

1 Derived *Homo sapiens* cis-eQTL regulation:
2 implications for brain evolution

3 Alejandro Andirkó^{1,2} and Cedric Boeckx^{1,2,3,*}

4 ¹University of Barcelona

5 ²University of Barcelona Institute of Complex Systems

6 ³ICREA

7 *Corresponding author: cedric.boeckx@ub.edu

8 **Abstract**

9 The high-quality sequence of the genomes of our extinct relatives, Ne-
10 anderthals and Denisovans, became recently public. At the same time,
11 we have seen the emergence of big databases of modern human genetic
12 variation. However, linking human genetic variation, neuronal phenotypes
13 and, eventually, behaviour, is only possible if we understand how variation
14 and genetic regulation interact. We used two publicly available datasets,
15 the GTEX cis-eQTL database (v7) and a catalog of high-frequency *Homo*
16 *Sapiens* specific alleles relative to the Neanderthals and Denisovan se-
17 quences, to understand how high-frequency *Homo Sapiens* derived alleles
18 affect gene expression regulation. The resulting dataset shows that genes
19 associated with brain development are affected by *Homo sapiens*-specific
20 eQTL in brain areas key in human evolution such as the cerebellum. We
21 also show that some of these eQTL overlap significantly with putative

22 regions of positive selection relative to archaic humans [Peyr gne et al.,
23 2017]. Additionally, we tested whether any of the variants are associated
24 with clinical conditions in modern human populations. These findings can
25 inform future experimental work and enrich current venues of research of
26 the *Homo Sapiens* brain evolution, such as the relationship between clini-
27 cal and evolutionary research and the recent expansion of the cerebellum
28 in *Homo Sapiens* [Gunz et al., 2010].

29 **Keywords**— Human evolution, eQTL, brain evolution

30 1 Introduction

31 The high-quality DNA sequencing of two Neanderthal individuals and a Deniso-
32 van from Altai and Vindija [Pr ufer et al., 2014, 2017, Meyer et al., 2012] has
33 opened numerous research avenues and opportunities for studying the evolution
34 of the *Homo sapiens* brain with unprecedented resolution and precision. Geo-
35 metric morphometric analysis on endocasts [Bruner et al., 2014, Gunz et al.,
36 2010] have already suggested that the differences between Neanderthal and
37 *Homo sapiens* skulls could be related to changes in neural tissue that might
38 in turn have had consequences for the evolution of human cognition. In par-
39 ticular, the temporal and the parietal lobes, as well as the cerebellum, have
40 been claimed to have expanded during the emergence of *Homo sapiens* evolu-
41 tion [Gunz et al., 2010]. The sequencing of ancient human genomes made it
42 possible perform selective sweep scans to detect areas of the genome that have
43 been significantly affected by natural selection after the split with Neanderthals
44 [Racimo et al., 2014, Peyr gne et al., 2017]. Much of the early efforts relied on
45 determining the function on the few missense mutations that are *Homo sapiens*-
46 specific. However, to characterize the effects of *Homo sapiens*-specific variants
47 we should target not only the variants that affect the structure and function of

48 the protein, but also those that regulate gene expression. Species-specific regu-
49 lation of genetic expression levels might also play a big role in determining the
50 modern human brain phenotype [Franchini and Pollard, 2017, Gokhman et al.,
51 2016].

52 In order to link genotype and brain phenotype, previous work has explored
53 the idea of connecting modern human variation data and brain evolution. A ma-
54 jor study [McCoy et al., 2017] explored the effects of Neanderthal and Denisovan
55 introgressed variants in 44 tissues and found reduced differential expression of
56 Neanderthal alleles in the cerebellum and the striatum. In a similar vein, an-
57 other study [Gunz et al., 2019] examined the effects of archaic introgression on
58 brain skull shape variability in modern human populations to determine which
59 variants are more associated with a globularized skull shape, a feature of *Homo*
60 *sapiens* skulls.

61 Expanding on these approaches, we aim to understand the effects of the de-
62 rived *Homo sapiens*-specific alleles in gene regulation in brain tissue. We took
63 advantage of a recent systematic review of high frequency changes in modern
64 humans [Kuhlwilm and Boeckx, 2019], which offered not only a catalog of *Homo*
65 *sapiens*-specific fixed missense mutations, but an exhaustive database with all
66 the derived *Homo sapiens* unique alleles in high frequency (90% or more cutoff).
67 We crossed this database with the GTEx version 7 database, which contains
68 information on Expression Quantitative Trait Loci (cis-eQTLs) accross tissues,
69 focusing exclusively on brain tissue. Compared to approaches that have used
70 the GTEx data in light of introgression [McCoy et al., 2017, Gunz et al., 2019],
71 our focus on the effect of the derived allele allows this study to span the whole
72 genome, as opposed to introgressed regions only. Additionally, the data gener-
73 ated, unlike in McCoy et al. [2017], doesn't focus on neanderthal-specific alleles
74 but on the effect of the derived, modern-specific allele (as defined in [Kuhlwilm

75 and Boeckx, 2019]).

76 Our results show differential regulation by derived cis-eQTLs in key brain
77 areas and circuits that are claimed to have changed significantly during the
78 emergence of *Homo sapiens*, such as the cerebellum and olfactory signalling
79 pathway [Bastir et al., 2011]. We also found genetic regulation of cellular pro-
80 cesses that can potentially impact neurodevelopment and disease, such as fo-
81 late one-carbon metabolism, aerobic glycolysis regulation, cell-cell adhesion and
82 regulation of cytoskeleton dynamics. Some of these processes, such as aerobic
83 glycolysis, have already been discussed in other studies in the context of human
84 evolution and human-specific diseases such as Alzheimer’s Disease [Bufill et al.,
85 2011]. While we limited the scope of our study to brain tissue, we found as
86 well eQTLs associated with neural crest cell development and the craniofacial
87 complex. This is of special interest, as the *Homo sapiens* face has a distinct re-
88 tracted profile compared to that of archaic humans [Lacruz et al., 2019]. Some
89 of the genes that affect brain development might exert an influence in adjacent
90 tissues [Boeckx, 2017].

91 Since the eQTLs affect genes previously identified as under selective sweep in
92 modern humans relative to archaic humans, we tested whether the eQTLs affect
93 genetic regions identified in human positive selection studies. We found *Homo*
94 *sapiens*-specific eQTLs overlap significantly with regions of positive selection
95 [Peyrégne et al., 2017] in a permutation test (n=10000) (Supp. Fig. 2). We
96 also explored whether the eQTL-associated genes were enriched in transcription
97 factor motifs with a known functionality, as well as their relationship with known
98 clinical conditions.

99 Overall, the results offer a landscape of *Homo sapiens* specific gene regu-
100 lation, enriching previously explored venues of research of the field of human
101 evolution studies and opening new possibilities for variant effect testing in the

102 context of human brain evolution.

103 2 Results

104 We used the publicly available GTEx v7 eQTLs lists, which consist of statis-
105 tically significant allele specific expression changes (as defined by GTEx [The
106 GTEx Consortium et al., 2015]) influencing gene expression dosage in 13 differ-
107 ent brain tissues from adult brain samples aged 20-60.

108 We first selected those eQTLs where the minor allele was also the ancestral
109 one (as defined by Kuhlwilm and Boeckx [2019]). The catalog of derived gene
110 changes in *Homo sapiens* we used for this study [Kuhlwilm and Boeckx, 2019]
111 imposes an arbitrary cutoff of a global 90% that allows some frequencies to be
112 under 90% as long as the overall mean fulfills the requirement. We processed
113 the data so that only those alleles that fulfill the 90% threshold in all metapop-
114 ulations (as detailed in Kuhlwilm and Boeckx [2019]) are included in order to
115 apply a more stringent filter. Finally, crossing this database with the GTEx
116 allele-specific genetic regulation datasets, we retrieved all the *Homo sapiens* de-
117 rived alleles that had a statistically significant effect in gene expression in any of
118 the adult human brain tissues and were in high frequency in modern population.

119 The resulting data includes 1,357 statistically significant unique SNPs associ-
120 ated with regulation of a total of 316 eGenes (i.e., genes affected by cis-regulation
121 by at least one SNP) whose major allele falls into at least 90% frequency in all
122 modern human metapopulations . As expected, we find some degree of func-
123 tional overlap: 266 of those 1,357 eQTLs regulate various genes in the same
124 tissue or act across different tissues. If we don't account for functional overlap
125 we count a total of 2,507 eQTLs across 13 brain tissues (Supp. Tabl. 1).

126 Each eQTL has a beta score assigned to it, i.e., a normalized score of the
127 modulatory effect of the SNP in the eGene normalized expression levels, as

rsID	Gene exon	Additional notes
rs17597625	<i>OR5I1</i>	Under positive selection in early selective sweep [Moreno-Estrada et al., 2007]
rs2917782	<i>STARD6</i>	Cerebellar development [Chang et al., 2012] and Alzheimer's Disease [Yin et al., 2019]
rs2917782	<i>PCNT</i>	Associated with Down Syndrome and dwarfism [Rauch et al., 2008]
rs74316182	<i>PLD4</i>	Influences myelination [Chiba et al., 2016]

Table 1: A summary of missense variants in high frequency in modern humans where the variant also differentially affects cis-regulation of gene expression.

128 detailed in the GTEx consortium documentation [The GTEx Consortium et al.,
129 2015]. In this case, the effect of an eQTL is understood as the relative gene
130 expression difference between the minor, ancestral allele and the high-frequency
131 derived allele. As shown in Figure 4 (C), chromosome 13 is underrepresented in
132 terms of eQTL regulation.

133 In terms of distance to the transcription starting site, eQTLs tend to accu-
134 mulate near the TSS (Transcription Start Site) of the regulated eGene, with a
135 slight upward tendency (Fig. 4 A). This is in accordance with the GTEx con-
136 sortium data report [The GTEx Consortium et al., 2015]: the GTEx consortium
137 data exhibits an overall tendency towards transcriptional regulation instead of
138 post-transcriptional regulation. We additionally report a tail of downstream
139 variants far away from the TSS.

140 When listed by number of occurrences, the gene that accumulates more vari-
141 ability from eQTL regulation is *METTL18* (296 SNPs), followed by *METTL14*
142 (72 SNPs, all affecting gene cis-regulation in the cerebellum), *POLI* (47), a
143 DNA repair gene [Jain et al., 2017], and *KATNAL2* (41), a gene implicated in
144 ciliogenesis and neurodevelopment in *Xenopus* [Willsey et al., 2018]. The full
145 list of occurrences per gene can be found in Supp. Table 1. Overall, close to
146 45% of SNPs accumulate in introns, consistently with the initial GTEx database
147 reports (Fig. 4 B) [The GTEx Consortium et al., 2015], while the second most
148 abundant subgroup is non-coding transcripts.

