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Genome-wide association study identifies genetic factors that modify age at onset in Machado-Joseph disease

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34 Abstract

35 Machado-Joseph disease (MJD/SCA3) is the most common form of dominantly inherited ataxia 36 worldwide. The disorder is caused by an expanded CAG repeat in the ATXN3 gene. Past studies 37 have revealed that the length of the expansion partly explains the disease age at onset (AO) variability of MJD, which is confirmed in this study. Using a total of 786 MJD patients from five 38 39 different geographical origins, a genome-wide association study (GWAS) was conducted to identify additional AO modifying factors that could explain some of the residual AO variability. 40 We identified nine suggestively associated loci ($P < 1 \times 10^{-5}$). These loci were enriched for genes 41 42 involved in vesicle transport, olfactory signaling, and synaptic pathways. Furthermore, associations between AO and the TRIM29 and RAG genes suggests that DNA repair mechanisms 43 might be implicated in MJD pathogenesis. Our study demonstrates the existence of several 44 additional genetic factors, along with CAG expansion, that may lead to a better understanding of 45 the genotype-phenotype correlation in MJD. 46

47 Keywords

48 Machado-Joseph disease, ATXN3, MJD/SCA3, age at onset, modifier, GWAS

49 Introduction

Machado-Joseph disease, also known as spinocerebellar ataxia type 3 (MJD/SCA3), is an autosomal dominant neurodegenerative disorder that is characterized by progressive cerebellar ataxia and pyramidal signs, which can be associated with a complex clinical picture and includes extrapyramidal signs or amyotrophy [1, 2]. MJD is caused by an abnormal CAG trinucleotide repeat expansion in exon 10 of the ataxin-3 gene (*ATXN3*), located at 14q32.1. Deleterious expansions consensually contain 61 to 87 CAG repeats, whereas wild type alleles range from 12 to 44 [2].

As with other diseases caused by repeat expansions, such as Huntington's disease (HD) and other 57 58 spinocerebellar ataxias, there is an inverse correlation between expanded repeat size and the age 59 at which pathogenesis leads to disease onset [3]. Depending on the cohort structure, the size of the repeat expansion explains 55 to 70% of the age at onset (AO) variability in MJD, suggesting the 60 existence of additional modifying factors [3,4]. Although several genetic factors have been 61 proposed as modifiers, such as CAG repeat size of normal ATXN3 (SCA3), HTT (HD), ATXN2 62 (SCA2) and ATN1 (DRPLA) alleles, APOE status, and expression level of HSP40 [4,5,6], these 63 64 were not replicated by subsequent studies [7, 8]. Since CAG tract profile and allelic frequencies of the potential modifier loci can have unique characteristics in different populations, large 65 collaborative studies are required to identify genetic modifiers in MJD, as well as replicate the 66 67 findings of such studies [8].

Previously, Genetic Modifiers of Huntington's Disease (GeM-HD) Consortium carried out a GWA
approach of HD individuals to reveal genetic modifiers of AO in HD [9,10]. A total of eleven [9]
and fourteen loci [10] were found to be associated with residual age at HD onset. In the present

study, we performed the first GWAS to identify some possible genetic modifiers of AO in MJD. First, we assessed the relationship between AO and size of the expanded (CAG_{exp}) and normal (CAG_{nor}) alleles, biological sex and geographical origin. Next, we determined a residual AO for each subject, which is the difference between the measured AO and the predicted/estimated AO from expanded CAG repeat size alone. Using the residuals as a quantitative phenotype for a GWAS, we looked for genetic factors that modulate AO in MJD.

77 Methods

78 Study subjects

A total of 786 MJD patients from five distinct geographical origins (Portugal, Brazil, North America, Germany and Australia) were included in the present study. The overall average age at onset (standard deviation) was 38 (\pm 1.82) years, with a 1:1 male to female ratio. All subjects provided informed consent, and the study was approved by the respective institutional review boards. Detailed cohort demographics are shown in Supplementary Table 1.

