetime generalized anxiety disorder according to the Composite International Diagnostic Interview question set^{42,43}. Cases did not report a diagnosis of schizophrenia, psychosis, attention-deficit/hyperactivity disorder (ADHD), autism, any eating disorder, or bipolar disorder. Controls were screened for any evidence of psychiatric or substance use disorders. Participants were limited to individuals of European ancestry, who were not excessively related, had no putative sex chromosome aneuploidy, and were not outliers for heterozygosity and genotype missingness. Stratified linear regressions were performed on ~7 million SNPs of high imputation quality (INFO > 0.9) with a minimum MAF \geq 0.01 in BGENIE v1.2 controlling for six ancestry PCs calculated on the European subsample of UK Biobank, assessment center, genotyping batch, and age.

GWAS of neuroticism

We performed sex-specific GWAS on neuroticism using the genotype data supplied by UK Biobank in males ($N = 142,875$) and in females ($N = 144,660$; total $N = 287,535$), following QC as described above and using PCs calculated on the European subsample of UK Biobank. The neuroticism phenotype was calculated as the sum score of neuroticism questions at the baseline assessment⁴⁴, corrected for age and sex-specific means and SDs from the UK population⁴⁵. In a second analysis, individuals were excluded if they reported any psychiatric illness resulting in 83,413 males and 73,946 females (total $N = 157,355$). Sex-stratified linear regressions were performed in PLINK using eight ancestry PCs and genotyping batch (as a factor) as covariates and later meta-analyzed using METAL⁴⁰.

Gene expression

We first investigated whether anorexia nervosa heritability was enriched in tissue/cell type specifically expressed genes using publicly available gene expression data: GTEx⁴⁶ (RNA-seq of macroscopic samples from multiple human tissues) and Cahoy⁴⁷ (mouse neural cell-types transcriptome database). Stratified LDSC estimated common variant heritability enrichment in the top 10% of specifically expressed genes in each tissue or cell type, taking into account confounders such as gene size, LD, and functionally enriched genomic regions (e.g., conserved regions across mammals⁴⁸). We followed a published method LDSC-SEG (LDSC applied to specifically expressed genes)⁴⁹. From the datasets, the method has derived a genome annotation corresponding to each tissue or cell type of interest, which contains the top 10% specifically expressed genes of the tissue or cell type together with 100 Kb windows on each side of the transcribed region of each genes. We obtained the derived genome annotations from the LDSC-SEG GitHub repository (<https://github.com/bulik/ldsc>).

Second, we created a new annotation by performing differential expression analyses among 9,970 single cells, previously clustered into 24 different cell types⁵⁰, from five different mouse brain regions. Briefly, we used the scran R package⁵¹ using the 50% of the genes with mean expression higher than the median to compute normalization factor for each single cell. The

normalization factors were computed after clustering cells using the scran package quickCluster function to account for cell type heterogeneity. We then performed 24 differential expression analysis using R package BPSC⁵² testing each cell type against the 23 other cell types using the computed normalization factors as a covariate. We then selected the top 10% most upregulated genes for each cell type and used the coordinates of these genes extended by 100 Kb on each side as an extra annotation in LDSC.

We used the “munge_sumstats.py” script built in the LDSC software^{31,32} to reformat the ANGI GWAS results. We then applied stratified LDSC regression to estimate heritability enrichment of the annotations⁴⁹ (tissue or cell type) conditional on 53 other annotations from the “baseline model” (e.g., conserved regions⁴⁸). We used regression weights computed from phase 3 of the 1000 Genome Project²⁹ with HLA regions removed (https://data.broadinstitute.org/alkesgroup/LDSCORE/1000G_Phase3_weights_hm3_no_MHC.tgz).

We primarily report the regression coefficient of each annotation (corresponding to gene expression of each tissue or cell type) and an associated *P* value. A positive regression coefficient suggests that the annotation contributes to the heritability of anorexia nervosa while accounting for the contributions of other annotations^{49,53}. The *P* value tests whether the regression coefficient was significantly positive (one-tailed), i.e., whether the contribution of the annotation is statistically significant. LDSC analyses are reported in **Supplementary Figs. 12-15**.

We also used MAGMA⁵⁴ (v1.06) (<https://ctg.cncr.nl/software/magma>), as done previously⁵⁰, to identify tissues/cell types underlying anorexia nervosa. GTEx data (V6p, median expression across individual per tissue) was downloaded from the GTEx website (<https://gtexportal.org/home>). Genes not expressed in any tissues (median = 0 for all tissues) were excluded. Gene expression from the different brain cell types was obtained as previously described. Briefly, we performed single cell RNA-seq from 9,970 single cells from five brain regions (neocortex, hippocampus, hypothalamus, striatum, midbrain, plus samples enriched for oligodendrocytes, dopaminergic neurons, and cortical parvalbuminergic interneurons), which allowed us to identify 24 cell types at level 1 (broad clustering) (pyramidal neurons, oligodendrocytes, etc...) and 149 cell types at level 2 (fine grained clustering) (pyramidal neurons from layer 6, layer 4, etc...).

For each gene expression dataset, we computed an index of specificity for each gene in each tissue/cell type by dividing the expression of a gene in a given tissue/cell type by the total expression of the gene in all tissues/cell types (range of specificity: 0-1). For each tissue/cell type, we then binned the specificity measure into 41 bins (0 representing a gene not expressed in the tissue/cell type, 1 gene in the bottom 2.5% quantile of specificity, ..., 40 genes that are in the 97.5% to 100% most specific genes in the tissue/cell type). We then used MAGMA to test for a positive correlation between binned tissue specificity and gene-level genetic association with anorexia nervosa for each tissue/cell type. The gene-level genetic association was computed with MAGMA (v1.06) using a window surrounding the gene by 35 kb upstream to 10 kb downstream of the gene. The gene-level association is computed by summing the association *P* value of SNPs located in the gene windows taking into account the LD structure of the region. MAGMA takes

into account confounders such as gene length, LD, and gene-gene correlation. MAGMA analyses are reported in **Supplementary Figs 9-11**.

