

Open access • Journal Article • DOI:10.1038/MP.2017.193

Genome-wide association study across European and African American ancestries identifies a SNP in DNMT3B contributing to nicotine dependence. — Source link

Dana B. Hancock, Yuelong Guo, Gunnar W Reginsson, Nathan C. Gaddis ...+37 more authors

Institutions: Research Triangle Park, deCODE genetics, Anschutz Medical Campus, Boston University ...+15 more institutions

Published on: 31 Aug 2018 - Molecular Psychiatry (Nature Publishing Group)

Topics: Nicotine, Smoking cessation, Genome-wide association study, Expression quantitative trait loci and Quantitative trait locus

Related papers:

- · Association studies of up to 1.2 million individuals yield new insights into the genetic etiology of tobacco and alcohol use
- Biological insights from 108 schizophrenia-associated genetic loci
- · Transancestral GWAS of alcohol dependence reveals common genetic underpinnings with psychiatric disorders
- · Genome-wide meta-analyses identify multiple loci associated with smoking behavior
- LD score regression distinguishes confounding from polygenicity in genome-wide association studies :











https://helda.helsinki.fi

Genome-wide association study across European and African American ancestries identifies a SNP in DNMT3B contributing to nicotine dependence

Hancock, D. B.

2018-09

 $\label{eq:hancock} Hancock \ , D \ B \ , Guo \ , Y \ , Reginsson \ , G \ W \ , Gaddis \ , N \ C \ , Lutz \ , S \ M \ , Sherva \ , R \ , Loukola \ , A \ , Minica \ , C \ C \ , Markunas \ , C \ A \ , Han \ , Y \ , Young \ , K \ A \ , Gudbjartsson \ , D \ F \ , Gu \ , F \ , McNeil \ , D \ W \ , Qaiser \ , B \ , Glasheen \ , C \ , Olson \ , S \ , Landi \ , M \ T \ , Madden \ , P \ A \ F \ , Farrer \ , L \ A \ , Vink \ , J \ , Saccone \ , N \ L \ , Neale \ , M \ C \ , Kranzler \ , H \ R \ , McKay \ , J \ , Hung \ , R \ J \ , Amos \ , C \ I \ , Marazita \ , M \ L \ , Boomsma \ , D \ I \ , Baker \ , T \ B \ , Gelernter \ , J \ , Kaprio \ , J \ , Caporaso \ , N \ E \ , Thorgeirsson \ , T \ E \ , Hokanson \ , J \ E \ , Bierut \ , L \ J \ , Stefansson \ , K \ & Johnson \ , E \ O \ 2018 \ , ' Genome-wide association study across European and African American ancestries identifies a SNP in DNMT3B contributing to nicotine dependence ' , Molecular Psychiatry \ , vol. 23 \ , no. 9 \ . https://doi.org/10.1038/mp.2017.193$

http://hdl.handle.net/10138/272981 https://doi.org/10.1038/mp.2017.193

unspecified published Version

Downloaded from Helda, University of Helsinki institutional repository.

This is an electronic reprint of the original article.

This reprint may differ from the original in pagination and typographic detail.

Please cite the original version.

www.nature.com/mp

ORIGINAL ARTICLE

Genome-wide association study across European and African American ancestries identifies a SNP in *DNMT3B* contributing to nicotine dependence

DB Hancock¹, Y Guo², GW Reginsson³, NC Gaddis⁴, SM Lutz⁵, R Sherva⁶, A Loukola⁷, CC Minica⁸, CA Markunas¹, Y Han⁹, KA Young¹⁰, DF Gudbjartsson^{3,11}, F Gu¹², DW McNeil^{13,14}, B Qaiser⁷, C Glasheen¹, S Olson¹⁵, MT Landi¹², PAF Madden¹⁶, LA Farrer^{6,17,18,19,20}, J Vink^{8,21}, NL Saccone²², MC Neale^{23,24}, HR Kranzler^{25,26}, J McKay²⁷, RJ Hung²⁸, CI Amos⁹, ML Marazita²⁹, DI Boomsma⁸, TB Baker³⁰, J Gelernter^{31,32,33,34}, J Kaprio^{7,35}, NE Caporaso¹², TE Thorgeirsson³, JE Hokanson¹⁰, LJ Bierut¹⁶, K Stefansson³ and EO Johnson³⁶

Cigarette smoking is a leading cause of preventable mortality worldwide. Nicotine dependence, which reduces the likelihood of quitting smoking, is a heritable trait with firmly established associations with sequence variants in nicotine acetylcholine receptor genes and at other loci. To search for additional loci, we conducted a genome-wide association study (GWAS) meta-analysis of nicotine dependence, totaling 38.602 smokers (28.677 Europeans/European Americans and 9925 African Americans) across 15 studies. In this largest-ever GWAS meta-analysis for nicotine dependence and the largest-ever cross-ancestry GWAS meta-analysis for any smoking phenotype, we reconfirmed the well-known CHRNA5-CHRNA3-CHRNB4 genes and further yielded a novel association in the DNA methyltransferase gene DNMT3B. The intronic DNMT3B rs910083-C allele (frequency = 44-77%) was associated with increased risk of nicotine dependence at $P = 3.7 \times 10^{-8}$ (odds ratio (OR) = 1.06 and 95% confidence interval (CI) = 1.04 - 1.07 for severe vs mild dependence). The association was independently confirmed in the UK Biobank (N = 48,931) using heavy vs never smoking as a proxy phenotype ($P = 3.6 \times 10^{-4}$, OR = 1.05, and 95% CI = 1.02–1.08). Rs910083-C is also associated with increased risk of squamous cell lung carcinoma in the International Lung Cancer Consortium (N = 60,586, meta-analysis P = 0.0095, OR = 1.05, and 95% CI = 1.01–1.09). Moreover, rs910083-C was implicated as a cis-methylation quantitative trait locus (QTL) variant associated with higher DNMT3B methylation in fetal brain (N = 166, $P = 2.3 \times 10^{-26}$) and a cis-expression QTL variant associated with higher DNMT3B expression in adult cerebellum from the Genotype-Tissue Expression project (N = 103, $P = 3.0 \times 10^{-6}$) and the independent Brain eQTL Almanac (N=134, P=0.028). This novel DNMT3B cis-acting QTL variant highlights the importance of genetically influenced regulation in brain on the risks of nicotine dependence, heavy smoking and consequent lung cancer.