149 We also found 15 missense mutations overlapping with cis-eQTLs, from

Tissue	Up	Down	Total	Tissue	Up	Down	Total
Adrenal gland	78	193	271	Cerebellar hemisphere	101	148	249
Amygdala	7	47	54	Hippocampus	7	44	51
BA9	22	64	86	Hypothalamus	12	74	86
BA24	82	102	184	Nucleus Accumbens	44	84	128
Caudate cortex	33	113	146	Pituitary gland	157	491	648
Cerebellum	115	348	463	Putamen	20	98	118
Substantia Nigra	3	20	23				

Table 2: Summary of up and downregulating high-frequency eQTLs in each brain tissue. BA = Brodmann’s Area.

150 which four affect genes that have appeared in the literature related to neurode-
151 velopment, evolution or disease (Table 1). Some of these missense mutations,
152 such as rs74316182, regulate different eGenes in each tissue (Fig. 4 B). Addi-
153 tionally, there is one stop gained (rs73600054) in a cilium and flagella related
154 gene (*CFAP157*) [Weidemann et al., 2016] and one lost (rs80336323) in *PSPN*
155 (persephin neurotrophic factor).

156 We found a splice region variant (rs17801742) in *COL2A1*, linked to Stickler
157 Syndrome (OMIM entry number: #108300) [Guo et al., 2017] and regulating
158 *PIGV*, a gene linked to neural crest cell regulation [Horn et al., 2014] that falls
159 within a region of modern human positive selection [Racimo et al., 2015]. The
160 full list of consequences can be consulted in Supp. Tab. 2.

161 2.1 Directional regulation of eGenes

162 To check if there was a significant directionality of regulation in any genetic
163 region or tissue we divided cis-eQTL variants in up and down effects per tis-
164 sue. (Fig. 4 C , Table 2). We performed a two sample Kolmogorov–Smirnov
165 test under the null hypothesis that the distribution of the high frequency *Homo*
166 *sapiens* derived eQTLs follows a similar distribution as the GTEx significant
167 eQTLs. The test does not reject that the upregulating eQTLs follow the same
168 distribution as the total GTEx significant variants per tissue, which would be

169 expected considering that the high-frequency eQTL list is a subset of the to-
170 tal of significant variants in each tissue. However, it reports a significant (p
171 $= 0.04979$, maximum $D^2 = 0.5$) difference in the distribution of the downregu-
172 lating eQTLs relative to the original GTEx v7 lists of significant variants per
173 tissue. Additionally, the data shows a numeric bias towards downregulating
174 eQTLs in all tissues, as visualized in Figure 3 C: in total, 681 alleles are linked
175 to gene upregulation, while 1,826 are linked to gene downregulation across tis-
176 sues. We tested whether the amount of downregulating to upregulating alleles
177 in the *Homo sapiens* derived data follows the same proportions as the down
178 to upregulating ratios in the unfiltered significant eQTL list with a chi-square
179 goodness-to-fit test. The results show a significant difference ($p < 0.05$) in the
180 down to upregulating ratio compared to the unfiltered GTEx list in all tissues
181 except Brodmann's Area 24 (p-values reported in Supp. Table 3). Note that
182 the allele specific expression measures the effect of the major allele in the GTEx
183 database. Since one of the filters we impose is an equivalency between derived
184 and major allele, in a downregulating eQTL the (ancestral) low frequency allele
185 is decreasing the dosage of the eGene relative to the high-frequency *Homo sapi-*
186 *ens*-derived allele. By contrast, upregulation of an eGene by an eQTL means
187 that the ancestral allele is increasing gene dosage relative to the high-frequency
188 derived allele. Given this, the strong bias towards downregulated eQTLs could
189 be reflecting our very limited sample of archaic human genomes. This is consis-
190 tent with previous results of a study that uses similar resources [McCoy et al.,
191 2017], which reports that Neanderthal alleles reduced genetic expression in the
192 modern human brain. Chromosome Y is depleted of eQTLs in our results for the
193 same reason, as there is no high-coverage genome of a male archaic individual
194 available at the moment (Fig. 4 C).

195 In terms of the relative proportion of eGenes to eQTLs, cerebellum and pi-

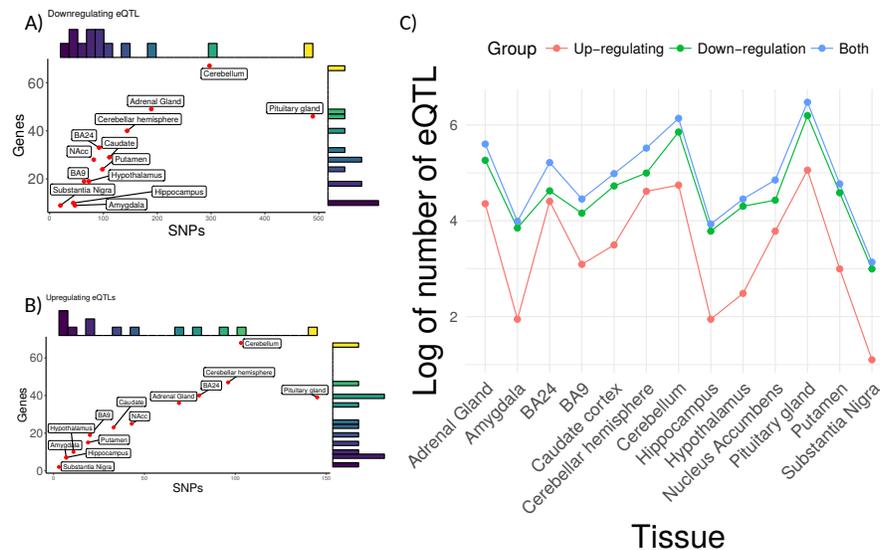


Figure 1: A) and B) Ratio of downregulating (A) and upregulating (B) eQTLs per genes affected in single tissues. C) Log value distribution in upregulating eQTL in human evolution, downregulating and total GTEx original data per tissue.

196 pituitary gland stand out both in up and downregulated eQTLs. (Figs. 3 A,
 197 B). While the cerebellar phenotype has been claimed to be derived in mod-
 198 ern humans [Gunz et al., 2019], the pituitary has not been highlighted by any
 199 experimental study regarding human evolution to our knowledge. However, a
 200 recent account of human evolution has put forward the idea that changes in the
 201 hypothalamic–pituitary–adrenal (HPA) axis could be driving the *Homo sapi-*
 202 *ens* sociocognitive profile (the so-called self-domestication process [Wrangham,
 203 2019]).

204 Some of the SNPs with the same regulatory directionality (up or down)
 205 accumulate in linkage disequilibrium regional clusters that affect specific eGenes,
 206 such as in *METTL18* or *METTL14*. (Fig. 4 C). There are also eGenes that are

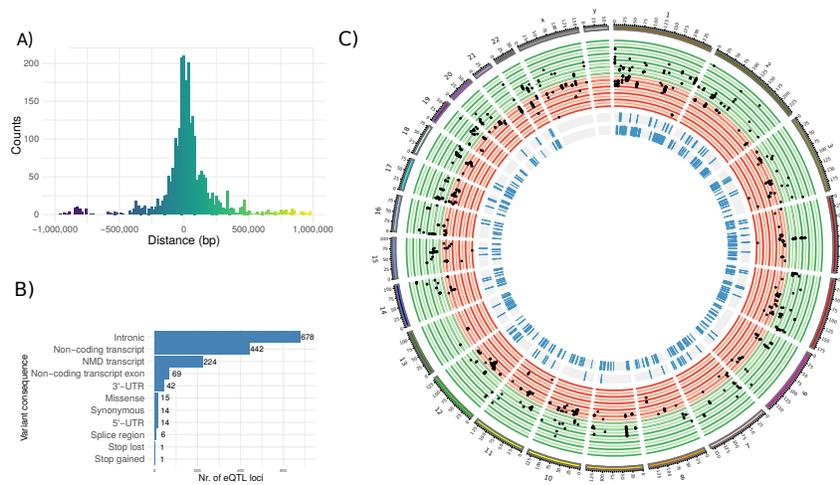


Figure 2: A) A bar plot showing the distribution of eQTL over distance to Transcription Starting Site B) Classification of the genetic consequences in our data C) A Circos plot showing the distribution along the genome of eQTLs. Each line denotes 0.5 steps in beta score (allele specific effects in gene expression), from 3 to -3. Red circles denote downregulation, green circles upregulation of eGenes. Inner rings, in blue: areas showing signals of positive selection relative to archaic humans in [Peyr egne et al., 2017] (innermost) and [Racimo et al., 2014] (outermost).

207 regulated exclusively in one tissue, both in up and downregulation of the archaic
 208 allele. Some cases where the derived allele is increasing the eGene expression in
 209 one tissue only are: *METTL14* (cerebellum), *INTS12* (adrenal gland), *STARD6*
 210 (Brodmann’s Area 24), *BCHE* (caudate) and *ENSG000000254547* (pituitary).

211 2.2 Overlap with regions of evolutionary significance

212 To identify whether previous studies could give additional evolutionary signifi-
 213 cance to any of the genes that are affected by regulation in our data, we have
 214 tested whether the eQTLs might regulate genes associated with signals of posi-

215 tive selection in modern humans versus archaics ([Peyr egne et al., 2017, Racimo
216 et al., 2014]). We ran a permutation test (n=10000) to see if the SNPs accumu-
217 late significantly in regions under positive selection relative to archaic humans
218 (Supp. Fig. 2). We found a significant ($p < 0.006$) correlation between eQTLs
219 and regions of positive selection as defined by [Peyr egne et al., 2017]. The same
220 test didn't show a significant overlap with an earlier positive selection study
221 [Racimo et al., 2014].

222 Overall, some eQTLs lie within regions of selective sweep, but the genes
223 that are affected by them aren't necessarily inside the sweep. It is possible
224 that gene expression modulation of genes outside the selected window is the one
225 driving the statistical significance of the positive selection scan, as it might be
226 targeting mutations that were on their own neutral but that affect translational
227 modification of gene expression levels outside the sweep region. Such a scenario
228 would apply for *PIGV*, *ZNF37A*, *ARID1A*, *GPN2*, *GOLGA4*, *UBR1*, *BAP1*,
229 *TMEM222*, *C3orf35*, *AC097637.2* and *AL117339.5*. An enrichment analysis
230 failed to identify a biological category for this subset of genes. Some of these
231 genes, including *PIGV*, also include other eQTLs that exert cis-regulation over
232 themselves. In order to find out if any of these genes harbouring eQTLs have had
233 a functional role in human evolution, we crossed our data with the results of Zhu
234 et al. [2018], a study that presented transcriptomic data of 16 brain tissues in
235 several developmental stages (see Supp. Tab. 5). We found that 22 of the genes
236 affected by eQTLs are also found in transcriptional heterochronic clusters, ie,
237 blocks of co-expressed genes that have changed their temporal expression profile
238 during the course of evolution relative to the macaque tissue. However, a gene
239 ontology enrichment analysis failed to return any relevant association.