Predicted tissue-specific gene expression

We predicted differential gene expression using S-PrediXcan v1.0⁵⁵ and genomic and transcriptomic reference data from the brain regions assayed in the GTEx project v7⁴⁶ and Depression Genes and Networks (DGN) whole-blood cohort⁵⁶. A total 258,158 gene-tissue pairs were tested (**Supplementary Table 13**). Significant genes were compared to genes in the gene-wise analysis performed with MAGMA.

A general note on multiple testing correction

Carrying out multiple testing runs the risk of inflating Type I error and increases the probability of false positive results. To manage this risk, we took the general approach of setting a conservative *a priori* P Bonferroni correction threshold. We did this generally on a within-analysis basis, since different analyses tested different underlying hypotheses, rather than paper-wide. Readers can identify the Bonferroni threshold used for a given analysis in Table and Figure notes.

Additional Results

Primary GWAS meta-analysis

Pre- and post-QC lambda, number of single-nucleotide polymorphisms (SNPs), and N_s in each of the final 33 datasets analyzed within the primary GWAS can be found in **Supplementary Table 1**. The meta-analysis LD intercept was 1.02 (s.e. = 0.01). The meta-analysis quantile-quantile (Q-Q) plot is shown in **Supplementary Fig. 1**. The LD score regression intercepts for each cohort after QC ranged from 0.98 (s.e. = 0.01) to 1.03 (s.e. = 0.01) (**Supplementary Table 1**). The genetic covariance intercepts in LD score bivariate regression analyses were close to 0 and indicated no evidence of sample overlap among the cohorts (**Supplementary Table 19**). Eight genome-wide significant loci were identified. Follow-up analyses included using genome-wide complex trait analysis (GCTA-COJO)⁵⁷ to conduct stepwise regression on markers with P values $< 10^{-8}$. GCTA-COJO identified eight independent signals taking into account the LD correlations between SNPs and running a conditional and joint analysis⁵⁸. All eight identified markers were equivalent to the SNPs resulting from distance- and LD-based clumping shown in **Table 1** (for conditional analyses please see **Supplementary Table 5**).

Genomic inflation and residual confounding

Here we provide detailed information to illustrate that confounding due to population stratification or other reasons was minimal in our primary GWAS. Two relevant parameters were estimated using linkage-disequilibrium (LD) score regression²⁹. Firstly, we estimated an LD intercept of 1.02 (s.e. = 0.01) for the meta-analysis and between 0.98-1.03 for individual datasets (**Supplementary Table 1**). The LD score intercept is significantly greater than one, but is in line with the expected small levels of inflation caused by sample size and heritability described in Loh et al.⁵⁹. A second measure, the attenuation ratio $[(\text{LDSC intercept} - 1) / (\text{mean } \chi^2 - 1)]$ for the meta-analysis was 0.07 (s.e. = 0.04) (**Supplementary Fig. 1**), also suggesting a lack of confounding. Together these suggest that deviation from the null was due to polygenic signal and not population structure or bias.

The overall inflation of summary statistics genome-wide or λ_{GC} for individual datasets post-QC ranged between 1.00-1.06 (**Supplementary Table 1**) and for the meta-analysis was 1.22 (**Supplementary Fig. 1**). Inflation of tests statistics is known to be due to a combination of polygenicity, uncorrected population stratification, and confounding. The larger λ_{GC} value observed for our GWAS meta-analysis is indeed expected at this sample size, trait polygenicity, and heritability⁵⁹. The LD score regression method for GWAS of highly polygenic phenotypes such as anorexia nervosa and large sample sizes separates the polygenic component (the slope) from population stratification and other systematic biases (estimated by the intercept and attenuation ratio)²⁹.

Correcting for λ_{GC} in large GWAS samples of polygenic phenotypes can cause loss of signal and power, as evidenced by the LD intercept of the GWAS summary statistics from the Genetic Investigation of Anthropometric Traits (GIANT) Consortium 2015 body mass index (BMI) paper (0.65)^{60,61}. Correction for λ_{GC} at the individual study level has also been shown to bias heritability estimates downward²⁹.

Previous hit

The chr12 locus identified in Duncan et al.¹ did not reach genome-wide significance. The OR for the index variant at this locus (rs4622308) was in the same direction in the present meta-analysis compared with Duncan et al. (present: A1 = C, A2 = T, OR = 1.06, s.e. = 0.01, $P = 7.02 \times 10^{-5}$; Duncan et al.: A1 = C, A2 = T, OR = 1.20, s.e. = 0.03, $P = 4.25 \times 10^{-9}$) and in 22 of the 33 cohorts ($z = 2.00$, $P = 0.05$, 2-tailed). To further assess this locus, a random-effects meta-analysis was conducted. Similar to the fixed-effect meta-analysis, the random-effects meta-analysis indicated that the effect of this locus was not genome-wide significant (OR = 1.06, s.e. = 0.05, $P = 0.02$) and showed evidence of heterogeneity, $I^2 = 53.7$ (**Supplementary Fig. 4**). There are many possible reasons why the result was not replicated, including winner's curse⁶², moderator variables given the heterogeneity observed (such as environmental risks), between-study heterogeneity in ascertainment, and differences in LD structure across cohorts in addition to true non-replication.