Molecular Psychiatry advance online publication, 3 October 2017; doi:10.1038/mp.2017.193

INTRODUCTION

Cigarette smoking is a leading cause of preventable death, resulting annually in nearly 6 million premature deaths worldwide. Smoking-related deaths are most often attributed to

increased rates of cancer, cardiovascular disease and chronic obstructive pulmonary disease.² Despite the well-known adverse health effects, an estimated 45.3 million US adults smoke cigarettes, of whom over 68% report wanting to quit.^{3,4}

¹Behavioral and Urban Health Program, Behavioral Health and Criminal Justice Division, RTI International, Research Triangle Park, NC, USA; ²Center for Genomics in Public Health and Medicine, RTI International, Research Triangle Park, NC, USA; 3deCODE Genetics/Amgen, Reykjavik, Iceland; 4Research Computing Division, RTI International, Research Triangle Park, NC, USA; 5Department of Biostatistics and Informatics, University of Colorado Anschutz Medical Campus, Aurora, CO, USA; 6Department of Medicine (Biomedical Genetics), Boston University School of Medicine, Boston, MA, USA; ⁷Institute for Molecular Medicine Finland (FIMM), University of Helsinki, Helsinki, Finland; ⁸Department of Biological Psychology, Vrije Universiteit, Amsterdam, The Netherlands; ⁹Biomedical Data Science, Geisel School of Medicine at Dartmouth, Hanover, NH, USA; ¹⁰Department of Epidemiology, University of Colorado Anschutz Medical Campus, Aurora, CO, USA; 11Department of Engineering and Natural Sciences, University of Iceland, Reykjavík, Iceland; 12Genetic Epidemiology Branch, Division of Cancer Epidemiology and Genetics, National Cancer Institute, National Institutes of Health, United States Department of Health and Human Services, Bethesda, MD, USA; 13 Department of Psychology, West Virginia University, Morgantown, WV, USA; 14 Department of Dental Practice and Rural Health, West Virginia University, Morgantown, WV, USA; ¹⁵Public Health Informatics Program, eHealth, Quality and Analytics Division, RTI International, Research Triangle Park, NC, USA; ¹⁶Department of Psychiatry, Washington University, St. Louis, MO, USA; ¹⁷Department of Neurology, Boston University School of Medicine, Boston, MA, USA; ¹⁸Department of Ophthalmology, Boston University School of Medicine, Boston, MA, USA; ¹⁹Department of Epidemiology, Boston University School of Public Health, Boston, MA, USA; ²⁰Department of Biostatistics, Boston University School of Public Health, Boston, MA, USA; ²¹Behavioural Science Institute, Radboud University, Nijmegen, The Netherlands; ²²Department of Genetics, Washington University, St. Louis, MO, USA; 23Virginia Institute for Psychiatric and Behavioral Genetics, Virginia Commonwealth University, Richmond, VA, USA; 24Department of Psychiatry, Virginia Commonwealth University, Richmond, VA, USA; 25Department of Psychiatry, University of Pennsylvania Perelman School of Medicine, Philadelphia, PA, USA; ²⁶Crescenz VA Medical Center, Philadelphia, PA, USA; ²⁷International Agency for Research on Cancer, World Health Organization, Lyon, France; ²⁸Lunenfeld-Tanenbaum Research Institute, Sinai Health System, University of Toronto, Toronto, ON, Canada; 29Center for Craniofacial and Dental Genetics, Department of Oral Biology, University of Pittsburgh, Pittsburgh, PA, USA; 30Center for Tobacco Research and Intervention, Department of Medicine, University of Wisconsin School of Medicine and Public Health, Madison, WI, USA; ³¹Department of Psychiatry, Yale University School of Medicine, New Haven, CT, USA; ³²Department of Genetics, Yale University School of Medicine, New Haven, CT, USA; 33 Department of Neuroscience, Yale University School of Medicine, New Haven, CT, USA; 34 VA CT Healthcare Center, Department of Psychiatry, West Haven, CT, USA; 35 Department of Public Health, Faculty of Medicine, University of Helsinki, Helsinki, Finland and 36Fellow Program and Behavioral Health and Criminal Justice Division, RTI International, Research Triangle Park, NC, USA. Correspondence: Dr DB Hancock, RTI International, 3040 East Cornwallis Road, PO Box 12194, Research Triangle Park, NC 27709, USA. E-mail: dhancock@rti.org

Received 28 March 2017; revised 14 July 2017; accepted 17 July 2017

Cigarette smoking is a complex multi-step behavior involving initiation, regular smoking, nicotine dependence, cessation and relapse. Some regular smokers maintain low-level smoking without developing symptoms of dependence,⁵ while others become heavily dependent smokers and experience the most difficulty with cessation and the highest risk of relapse.^{6,7} Nicotine dependence has high heritability (estimates up to 75%),^{8,9} and besides reducing the likelihood of quitting smoking, it is predictive of withdrawal severity,¹⁰ response to treatment¹¹ and smoking-related health outcomes.^{12,13}

Genome-wide association study (GWAS) analyses of nicotine dependence phenotypes ^{14–24} have firmly established associations with several loci, including nicotinic acetylcholine receptor genes on chromosomes 15g25 (CHRNA5-CHRNA3-CHRNB4), (CHRNB3-CHRNA6) and 20q13 (CHRNA4). The largest GWAS metaanalyses relied on widely ascertained phenotypes such as cigarettes per day (CPD), 16-18 which represents only one of several components of nicotine dependence.²⁵ Focusing GWAS on nicotine dependence rather than CPD may improve statistical power for identifying variants that influence the broader construct of dependence.¹⁹ This idea is supported by our prior nicotine dependence GWAS meta-analysis (total N=17,074 ever-smokers of European/European American ancestry (EUR)]) that discovered associations with CHRNA4 single-nucleotide polymorphisms (SNPs) that were driven by time to first cigarette in the morning (TTFC) and had not been detected in GWAS meta-analyses of CPD with much larger sample sizes.²³ To improve statistical power further and to search for additional loci, we more than doubled our sample size to perform the largest GWAS meta-analysis of nicotine dependence to date, including 38,602 ever-smokers (28,677 of EUR and 9925 of African American (AA) ancestries) across 15 studies. We extended our study to include correlations with DNA methylation (DNAm) and RNA expression (RNAexp) of nearby genes across human brain tissues and evaluated associations with a critical smoking-related outcome: lung cancer (total N = 81,821 cases and controls).