240 We also tested whether any of the eQTLs fell within deserts of introgression,
241 i.e., genetic windows of at least 1,000 Mb that have resisted genetic introgression

242 from archaic humans to *Homo sapiens* ([Sankararaman et al., 2016]), as well as
243 within regions of Neanderthal or Denisovan introgression ([McCoy et al., 2017]).
244 The results show that none of our data entries overlap with these regions of
245 special evolutionary interest, and that our data points are not redundant relative
246 to other studies with a similar scope [McCoy et al., 2017]. This is most probably
247 due to the frequency cutoff imposed on the data, as introgression fragments are
248 generally low in frequency (unless adaptative).

249 **2.3 Mechanisms of regulation**

250 We hypothesized that part of the regulation of eGenes could be driven by tran-
251 scription factor binding site motif variability. We found that there is a number
252 of binding site motif enrichments in genes that are affected by eQTLs in different
253 tissues. We ran a Transcription Factor Binding Site enrichment of each tissue
254 (including both up and down eQTLs regulation) through SNP2TFBS [Kumar
255 et al., 2017]. This tool takes as input a list of rsIDs and maps each SNP to the
256 Transcription Factor Binding Site (TFBS) JASPAR database.

257 We found a significant enrichment of *RUNX1* TFBS in the cerebellum.
258 *MEIS1* and *RUNX1* regulate Hedgehog signaling through the *ATOH1* transcrip-
259 tion factor [Owa et al., 2018]. The Hedgehog signaling pathway lies at the heart
260 of the developmental program of the cerebellum [De Luca et al., 2016]. Gene
261 variants affecting this pathway might have consequences for the development of
262 cerebellar tissue. Within our eQTL list affecting gene expression in the cere-
263 bellum we focused on *METTL14*. *METTL14* is of special relevance because it
264 accumulates 72 tissue-specific eQTLs in cerebellar tissue. We hypothesized that
265 we might find that transcription factors binding to sites in *METTL14* are related
266 to cerebellar development, since the cerebellum is known to have expanded re-
267 cently in human evolution [Gunz et al., 2010]. We found that *METTL14* SNPS

268 accumulate an enrichment of *MEIS1*, *RUNX1* and retinoic acid (*RXR-RAR*)
269 (Supp. Fig. 3), possibly driving part of the statistical significance in the overall
270 tissue. Additionally, we found other tissues affected by derived eQTLs enriched
271 in transcription factor binding motifs involved in development and disease. The
272 adrenal gland and Brodmann's Area 9 derived eQTLs are enriched in *EGR1*
273 binding sites. *EGR1* is a brain growth factor that plays a key role in stress
274 regulation, schizophrenia and synaptic plasticity [Duclot and Kabbaj, 2017].
275 We also found that an enrichment in *NFATC2* TFBS in cerebellar hemisphere
276 and hypothalamus. *NFATC2* affects neuronal development and axonal growth
277 [Nguyen and Di Giovanni, 2008]. Finally, we found *STAT6*, a regulator of the
278 neurogenesis in *Xenopus* upon Amyloid-beta42 administration [Bhattarai et al.,
279 2016], in the pituitary.

280 Additionally, we explored whether derived eQTLs overlapped with any know
281 human miRNA or miRNA seeds (as defined in [Branco et al., 2018]), but found
282 no overlap with our data.

283 **2.4 Clinical data**

284 The ultimate goal of the identification of these derived eQTLs should be bridg-
285 ing genotypical variation with the evolution of phenotypes. By crossing our
286 data with the NCBI Clinvar database [Landrum et al., 2018] — a record of
287 associations of SNPs with a broad range of clinical conditions — we expected
288 to identify the effects of the eQTLs under specific genetic backgrounds to see if
289 they might affect developmental pathways of the brain (full data in Supp. Tab.
290 4).

291 Our data shows that a series of conditions associated with changes in the
292 craniofacial structure are linked to some of the modern human eQTLs. Two of
293 the downregulating eQTLs (rs17801742, rs41317939) are known to be related

294 to Stickler Syndrome (OMIM entry number: #108300), a condition that affects
295 the development of the skull and face, among other effects. Both SNPs are
296 downregulating *PFKM* expression levels in the cerebellum, a gene associated to
297 aerobic glycolysis metabolism usually found in muscles. No known link exists
298 between *PKFM* and Stickler Syndrome.

299 Another eQTL in our data is related to microcephalic osteodysplastic primor-
300 dial dwarfism (OMIM: #210720), and epileptic encephalopathy (rs79305633).
301 One of the key characteristics of microcephalic osteodysplastic primordial dwarfism
302 is microcephaly, with the retracted development of the brain starting prenatally
303 [Reference]. Other pathologies linked to the skull-brain co-development found
304 in our data are Noonan Syndrome (OMIM: #613224) and Cowden syndrome
305 (OMIM: #615108). The form of Noonan syndrome linked to the eQTL is char-
306 acterized by macrocephaly and changes in the craniofacial profile, sometimes
307 accompanied of cognitive consequences. Cowden syndrome is characterized by
308 macrocephaly as well, and it is usually linked to variants in the gene coding for
309 PTEN, a key protein in brain development. Interestingly, *PTEN* harbors an ex-
310 cess of modern-specific mutations compared to archaic humans [Kuhlwilm and
311 Boeckx, 2019]. Other variants of interest are one eQTL linked to Charcot-Marie
312 tooth disease (OMIM: #606482), variants associated with Parkinson disease
313 (OMIM: # 260300) and cerebellar ataxia (OMIM: #610743).

314 Despite these associations, a note of caution should be raised here as we don't
315 fully understand the genetic background that causes these complex disorders.
316 The eQTL-disease relationship is not necessarily causal but could contribute to
317 the epistatic background of tissues of interest.

318 **2.5 Clustered GO analysis**

319 To shed light on whether genes affected by eQTLs affect specific networks of
320 genes we performed a gene ontology (GO) clustered analysis with Metascape
321 [Zhou et al., 2019]. Metascape performs an enrichment analysis that organizes
322 hierarchically GO by establishing a measure of distance between each category
323 and then organizing resulting GO nodes in networks according to the distance
324 between each node. We performed this analysis with a list of all the genes that
325 are affected by differential regulation across all tissues. The results of the anal-
326 ysis (Supp. Fig. 1) reveal a significant number of terms related to the olfactory
327 signalling pathway, cell-cell adhesion and negative regulation of microtubule de-
328 polymerization. Geometric morphometrics studies that compared *Homo sapi-*
329 *ens* and Neanderthals [Bastir et al., 2011, Kochiyama et al., 2018] had already
330 identified that *Homo sapiens* has significantly larger olfactory bulbs compared
331 to Neanderthals. Cell-cell adhesion also has been found enriched previously
332 in a motif macroanalysis of the Neanderthal and Denisovan genomes [Cserhati
333 et al., 2018]. Cell adhesion is an important process in human-specific cortical
334 folding and apical progenitors cell cycle [Mora-Bermúdez et al., 2016, Wianny
335 et al., 2018]. Microtubule dynamics play an important role in neurodegenerative
336 diseases such as Alzheimer’s [Hernández et al., 2013] but also in learning and
337 memory [Dent, 2017].

338 **3 Discussion**

339 Previous work such as [McCoy et al., 2017, Gunz et al., 2019] had already high-
340 lighted molecular mechanisms of evolutionary change in highly-derived tissues
341 in *Homo sapiens*. Here we propose that such changes also rely on regulatory
342 changes in human evolution in several tissues. Highly derived structures in-

343 identified in previous literature, such as the cerebellum [Gunz et al., 2010] and
344 the olfactory bulbs [Kochiyama et al., 2018] are prominent in our data. The
345 GO clustered analysis revealed a significant enrichment in olfactory signalling
346 pathway related genes in our data. An early study on the subject [Bastir et al.,
347 2011] suggests that the olfactory signalling pathway overlaps significantly with
348 the memory and learning processing circuitry. Differential regulation of the ol-
349 factory circuitry might have consequences for derived aspects of *Homo sapiens*
350 cognition.

351 We also found that the cerebellum accumulates more differential regulation
352 than the rest of tissues (Figs. 3 A, B). In terms of molecular mechanisms, our
353 results highlight *METTL14* as a central node of regulation of cerebellar devel-
354 opment. *METTL14* is the second gene in terms of number of eQTLs. However,
355 in this case the differential regulation is tissue specific and accumulates in the
356 cerebellum, making it a strong candidate for one of the genes underlying the
357 cerebellar growth phenotype. *METTL14*, along with *METTL3*, is part of a
358 complex that regulates expression of N6-methyladenosine (m6a), a epitransla-
359 tional RNA modifier that has an important role in cerebellar development
360 in mice [Ma et al., 2018].

361 Supporting the idea of a brain-face coregulatory genetic network, the data
362 shows other genes related to the craniofacial complex expressed in the brain.
363 *PIGV* and *WHSC1* are involved in hyperphosphatasia-mental retardation syn-
364 drome (OMIM: #239300) [Horn et al., 2014] and Wolf-Hirschhorn syndrome
365 (OMIM: #194190) [Yu et al., 2017], respectively. These two disorders are char-
366 acterized by distinctive craniofacial profiles. Both *PIGV* and *WHSC1* are up-
367 regulated in *Homo sapiens* in our data. Additionally, *PIGV* is associated with
368 a signal of positive selection relative to the archaic human genomes [Racimo
369 et al., 2015]. The way the brain and the craniofacial complex share some of the

370 genetic regulatory networks is reflected in the amount of *Homo sapiens* derived
371 eQTLs in the brain linked with disorders that affect the fusion of the skull
372 bones or typical deviations of the craniofacial disorder such as cleft palate (see
373 section 2.4). These findings open up the possibility of using modern population
374 clinical data to establish genetic regulatory networks that can inform about the
375 molecular evolution of the *Homo sapiens* species, and viceversa.

376 Other aspects of our results can be also found at the crossroads between
377 clinical data and evolutionary studies. Some of the eQTLs affect differential reg-
378 ulation of both metabolic programming and microtubule dynamics. Our data
379 shows distinct regulation of genes such *GAPDH* and *PFKM*. Glyceraldehyde-
380 3-phosphate dehydrogenase (*GAPDH*) is one of the enzymes of the aerobic
381 glycolysis metabolic chain. The idea that aerobic glycolysis may have played
382 a role in human evolution was proposed before in [Bufill et al., 2011]. Modern
383 humans have a protracted neuronal development [Somel et al., 2009], a pheno-
384 type that has been suggested [Bufill et al., 2011, Bauernfeind et al., 2014] to
385 be associated with a species-specific metabolic program involving an upregula-
386 tion of aerobic glycolysis. Aerobic glycolysis correlates with key stages of the
387 development of the human brain, such as myelination, synaptogenesis and ax-
388 onal elongation [Bauernfeind et al., 2014]. Changes in brain metabolism could
389 also be underlying human specific neurodegenerative diseases correlated with
390 metabolic failures, such as Alzheimer’s disease [Bufill et al., 2013]. *GAPDH* is
391 known to provide glycolytic energy in fast axonal transport during vesicle traf-
392 ficking [Zala et al., 2013] and is implicated in Alzheimer’s disease, among other
393 neurodegenerative diseases [El Kadmiri et al., 2014].