84 Assessment of the *ATXN3* CAG repeat length

A singleplex polymerase chain reaction was performed to determine the length of the CAG_{exp} and CAG_{nor} alleles at exon 10 of *ATXN3* [11]. The final volume for each assay was 10 μ L: 7.5 ng of gDNA, 0.2 μ M of each primer, 5 μ L of Taq PCR Master Mix Kit Qiagen®, 1 μ L of Q-Solution from Qiagen® and H₂O. Fragment length analysis was done using ABIPrism 3730xl sequencer (Applied Biosystems®, McGill University and Genome Québec Innovation Centre) and GeneMapper software [12]. A stepwise regression model was performed to assess the correlation between AO and CAG_{exp} size, as well as gender, origin, CAG_{nor} size, and interaction between these

variables. Residual AO was calculated for each subject by subtracting individual's expected AO
based upon CAG_{exp} size from actual AO, to be used as the primary phenotype for following genetic
approach.

95 Genotyping, quality control and imputation

96 Samples were genotyped using the Global Screening Array v.1.0 from Illumina (636,139 markers). 97 Sample-based (missingness, relatedness, sex, and multidimensional scaling analysis) and SNPbased quality assessments (missingness, Hardy-Weinberg equilibrium, and minor allele 98 99 frequency) were conducted using PLINK version 1.9 [13]. In sample level QC, samples were 100 excluded with one or more of the following: high missingness (missingness rate > 0.05), close 101 relationship (pi-hat value > 0.2), discrepancy between genetically-inferred sex and reported sex, 102 population outliers (deviation \geq 4 SD from the population mean in multidimensional scaling analysis). All SNPs were checked for marker genotyping call rate (> 98%), minor allele frequency 103 (MAF) > 0.05, and HWE (p-value threshold = 1.0×10^{-5}). 104

Phasing and imputation were performed using SHAPEIT [14] and PBWT [15] pipelines, implemented on the Sanger Imputation Service [16]. Haplotype Reference Consortium (HRC) reference panel r1.1 containing 64,940 human haplotypes at 40,405,505 genetic markers were used as the reference panel. Imputed variants with an allele count of 30 (MAF > 0.02), an imputation quality score above 0.3 and an HWE p-value of > 1.0×10^{-5} were included for subsequent analysis.

110 Genome-wide association analysis

A genome-wide linear mixed model based association analysis was conducted using GCTA
version 1.91.7 [17]. Residual AO was modelled as a function of minor allele count of the test SNP,

sex, and the first three principal components based on the scree plot (Supplementary Figure 1).
The --mlma-loco option, which takes into account the difference in allele frequency between
populations, was used to control for population structure. QQ plots and Manhattan plots were
generated in FUMA v.1.3.4 [18]. Regional association plots were generated using LocusZoom
[19] (Supplementary Figure 3).

118 Functional annotation of SNPs

Genomic risk loci were defined using SNP2GENE function implemented in FUMA. Independent suggestive SNPs ($P < 1 \times 10^{-5}$) with a threshold of $r^2 < 0.6$ were selected within a 250 kb window. The UK Biobank release 2 European population consisting of randomly selected 10,000 subjects was used as the reference population panel. The ANNOVAR [20] categories and combined annotation-dependent depletion (CADD) [21] scores were obtained from FUMA for functional annotation. Functionally annotated variants were mapped to genes based on genomic position using FUMA positional mapping tool.

126 Pathway analysis

127 To identify known biological pathways and gene sets at the associated loci, an enrichment using public datasets containing Gene 128 approach was applied Ontology (GO. http://geneontology.org), the Kyoto Encyclopaedia of Genes and Genomes (KEGG, 129 130 https://www.genome.jp/kegg) and Reactome (https://reactome.org) pathways. The primary enrichment analysis was performed using the i-GSEA4GWAS v2 [22]. It uses a candidate list of 131 a genome-wide set of genes mapped within the SNP loci and ranks them based on the strength of 132 their association with the phenotype. Genes were mapped within 20 kb up or downstream of the 133 SNPs with a P < 0.05. Gene and pathway sets meeting a false discovery rate (FDR)-corrected q-134

value < 0.05 were regarded as significantly associated with high confidence, and q-value < 0.25135 was regarded to be possibly associated with the phenotype of interest. We performed a secondary 136 gene-based association test using the Versatile Gene-based Association Study (VEGAS) [23] 137 algorithm that controls the number of SNPs in each gene and the linkage disequilibrium (LD) 138 between these SNPs using the HapMap European population. As a third algorithm to identify 139 140 enriched pathways, we used Pathway Scoring Algorithm (PASCAL) [24], which controls for potential bias from gene size, SNP density, as well as LD. ClueGO [25] and CluePedia [26] plug-141 ins in Cytoscape were employed to visualize identified pathways and their clustering. 142