MATERIALS AND METHODS

Study protocols received institutional review board approval at their respective sites. All study participants provided written informed consent. We included the five studies from our prior GWAS meta-analysis²³ and 10 additional studies. Details of their study design, genotyping, quality control, 1000 Genomes (1000G) imputation and analysis are provided in Supplementary Methods and Supplementary Table 1.

Nicotine dependence phenotype

We included studies with SNP genotypes and Fagerström Test for Nicotine Dependence (FTND) data collected among smokers. FTND scores range from 0 (no dependence) to 10 (highest dependence level). As before, we used FTND to categorize nicotine dependence as mild (scores 0–3), moderate (scores 4–6) or severe (scores 7–10). Two of the 15 studies additionally included low-intensity smokers who reported CPD as \leq 10 but had no data available on the other FTND items and were defined as mildly dependent. Concordance rates between these FTND and CPD categories showed minimal phenotype misclassification (Supplementary Methods).

Nicotine dependence GWAS meta-analysis and independent follow-up

We used linear regression to test SNP/indel associations with categorical nicotine dependence (mild=0/moderate=1/severe=2) in each separate study and ancestry group. Covariates included age, sex, principal components and study-specific covariates (as needed); additional adjustment for family structure was made in studies with relatives included (Supplementary Methods).

We combined GWAS results, using METAL²⁷ with fixed-effects inverse variance-weighting meta-analysis, across all studies with FTND data to maximize statistical power. Genomic control was applied to the deCODE results to adjust for inflation due to relatedness among participants; all

other studies had low inflation values ($\lambda \leq 1.02$). We excluded SNPs/indels with minor allele frequency < 1% in 1000G EUR or African (AFR) panels, depending on the ancestry group analyzed. The standard threshold $(P < 5 \times 10^{-8})$, originally based on 1 million independent tests genomewide as computed using HapMap-based imputation for EUR studies, ²⁸ has been validated for 1000G-imputed GWAS of common variants.²⁹ Rather than imposing a more stringent, yet to be consistently determined, threshold when analyzing common variants across EUR and AA studies. $^{29-32}$ we carried forward novel variation implicated at $P < 5 \times 10^{-8}$ and relied on confirmation in an independent study to declare genomewide significance. For this confirmation step, we utilized UK Biobank (N = 48,931 EUR participants) results with heavy, defined as pack-years ([CPD/20 cigarettes per pack] × years smoked) ≥ 10, vs never smoking as a proxy phenotype.²⁴ This prior GWAS was designed as a nested case-control analysis that sampled the extremes of smoking behavior, and thus did not encompass light smoking.

Regional association plots were created using LocusZoom³³ with linkage disequilibrium (LD) estimates of r^2 and D' based on 1000G EUR and AFR panels. Allele frequencies were weighted by sample size. Odds ratio (OR) estimates were computed using the β estimate from the SNP term in the linear regression model ($e^{[2\times\beta]}$ for severe vs mild dependence) and then compared across studies using the Forest Plot Viewer.³⁴ Heterogeneity across studies was assessed using the I^2 index.³⁵

SNP associations with DNAm and RNAexp

The top novel nicotine dependence-associated variant was assessed for *cis*-acting regulatory effects, using previously published methylation quantitative trait locus (meQTL) and expression QTL (eQTL) studies of postmortem brain tissues (Supplementary Methods). We used meQTL results from 166 fetal brain samples in the Human Developmental Biology Resource and UK Medical Research Council Brain Banks network; these samples were not dissected further into different tissues. S SNP genotypes were imputed using 1000G, and DNAm was measured using the Illumina HumanMethylation450 BeadChip. The QTL results were obtained at http://epigenetics.essex.ac.uk/mQTL/.

We assessed SNP associations with RNAexp of nearby genes (±1 Mb around the transcription start sites) across adult brain tissues from the Genotype-Tissue Expression (GTEx; https://gtexportal.org) project³⁷ with 1000G-imputed genotypes and RNA sequencing measures from up to 103 deceased persons. We followed up *cis*-eQTL SNP results in the Brain eQTL Almanac data set (*N*=134 EUR deceased persons, http://www.braineac.org/), using RNAexp probes measured on Affymetrix Human Exon 1.0 ST arrays across brain tissues, many of which overlap with tissues available in GTEx.³⁸

SNP associations with lung cancer

The top novel nicotine dependence-associated variant was tested for association with lung cancer using 27,349 cases and 54,472 controls from > 30 EUR studies in the Transdisciplinary Research for Cancer in Lung of the International Lung Cancer Consortium (TRICL-ILCCO). Their genotyping, quality control, 1000G imputation and statistical analyses were previously described.³⁹ Briefly, the SNP associations were derived from a meta-analysis of logistic regression model results that compared lung cancer cases and controls, adjusting for age, sex and the first two principal components. Histological subtypes of adenocarcinoma and squamous cell lung carcinoma cases were compared, separately, to controls. We removed the EAGLE study from the original set of lung cancer studies to avoid overlap in the nicotine dependence and lung cancer meta-analyses. All lung cancer analyses included ever- and never-smokers; our follow-up association testing made additional adjustments for smoking history (ever vs never) and pack-years.

RESULTS

We performed GWAS analyses across 15 studies, totaling 38,602 (28,677 EUR and 9925 AA) ever-smokers (Supplementary Table 2), with nicotine dependence defined as mild (N=17,796; 46.1%), moderate (N=13,527; 35.0%) or severe (N=7279; 18.9%). More than 99% of the participants were ≥ 18 years old. Males constituted 53.2% of the total sample size.