394 Our data also shows that genes affected by eQTLs are enriched in the regu-
395 lation of microtubule dynamics. Microtubule depolymerization plays a crucial
396 role in neuronal development and synaptic plasticity [Dent, 2017], as well as in

397 the pathogenesis of Alzheimer’s Disease [Hernández et al., 2013]. Among the
398 genes affecting microtubule dynamics we found one that is associated to a cellu-
399 lar phenotype in the same tissue where the human eQTL is affecting expression
400 levels: *FMN1* in the hippocampus. Overexpression of *FMN1* promotes primary
401 dendritic development in the hippocampus [Simon-Arecas et al., 2011]. Whether
402 *FMN1* overexpression would have the same effect in human hippocampus is cur-
403 rently unknown, but this example shows how the present study can open new
404 testable hypotheses for human evolution studies concerning understudied sub-
405 cortical structures.

406 All in all, our work shows the potential of using human variation databases
407 as a valuable point of entry to bridge genotype and phenotype in brain evolution
408 studies.

409 4 Methods

410 We accessed the *Homo sapiens* high-frequency variant annotation data from
411 [Kuhlwilm and Boeckx, 2019]. The data is publicly available in [https://
412 doi.org/10.6084/m9.figshare.8184038](https://doi.org/10.6084/m9.figshare.8184038). We then applied more stringent cri-
413 teria to the high-frequency criteria [Kuhlwilm and Boeckx, 2019]. The study in
414 Kuhlwilm and Boeckx [2019] defines an arbitrary cutoff point of *Homo sapiens*
415 derived 90% frequency based on previous work, but the cutoff is global instead
416 of relative to metapopulation allele frequency, i.e., it is required that the global
417 frequency of an allele be more than or equal to 90%, but specific populations
418 can have lower frequencies. Since the data itself already included metapopula-
419 tion frequency, we applied a more rigorous filter and removed any alleles that
420 where below 90% in any of the considered metapopulations. A full description
421 of the methods to create these annotations can be obtained from the origi-
422 nal paper [Kuhlwilm and Boeckx, 2019]. We also obtained the publicly avail-

423 able single-tissue cis-eQTLs eGene and significant variant-gene associations (v7)
424 (<https://gtexportal.org/home/datasets>). We matched eQTL ID to rs IDs
425 with a lookup table provided by GTEx. The chi-square goodness-to-fit and the
426 Kolmogorov–Smirnov test were performed in R. To investigate the allelic conse-
427 quences and the clinvar data we used the Biomart tool [Zerbino et al., 2018]. We
428 performed the gene ontology cluster enrichment analysis with Metascape [Zhou
429 et al., 2019] and the transcription factor enrichment with SNP2TFBS [Kumar
430 et al., 2017]. We generated all the main figures with the ggplot2 R package
431 [Wickham, 2009] and Circos [Krzywinski et al., 2009]. Supplementary Figure 1,
432 2 and 3 where generated by Metascape [Zhou et al., 2019], RegioneR [Gel et al.,
433 2015] and SNP2TFBS [Kumar et al., 2017] respectively. The miRNA data was
434 extracted from the supplementary tables 6 and 7 of Branco et al. [2018]. We
435 used the supplementary table S5 from Racimo et al. [2014], the supplemen-
436 tary table S2 from Peyrégne et al. [2017] for the human selective sweep data,
437 and the data from Sankararaman et al. [2016] for the deserts of introgression.
438 The evolutionary clusters data comes from the supplementary table S20 of Zhu
439 et al. [2018]. We performed the permutation test (n=1000) with the R package
440 RegioneR [Gel et al., 2015].

441 **Data acces**

442 All raw data is publicly available as described in Methods. Processed single-
443 tissue data and clinical data can be found here: [https://github.com/AGMAndirko/](https://github.com/AGMAndirko/GTEx_project)
444 [GTEx_project](https://github.com/AGMAndirko/GTEx_project) and in the supplementary materials.

445 **Acknowledgments**

446 The Genotype-Tissue Expression (GTEx) Project was supported by the Com-
447 mon Fund of the Office of the Director of the National Institutes of Health, and
448 by NCI, NHGRI, NHLBI, NIDA, NIMH, and NINDS. The data used for the
449 analyses described in this manuscript were obtained from the GTEx Portal on
450 05/15/19.

451 **Author Contributions**

452 Conceptualization: CB & AA; Data Curation: AA; Formal Analysis: AA; Fund-
453 ing Acquisition: CB; Investigation: CB & AA; Methodology: CB & AA; Soft-
454 ware: AA; Supervision: CB; Visualization: CB & AA; Writing — Original Draft
455 Preparation: CB & AA; Writing — Review & Editing: CB & AA.

456 **Disclosure declaration**

457 The authors declare no conflict of interest.

458 **Funding statement**

459 AA acknowledges financial support from the Spanish Ministry of Economy
460 and Competitiveness and the European Social Fund (BES-2017-080366). CB
461 acknowledges financial support from the Spanish Ministry of Economy and

462 Competitiveness/FEDER (grant FFI2016-78034-C2-1-P), the Marie Curie In-
463 ternational Reintegration Grant from the European Union (PIRG-GA-2009-
464 256413), research funds from the Fundació Bosch i Gimpera, the MEXT/JSPS
465 Grant-in-Aid for Scientific Research on Innovative Areas 4903 (Evolinguistics:
466 JP17H06379), and support from the Generalitat de Catalunya (2017-SGR-341).

467 **Competing interest**

468 Authors declare NO competing financial or non-financial interest.

469 **References**

470 Markus Bastir, Antonio Rosas, Philipp Gunz, Angel Peña-Melian, Giorgio
471 Manzi, Katerina Harvati, Robert Kruszynski, Chris Stringer, and Jean-
472 Jacques Hublin. Evolution of the base of the brain in highly encephalized hu-
473 man species. *Nature Communications*, 2(1):588, September 2011. ISSN 2041-
474 1723. doi: 10.1038/ncomms1593. URL [http://www.nature.com/articles/
475 ncomms1593](http://www.nature.com/articles/ncomms1593).

476 Amy L. Bauernfeind, Sarah K. Barks, Tetyana Duka, Lawrence I. Grossman,
477 Patrick R. Hof, and Chet C. Sherwood. Aerobic glycolysis in the primate
478 brain: reconsidering the implications for growth and maintenance. *Brain
479 Structure and Function*, 219(4):1149–1167, July 2014. ISSN 1863-2661. doi:
480 10.1007/s00429-013-0662-z. URL [https://doi.org/10.1007/s00429-013-
481 0662-z](https://doi.org/10.1007/s00429-013-
481 0662-z).

482 Prabesh Bhattarai, Alvin Kuriakose Thomas, Mehmet Ilyas Cosacak, Chris-
483 tos Papadimitriou, Violeta Mashkaryan, Cynthia Froc, Susanne Reinhardt,
484 Thomas Kurth, Andreas Dahl, Yixin Zhang, and Caghan Kizil. IL4/STAT6
485 Signaling Activates Neural Stem Cell Proliferation and Neurogenesis upon

486 Amyloid- 42 Aggregation in Adult Zebrafish Brain. *Cell Reports*, 17(4):941–
487 948, October 2016. ISSN 22111247. doi: 10.1016/j.celrep.2016.09.075. URL
488 <https://linkinghub.elsevier.com/retrieve/pii/S2211124716313316>.

489 Cedric Boeckx. The language-ready head: Evolutionary considerations.
490 *Psychonomic Bulletin & Review*, 24(1):194–199, February 2017. ISSN
491 1069-9384, 1531-5320. doi: 10.3758/s13423-016-1087-5. URL [http://](http://link.springer.com/10.3758/s13423-016-1087-5)
492 link.springer.com/10.3758/s13423-016-1087-5.

493 Paulo R. Branco, Gilderlanio S. de Araújo, Júnior Barrera, Guilherme Suarez-
494 Kurtz, and Sandro José de Souza. Uncovering association networks through
495 an eQTL analysis involving human miRNAs and lincRNAs. *Scientific Reports*,
496 8(1):15050, December 2018. ISSN 2045-2322. doi: 10.1038/s41598-018-33420-
497 z. URL <http://www.nature.com/articles/s41598-018-33420-z>.

498 Emiliano Bruner, Jos  Manuel de la Cu tara, Michael Masters, Hideki
499 Amano, and Naomichi Ogihara. Functional craniology and brain evolu-
500 tion: from paleontology to biomedicine. *Frontiers in Neuroanatomy*, 8,
501 April 2014. ISSN 1662-5129. doi: 10.3389/fnana.2014.00019. URL [http://](http://journal.frontiersin.org/article/10.3389/fnana.2014.00019/abstract)
502 journal.frontiersin.org/article/10.3389/fnana.2014.00019/abstract.

503 Enric Bufill, Jordi Agust , and Rafael Blesa. Human neoteny revisited: The
504 case of synaptic plasticity. *American Journal of Human Biology*, 23(6):729–
505 739, November 2011. ISSN 10420533. doi: 10.1002/ajhb.21225. URL [http:](http://doi.wiley.com/10.1002/ajhb.21225)
506 [//doi.wiley.com/10.1002/ajhb.21225](http://doi.wiley.com/10.1002/ajhb.21225).

507 Enric Bufill, Rafael Blesa, and Jordi Agust . Alzheimer’s disease: an evolution-
508 ary approach. *Journal of Anthropological Sciences*, (91):135–157, 2013. ISSN
509 1827-4765. doi: 10.4436/JASS.91001. URL [http://www.isita-org.com/](http://www.isita-org.com/jass/Contents/ContentsVol91.htm)
510 [jass/Contents/ContentsVol91.htm](http://www.isita-org.com/jass/Contents/ContentsVol91.htm).

511 In Youb Chang, Takbum Ohn, Gil Seok Ko, Young Yoon, Jung Woo Kim,
512 and Sang Pil Yoon. Immunolocalization of steroidogenic acute regula-
513 tory protein-related lipid transfer (START) domain-containing proteins in
514 the developing cerebellum of normal and hypothyroid rats. *Journal of*
515 *Chemical Neuroanatomy*, 43(1):28–33, January 2012. ISSN 08910618. doi:
516 10.1016/j.jchemneu.2011.10.003. URL [https://linkinghub.elsevier.com/](https://linkinghub.elsevier.com/retrieve/pii/S0891061811001530)
517 [retrieve/pii/S0891061811001530](https://linkinghub.elsevier.com/retrieve/pii/S0891061811001530).

518 Terumasa Chiba, Yoshinori Otani, Yoshihide Yamaguchi, Tomoko Ishibashi,
519 Akiko Hayashi, Kenji F. Tanaka, Maya Yamazaki, Kenji Sakimura, and
520 Hiroko Baba. Microglial phospholipase D4 deficiency influences myelina-
521 tion during brain development. *Proceedings of the Japan Academy, Se-*
522 *ries B*, 92(7):237–254, 2016. ISSN 0386-2208, 1349-2896. doi: 10.2183/
523 pjab.92.237. URL [https://www.jstage.jst.go.jp/article/pjab/92/7/](https://www.jstage.jst.go.jp/article/pjab/92/7/92_PJA9207B-01/_article)
524 [92_PJA9207B-01/_article](https://www.jstage.jst.go.jp/article/pjab/92/7/92_PJA9207B-01/_article).