Chromosome X

The separate analysis of chrX included $n = 14,915$ cases and $n = 27,854$ controls. There were no genome-wide ($P < 5 \times 10^{-8}$) or suggestive ($P < 1 \times 10^{-5}$) significant loci.

Female-only secondary GWAS

A supplementary analysis conducted on female cases and controls only—to determine the similarity of results to the main GWAS analysis which included females and males—had 14,896 cases and 27,865 controls. The female-only GWAS revealed one genome-wide significant locus ($P < 5 \times 10^{-8}$) on chr3 (rs9812977; 48.2-49 Mb; $P = 1.31 \times 10^{-9}$; OR = 1.08; 95% CI: 1.03-1.14), which was the top locus in the main GWAS analysis.

eQTL and Hi-C interactions

For locus 1 (multigenic, chr3:47.5-51.3 Mb, **Supplementary Fig. 5a**), our data implicate 100 brain-expressed genes. Locus 1 is gene-dense with a large number of brain eQTLs and regulatory chromatin interactions. Notably, 16 genes with regulatory chromatin interactions mapping to the locus lie outside the LD-defined locus. Locus 3 (multigenic, chr2:53.8-54.3 Mb, **Supplementary Fig. 5c**) is less complex than locus 1. Nonetheless, we implicate 12 genes, six within and 6 outside the LD-defined locus.

For all four single-gene loci, eQTL and/or chromatin interaction connections implicated the gene intersecting the locus. For locus 2 (chr11:114.9-115.4 Mb, **Supplementary Fig. 5b**), both eQTL and regulatory chromatin interaction data confirmed the connection to the locus-intersecting gene *CADMI*. Genetic variants near to *CADMI* (cell adhesion molecule 1) have been implicated by GWAS for body mass and age at menarche. *CADMI* protein levels appear to be elevated in the hypothalamus of BMI risk variant carriers. Obese mice have been reported to show elevated *CADMI* expression and *CADMI* knockout mice show reduced body weight⁶³. For locus 4 (chr10:131.2-131.4 Mb, **Supplementary Fig. 5d**), eQTL data connected to the locus-intersecting gene *MGMT*. *MGMT* (O-6-methylguanine-DNA methyltransferase) encodes an epigenetic regulator important in multiple cancers, including glioblastoma. For locus 5 (chr3:70.6-71.0 Mb, **Supplementary Fig. 5e**), regulatory chromatin interaction data confirmed the connection to the

locus-intersecting gene *FOXP1* (forkhead box P1), which encodes a transcription factor in the forkhead box family. Brain-specific deletion of *FOXP1* results in defects in striatal development and changes in the hippocampus⁶⁴. Interestingly, *FOXP1* knockout mice exhibit a significant reduction in body weight as compared to littermate controls. For locus 5, eQTL data also suggested a functional connection to the more distal *GPR27*, which encodes an orphan G-protein coupled receptor, and is highly expressed in the brain⁶⁵. It has recently been associated with insulin secretion⁶⁶. Finally, for locus 6 (chr1:96.6-97.2 Mb, **Supplementary Fig. 5f**), both eQTL and regulatory chromatin interaction data confirmed the connection to the locus-intersecting gene *PTBP2*. The protein encoded by this gene binds to intronic polypyrimidine clusters in pre-mRNA molecules and is implicated in controlling the assembly of other splicing-regulatory proteins.

One intergenic locus (locus 8, chr5:93.9-95.0 Mb, **Supplementary Fig. 5h**) had eQTL connections to *PROS1* and *ARL13B*. *PROS1* encodes an anticoagulation factor that plays a role in blood-brain-barrier integrity⁶⁷. *ARL13B* encodes a member of the ADP-ribosylation factor-like (ARL) small Ras GTPase family. ARL13B protein is expressed in the cilia of all organs; mutations are associated with Joubert syndrome 8, which like other ciliopathies has been associated with obesity^{68,69}. Mutations are also associated with intellectual disability (University of Chicago's Intellectual Disability Exome Panel). Thomas et al.⁶⁹ found *ARL13B* expression throughout the developing human brain. Additionally, they identified *ARL13B* expression in primary cilia of hypothalamic neurons of newborn mice. Furthermore, Higginbotham et al.⁷⁰ found that mutant ARL13B disrupts the development and migration of interneurons.

For locus 7 (intergenic, chr5:24.9-25.3 Mb, **Supplementary Fig. 5g**), there were no supporting eQTL or regulatory chromatin interactions.

Multi-trait conditional and joint analysis (mt-COJO)

Seven of the eight genome-wide significant loci showed only slight changes in their effect sizes (i.e., betas) after conditioning on education years⁷¹, type 2 diabetes⁷², high-density lipoprotein (HDL) cholesterol⁷³, BMI (Hübel, Gaspar, Coleman, Hanscombe, Purves...Breen, unpublished report; see **Additional Methods**), schizophrenia²⁸, or neuroticism³⁵. The results suggest that the loci are independent of the traits on which they were conditioned and that the traits may not share genetic association at these loci. The association of locus 2 on chr11 tagged by rs6589488 with anorexia nervosa may not be independent of genetic associations with type 2 diabetes as the beta was diminished after conditioning. This suggests, at locus 2, that the association with anorexia nervosa may not be independent of genetic underpinnings of glycemic alterations seen in type 2 diabetes.