Our GWAS meta-analysis tested nearly 18 million genotyped and 1000G-imputed SNPs/indels for association with mild/

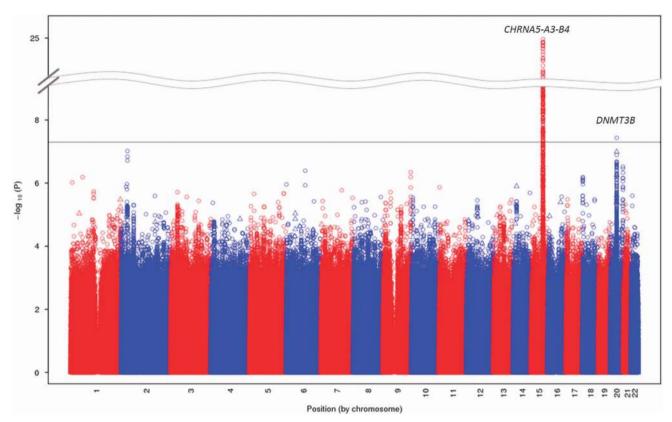


Figure 1. Manhattan plot of SNP and indel associations with nicotine dependence from GWAS meta-analysis across 15 studies (total N = 38,602 European/European Americans and African Americans). The $-\log_{10}$ meta-analysis P-values are plotted by chromosomal position of SNPs (depicted as circles) and indels (depicted as triangles). The genome-wide statistical significance threshold ($P < 5 \times 10^{-8}$) is shown as a solid black line. GWAS, genome-wide association study; SNP, single-nucleotide polymorphism.

moderate/severe dependence (λ = 1.027, Supplementary Figure 1A). We observed SNP/indel associations at meta-analysis $P < 5 \times 10^{-8}$ at two loci (Figure 1): the well-known chromosome 15q25 locus spanning *CHRNA5-CHRNA3-CHRNB4* (smallest meta-analysis P = 1.3 × 10 $^{-25}$) and a novel chromosome 20q11 locus spanning the DNA (cytosine-5-)-methyltransferase 3 beta (*DNMT3B*) gene (smallest meta-analysis P = 3.7 × 10 $^{-8}$). No additional loci were identified at P < 5 × 10 $^{-8}$ in the ancestry-specific GWAS meta-analyses (Supplementary Figures 1B–C and 2A–B).

DNMT3B SNP associations with nicotine dependence

Rs910083, an intronic *DNMT3B* SNP, was identified across both ancestries: meta-analysis $P = 3.7 \times 10^{-8}$ and β (standard error) = 0.032 (0.0057) for the C allele (Table 1), corresponding to OR (95% confidence interval (CI)) of 1.06 (1.04–1.07) for severe vs mild dependence (Supplementary Figure 3). Rs910083 was imputed well (quality scores = 0.98–1), and it showed no evidence of heterogeneity across studies (P = 0.71).

Rs910083-C, the minor allele for EUR (frequency = 44%) but the major allele for AA (frequency = 77%), is associated with increased nicotine dependence risk (Table 1): EUR-specific meta-analysis $P=4.1\times10^{-5}$ and OR (95% CI) = 1.06 (1.03–1.08); and AA-specific meta-analysis $P=7.3\times10^{-5}$ and OR (95% CI) = 1.10 (1.05–1.15). Many SNPs/indels were in moderate to high LD ($r^2>0.4$) with rs910083 in 1000G EUR, spanning 220 kb (chr20:31,268,924–31,488,466) and including *DNMT3B* and its neighboring genes, microtubule-associated protein, RP/EB family, member 1 (*MAPRE1*) and COMM domain containing 7 (*COMMD7*). However, in 1000G AFR, SNPs in LD ($r^2>0.4$) with rs910083 were localized to a 47 kb

region (chr20:31,356,560–31,403,394) including only *DNMT3B* (Figures 2A–B, NCBI build 37 positions).

No DNMT3B variants have been implicated previously for any substance use disorder (SUD) phenotype. Upstream of DNMT3B, chromosome 20g11 also harbors the nucleolar protein 4-like (NOL4L) gene, which was reported at genome-wide significance for heavy vs never smoking in the UK Biobank for the indel rs57342388.²⁴ This indel was associated at meta-analysis P = 0.0017 in our study (Table 1): OR (95% CI) = 1.04 (1.02–1.07) for severe vs mild dependence for the insertion allele, consistent with the prior result. Rs57342388 is located 216 kb upstream of our top DNMT3B SNP rs910083. The two variants are weakly correlated ($r^2 = 0.11$ in 1000G EUR where minor allele frequency = 2% for rs57342388 vs 18% for rs910083, $r^2 = 0.0022$ in AFR where minor allele frequency = 19% vs 42%) but are in moderate to high LD (D' = 0.57 in EUR, D' = 1 in AFR). In follow-up testing with both SNPs included in the same model, both were associated with nicotine dependence (meta-analysis $P = 1.7 \times 10^{-6}$ for rs910083 and 8.3×10^{-3} for rs57342388), showing that our observed DNMT3B association signal is not explained by the previously reported NOL4L signal.

We tested rs910083-C for association with each of the specific FTND items, as presented in Supplementary Table 3. The rs910083-C association was driven most strongly by TTFC (meta-analysis $P=1.2\times10^{-4}$). Its next most significantly associated FTND item was CPD (meta-analysis P=0.0011). TTFC is an indicator of withdrawal severity upon awakening⁴⁰ and behavioral automaticity (habitual smoking without awareness or cognitive control⁷). Although TTFC has its distinct features (for example, strongest