525 Matyas F. Cserhati, Mary-Ellen Mooter, Lauren Peterson, Benjamin Wicks,
526 Peng Xiao, Mark Pauley, and Chittibabu Guda. Motifome compari-
527 son between modern human, Neanderthal and Denisovan. *BMC Ge-*
528 *nomics*, 19(1):472, December 2018. ISSN 1471-2164. doi: 10.1186/s12864-
529 018-4710-1. URL [https://bmcbgenomics.biomedcentral.com/articles/](https://bmcbgenomics.biomedcentral.com/articles/10.1186/s12864-018-4710-1)
530 [10.1186/s12864-018-4710-1](https://bmcbgenomics.biomedcentral.com/articles/10.1186/s12864-018-4710-1).

531 Annarita De Luca, Valentina Cerrato, Elisa Fucà, Elena Parmigiani, Annal-
532 isa Buffo, and Ketty Leto. Sonic hedgehog patterning during cerebellar de-
533 velopment. *Cellular and Molecular Life Sciences*, 73(2):291–303, January
534 2016. ISSN 1420-682X, 1420-9071. doi: 10.1007/s00018-015-2065-1. URL
535 <http://link.springer.com/10.1007/s00018-015-2065-1>.

536 Erik W. Dent. Of microtubules and memory: implications for microtubule

537 dynamics in dendrites and spines. *Molecular Biology of the Cell*, 28(1):1–8,
538 2017. ISSN 1939-4586. doi: 10.1091/mbc.E15-11-0769.

539 Florian Duclot and Mohamed Kabbaj. The Role of Early Growth Re-
540 sponse 1 (EGR1) in Brain Plasticity and Neuropsychiatric Disorders. *Frontiers in Behavioral Neuroscience*, 11, March 2017. ISSN 1662-5153.
541 doi: 10.3389/fnbeh.2017.00035. URL [http://journal.frontiersin.org/
542 article/10.3389/fnbeh.2017.00035/full](http://journal.frontiersin.org/article/10.3389/fnbeh.2017.00035/full).
543

544 N. El Kadmiri, I. Slassi, B. El Moutawakil, S. Nadifi, A. Tadevosyan,
545 A. Hachem, and A. Soukri. Glyceraldehyde-3-phosphate dehydrogenase
546 (GAPDH) and Alzheimer’s disease. *Pathologie Biologie*, 62(6):333–336, De-
547 cember 2014. ISSN 03698114. doi: 10.1016/j.patbio.2014.08.002. URL
548 <https://linkinghub.elsevier.com/retrieve/pii/S0369811414001278>.

549 Lucía F. Franchini and Katherine S. Pollard. Human evolution: the non-coding
550 revolution. *BMC Biology*, 15(1), December 2017. ISSN 1741-7007. doi:
551 10.1186/s12915-017-0428-9. URL [http://bmcbiol.biomedcentral.com/
552 articles/10.1186/s12915-017-0428-9](http://bmcbiol.biomedcentral.com/articles/10.1186/s12915-017-0428-9).

553 Bernat Gel, Anna Díez-Villanueva, Eduard Serra, Marcus Buschbeck, Miguel A.
554 Peinado, and Roberto Malinverni. regioneR: an R/Bioconductor pack-
555 age for the association analysis of genomic regions based on permutation
556 tests. *Bioinformatics*, page btv562, September 2015. ISSN 1367-4803, 1460-
557 2059. doi: 10.1093/bioinformatics/btv562. URL [https://academic.oup.com/
558 bioinformatics/article-lookup/doi/10.1093/bioinformatics/btv562](https://academic.oup.com/bioinformatics/article-lookup/doi/10.1093/bioinformatics/btv562).

559 David Gokhman, Eran Meshorer, and Liran Carmel. Epigenetics: It’s Get-
560 ting Old. Past Meets Future in Paleoeugenetics. *Trends in Ecology &
561 Evolution*, 31(4):290–300, April 2016. ISSN 01695347. doi: 10.1016/

562 j.tree.2016.01.010. URL [https://linkinghub.elsevier.com/retrieve/](https://linkinghub.elsevier.com/retrieve/pii/S0169534716000252)
563 [pii/S0169534716000252](https://linkinghub.elsevier.com/retrieve/pii/S0169534716000252).

564 Philipp Gunz, Simon Neubauer, Bruno Maureille, and Jean-Jacques Hublin.
565 Brain development after birth differs between Neanderthals and modern hu-
566 mans. *Current Biology*, 20(21):R921–R922, November 2010. ISSN 09609822.
567 doi: 10.1016/j.cub.2010.10.018. URL [https://linkinghub.elsevier.com/](https://linkinghub.elsevier.com/retrieve/pii/S0960982210012820)
568 [retrieve/pii/S0960982210012820](https://linkinghub.elsevier.com/retrieve/pii/S0960982210012820).

569 Philipp Gunz, Amanda K. Tilot, Katharina Wittfeld, Alexander Teumer,
570 Chin Yang Shapland, Theo G.M. van Erp, Michael Dannemann, Benjamin
571 Vernot, Simon Neubauer, Tulio Guadalupe, Guillén Fernández, Han G. Brun-
572 ner, Wolfgang Enard, James Fallon, Norbert Hosten, Uwe Völker, Antonio
573 Profico, Fabio Di Vincenzo, Giorgio Manzi, Janet Kelso, Beate St. Pour-
574 cain, Jean-Jacques Hublin, Barbara Franke, Svante Pääbo, Fabio Macciardi,
575 Hans J. Grabe, and Simon E. Fisher. Neandertal Introgression Sheds Light
576 on Modern Human Endocranial Globularity. *Current Biology*, 29(1):120–
577 127.e5, January 2019. ISSN 09609822. doi: 10.1016/j.cub.2018.10.065. URL
578 <https://linkinghub.elsevier.com/retrieve/pii/S0960982218314702>.

579 Long Guo, Nursel H Elcioglu, Zheng Wang, Yasemin K Demirkol, Pinar Is-
580 guven, Naomichi Matsumoto, Gen Nishimura, Noriko Miyake, and Shiro
581 Ikegawa. Novel and recurrent COL11a1 and COL2a1 mutations in the
582 Marshall–Stickler syndrome spectrum. *Human Genome Variation*, 4(1):
583 17040, December 2017. ISSN 2054-345X. doi: 10.1038/hgv.2017.40. URL
584 <http://www.nature.com/articles/hgv201740>.

585 Félix Hernández, Esther García-García, and Jesús Avila. Microtubule Depoly-
586 merization and Tau Phosphorylation. *Journal of Alzheimer’s Disease*, 37(3):
587 507–513, September 2013. ISSN 18758908, 13872877. doi: 10.3233/JAD-

588 130545. URL [http://www.medra.org/servlet/aliasResolver?alias=](http://www.medra.org/servlet/aliasResolver?alias=iospress&doi=10.3233/JAD-130545)
589 [iospress&doi=10.3233/JAD-130545](http://www.medra.org/servlet/aliasResolver?alias=iospress&doi=10.3233/JAD-130545).

590 Denise Horn, Dagmar Wiczorek, Kay Metcalfe, Ivo Barić, Lidija Paležac, Mario
591 Ćuk, Danijela Petković Ramadža, Ulrike Krüger, Stephanie Demuth, Wol-
592 fram Heinritz, Tobias Linden, Jens Koenig, Peter N Robinson, and Peter
593 Krawitz. Delineation of PIGV mutation spectrum and associated phenotypes
594 in hyperphosphatasia with mental retardation syndrome. *European Jour-*
595 *nal of Human Genetics*, 22(6):762–767, June 2014. ISSN 1018-4813, 1476-
596 5438. doi: 10.1038/ejhg.2013.241. URL [http://www.nature.com/articles/](http://www.nature.com/articles/ejhg2013241)
597 [ejhg2013241](http://www.nature.com/articles/ejhg2013241).

598 Rinku Jain, Jayati Roy Choudhury, Angeliki Buku, Robert E. Johnson,
599 Louise Prakash, Satya Prakash, and Aneel K. Aggarwal. Mechanism of
600 error-free DNA synthesis across N1-methyl-deoxyadenosine by human DNA
601 polymerase- . *Scientific Reports*, 7(1):43904, April 2017. ISSN 2045-2322. doi:
602 10.1038/srep43904. URL <http://www.nature.com/articles/srep43904>.

603 Takanori Kochiyama, Naomichi Ogihara, Hiroki C. Tanabe, Osamu Kondo,
604 Hideki Amano, Kunihiro Hasegawa, Hiromasa Suzuki, Marcia S. Ponce de
605 León, Christoph P. E. Zollikofer, Markus Bastir, Chris Stringer, Norihiro
606 Sadato, and Takeru Akazawa. Reconstructing the Neanderthal brain us-
607 ing computational anatomy. *Scientific Reports*, 8(1):6296, April 2018. ISSN
608 2045-2322. doi: 10.1038/s41598-018-24331-0. URL [http://www.nature.com/](http://www.nature.com/articles/s41598-018-24331-0)
609 [articles/s41598-018-24331-0](http://www.nature.com/articles/s41598-018-24331-0).

610 M. Krzywinski, J. Schein, I. Birol, J. Connors, R. Gascoyne, D. Horsman, S. J.
611 Jones, and M. A. Marra. Circos: An information aesthetic for comparative
612 genomics. *Genome Research*, 19(9):1639–1645, September 2009. ISSN 1088-

613 9051. doi: 10.1101/gr.092759.109. URL [http://genome.cshlp.org/cgi/doi/](http://genome.cshlp.org/cgi/doi/10.1101/gr.092759.109)
614 [10.1101/gr.092759.109](http://genome.cshlp.org/cgi/doi/10.1101/gr.092759.109).

615 Martin Kuhlwilm and Cedric Boeckx. A catalog of single nucleotide changes
616 distinguishing modern humans from archaic hominins. *Scientific Reports*, 9
617 (1):8463, December 2019. ISSN 2045-2322. doi: 10.1038/s41598-019-44877-x.
618 URL <http://www.nature.com/articles/s41598-019-44877-x>.

619 Sunil Kumar, Giovanna Ambrosini, and Philipp Bucher. SNP2tfbs – a
620 database of regulatory SNPs affecting predicted transcription factor bind-
621 ing site affinity. *Nucleic Acids Research*, 45(D1):D139–D144, January 2017.
622 ISSN 0305-1048, 1362-4962. doi: 10.1093/nar/gkw1064. URL [https:](https://academic.oup.com/nar/article-lookup/doi/10.1093/nar/gkw1064)
623 [//academic.oup.com/nar/article-lookup/doi/10.1093/nar/gkw1064](https://academic.oup.com/nar/article-lookup/doi/10.1093/nar/gkw1064).