Clinical investigations

Anorexia nervosa subtype

One potential source of genetic heterogeneity lies in differing clinical presentations of anorexia nervosa (i.e., with or without the presence of binge eating). This was not supported in preliminary analyses. The SNP-based genetic correlation (SNP- r_g) between the anorexia nervosa subtypes was 0.74 (s.e. = 0.16; $P = 1.74 \times 10^{-6}$). To test for heterogeneity in the genetic variation

associated with these two subtypes, we tested whether their SNP- r_g was significantly different from unity. We used a block jackknife approach using the LD score regression software (LDSC) v1.0²⁹. The genetic correlation between anorexia nervosa with and without binge eating was not significantly different from 1 ($P = 0.08$). There were no significant differences in the mean polygenic risk score (PRS) between subtypes in the three cohorts for which anorexia nervosa subtype data were available (**Supplementary Fig. 6**). We also calculated SNP- r_g by anorexia nervosa subtype (**Supplementary Table 9**) and found no significant differences in SNP- r_g with external traits, although small sample sizes limit interpretation. In summary, our preliminary subtype analyses do not indicate significant differences in the genetic architectures of anorexia nervosa with and without binge eating; however, larger sample sizes are necessary for confirmation.

Males with anorexia nervosa

The number of males identified on the basis of genotype sex in the meta-analysis was 447 cases and 20,347 controls. Anorexia nervosa PRS scores derived using the female-only GWAS were associated with a higher risk of anorexia nervosa in males. Those at the highest decile had 4.13 (95% CI: 2.58, 6.62) times the odds of lifetime anorexia nervosa compared with those at the lowest decile. Anorexia nervosa PRS accounted for $\sim 1.8\%$ of the total variance in anorexia nervosa in males for the discovery cohort P threshold (pT) = 0.5, comparable to $\sim 1.7\%$ at P threshold (pT) = 0.5 in the cohort as whole. Taken together, these preliminary results do not provide evidence for a major sex-specific difference in the common genetic architecture of anorexia nervosa, although our conclusions are limited due to the small sample size.

Within-trait prediction: polygenic risk scoring

We observed that across cohorts, anorexia nervosa cases were more likely to be in the PRS higher deciles than controls based on their anorexia nervosa genetic load (**Supplementary Fig. 16**). Visual inspection of the decile plots showed that the score distributions across deciles were relatively uniform across target datasets; thus, there was no indication of any extreme outlying datasets. PRS applied to the largest target dataset, *sedk*, showed that those at the highest decile had 2.59 times higher odds (95% CI: 2.12-3.18) of lifetime anorexia nervosa compared with those at the lowest decile. The results also provide evidence of the replicability of the main GWAS results (**Supplementary Fig. 16**).

Gene-wise analysis

Results of MAGMA gene-wise associations are reported in **Supplementary Table 11**. Seventy-nine genes were Bonferroni-significant (threshold = 2.62×10^{-6}), and 506 had a Benjamini and Hochberg⁷⁴ FDR q value < 0.05 . None of the 79 Bonferroni-significant genes are part of the MHC, but 57 genes are located in a gene-rich locus on chr3. The top genes on chr3 were *NCKIPSD*, *CELSR3*, and *IP6K2*. Through the MAGMA analysis we identified 16 additional genes which were not annotated via clumping. These 16 genes are located in loci on chr 1, 2, 3, 7, 10, 11, 12, 17, and 22. Some of these Bonferroni-significant genes have been indicated to be involved in glycemic and metabolic disease phenotypes (*CTNNA1*⁷⁵; *EXOC4*⁷⁶; *FAM19A2*⁷⁷; *VAMP2*⁷⁸).

Pathway analysis

The Bonferroni-significant pathway (**Supplementary Table 12**)

GO:positive_regulation_of_embryonic_development (Gene Ontology, 32 genes, 6.31×10^{-6}) has two Bonferroni-significant genes: *DAG1* and *CTNNB1*. *CTNNB1* encodes beta-catenin, an essential component of the canonical Wnt signaling pathway. Beta-catenin can regulate neuronal progenitor proliferation and affect cortical size⁷⁹. *CTNNB1* expression has been associated with adipogenesis, glucose metabolism, and obesity⁷⁵. Metabolic diseases including obesity and type 2 diabetes are influenced by genetic and functional variations in the Wnt signalling pathway⁷⁵. *DAG1* encodes dystroglycan, an essential member of the dystrophin-glycoprotein complex that has been mainly studied in the context of muscular dystrophies. Dystroglycan and other members of the dystrophin-glycoprotein complex are also found in neurons and glia and their disruption has been linked to intellectual disability and altered brain development⁸⁰. Specific ablation of dystroglycan in neurons or glia results in distinct phenotypes⁸¹. Dystroglycan is expressed in pyramidal cells of the cortex and hippocampus, where it appears to be essential for the establishment and maintenance of CCK-positive basket cell terminals⁸².

Tissue and cell type analyses

We first used partitioned heritability analyses in LDSC using annotation of elements in the genome with specific functions. Considering general annotations, enrichment in conserved regions was the main finding (enrichment (s.e.) = 24.97 (3.29), $P = 3.32 \times 10^{-11}$, **Supplementary Fig. 7**). Partitioned heritability analysis was then used to test for cell type-specific enrichment in the GWAS of anorexia nervosa among 10 cell type groups: adrenal and pancreas; cardiovascular; CNS; connective and bone; gastrointestinal; immune and hematopoietic; kidney; liver; skeletal muscle; and other tissue (which includes adipose tissue). The CNS cell type group showed a 2.8-fold significant enrichment (**Supplementary Fig. 8**).