Table 1.	SNP association	Table 1. SNP associations in the novel DNM73B gene and previously reported genes that were identified at $P < 5 \times 10^{-7}$ in the nicotine dependence GWAS meta-analysis	ne and previously repor	ted genes that	were identifi	ied at $P < 5 \times 10$	⁻⁷ in the nic	otine dependence	e GWAS meta-ar	ıalysis	
Chr	Gene/nearest gene	Prior GWAS reported smoking phenotype	SNP (allele)	N	icotine depen	ıdence meta-anaı	lyses among e	Nicotine dependence meta-analyses among ever regular smokers	.5	UK Biobank heavy vs never smokers $(N = 48,931)^{24}$	1vy vs never 48,931) ²⁴
				EUR studies (N=28,677)	= 28,677)	AA studies (N = 9925)	V = 9925)	EUR and AA studies ($N = 38,602$)	ies (N = 38,602)		
				β (SE)	۵	β (SE)	۵	β (SE)	۵	β (SE)	Ρa
20q11	20q11 DNMT3B NOL4L	None rs910083 (C) Heavy vs never smoking ²⁴ rs57342388 (CACGG)	rs910083 (C) rs57342388 (CACGG)	0.027 (0.0065) 0.021 (0.0075)	4.2×10^{-5} 0.0059	0.047 (0.012) 0.032 (0.020)	7.3×10^{-5} 0.10	0.032 (0.0057) 0.022 (0.0070)	3.7×10^{-8} 0.0017	0.048 (0.014) 0.094 (0.016)	3.6×10^{-4} 4.7×10^{-9}
20q13	20q13 CHRNA4	None Nicotine dependence ²³ Heavy vs. never smoking ²⁴	rs6062901 ^b (G) rs2273500 ^b (C) rs11697662 ^b (C)	0.043 (0.0088) 0.046 (0.0098)	8.3×10^{-7} 2.0×10^{-6} 9.4×10^{-5}	0.022 (0.010) 0.0085 (0.016) 0.013 (0.011)	0.035 0.59	0.034 (0.0067) 0.036 (0.0083) 0.025 (0.0066)	3.0×10^{-7} 1.6×10^{-5} 1.0×10^{-4}	0.048 (0.018) 0.091 (0.019)	0.0075 1.4×10^{-6} 1.0×10^{-7}
9q34 DBH	рвн	Note	rs56116178 ^c (G) rs111280114 ^c (G) rs3025343 ^c (A)	0.053 (0.010) 0.052 (0.010) 0.043 (0.0099)	4.5×10 ⁻⁷ 6.4×10 ⁻⁷ 1.7×10 ⁻⁵	S V V V) 	0.053 (0.003) 0.052 (0.0098) 0.042 (0.0098)	4.5 × 10 ⁻⁷ 2.5 × 10 ⁻⁵ 2.5 × 10 ⁻⁵	0.097 (0.022) 0.099 (0.022) 0.091 (0.021)	9.3 × 10 ⁻⁶ 6.0 × 10 ⁻⁶ 1.1 × 10 ⁻⁵

EUR studies, = 0.72-0.92 in European/European American; GWAS, genome-wide association study; LD, linkage disequilibrium; SE, standard error; SNP, single-nucleotide polymorphism. The SNP with the smallest nicotine dependence GWAS meta-analysis P is shown, along with nearby SNPs previously reported in other GWAS of smoking phenotypes. Results from the UK Biobank heavy vs never smoking originally implicated in control being applied. PRs2773500 and rs11697662 were originally impli 3-0.84 with rs2273500 and rs11697662 in 1000G EUR and r^2 = 0.11-0.51 1%. ^aP-values correspond to genomic allele frequency $\,<$ 1%. $^{
m a}P_{
m -}$ in 1000G EUR). Rs6062901 not available due to minor LD ($r^2 = 0.64$ and D' = 0.95 i rs111280114 had Abbreviations: AA, African American; EUR, ΝĄ GWAS are also presented. where they exist 000G AFR.

predictor of cessation success among the FTND items), it is correlated with heaviness of smoking as captured by CPD.⁷

DNMT3B SNP association with heavy smoking in an independent study

Using heavy vs never smoking GWAS results from the UK Biobank (N=24,457 heavy and 24,474 never smokers), 24 we found that rs910083-C is associated at $P=3.6\times10^{-4}$ and OR (95% CI) = 1.05 (1.02–1.08) for heavier smoking. Although effect sizes were small, its associations with risks of nicotine dependence and heavy smoking were consistent (Table 1).

DNMT3B SNP associations with DNAm and RNAexp

Using a genome-wide meQTL study of 166 fetal brain samples, 36 we found that rs910083-C associates with higher DNAm levels of the probe cg13636640, located 252 base pairs upstream of the DNMT3B gene (β =0.082, P=2.3 \times 10 $^{-26}$). This probe was the only one observed in the region with significant mQTL variants (Supplementary Figure 4). There were a total of 140 SNPs associated with DNAm of this probe (smallest P=1.8 \times 10 $^{-45}$), and we observed associations of these SNPs with nicotine dependence with meta-analysis P-values ranging from 3.66×10^{-8} for rs910083 to 0.0051 (Supplementary Table 4). To our knowledge, there are no similar data to assess cis-meQTL effects in brain tissues from adults with no psychiatric disease. However, we assessed the cis-acting effect of rs910083 on RNAexp across several adult brain tissues using two independent data sets, GTEx 37 followed by Brain eQTL Almanac. 38

Across the 13 brain tissues in GTEx, we observed the highest DNMT3B gene expression levels in cerebellum (median log₁₀ RPKM = 0.304) and cerebellar hemisphere (median log₁₀ RPKM = 0.298), compared to median log_{10} RPKM ≤ -0.325 for all other brain tissues (Supplementary Figure 5). Moreover, across the brain tissues, rs910083 was most significantly associated with DNMT3B gene expression levels in cerebellum ($P = 3.0 \times 10^{-6}$) and cerebellar hemisphere ($P = 7.0 \times 10^{-7}$), with the C allele associated with higher DNMT3B expression (Figure 3 and Supplementary Table 5). We replicated this pattern in the Brain eQTL Almanac, where DNMT3B mRNA transcript expression levels were highest in cerebellar cortex (Supplementary Figure 6)—the outer layer of the cerebellum that comprises most of its volume. Consistent with GTEx, rs910083-C was associated with increased DNMT3B mRNA transcript expression levels specifically in cerebellar cortex (P = 0.028).

Beyond the brain tissues in GTEx, rs910083 is associated with RNAexp of other genes within 1 MB (Supplementary Table 6): MAPRE1 (smallest $P=7.3\times10^{-17}$ in sun exposed skin), COMMD7 (smallest $P=1.0\times10^{-6}$ in colon) and BPI fold containing family B, member 2 (BPIFB4, smallest $P=5.4\times10^{-5}$ in artery).

DNMT3B SNP associations with lung cancer

We assessed rs910083 for association with lung cancer using a GWAS meta-analysis of EUR studies from TRICL-ILCCO³⁹ (N=27,349 cases and 54,472 controls, Supplementary Table 7). Rs910083-C is significantly associated with increased risk of squamous cell carcinoma (N=6937 cases and 53,649 controls, meta-analysis P=0.0095 and OR [95% CI]=1.05 [1.01–1.09]), consistent with the increased nicotine dependence risk. Rs910083 is not associated with adenocarcinoma.