624 Rodrigo S. Lacruz, Chris B. Stringer, William H. Kimbel, Bernard Wood, Kate-
625 rina Harvati, Paul O’Higgins, Timothy G. Bromage, and Juan-Luis Arsuaga.
626 The evolutionary history of the human face. *Nature Ecology & Evolution*, 3
627 (5):726–736, May 2019. ISSN 2397-334X. doi: 10.1038/s41559-019-0865-7.
628 URL <http://www.nature.com/articles/s41559-019-0865-7>.

629 Melissa J Landrum, Jennifer M Lee, Mark Benson, Garth R Brown, Chen
630 Chao, Shanmuga Chitipiralla, Baoshan Gu, Jennifer Hart, Douglas Hoff-
631 man, Wonhee Jang, Karen Karapetyan, Kenneth Katz, Chunlei Liu, Zenith
632 Maddipatla, Adriana Malheiro, Kurt McDaniel, Michael Ovetsky, George
633 Riley, George Zhou, J Bradley Holmes, Brandi L Kattman, and Donna R
634 Maglott. ClinVar: improving access to variant interpretations and sup-
635 porting evidence. *Nucleic Acids Research*, 46(D1):D1062–D1067, January
636 2018. ISSN 0305-1048, 1362-4962. doi: 10.1093/nar/gkx1153. URL [http:](http://academic.oup.com/nar/article/46/D1/D1062/4641904)
637 [//academic.oup.com/nar/article/46/D1/D1062/4641904](http://academic.oup.com/nar/article/46/D1/D1062/4641904).

638 Chunhui Ma, Mengqi Chang, Hongyi Lv, Zhi-Wei Zhang, Weilong Zhang,

639 Xue He, Gaolang Wu, Shunli Zhao, Yao Zhang, Di Wang, Xufei Teng,
640 Chunying Liu, Qing Li, Arne Klungland, Yamei Niu, Shuhui Song, and
641 Wei-Min Tong. RNA m6a methylation participates in regulation of post-
642 natal development of the mouse cerebellum. *Genome Biology*, 19(1):
643 68, December 2018. ISSN 1474-760X. doi: 10.1186/s13059-018-1435-
644 z. URL [https://genomebiology.biomedcentral.com/articles/10.1186/
645 s13059-018-1435-z](https://genomebiology.biomedcentral.com/articles/10.1186/s13059-018-1435-z).

646 Rajiv C. McCoy, Jon Wakefield, and Joshua M. Akey. Impacts of Neanderthal-
647 Introgressed Sequences on the Landscape of Human Gene Expression.
648 *Cell*, 168(5):916–927.e12, February 2017. ISSN 00928674. doi: 10.1016/
649 j.cell.2017.01.038. URL [https://linkinghub.elsevier.com/retrieve/pii/
650 S0092867417301289](https://linkinghub.elsevier.com/retrieve/pii/S0092867417301289).

651 Matthias Meyer, Martin Kircher, Marie-Theres Gansauge, Heng Li, Fernando
652 Racimo, Swapan Mallick, Joshua G. Schraiber, Flora Jay, Kay Prüfer, Ce-
653 sare de Filippo, Peter H. Sudmant, Can Alkan, Qiaomei Fu, Ron Do, Nadin
654 Rohland, Arti Tandon, Michael Siebauer, Richard E. Green, Katarzyna Bryc,
655 Adrian W. Briggs, Udo Stenzel, Jesse Dabney, Jay Shendure, Jacob Kit-
656 zman, Michael F. Hammer, Michael V. Shunkov, Anatoli P. Derevianko, Nick
657 Patterson, Aida M. Andrés, Evan E. Eichler, Montgomery Slatkin, David
658 Reich, Janet Kelso, and Svante Pääbo. A High-Coverage Genome Sequence
659 from an Archaic Denisovan Individual. *Science*, 338(6104):222–226, Octo-
660 ber 2012. ISSN 0036-8075, 1095-9203. doi: 10.1126/science.1224344. URL
661 <http://science.sciencemag.org/content/338/6104/222>.

662 Felipe Mora-Bermúdez, Farhath Badsha, Sabina Kanton, J Gray Camp, Ben-
663 jamin Vernot, Kathrin Köhler, Birger Voigt, Keisuke Okita, Tomislav Mari-
664 cic, Zhisong He, Robert Lachmann, Svante Pääbo, Barbara Treutlein, and
665 Wieland B Huttner. Differences and similarities between human and chim-

666 panzee neural progenitors during cerebral cortex development. *eLife*, 5:
667 e18683, September 2016. ISSN 2050-084X. doi: 10.7554/eLife.18683. URL
668 <https://elifesciences.org/articles/18683>.

669 A. Moreno-Estrada, F. Casals, A. Ramirez-Soriano, B. Oliva, F. Calafell,
670 J. Bertranpetit, and E. Bosch. Signatures of Selection in the Human Ol-
671 factory Receptor OR5i1 Gene. *Molecular Biology and Evolution*, 25(1):
672 144–154, November 2007. ISSN 0737-4038, 1537-1719. doi: 10.1093/
673 molbev/msm240. URL [https://academic.oup.com/mbe/article-lookup/
674 doi/10.1093/molbev/msm240](https://academic.oup.com/mbe/article-lookup/doi/10.1093/molbev/msm240).

675 Tuan Nguyen and Simone Di Giovanni. NFAT signaling in neural devel-
676 opment and axon growth. *International Journal of Developmental Neu-*
677 *roscience*, 26(2):141–145, April 2008. ISSN 07365748. doi: 10.1016/
678 j.ijdevneu.2007.10.004. URL [https://linkinghub.elsevier.com/retrieve/
679 pii/S0736574807001608](https://linkinghub.elsevier.com/retrieve/pii/S0736574807001608).

680 Tomoo Owa, Shinichiro Taya, Satoshi Miyashita, Mariko Yamashita, Toma
681 Adachi, Koyo Yamada, Miwa Yokoyama, Shogo Aida, Tomoki Nish-
682 ioka, Yukiko U. Inoue, Ryo Goitsuka, Takuro Nakamura, Takayoshi In-
683 oue, Kozo Kaibuchi, and Mikio Hoshino. Meis1 Coordinates Cerebel-
684 lar Granule Cell Development by Regulating Pax6 Transcription, BMP
685 Signaling and Atoh1 Degradation. *The Journal of Neuroscience*, 38(5):
686 1277–1294, January 2018. ISSN 0270-6474, 1529-2401. doi: 10.1523/
687 JNEUROSCI.1545-17.2017. URL [http://www.jneurosci.org/lookup/doi/
688 10.1523/JNEUROSCI.1545-17.2017](http://www.jneurosci.org/lookup/doi/10.1523/JNEUROSCI.1545-17.2017).

689 Stéphane Peyrégne, Michael James Boyle, Michael Dannemann, and Kay Prüfer.
690 Detecting ancient positive selection in humans using extended lineage sorting.
691 *Genome Research*, 27(9):1563–1572, September 2017. ISSN 1088-9051, 1549-

692 5469. doi: 10.1101/gr.219493.116. URL <http://genome.cshlp.org/lookup/>
693 [doi/10.1101/gr.219493.116](http://doi.org/10.1101/gr.219493.116).

694 Kay Prüfer, Fernando Racimo, Nick Patterson, Flora Jay, Sriram Sankararam,
695 Susanna Sawyer, Anja Heinze, Gabriel Renaud, Peter H. Sudmant,
696 Cesare de Filippo, Heng Li, Swapan Mallick, Michael Dannemann, Qiaomei
697 Fu, Martin Kircher, Martin Kuhlwilm, Michael Lachmann, Matthias Meyer,
698 Matthias Ongyerth, Michael Siebauer, Christoph Theunert, Arti Tandon,
699 Priya Moorjani, Joseph Pickrell, James C. Mullikin, Samuel H. Vohr,
700 Richard E. Green, Ines Hellmann, Philip L. F. Johnson, Hélène Blanche,
701 Howard Cann, Jacob O. Kitzman, Jay Shendure, Evan E. Eichler, Ed S.
702 Lein, Trygve E. Bakken, Liubov V. Golovanova, Vladimir B. Doronichev,
703 Michael V. Shunkov, Anatoli P. Derevianko, Bence Viola, Montgomery
704 Slatkin, David Reich, Janet Kelso, and Svante Pääbo. The complete genome
705 sequence of a Neanderthal from the Altai Mountains. *Nature*, 505(7481):43–
706 49, January 2014. ISSN 0028-0836, 1476-4687. doi: 10.1038/nature12886.
707 URL <http://www.nature.com/articles/nature12886>.

708 Kay Prüfer, Cesare de Filippo, Steffi Grote, Fabrizio Mafessoni, Petra Kor-
709 lević, Mateja Hajdinjak, Benjamin Vernot, Laurits Skov, Pinghsun Hsieh,
710 Stéphane Peyrégne, David Reher, Charlotte Hopfe, Sarah Nagel, Tomislav
711 Maricic, Qiaomei Fu, Christoph Theunert, Rebekah Rogers, Pontus
712 Skoglund, Manjusha Chintalapati, Michael Dannemann, Bradley J. Nelson,
713 Felix M. Key, Pavao Rudan, Željko Kućan, Ivan Gušić, Liubov V. Golo-
714 vanova, Vladimir B. Doronichev, Nick Patterson, David Reich, Evan E.
715 Eichler, Montgomery Slatkin, Mikkel H. Schierup, Aida M. Andrés, Janet
716 Kelso, Matthias Meyer, and Svante Pääbo. A high-coverage Neandertal
717 genome from Vindija Cave in Croatia. *Science*, 358(6363):655–658, Novem-

718 ber 2017. ISSN 0036-8075, 1095-9203. doi: 10.1126/science.aa01887. URL
719 <http://www.sciencemag.org/lookup/doi/10.1126/science.aa01887>.

720 Fernando Racimo, Martin Kuhlwilm, and Montgomery Slatkin. A Test for
721 Ancient Selective Sweeps and an Application to Candidate Sites in Modern
722 Humans. *Molecular Biology and Evolution*, 31(12):3344–3358, December 2014.
723 ISSN 1537-1719, 0737-4038. doi: 10.1093/molbev/msu255. URL [https://](https://academic.oup.com/mbe/article-lookup/doi/10.1093/molbev/msu255)
724 academic.oup.com/mbe/article-lookup/doi/10.1093/molbev/msu255.

725 Fernando Racimo, Sriram Sankararaman, Rasmus Nielsen, and Emilia Huerta-
726 Sánchez. Evidence for archaic adaptive introgression in humans. *Nature*
727 *Reviews Genetics*, 16(6):359–371, June 2015. ISSN 1471-0056, 1471-0064.
728 doi: 10.1038/nrg3936. URL <http://www.nature.com/articles/nrg3936>.