143 **Results**

144 The inverse correlation between CAG_{exp} and age at onset

145 In the first phase of the study, the expanded ATXN3-CAG repeat lengths of 786 MJD patients were assessed. The mean (SD) CAG_{exp} size were Australia: 68.2 (±3.3), Brazil: 74.3 (3.9), Germany: 146 147 72.9 (\pm 3.6), North America: 73 (\pm 4.3) and Portugal: 72 (\pm 4.0). Next, the relationship between AO and CAG_{exp} size, CAG_{nor} size, sex and ethnicity was examined (Supplementary Table 1). The 148 previously observed negative correlation between ATXN3 CAGexp size and AO [3] was confirmed 149 (Pearson's correlation coefficient $R^2 = 0.62$) (Figure 1). The CAG_{nor} size (P = 0.39), sex (P = 0.02) 150 and geographic origin (P [Brazil] = 0.38, P [Germany] = 0.38, P [North America] = 0.33, P 151 [Portugal] = 0.29) were not significant and their addition had little contribution to the model (ΔR^2 152 = 0.0072). Residual AO for each sample was calculated and used as a quantitative phenotype to 153 identify the modifiers of AO. The distribution of residual AO was close a theoretical normal 154 155 distribution (Figure 1).

156 Genome-wide association study

After post-imputation quality assessments, a total of 700 individuals with genotyping information 157 for 6,716,580 variants remained for GWAS. The resulting Manhattan plots and quantile-quantile 158 (QQ) plots are shown in Figure 2. The genomic inflation factor was close to one ($\lambda = 0.98$), 159 indicating the p-values were not inflated. No association signal was identified meeting genome-160 wide significance ($P < 5 \times 10^{-8}$, the genome-wide Bonferroni-corrected significance threshold); 161 however, genome-wide suggestive associations ($P < 1 \times 10^{-5}$) with 204 variants across 9 loci were 162 identified (Supplementary Table 3). The most significantly associated SNP at each locus are shown 163 164 in Table 1. Positional gene mapping aligned SNPs to 17 genes by their genomic location. Fourteen 165 of the 204 variants had a Combined Annotation Dependent Depletion (CADD)-PHRED score higher than the suggested threshold for deleterious SNPs (12.37), arguing the given loci have a 166 functional role [27]. 167

168 Interaction analysis between CAG_{exp} and SNP genotype

To assess a possible interaction between CAG_{exp} size and the variants identified, each of the nine variants was added to the initial linear regression, modelling AO as a function of CAG_{exp} size, SNP, sex, the first three principal components, CAG_{nor} size, and interaction of $SNP:CAG_{exp}$. Association of each independent SNP with AO revealed nominally significant p-values. Among the nine variants, only rs585809 (mapped to *TRIM29*) had a significant interaction with CAG_{exp} (P = 0.01), suggesting that rs585809 might modulate AO through this epistatic interaction on CAG_{exp} .

175 Association of HD-AO modifier variants in MJD

Association of previously identified HD-AO modifier loci in MJD were assessed. Among the 25
HD-AO modifier variants in 17 loci, a total of 18 variants (MAF > 0.02) in 12 loci were tested in
this study (Supplementary Table 4). None of these HD-AO modifiers reached the genome-wide

suggestive threshold. However, two variants rs144287831 (P = 0.02, effect size = - 0.98) and rs1799977 (P = 0.02, effect size = - 0.98) in the *MLH1* locus were found to be nominally associated with a later AO in MJD.

182 Pathway and gene-set enrichment analysis

183 A gene-set enrichment and pathway analysis was conducted using i-GSEA4GWAS. Various 184 approaches and algorithms are currently in use to conduct similar analyses. To be able to make 185 better comparisons with other studies that may use different approaches, we performed a secondary 186 gene-set enrichment and pathway analysis using the VEGAS2 and PASCAL software (Supplementary Tables 5-7). We also used these results for replication purposes in our own study. 187 188 A total of 13 overrepresented pathways were found, after FDR-multiple testing correction (q-value 189 < 0.05) in the primary GSEA analysis and replicated using at least one of the secondary gene-set enrichment algorithms (Table 2). Overall, the most significantly enriched gene-sets and pathways 190 191 were vesicle transport, olfactory signaling, and synaptic pathways. Visualization and clustering of pathways are shown in Figure 3. 192

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194 **Discussion**

Using five cohorts from different geographical origins, we performed the first GWAS to examine the presence of genetic factors that could modify AO in MJD. We identified a total of nine loci that were potentially associated with either an earlier or later AO. Concomitantly, we confirmed the previously observed negative correlation between CAG_{exp} and AO [3]. It was shown previously that normal *ATXN3* allele (CAG_{nor}) had a significant influence on AO of MJD [28]; however,

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several studies did not replicate this effect [6,8]. Indeed, we did not observe an association between CAG_{nor} and AO. However, it had little contribution to our model, with a minor difference in the correlation coefficient ($\Delta R^2 = 0.0012$).