We evaluated the effect of adjusting for smoking on the rs910083 association with squamous cell carcinoma in the studies with smoking data readily available. We found that the level of statistical significance and magnitude of association were both weakened with adjustment for ever/never smoking and packyears (Supplementary Table 8), suggesting that the rs910083

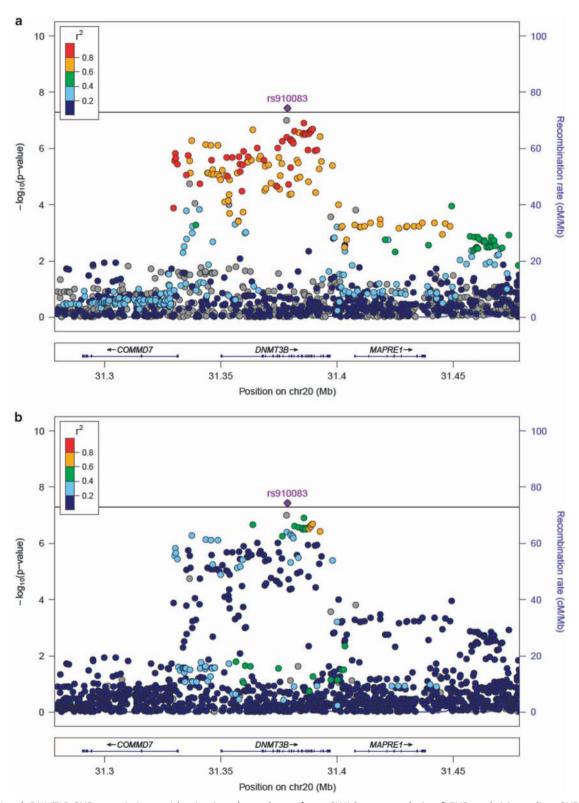


Figure 2. Novel *DNMT3B* SNP associations with nicotine dependence from GWAS meta-analysis of EUR and AA studies. SNP and indel associations are shown across *DNMT3B* and its 100 kb flanking region (NCBI build 37 positions presented). r^2 values between the top SNP rs910083 and all other SNPs are shown in reference to 1000 Genomes panels: (a) EUR and (b) AFR. Indels with missing r^2 values are indicated in gray. The *P*-value threshold of 5×10^{-8} is marked by the solid black line. AA, African American; EUR, European/European American; GWAS, genome-wide association study; SNP, single-nucleotide polymorphism.

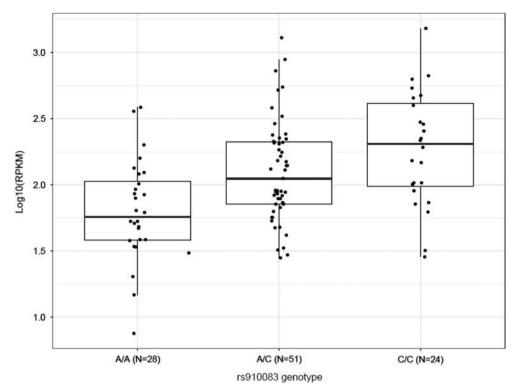


Figure 3. Normalized *DNMT3B* gene expression levels as a function of rs910083 genotype in cerebellum from the Genotype-Tissue Expression (GTEx) project. The box lines mark the first quartile, median and third quartile; and the whiskers are marked by the highest and lowest data points within the 1.5 × inter-quartile range (third–first quartile) to show outliers that fall outside of these boundaries.

association with squamous cell carcinoma is mediated by smoking.

DISCUSSION

This largest-ever GWAS meta-analysis for nicotine dependence, and the largest-ever cross-ancestry GWAS meta-analysis for any smoking phenotype, identified rs910083 as a novel SNP that regulates DNMT3B in human brain and contributes to risks of nicotine dependence and heavy smoking. Rs910083 was discovered via meta-analysis of two ancestry groups; the association signal includes SNPs in high LD with rs910083 across the COMMD7, DNMT3B and MAPRE1 genes in EUR ancestry, but LD is localized to the DNMT3B gene in AA ancestry. Moreover, rs910083 was implicated as a cis-acting QTL SNP that influences DNMT3B DNAm in fetal brain and DNMT3B RNAexp in adult cerebellum, with the C allele being associated with higher DNAm and RNAexp levels. While this pattern might contrast the traditional view of higher DNAm being correlated with lower RNAexp, the observed effects reflect temporal and spatial differences, thus limiting our ability to draw direct correlations. Nonetheless, genome-wide QTL comparisons in human brain have shown that almost half of SNPs that act as both an eQTL and meQTL show the same direction of association for DNAm and RNAexp, 41 as we observed here for *DNMT3B* and before for *CHRNA5*.42 The previously established nicotinic acetylcholine receptor genes also harbor nicotine dependence-associated SNPs with important consequences for gene regulation, including noncoding SNPs that correlate with DNAm, 42 splicing, 23 and/or RNAexp43,44 in brain tissues that are frequently studied for nicotine and other SUDs because of their role in primary reward pathways and executive function, such as prefrontal cortex. 45 This newly identified *DNMT3B* SNP association highlights changes in DNAm in fetal brain and with RNAexp specifically in cerebellum, a part of the brain that has often been overlooked despite some indications for its involvement in the neurobiology of addiction. $^{46-48}$

DNMT3B encodes a DNA methyltransferase prominently involved in de novo DNAm that establishes patterns early in development; it may also contribute to maintenance DNAm. 49 Although initially reported for maintaining DNAm at imprinted loci,⁵⁰ later evidence showed that other members of the DNMT family (DNMT3A and DNMT3L) are required for imprinting, while DNMT3B may not play an essential role.51 Mouse models have shown that complete loss of DNMT3B function is embryonically lethal.⁵⁰ However, recessive inheritance of rare mutations that render DNMT3B partially functional are known to cause Immunodeficiency, Centromeric instability and Facial dysmorphism syndrome, which manifests with growth and neurodevelopmental abnormalities. 49 Because of the critical role that DNMT3B plays in establishing methylation, altered expression has been associated with Immunodeficiency, Centromeric instability and Facial dysmorphism syndrome at >700 genes involved in brain development and other processes.⁵² These and other genes that are regulated by DNMT3B methylation represent candidate genes that may directly contribute to nicotine dependence susceptibility.