We next investigated whether there were tissue or cell type associations with anorexia nervosa using gene expression data to annotate gene sets characteristic of specific cells or tissues (for details see⁵⁰). Gene expression datasets used were: GTEx (RNA-seq of samples from multiple human tissues); gene expression from neurons; astrocytes and oligodendrocytes from developing and mature mouse forebrain⁴⁷; and gene expression from 149 mouse brain cell types (KI level 1: 24 broad categories, KI level 2: 149 cell types underlying the 24 broad categories)⁵⁰.

Using MAGMA⁵⁴, the majority of brain tissues in GTEx were significantly associated with anorexia nervosa (**Supplementary Fig. 9**), the strongest hits being brain cerebellum and brain cerebellar hemispheres. Enrichment in muscle-skeletal tissues also appeared likely (**Supplementary Fig. 9**). Medium spiny neurons and pyramidal neurons from the CA1 region of the hippocampus were significantly associated with anorexia nervosa at a Bonferroni threshold in the 24 level 1 cell types (**Supplementary Fig. 10**). Among the 149 level 2 cell types, pyramidal neurons from the CA1 region of the hippocampus and pyramidal neurons from the somatosensory cortex layer 5 were Bonferroni-significant (**Supplementary Fig. 11**).

Medium spiny neurons (MSNs) are dopaminergic and inhibitory. They are the primary cell type of the striatum. The dorsal striatum has been linked to feeding behaviors including food

motivation and reward⁸³. D1R-medium-spiny-neurons (medium spiny neurons that express the D1-type dopamine receptor) afferents have been reported to be the primary source of nucleus accumbent inhibition to the lateral hypothalamus⁸⁴. Furthermore, in that study, inhibition of D1R-MSNs increased feeding, while activation decreased feeding⁸⁴. Pyramidal neurons are excitatory and are the primary excitatory cell type found in cortical structures⁸⁵. Kim et al.⁸⁶ recently found that *PPP1R1B*-expressing pyramidal neurons from the basolateral amygdala project to DR1-expressing central amygdala neurons, which are known to modulate appetitive behaviors.

Using the LDSC partitioned heritability approach⁵³, no significant signal was found in the GTEx database for tissues (**Supplementary Figs. 12-13**), in the Cahoy database for cells (**Supplementary Fig. 14**), or in the single-cell RNA-sequencing database (**Supplementary Fig. 15**). Nevertheless, in GTEx, the enrichment of heritability was more common in cell types in the brain, although enrichment in muscle-skeletal tissues was also evident, though no cell type reached significance (**Supplementary Fig. 13**). In the Cahoy database, the enrichment appeared more common in astrocytes compared with neurons and oligodendrocytes, although none of these reached statistical significance after Bonferroni correction (**Supplementary Fig. 14**). In the single-cell RNA-sequencing database, signal was strongest in neuroblasts (**Supplementary Fig. 15**).

Predicted tissue-specific gene expression

The PrediXcan analysis suggested significant effects on the expression of 36 genes across 44 GTEx tissues ($P \leq 1.94 \times 10^{-7}$; **Supplementary Table 13**). *MGMT* is located on chr10 (131.2-131.4 Mb). The majority of others are located within the multi-genic region on chr3. Downregulation and upregulation are presented in **Supplementary Table 13**.

Cross-trait analysis

Genetic correlations

Full results are shown in **Supplementary Table 10** and Bonferroni-significant results are shown graphically in **Fig. 2**. In instances where the same phenotype appears in multiple study sources, the main manuscript and **Fig. 2** report the result from the study source with the largest sample and/or of European ancestry.

Generalized summary data-based Mendelian randomization (GSMR)

BMI. We used GSMR⁸⁷ to investigate causal associations between BMI and anorexia nervosa using an extension of GCTA⁵⁷ (**Supplementary Table 16**). A one standard deviation (*SD*) decrease in genetically-estimated BMI increased the risk for anorexia nervosa by 4% (OR = 1.04; s.e. = 0.01; $P_{\text{GSMR}} = 0.008$) while an increase in genetically-estimated anorexia nervosa had a BMI-lowering effect ($b_{xy} = -0.28$, s.e. = 0.07, $P_{\text{GSMR}} = 5.15 \times 10^{-5}$). We only used eight SNPs to build the multiple SNP instrument for anorexia nervosa as an exposure and, hence, these results should be interpreted with caution.

To further separate effects of BMI from the anorexia nervosa phenotype, we used GCTA-mtCOJO^{57,87} (multi-trait-based conditional & joint analysis using GWAS summary data) to adjust the anorexia nervosa GWAS summary statistics for BMI, using BMI summary statistics

from our UK Biobank analysis, which excluded individuals with a mental health diagnosis or taking a psychiatric or weight changing medication (see **Additional Methods**) and re-ran the GSMR analysis. The anorexia nervosa and BMI GWAS were performed on independent samples. As expected, after conditioning on BMI, the bidirectional pattern was no longer observable with $OR_{BMI \rightarrow AnorexiaNervosaAdjBMI} = 1.00$ (s.e. = 0.01; $P_{GSMR} = 0.78$). However, the putative causal association from AnorexiaNervosaAdjBMI to BMI was still statistically significant ($b_{xy} = -0.22$, s.e. = 0.08, $P_{GSMR} = 0.004$). The results are consistent with a causal relationship not due to pleiotropy (in the case of the anorexia nervosa \rightarrow BMI effect) between these two traits.