729 A. Rauch, C. T. Thiel, D. Schindler, U. Wick, Y. J. Crow, A. B. Ekici, A. J.
730 van Essen, T. O. Goecke, L. Al-Gazali, K. H. Chrzanowska, C. Zweier, H. G.
731 Brunner, K. Becker, C. J. Curry, B. Dallapiccola, K. Devriendt, A. Dor-
732 fler, E. Kinning, A. Megarbane, P. Meinecke, R. K. Semple, S. Spranger,
733 A. Toutain, R. C. Trembath, E. Voss, L. Wilson, R. Hennekam, F. de Zegher,
734 H.-G. Dorr, and A. Reis. Mutations in the Pericentrin (PCNT) Gene
735 Cause Primordial Dwarfism. *Science*, 319(5864):816–819, February 2008.
736 ISSN 0036-8075, 1095-9203. doi: 10.1126/science.1151174. URL [http:](http://www.sciencemag.org/cgi/doi/10.1126/science.1151174)
737 [//www.sciencemag.org/cgi/doi/10.1126/science.1151174](http://www.sciencemag.org/cgi/doi/10.1126/science.1151174).

738 Genetics Home Reference. Genetics Home Reference, Your Guide to Under-
739 standing Genetic Conditions. URL <https://ghr.nlm.nih.gov/>.

740 Sriram Sankararaman, Swapan Mallick, Nick Patterson, and David Reich. The
741 Combined Landscape of Denisovan and Neanderthal Ancestry in Present-
742 Day Humans. *Current Biology*, 26(9):1241–1247, May 2016. ISSN 09609822.

743 doi: 10.1016/j.cub.2016.03.037. URL [https://linkinghub.elsevier.com/
744 retrieve/pii/S0960982216302470](https://linkinghub.elsevier.com/retrieve/pii/S0960982216302470).

745 Julia Simon-Areces, Ana Dopazo, Markus Dettenhofer, Alfredo Rodriguez-
746 Tebar, Luis Miguel Garcia-Segura, and Maria-Angeles Arevalo. Formin1
747 Mediates the Induction of Dendritogenesis and Synaptogenesis by Neuro-
748 genin3 in Mouse Hippocampal Neurons. *PLoS ONE*, 6(7):e21825, July
749 2011. ISSN 1932-6203. doi: 10.1371/journal.pone.0021825. URL [http:
750 //dx.plos.org/10.1371/journal.pone.0021825](http://dx.plos.org/10.1371/journal.pone.0021825).

751 M. Somel, H. Franz, Z. Yan, A. Lorenc, S. Guo, T. Giger, J. Kelso, B. Nickel,
752 M. Dannemann, S. Bahn, M. J. Webster, C. S. Weickert, M. Lachmann,
753 S. Paabo, and P. Khaitovich. Transcriptional neoteny in the human brain.
754 *Proceedings of the National Academy of Sciences*, 106(14):5743–5748, April
755 2009. ISSN 0027-8424, 1091-6490. doi: 10.1073/pnas.0900544106. URL [http:
756 //www.pnas.org/cgi/doi/10.1073/pnas.0900544106](http://www.pnas.org/cgi/doi/10.1073/pnas.0900544106).

757 The GTEx Consortium, K. G. Ardlie, D. S. Deluca, A. V. Segre, T. J. Sulli-
758 van, T. R. Young, E. T. Gelfand, C. A. Trowbridge, J. B. Maller, T. Tuki-
759 ainen, M. Lek, L. D. Ward, P. Kheradpour, B. Iriarte, Y. Meng, C. D.
760 Palmer, T. Esko, W. Winckler, J. N. Hirschhorn, M. Kellis, D. G. MacArthur,
761 G. Getz, A. A. Shabalina, G. Li, Y.-H. Zhou, A. B. Nobel, I. Rusyn, F. A.
762 Wright, T. Lappalainen, P. G. Ferreira, H. Ongen, M. A. Rivas, A. Bat-
763 tle, S. Mostafavi, J. Monlong, M. Sammeth, M. Mele, F. Reverter, J. M.
764 Goldmann, D. Koller, R. Guigo, M. I. McCarthy, E. T. Dermitzakis, E. R.
765 Gamazon, H. K. Im, A. Konkashbaev, D. L. Nicolae, N. J. Cox, T. Flutre,
766 X. Wen, M. Stephens, J. K. Pritchard, Z. Tu, B. Zhang, T. Huang, Q. Long,
767 L. Lin, J. Yang, J. Zhu, J. Liu, A. Brown, B. Mestichelli, D. Tidwell, E. Lo,
768 M. Salvatore, S. Shad, J. A. Thomas, J. T. Lonsdale, M. T. Moser, B. M.
769 Gillard, E. Karasik, K. Ramsey, C. Choi, B. A. Foster, J. Syron, J. Fleming,

770 H. Magazine, R. Hasz, G. D. Walters, J. P. Bridge, M. Miklos, S. Sulli-
771 van, L. K. Barker, H. M. Traino, M. Mosavel, L. A. Siminoff, D. R. Val-
772 ley, D. C. Rohrer, S. D. Jewell, P. A. Branton, L. H. Sobin, M. Barcus,
773 L. Qi, J. McLean, P. Hariharan, K. S. Um, S. Wu, D. Tabor, C. Shive,
774 A. M. Smith, S. A. Buia, A. H. Undale, K. L. Robinson, N. Roche, K. M.
775 Valentino, A. Britton, R. Burges, D. Bradbury, K. W. Hambright, J. Seleski,
776 G. E. Korzeniewski, K. Erickson, Y. Marcus, J. Tejada, M. Taherian, C. Lu,
777 M. Basile, D. C. Mash, S. Volpi, J. P. Struewing, G. F. Temple, J. Boyer,
778 D. Colantuoni, R. Little, S. Koester, L. J. Carithers, H. M. Moore, P. Guan,
779 C. Compton, S. J. Sawyer, J. P. Demchok, J. B. Vaught, C. A. Rabiner,
780 N. C. Lockhart, K. G. Ardlie, G. Getz, F. A. Wright, M. Kellis, S. Volpi,
781 and E. T. Dermitzakis. The Genotype-Tissue Expression (GTEx) pilot anal-
782 ysis: Multitissue gene regulation in humans. *Science*, 348(6235):648–660,
783 May 2015. ISSN 0036-8075, 1095-9203. doi: 10.1126/science.1262110. URL
784 <http://www.sciencemag.org/cgi/doi/10.1126/science.1262110>.

785 Marina Weidemann, Karin Schuster-Gossler, Michael Stauber, Christoph
786 Wrede, Jan Hegermann, Tim Ott, Karsten Boldt, Tina Beyer, Katrin
787 Serth, Elisabeth Kremmer, Martin Blum, Marius Ueffing, and Achim
788 Gossler. CFAP157 is a murine downstream effector of FOXJ1 that is
789 specifically required for flagellum morphogenesis and sperm motility. *De-*
790 *velopment*, 143(24):4736–4748, December 2016. ISSN 0950-1991, 1477-9129.
791 doi: 10.1242/dev.139626. URL [http://dev.biologists.org/lookup/doi/](http://dev.biologists.org/lookup/doi/10.1242/dev.139626)
792 [10.1242/dev.139626](http://dev.biologists.org/lookup/doi/10.1242/dev.139626).

793 Florence Wianny, Henry Kennedy, and Colette Dehay. Bridging the Gap be-
794 tween Mechanics and Genetics in Cortical Folding: ECM as a Major Driving
795 Force. *Neuron*, 99(4):625–627, August 2018. ISSN 08966273. doi: 10.1016/

796 j.neuron.2018.08.012. URL [https://linkinghub.elsevier.com/retrieve/](https://linkinghub.elsevier.com/retrieve/pii/S0896627318306858)
797 [pii/S0896627318306858](https://linkinghub.elsevier.com/retrieve/pii/S0896627318306858).

798 Hadley Wickham. *Ggplot2: elegant graphics for data analysis*. Use R! Springer,
799 New York, 2009. ISBN 978-0-387-98140-6. OCLC: ocn382399721.

800 Helen Rankin Willsey, Peter Walentek, Cameron R.T. Exner, Yuxiao Xu,
801 Andrew B. Lane, Richard M. Harland, Rebecca Heald, and Niovi San-
802 tama. Katanin-like protein *Katnal2* is required for ciliogenesis and brain
803 development in *Xenopus* embryos. *Developmental Biology*, 442(2):276–287,
804 October 2018. ISSN 00121606. doi: 10.1016/j.ydbio.2018.08.002. URL
805 <https://linkinghub.elsevier.com/retrieve/pii/S0012160618302628>.

806 Richard W. Wrangham. *The goodness paradox: the strange relationship between*
807 *virtue and violence in human evolution*. Pantheon Books, New York, first
808 edition edition, 2019. ISBN 978-1-101-87090-7.

809 Jiajun Yin, Wei Feng, Hongwei Yuan, Jianmin Yuan, Yue Wu, Xiaowei Liu,
810 Chunhui Jin, and Zaohuo Cheng. Association analysis of polymorphisms
811 in *STARD6* and near *ECHDC3* in alzheimer’s disease patients carrying the
812 *APOE*-epsilon4 allele. *Neuropsychiatric Disease and Treatment*, Volume 15:
813 213–218, January 2019. ISSN 1178-2021. doi: 10.2147/NDT.S186705. URL
814 [https://www.dovepress.com/association-analysis-of-polymorphisms-](https://www.dovepress.com/association-analysis-of-polymorphisms-in-stard6-and-near-echdc3-in-alz-peer-reviewed-article-NDT)
815 [in-stard6-and-near-echdc3-in-alz-peer-reviewed-article-NDT](https://www.dovepress.com/association-analysis-of-polymorphisms-in-stard6-and-near-echdc3-in-alz-peer-reviewed-article-NDT).

816 Chuan Yu, Xiaomin Yao, Linjie Zhao, Ping Wang, Qian Zhang, Chengjian
817 Zhao, Shaohua Yao, and Yuquan Wei. Wolf–Hirschhorn Syndrome Can-
818 didate 1 (*whsc1*) Functions as a Tumor Suppressor by Governing Cell
819 Differentiation. *Neoplasia*, 19(8):606–616, August 2017. ISSN 14765586.
820 doi: 10.1016/j.neo.2017.05.001. URL [https://linkinghub.elsevier.com/](https://linkinghub.elsevier.com/retrieve/pii/S1476558617300155)
821 [retrieve/pii/S1476558617300155](https://linkinghub.elsevier.com/retrieve/pii/S1476558617300155).

822 Diana Zala, Maria-Victoria Hinckelmann, Hua Yu, Marcel Menezes Lyra da
823 Cunha, Géraldine Liot, Fabrice P. Cordelières, Sergio Marco, and Frédéric
824 Saudou. Vesicular glycolysis provides on-board energy for fast axonal trans-
825 port. *Cell*, 152(3):479–491, January 2013. ISSN 1097-4172. doi: 10.1016/
826 j.cell.2012.12.029.