In our GWAS, the strongest signal is for the rs11529293 variant ($P = 3.30 \times 10^{-6}$) within the 203 Cllorf72 and RAG loci at 11p12. Within this locus, two RAG genes, recombination-activating 204 genes RAG1 and RAG2, were shown to be implicated in DNA damage response and DNA repair 205 206 machineries [29,30]. The rs585809 variant, which was mapped to the TRIM29 gene, was found to 207 interact with CAG_{exp} , suggesting that it might have an effect on AO through this interaction. Both 208 RAG and TRIM29 loci were identified as AO-hastening modifiers. TRIM29 encodes for tripartite 209 motif protein 29, which is implicated in mismatch repair and double strand breaks pathways 210 [31,32]. TRIM29 is involved both upstream and downstream of these pathways, in the regulation 211 of DNA repair proteins into chromatin by mediating the interaction between them. One of these 212 DNA repair proteins is MLH1, which is implicated in mismatch repair complex [32]. Previously, 213 the *MLH1* locus was identified as an AO modifier in another neurodegenerative disease caused by 214 CAG repeat expansion, Huntington's disease [9,10,33]. Additionally, in a genome-wide genetic 215 screening study, MLH1-knock out was shown to modify the somatic expansion of the CAG repeat 216 and slow the pathogenic process in HD mouse model [34]. Overall, the association of TRIM29 and 217 RAG loci suggests that DNA repair mechanisms may be implicated in the alteration of AO of MJD, as well as HD, and may have a role in the pathogenesis of other CAG repeat diseases. Interestingly, 218 219 in a previous study, we found variants in three transcription-coupled repair genes (ERCC6, RPA, 220 and CDK7) associated with different CAG instability patterns in MJD [35].

We identified gene-sets enriched in olfactory signaling, vesicle transport, and synaptic pathways.Olfactory dysfunction is one of the main non-motor symptoms that was already described in

223 patients with MJD [36,37]. In a previous study, transplantation of olfactory ensheathing cells, which are specialized glial cells of the primary olfactory system, were found to improve motor 224 function in an MJD mice model, and were suggested as a novel potential strategy for MJD 225 treatment [38]. Vesicle transport and synaptic pathways were also implicated in MJD, as well as 226 227 in other neurodegenerative diseases [39,40]. An interruption of synaptic transmission caused by 228 an expanded polyglutamine repeat and mutant ataxin-3 aggregates were shown in *Drosophila* and Caenorhabditis elegans models of MJD. Therefore, the interaction between synaptic vesicles and 229 230 mutant aggregates supports the role of synaptic vesicle transport in the pathogenesis of MJD 231 [41,42]. Overall, we suggest that these gene-sets and pathways might construct a larger molecular network that could modulate the AO in MJD. 232

In summary, our study identified nine genetic loci that may modify the AO of MJD. Identification of *TRIM29* and *RAG* genetic variants, as well as our gene-set enrichment analyses, implicated DNA repair, olfactory signaling, synaptic, and vesicle transport pathways in the pathogenesis of MJD. Although we used different cohorts from five distinct geographical ethnicities, a replication study in similar or additional populations would add valuable evidence to support our findings.