Type 2 diabetes. We also investigated the causal relationship between Type 2 diabetes and anorexia nervosa using GSMR. The association with Type 2 diabetes as an exposure and anorexia nervosa as an outcome was not statistically significant ($b_{xy} = -0.02$, s.e. = 0.03, $P_{GSMR} = 0.035$), nor was the association with anorexia nervosa as an exposure and Type 2 diabetes as an outcome ($b_{xy} = -0.09$, s.e. = 0.09, $P_{GSMR} = 0.30$). These results do not support either phenotype as having a putative causal effect on the other. The analysis with anorexia nervosa as an exposure had only 7 instrumental variables rather than the recommended minimum of 10, therefore results are to be interpreted cautiously and may change once the anorexia nervosa GWAS sample size increases.

References

1. Duncan, L. *et al.* Significant locus and metabolic genetic correlations revealed in genome-wide association study of anorexia nervosa. *Am. J. Psychiatry* **174**, 850-858 (2017).
2. Boraska, V. *et al.* A genome-wide association study of anorexia nervosa. *Mol. Psychiatry* **19**, 1085-1094 (2014).
3. Kaye, W.H. *et al.* A search for susceptibility loci for anorexia nervosa: methods and sample description. *Biol. Psychiatry* **47**, 794-803 (2000).
4. Kaye, W.H. *et al.* Genetic analysis of bulimia nervosa: methods and sample description. *Int. J. Eat. Disord.* **35**, 556-570 (2004).
5. Reba, L. *et al.* Relationships between features associated with vomiting in purging-type eating disorders. *Int. J. Eat. Disord.* **38**, 287-294 (2005).
6. Wang, K. *et al.* A genome-wide association study on common SNPs and rare CNVs in anorexia nervosa. *Mol. Psychiatry* **16**, 949-959 (2011).
7. Hou, L. *et al.* Genome-wide association study of 40,000 individuals identifies two novel loci associated with bipolar disorder. *Hum. Mol. Genet.* **25**, 3383-3394 (2016).
8. Mühleisen, T.W. *et al.* Genome-wide association study reveals two new risk loci for bipolar disorder. *Nat. Commun.* **5**, 3339 (2014).
9. Stahl, E. *et al.* Genomewide association study identifies 30 loci associated with bipolar disorder. *bioRxiv*, 173062.1 (2018).
10. McKay, J.D. *et al.* A genome-wide association study of upper aerodigestive tract cancers conducted within the INHANCE consortium. *PLoS Genet.* **7**, e1001333 (2011).
11. Kirk, K.M. *et al.* The Anorexia Nervosa Genetics Initiative: study description and sample characteristics of the Australian and New Zealand arm. *Aust. N. Z. J. Psychiatry* **51**, 583-594 (2017).
12. Thornton, L.M. *et al.* The Anorexia Nervosa Genetics Initiative (ANGI): overview and methods. *Contemp. Clin. Trials* **74**, 61-69 (2018).
13. Olsen, C.M. *et al.* Cohort profile: the QSkin Sun and Health Study. *Int. J. Epidemiol.* **41**, 929-929i (2012).
14. World Health Organization. *ICD-10: international statistical classification of diseases and related health problems: 10th revision*, (World Health Organization, Geneva, 1992).
15. Mors, O., Perto, G.P. & Mortensen, P.B. The Danish Psychiatric Central Research Register. *Scand. J. Public Health* **39**, 54-57 (2011).
16. Pedersen, C.B. *et al.* The iPSYCH2012 case-cohort sample: new directions for unravelling genetic and environmental architectures of severe mental disorders. *Mol. Psychiatry* **23**, 6-14 (2018).
17. Illumina. *Illumina GenCall Data Analysis Software*, Number of (2005) Available at: https://support.illumina.com/content/dam/illumina-marketing/documents/products/technotes/technote_gencall_data_analysis_software.pdf (Accessed: March 18, 2019).
18. Wysoker, J.N. *et al.* Integrated genotype calling and association analysis of SNPs, common copy number polymorphisms and rare CNVs. *Nat. Genet.* **40**, 1253-1260 (2008).