DNMT3B has not previously been connected with the biology underlying the risk of nicotine dependence or any other SUD. However, there is evidence that *in vitro* cigarette smoke exposure leads to increased DNMT3B expression in human respiratory epithelial cells, and DNMT3B overexpression results in downstream hypermethylation that has been widely implicated in lung cancer.⁵³ Given this indication, DNMT inhibition has been an active area of research for cancer treatment; two inhibitor agents are currently approved by the US Food and Drug Administration (decitabine, which shows high affinity for DNMT3A/3B over DNMT1,⁵⁴ and azacytidine), and at least one other promising agent (zebularine) awaits clinical trial testing.⁵⁵ With the DNMT3B variant discovery for nicotine dependence, DNMT3B inhibition

may merit future study for smoking cessation treatment. Rs910083 resides in an active promoter marked by H3K9ac in several brain regions examined in the Roadmap Epigenomics Project, ⁵⁶ and it associates with *DNMT3B* RNAexp specifically in the cerebellum. This finding does not negate the importance of other brain regions known to be involved in SUDs but brings up the possibility of altered gene regulation in the cerebellum contributing to the complex neurobiological pathway leading to dependence. A main function of the cerebellum is motor coordination, but it is also involved in non-motor functions relevant to SUD, including reward. ^{46–48} The cerebellum responds to acute and long-term exposures to nicotine ^{57–62} and other substances, ⁴⁷ and it makes functional connections with the prefrontal cortex and other brain tissues that are widely recognized for their involvement in SUDs. ⁴⁷

Our discovery of the rs910083 association with nicotine dependence was made possible by assembling the largest possible sample size of FTND studies comprised of ever-smokers to maximize statistical power. No large FTND studies were left for replication, but because CPD is a central component of both the FTND and the heavy vs never smoking definition used in the UK Biobank, we tested for independent confirmation of the novel association with heavy vs never smoking and found that rs910083-C also conferred risk in the UK Biobank.

Larger GWAS meta-analyses have been reported for other smoking phenotypes such as ever vs never smoking, but these studies were comprised only of EUR participants and based on HapMap imputation. 16–18 Rs910083, a 1000G-imputed SNP, was not captured in these studies, but 9 of the 18 DNMT3B SNPs associated with nicotine dependence at meta-analysis $P < 5 \times 10^{-7}$ in the present study were HapMap-imputed (Supplementary Table 9); these SNPs were in strong LD with rs910083 among EURs ($r^2 = 0.78-0.99$ in 1000G EUR) but weaker LD among AAs $(r^2 = 0.29-0.76 \text{ in } 1000\text{G AFR})$. Using results from the largest GWAS meta-analysis of CPD (Tobacco and Genetics (TAG) consortium, N = 38.181 EUR ever smokers independent of the ones included here), we found that the nine HapMap-imputed DNMT3B SNPs were associated with CPD at P-values ranging from 0.027 to 0.059 and a consistent direction of association with nicotine dependence; in comparison, P-values for ever vs never smoking (N = 74,035 in the TAG consortium) ranged from 0.049 to 0.34 (Supplementary Table 9). We caution that the best *DNTM3B* signal in the TAG consortium was observed for CPD at only nominal significance (smallest P = 0.027), despite having a nearly equivalent sample size as our study. However, our study yielding more statistically significant DNMT3B SNP associations with nicotine dependence (smallest $P = 3.7 \times 10^{-8}$) is likely due to a combination of factors, including (1) reliance on FTND, a multidimensional phenotype that encompasses CPD and other important features of smoking behavior including TTFC that drove the rs910083 association, and (2) 1000G imputation which has been shown to strengthen association signals for some loci due to the finer mapping available. 63,64 We have similarly observed more statistically significant associations with nicotine dependence and stronger effect sizes, compared to CPD, in prior studies of CHRNB3¹⁹ and CHRNA4.²³

Until now, the only common *DNMT3B* variant implicated by GWAS was identified in a study of inflammatory bowel disease (intronic rs4911259). 65 This SNP was associated with nicotine dependence in our meta-analysis ($P\!=\!3.5\times10^{-6}$) and is in LD with our top SNP rs910083 ($r^2\!=\!0.76$ and $D'\!=\!1$ in 1000G EUR, $r^2\!=\!0.16$ and $D'\!=\!1$ in 1000G AFR). Cigarette smoking is the environmental factor most consistently associated with inflammatory bowel disease, 66 and these shared SNP association signals suggest that *DNMT3B* SNPs may exert pleiotropic effects. Alternatively, it is possible that smoking mediates the *DNMT3B* SNP association with inflammatory bowel disease, but the inflammatory bowel disease GWAS did not include adjustment for smoking. 65

Beyond finding *DNMT3B* and (as expected) *CHRNA5-CHRNA3-CHRNB4*, our GWAS meta-analysis resulted in $P < 5 \times 10^{-7}$ for two other loci previously implicated in smoking, *CHRNA4* on chromosome $20q13^{23,24}$ and dopamine β -hydroxylase (*DBH*) on chromosome $9q34.^{17,24}$ Our prior nicotine dependence GWAS meta-analysis of EUR studies identified the *CHRNA4* splice site SNP rs2773500.²³ Rare *CHRNA4* variants have also been found to associate with nicotine dependence.^{67,68} Our study supported common *CHRNA4* SNP associations among EUR samples, but no association was detected for these SNPs among AA studies (Table 1).

DBH is a strong functional candidate for influencing nicotine dependence. The dopaminergic system lies at the core of the brain's reward pathway, and the DBH enzyme converts dopamine into norepinephrine. An upstream DBH SNP (rs3025343) was identified in a GWAS of smoking cessation (current vs former smokers)^{16–18} and later independently replicated.^{69,70} Consistent with rs3025343-A being associated with reduced success of quitting smoking, its phenotypic profile has been expanded to include associations with: (1) heavier smoking (N=48,931 in the UK Biobank, P=1.2×10⁻⁵)²⁴; (2) higher FTND scores (N=1430 EUR participants, P=0.023)⁷¹; and (3) higher nicotine dependence risk in our EUR studies (N = 28,677, meta-analysis $P = 1.7 \times 10^{-5}$). Smaller P-values were found for other 1000G-imputed upstream DBH SNPs in the UK Biobank and our study (rs111280114 and rs56116178, respectively: Table 1); the minor alleles of these SNPs were similarly associated with increased risks among EUR studies (Supplementary Figure 7 for rs56116178). Because these DBH SNPs all occur at < 1% frequency among AAs, studying DBH variation on nicotine dependence risk in this ancestry group will require larger sample sizes or an alternative study design.