239 **Description of Supplemental Data**

240 Supplemental Data include three figures and seven tables

241 **Declaration of Interests**

242 The authors declare no conflict of interest.

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SNP	Chr	Position (GRCh37)	Nearest gene	Minor allele	Major allele	MJD MAF	1KGP MAF	b (SNP effect)	P-value
rs62171220	2	137802855	THSD7B	G	С	0.13	0.11	2.71	4.45×10^{-6}
rs2067390	2	191209028	HIBCH, INPP1	А	Т	0.04	0.06	4.74	6.39×10^{-6}
rs144891322	5	85135387	RPL5P17,	С	Т	0.02	0.007	6.10	$5.18 imes10^{-6}$
rs11529293	11	36855388	Cllorf74, RAG1,	Т	С	0.14	0.26	-2.71	$3.30 imes 10^{-6}$
			RAG2						
rs7480166	11	42984753	HNRNPKP3	А	G	0.40	0.40	-1.86	$4.17 imes 10^{-6}$
rs585809	11	119949979	TRIM29	Т	С	0.06	0.17	-3.76	$9.50 imes 10^{-6}$
rs72660056	13	113507543	ATP11A	А	G	0.08	0.05	-3.29	$3.94 imes 10^{-6}$
rs11857349	15	99924857	TTC23, SYNM,	G	А	0.04	0.02	-4.58	3.43×10^{-6}
			LRRC28						
rs8141510	22	42821185	NFAM1, CYP2D6,	C	Т	0.43	0.49	1.83	$3.94 imes 10^{-6}$
			NAGA, NDUFA6	С					

Table 1. Suggestive loci associated with residual age at onset in MJD. Chr: chromosome, MAF: minor allele frequency, 1KGP: 1000
 Genomes Project

Pathway	Description	p-value (GSEA)	q-value (GSEA)	p-value (VEGAS)	permuted p- value (VEGAS)	p-value (PASCAL)
GO:0030133	transport vesicle	$< 1.0 \text{ x } 10^{-3}$	8.20 x 10 ⁻³	6.15 x 10 ⁻⁴⁰	4.46 x 10 ⁻¹	6.70 x 10 ⁻³
KEGG:04740	olfactory transduction	$< 1.0 \text{ x } 10^{-3}$	8.30 x 10 ⁻³	NA	NA	3.89 x 10 ⁻⁴
R-HSA:381753	olfactory signaling pathway	$< 1.0 \text{ x } 10^{-3}$	8.80 x 10 ⁻³	1.10 x 10 ⁻²⁷	7.71 x 10 ⁻¹	2.51 x 10 ⁻⁴
GO:0044456	synapse part	$< 1.0 \text{ x } 10^{-3}$	9.30 x 10 ⁻³	1.25 x 10 ⁻¹⁸²	< 1.0 x 10 ⁻⁶	$< 1.0 \text{ x } 10^{-7}$
R-HSA:74217	purine salvage	$< 1.0 \text{ x } 10^{-3}$	1.06 x 10 ⁻²	1.06 x 10 ⁻²	2.15 x 10 ⁻¹	6.48 x 10 ⁻³
GO:0045202	synapse	$< 1.0 \text{ x } 10^{-3}$	1.15 x 10 ⁻²	1.15 x 10 ⁻²	$< 1.0 \text{ x } 10^{-6}$	< 1.0 x 10 ⁻⁷
GO:0004177	aminopeptidase activity	$< 1.0 \text{ x } 10^{-3}$	1.50 x 10 ⁻²	1.50 x 10 ⁻²	3.41 x 10 ⁻¹	1.24 x 10 ⁻²
GO:0008238	exopeptidase activity	$< 1.0 \text{ x } 10^{-3}$	1.80 x 10 ⁻²	1.80 x 10 ⁻²	2.80 x 10 ⁻²	8.31 x 10 ⁻³
GO:0006898	receptor mediated endocytosis	$< 1.0 \text{ x } 10^{-3}$	2.25 x 10 ⁻²	2.25 x 10 ⁻²	2.03 x 10 ⁻¹	6.64 x 10 ⁻³
GO:0016917	GABA receptor activity	$< 1.0 \text{ x } 10^{-3}$	2.26 x 10 ⁻²	2.26 x 10 ⁻²	1.30 x 10 ⁻⁴	2.30 x 10 ⁻⁵
GO:0030140	trans Golgi network transport vesicle	< 1.0 x 10 ⁻³	2.36 x 10 ⁻²	2.36 x 10 ⁻²	2.80 x 10 ⁻²	1.28 x 10 ⁻¹
GO:0009725	response to hormone stimulus	< 1.0 x 10 ⁻³	2.73 x 10 ⁻²	2.73 x 10 ⁻²	1.32 x 10 ⁻¹	1.30 x 10 ⁻⁴
GO:0030425	Dendrite	< 1.0 x 10 ⁻³	3.86 x 10 ⁻²	3.86 x 10 ⁻²	$< 1.0 \text{ x } 10^{-6}$	< 1.0 x 10 ⁻⁷

Table 2. Pathways significant after multiple-correction ($q < 5 \times 10^{-2}$) in the primary GSEA analysis and replicated using at least one of the secondary gene-set enrichment algorithms. NA means that the pathway was not enriched by at least two significant genes in VEGAS.