19. Swedish Association of Local Authorities and Regions. *National healthcare quality registries in Sweden*, Number of (Edita, Stockholm, 2007) Available at: <https://webbutik.skl.se/bilder/artiklar/pdf/7164-096-7.pdf> (Accessed: March 18, 2019).
20. Almqvist, C. *et al.* LifeGene—a large prospective population-based study of global relevance. *Eur. J. Epidemiol.* **26**, 67-77 (2011).
21. American Psychiatric Association. *Diagnostic and statistical manual of mental disorders, text revision*, (American Psychiatric Association, Washington, DC, 2000).
22. Jonassaint, C.R. *et al.* Absence of association between specific common variants of the obesity-related FTO gene and psychological and behavioral eating disorder phenotypes. *Am. J. Med. Genet. B Neuropsychiatr. Genet.* **156B**, 454-461 (2011).
23. Sudlow, C. *et al.* UK biobank: an open access resource for identifying the causes of a wide range of complex diseases of middle and old age. *PLoS Med.* **12**, e1001779 (2015).
24. Smoller, J.W. *et al.* Psychiatric genetics and the structure of psychopathology. *Mol. Psychiatry* **24**, 409-420 (2019).
25. Hudson, J.I., Hiripi, E., Pope, H.G. & Kessler, R.C. The prevalence and correlates of eating disorders in the National Comorbidity Survey Replication. *Biol. Psychiatry* **61**, 348-358 (2007).
26. Ripke, S. *et al.* Genome-wide association analysis identifies 13 new risk loci for schizophrenia. *Nat. Genet.* **45**, 1150-1159 (2013).
27. Ripke, S. *et al.* Genome-wide association study identifies five new schizophrenia loci. *Nat. Genet.* **43**, 969-976 (2011).
28. Schizophrenia Working Group of the Psychiatric Genomics Consortium *et al.* Biological insights from 108 schizophrenia-associated genetic loci. *Nature* **511**, 421-427 (2014).
29. 1000 Genomes Project Consortium *et al.* An integrated map of genetic variation from 1,092 human genomes. *Nature* **491**, 56-65 (2012).
30. International HapMap 3 Consortium. Integrating common and rare genetic variation in diverse human populations. *Nature* **467**, 52-58 (2010).
31. Bulik-Sullivan, B.K. *et al.* LD Score regression distinguishes confounding from polygenicity in genome-wide association studies. *Nat. Genet.* **47**, 291-295 (2015).
32. Bulik-Sullivan, B. *et al.* An atlas of genetic correlations across human diseases and traits. *Nat. Genet.* **47**, 1236-1241 (2015).
33. American Psychiatric Association. *Diagnostic and statistical manual of mental disorders (DSM-5®)*, (American Psychiatric Association, Washington, DC, 2013).
34. Sullivan, P.F., Bulik, C.M. & Kendler, K.S. Genetic epidemiology of bingeing and vomiting. *Br. J. Psychiatry* **173**, 75-79 (1998).
35. Hübel, C. *et al.* Genomics of body fat percentage may contribute to sex bias in anorexia nervosa. *Am. J. Med. Genet. B Neuropsychiatr. Genet.* **in press**.
36. Purves, K.L. *et al.* The common genetic architecture of anxiety disorders. *bioRxiv*, 203844 (2017).
37. McCarthy, S. *et al.* A reference panel of 64,976 haplotypes for genotype imputation. *Nat. Genet.* **48**, 1279 (2016).
38. Bycroft, C. *et al.* The UK Biobank resource with deep phenotyping and genomic data. *Nature* **562**, 203-209 (2018).
39. Townsend, P. Deprivation. *J. Soc. Policy* **16**, 125-146 (1987).
40. Willer, C.J., Li, Y. & Abecasis, G.R. METAL: fast and efficient meta-analysis of genomewide association scans. *Bioinformatics* **26**, 2190-2191 (2010).

41. Doherty, A. *et al.* Large scale population assessment of physical activity using wrist worn accelerometers: the UK Biobank study. *PLoS One* **12**, e0169649 (2017).
42. Kessler, R.C., Andrews, G., Mroczek, D., Ustun, B. & Wittchen, H.-U. The World Health Organization composite international diagnostic interview short-form (CIDI-SF). *Int. J. Methods Psychiatr. Res.* **7**, 171-185 (1998).
43. Davis, K.A. *et al.* Mental health in UK Biobank: development, implementation and results from an online questionnaire completed by 157 366 participants. *BJPsych Open* **4**, 83-90 (2018).
44. Smith, D.J. *et al.* Prevalence and characteristics of probable major depression and bipolar disorder within UK biobank: cross-sectional study of 172,751 participants. *PloS One* **8**, e75362 (2013).
45. Eysenck, S.B., Eysenck, H.J. & Barrett, P. A revised version of the psychoticism scale. *Pers. Individ. Dif.* **6**, 21-29 (1985).
46. GTEx Consortium. Genetic effects on gene expression across human tissues. *Nature* **550**, 204-213 (2017).
47. Cahoy, J.D. *et al.* A transcriptome database for astrocytes, neurons, and oligodendrocytes: a new resource for understanding brain development and function. *J. Neurosci.* **28**, 264-278 (2008).
48. Lindblad-Toh, K. *et al.* A high-resolution map of human evolutionary constraint using 29 mammals. *Nature* **478**, 476-482 (2011).
49. Finucane, H. *et al.* Heritability enrichment of specifically expressed genes identifies disease-relevant tissues and cell types. *Nat. Genet.* **50**, 621-629 (2018).
50. Skene, N.G. *et al.* Genetic identification of brain cell types underlying schizophrenia. *Nat. Genet.* **50**, 825-833 (2018).
51. Lun, A.T.L., McCarthy, D.J. & Marioni, J.C. A step-by-step workflow for low-level analysis of single-cell RNA-seq data with Bioconductor. *F1000Res.* **5**(2016).
52. Vu, T.N. *et al.* Beta-poisson model for single-cell RNA-seq data analyses. *Bioinformatics* **32**, 2128-2135 (2016).
53. Finucane, H.K. *et al.* Partitioning heritability by functional annotation using genome-wide association summary statistics. *Nat. Genet.* **47**, 1228-1235 (2015).
54. de Leeuw, C.A., Mooij, J.M., Heskes, T. & Posthuma, D. MAGMA: generalized gene-set analysis of GWAS data. *PLoS Comput. Biol.* **11**, e1004219 (2015).
55. Barbeira, A.N. *et al.* Exploring the phenotypic consequences of tissue specific gene expression variation inferred from GWAS summary statistics. *Nat. Commun.* **9**, 1825 (2018).
56. Battle, A. *et al.* Characterizing the genetic basis of transcriptome diversity through RNA-sequencing of 922 individuals. *Genome Res.* **24**, 14-24 (2014).
57. Yang, J., Lee, S.H., Goddard, M.E. & Visscher, P.M. GCTA: a tool for genome-wide complex trait analysis. *Am. J. Hum. Genet.* **88**, 76-82 (2011).
58. Yang, J. *et al.* Conditional and joint multiple-SNP analysis of GWAS summary statistics identifies additional variants influencing complex traits. *Nat. Genet.* **44**, 369-75, S1-3 (2012).
59. Loh, P.-R. *et al.* Efficient Bayesian mixed-model analysis increases association power in large cohorts. *Nat. Genet.* **47**, 284-290 (2015).
60. Locke, A.E. *et al.* Genetic studies of body mass index yield new insights for obesity biology. *Nature* **518**, 197 (2015).