827 Daniel R Zerbino, Premanand Achuthan, Wasiu Akanni, M Ridwan Amode,
828 Daniel Barrell, Jyothish Bhai, Konstantinos Billis, Carla Cummins, Astrid
829 Gall, Carlos García Girón, Laurent Gil, Leo Gordon, Leanne Haggerty, Erin
830 Haskell, Thibaut Hourlier, Osagie G Izuogu, Sophie H Janacek, Thomas
831 Juettemann, Jimmy Kiang To, Matthew R Laird, Ilias Lavidas, Zhicheng
832 Liu, Jane E Loveland, Thomas Maurel, William McLaren, Benjamin Moore,
833 Jonathan Mudge, Daniel N Murphy, Victoria Newman, Michael Nuhn, Denye
834 Ogeh, Chuang Kee Ong, Anne Parker, Mateus Patricio, Harpreet Singh
835 Riat, Helen Schuilenburg, Dan Sheppard, Helen Sparrow, Kieron Taylor,
836 Anja Thormann, Alessandro Vullo, Brandon Walts, Amonida Zadissa, Adam
837 Frankish, Sarah E Hunt, Myrto Kostadima, Nicholas Langridge, Fergal J Mar-
838 tin, Matthieu Muffato, Emily Perry, Magali Ruffier, Dan M Staines, Stephen J
839 Trevanion, Bronwen L Aken, Fiona Cunningham, Andrew Yates, and Paul
840 Flicek. Ensembl 2018. *Nucleic Acids Research*, 46(D1):D754–D761, Jan-
841 uary 2018. ISSN 0305-1048, 1362-4962. doi: 10.1093/nar/gkx1098. URL
842 <http://academic.oup.com/nar/article/46/D1/D754/4634002>.

843 Yingyao Zhou, Bin Zhou, Lars Pache, Max Chang, Alireza Hadj Khodabakhshi,
844 Olga Tanaseichuk, Christopher Benner, and Sumit K. Chanda. Metas-
845 cape provides a biologist-oriented resource for the analysis of systems-level
846 datasets. *Nature Communications*, 10(1):1523, December 2019. ISSN 2041-
847 1723. doi: 10.1038/s41467-019-09234-6. URL [http://www.nature.com/
848 articles/s41467-019-09234-6](http://www.nature.com/articles/s41467-019-09234-6).

849 Ying Zhu, André M. M. Sousa, Tianliuyun Gao, Mario Skarica, Mingfeng
850 Li, Gabriel Santpere, Paula Esteller-Cucala, David Juan, Luis Ferrández-
851 Peral, Forrest O. Gulden, Mo Yang, Daniel J. Miller, Tomas Marques-Bonet,
852 Yuka Imamura Kawasawa, Hongyu Zhao, and Nenad Sestan. Spatiotempo-
853 ral transcriptomic divergence across human and macaque brain development.
854 *Science*, 362(6420):eaat8077, December 2018. ISSN 0036-8075, 1095-9203.
855 doi: 10.1126/science.aat8077. URL <http://www.sciencemag.org/lookup/>
856 [doi/10.1126/science.aat8077](https://doi.org/10.1126/science.aat8077).

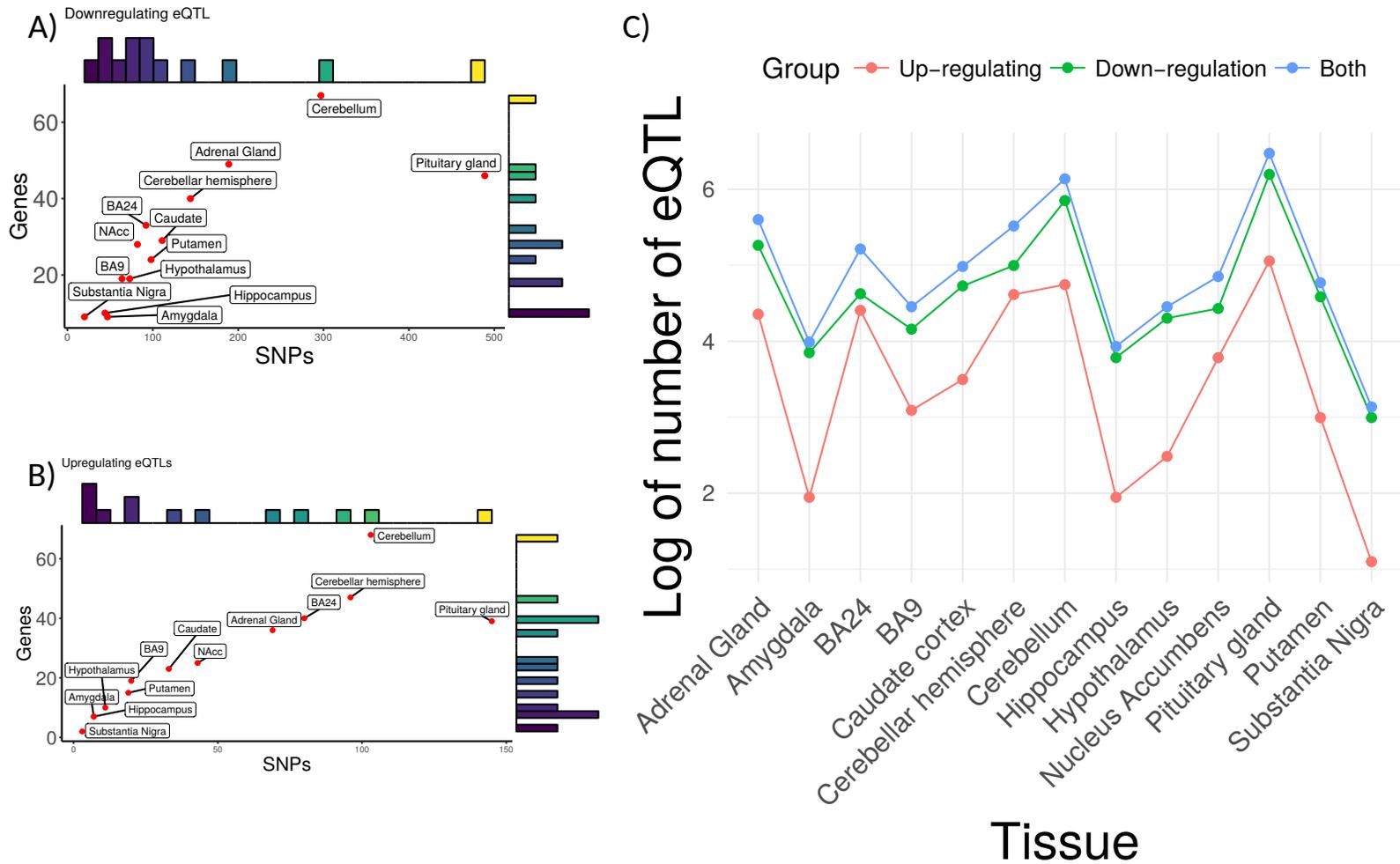


Figure 3: A) and B) Ratio of downregulating (A) and upregulating (B) eQTLs per genes affected in single tissues. C) Log value distribution in upregulating eQTL in human evolution, downregulating and total GTEx original data per tissue.

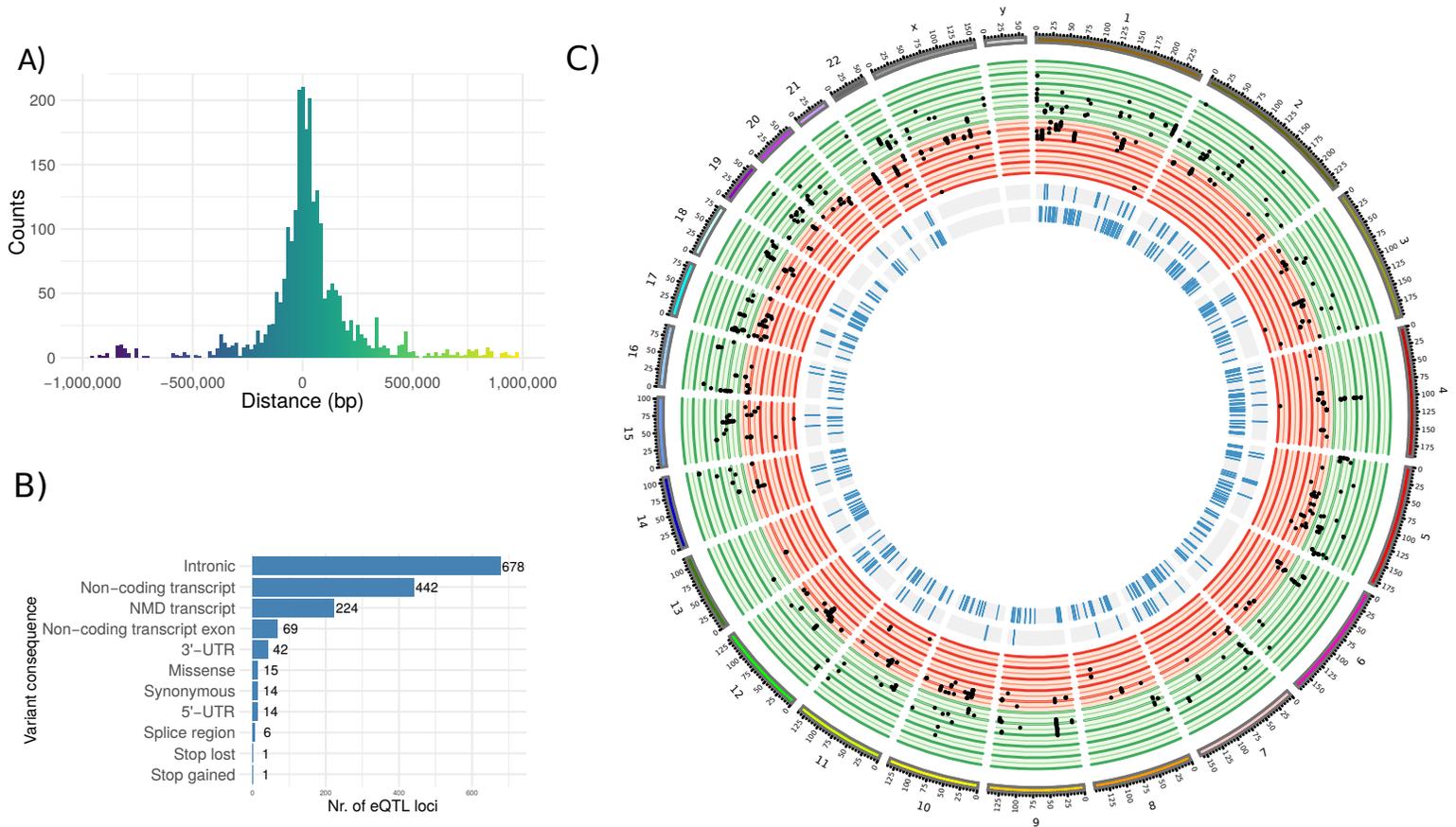


Figure 4: A) A bar plot showing the distribution of eQTL over distance to Transcription Starting Site B) Classification of the genetic consequences in our data C) A Circos plot showing the distribution along the genome of eQTLs. Each line denotes 0.5 steps in beta score (allele specific effects in gene expression), from 3 to -3. Red circles denote downregulation, green circles upregulation of eGenes. Inner rings, in blue: areas showing signals of positive selection relative to archaic humans in [Peyrégne et al., 2017] (innermost) and [Racimo et al., 2014] (outermost).