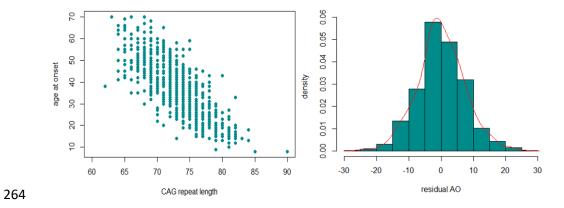


Figure 1. The inverse correlation between CAG_{exp} and AO (left) and the distribution of residual AO (right) observed in our MJD cohort.

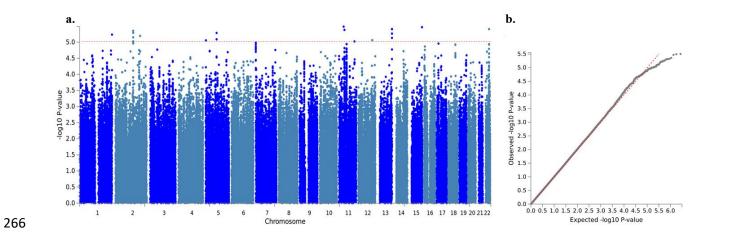
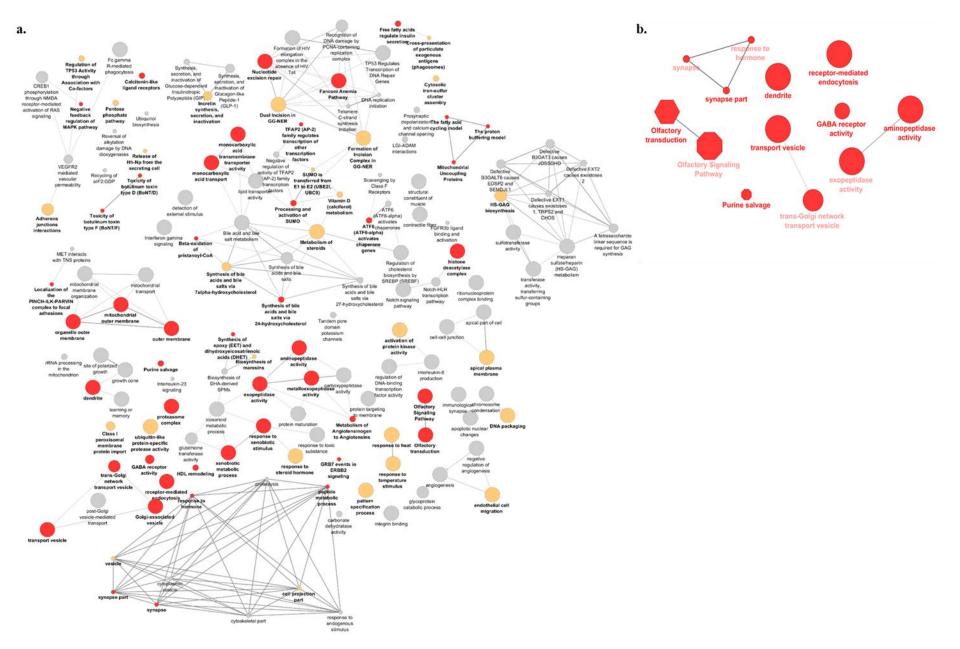


Figure 2. Manhattan plot (a) and QQ plot (b) of the GWAS for residual AO of MJD. Imputed using the HRC panel, 6,716,580 variants that passed QC are included in the plot. The x-axis shows the physical position along the genome. The y-axis shows the $-\log_{10}(p-value)$ for association. The red line indicates the level of genome-wide suggestive association ($P = 1 \times 10^{-5}$).



- Figure 3. Visualization of the gene-sets and pathways enriched in primary GSEA analysis (a) and replicated in VEGAS and PASCAL
- (b). The size of the nodes corresponds to the number of the genes associated with a term. The significance is represented by the color of
- the nodes (P < 0.05, 0.05 < P < 0.1 and P > 0.1 are represented by red, yellow and gray, respectively).

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