61. Yang, J. *et al.* Genomic inflation factors under polygenic inheritance. *Eur. J. Hum. Genet.* **19**, 807-812 (2011).
62. Ioannidis, J.P., Ntzani, E.E., Trikalinos, T.A. & Contopoulos-Ioannidis, D.G. Replication validity of genetic association studies. *Nat. Genet.* **29**, 306-309 (2001).
63. Rathjen, T. *et al.* Regulation of body weight and energy homeostasis by neuronal cell adhesion molecule 1. *Nat. Neurosci.* **20**, 1096-1103 (2017).
64. Bacon, C. *et al.* Brain-specific Foxp1 deletion impairs neuronal development and causes autistic-like behaviour. *Mol. Psychiatry* **20**, 632-639 (2015).
65. Matsumoto, M. *et al.* An evolutionarily conserved G-protein coupled receptor family, SREB, expressed in the central nervous system. *Biochem. Biophys. Res. Commun.* **272**, 576-582 (2000).
66. Ku, G.M., Pappalardo, Z., Luo, C.C., German, M.S. & McManus, M.T. An siRNA screen in pancreatic beta cells reveals a role for Gpr27 in insulin production. *PLoS Genet.* **8**, e1002449 (2012).
67. Zhu, D. *et al.* Protein S controls hypoxic/ischemic blood-brain barrier disruption through the TAM receptor Tyro3 and sphingosine 1-phosphate receptor. *Blood* **115**, 4963-4972 (2010).
68. Cantagrel, V. *et al.* Mutations in the cilia gene ARL13B lead to the classical form of Joubert syndrome. *Am. J. Hum. Genet.* **83**, 170-179 (2008).
69. Thomas, S. *et al.* Identification of a novel ARL13B variant in a Joubert syndrome-affected patient with retinal impairment and obesity. *Eur. J. Hum. Genet.* **23**, 621-627 (2015).
70. Higginbotham, H. *et al.* Arl13b in primary cilia regulates the migration and placement of interneurons in the developing cerebral cortex. *Dev. Cell* **23**, 925-938 (2012).
71. Okbay, A. *et al.* Genome-wide association study identifies 74 loci associated with educational attainment. *Nature* **533**, 539-542 (2016).
72. Morris, A.P. *et al.* Large-scale association analysis provides insights into the genetic architecture and pathophysiology of type 2 diabetes. *Nat. Genet.* **44**, 981-990 (2012).
73. Teslovich, T.M. *et al.* Biological, clinical and population relevance of 95 loci for blood lipids. *Nature* **466**, 707-713 (2010).
74. Benjamini, Y. & Hochberg, Y. Controlling the false discovery rate: a practical and powerful approach to multiple testing. *J. R. Stat. Soc. Series B Stat. Methodol.* **57**, 289-300 (1995).
75. Morikawa, T. *et al.* Prospective analysis of body mass index, physical activity, and colorectal cancer risk associated with beta-catenin (CTNNB1) status. *Cancer Res.* **73**, 1600-1610 (2013).
76. Laramie, J.M. *et al.* Polymorphisms near EXOC4 and LRGUK on chromosome 7q32 are associated with type 2 Diabetes and fasting glucose; the NHLBI Family Heart Study. *BMC Med. Genet.* **9**, 46 (2008).
77. Walford, G.A. *et al.* Genome-wide association study of the modified Stumvoll Insulin Sensitivity Index identifies BCL2 and FAM19A2 as novel insulin sensitivity loci. *Diabetes* **65**, 3200-3211 (2016).
78. Dhar, M.S., Yuan, J.S., Elliott, S.B. & Sommardahl, C. A type IV P-type ATPase affects insulin-mediated glucose uptake in adipose tissue and skeletal muscle in mice. *J. Nutr. Biochem.* **17**, 811-820 (2006).

79. Chenn, A. & Walsh, C.A. Regulation of cerebral cortical size by control of cell cycle exit in neural precursors. *Science* **297**, 365-369 (2002).
80. Waite, A., Brown, S.C. & Blake, D.J. The dystrophin–glycoprotein complex in brain development and disease. *Trends Neurosci.* **35**, 487-496 (2012).
81. Satz, J.S. *et al.* Distinct functions of glial and neuronal dystroglycan in the developing and adult mouse brain. *J. Neurosci.* **30**, 14560-14572 (2010).
82. Früh, S. *et al.* Neuronal dystroglycan is necessary for formation and maintenance of functional CCK-positive basket cell terminals on pyramidal cells. *J. Neurosci.* **36**, 10296-10313 (2016).
83. Berridge, K.C., Ho, C.Y., Richard, J.M. & DiFeliceantonio, A.G. The tempted brain eats: pleasure and desire circuits in obesity and eating disorders. *Brain Res.* **1350**, 43-64 (2010).
84. O'Connor, E.C. *et al.* Accumbal D1R neurons projecting to lateral hypothalamus authorize feeding. *Neuron* **88**, 553-564 (2015).
85. Spruston, N. Pyramidal neurons: dendritic structure and synaptic integration. *Nat. Rev. Neurosci.* **9**, 206-221 (2008).
86. Kim, J., Zhang, X., Muralidhar, S., LeBlanc, S.A. & Tonegawa, S. Basolateral to central amygdala neural circuits for appetitive behaviors. *Neuron* **93**, 1464-1479.e5 (2017).
87. Zhu, Z. *et al.* Causal associations between risk factors and common diseases inferred from GWAS summary data. *Nat. Commun.* **9**, 224 (2018).