Nicotine dependence-associated variants in CHRNA5-CHRNA3-CHRNB4^{44,72} and CHRNA4^{23,67} have been previously shown to associate with lung cancer and other smoking-related diseases. Our study shows that the nicotine dependence-associated SNPs in DNMT3B and DBH are also associated with lung cancer (Supplementary Table 7). These findings may reflect the SNPs acting indirectly on lung through their influence on smoking (Supplementary Table 8). Alternatively, because DBH is expressed in the lung⁷³ and *DNMT3B* overexpression has been shown in lung cancer, we cannot exclude the possibility that either of these SNPs act directly to promote lung cancer through an unknown mechanism.⁵³ The *DNMT3B* and *DBH* SNPs were both associated with squamous cell lung carcinoma. This histological subtype has a strong association with smoking and occurs infrequently in never-smokers. In contrast, neither SNP is associated with adenocarcinoma, a subtype that has a weaker association with smoking⁷⁴ and an increasing prevalence over time among neversmokers.⁷⁵ Histology-specific associations are not uncommon for lung cancer genetic loci.39

Our findings expand the known genetic architecture of nicotine dependence, by showing that the *DNMT3B* SNP rs910083 increases the likelihood of developing nicotine dependence as observed across two different ancestries, smoking heavily, and consequently incurring a heightened risk of lung cancer.^{53,55} The convergence of prior and current findings indicate that the complex neurobiology underlying nicotine dependence involves several sequence variants with functional and regulatory effects across distinct brain tissues.

CONFLICT OF INTEREST

Dr Bierut and the spouse of Dr Saccone are listed as inventors on U.S. Patent 8,080,371, 'Markers for Addiction' covering the use of certain SNPs in determining the diagnosis, prognosis and treatment of addiction. Authors listed with the affiliation deCODE Genetics/AMGEN are employees of deCODE genetics/AMGEN. Although unrelated to this research, Dr Kranzler has been a consultant or advisory board member for Lundbeck and Indivior and is a member of the American Society of

Clinical Psychopharmacology's Alcohol Clinical Trials Initiative, which was supported in the last 3 years by AbbVie, Alkermes, Ethypharm, Indivior, Lilly, Lundbeck, Otsuka, Pfizer, Arbor and Amygdala Neurosciences. Dr Kaprio has consulted for Pfizer in 2012–2014 on nicotine dependence. The remaining authors declare no conflict of interest

ACKNOWLEDGMENTS

We thank the many study participants. We also thank Michael E. Hall for reviewing the manuscript. This work was supported by the National Institute on Drug Abuse grant numbers R01 DA035825, R01 DA036583 and R01 DA042090. Acknowledgments for the nicotine dependence studies are included in the Supplementary Information. Funding for lung cancer studies was provided by the National Cancer Institute grant number U19 CA148127.

REFERENCES

- 1 World Health Organization. WHO report on the global tobacco epidemic, 2011: Warning about the dangers of tobacco. World Health Organization: Geneva, Switzerland. 2011.
- 2 U.S. Department of Health and Human Services. The health consequences of smoking—50 years of progress: a Report of the Surgeon General. U.S. Department of Health and Human Services: Atlanta, GA, 2014.
- 3 Centers for Disease Control and Prevention (CDC). Vital signs: current cigarette smoking among adults aged > 18 years—United States, 2005–2010. MMWR Morb Mortal Wklv Rep. 2011: 60: 1207–1212.
- 4 Centers for Disease Control and Prevention (CDC). Quitting smoking among adults—United States, 2001–2010. MMWR Morb Mortal Wkly Rep 2011; 60: 1513–1519.
- 5 Shiffman S, Tindle H, Li X, Scholl S, Dunbar M, Mitchell-Miland C. Characteristics and smoking patterns of intermittent smokers. *Exp Clin Psychopharmacol* 2012; 20: 264–277.
- 6 Breslau N, Johnson EO. Predicting smoking cessation and major depression in nicotine-dependent smokers. *Am J Public Health* 2000; **90**: 1122–1127.
- 7 Baker TB, Piper ME, McCarthy DE, Bolt DM, Smith SS, Kim SY et al. Time to first cigarette in the morning as an index of ability to quit smoking: implications for nicotine dependence. Nicotine Tob Res 2007; 9(Suppl 4): S555–S570.
- 8 Sullivan PF, Kendler KS. The genetic epidemiology of smoking. *Nicotine Tob Res* 1999; 1: 551–557.
- 9 Vink JM, Willemsen G, Boomsma Dl. Heritability of smoking initiation and nicotine dependence. Behav Genet 2005; 35: 397–406.
- 10 Baker TB, Piper ME, Schlam TR, Cook JW, Smith SS, Loh WY et al. Are tobacco dependence and withdrawal related amongst heavy smokers? Relevance to conceptualizations of dependence. J Abnorm Psychol 2012; 121: 909–921.
- 11 Zelman DC, Brandon TH, Jorenby DE, Baker TB. Measures of affect and nicotine dependence predict differential response to smoking cessation treatments. J Consult Clin Psychol 1992; 60: 943–952.
- 12 Gu F, Wacholder S, Kovalchik S, Panagiotou OA, Reyes-Guzman C, Freedman ND et al. Time to smoke first morning cigarette and lung cancer in a case-control study. J Natl Cancer Inst 2014; 106: dju118.
- 13 Guertin KA, Gu F, Wacholder S, Freedman ND, Panagiotou OA, Reyes-Guzman C et al. Time to first morning cigarette and risk of chronic obstructive pulmonary disease: smokers in the PLCO Cancer Screening Trial. PLoS One 2015; 10: e0125973.
- 14 Bierut LJ, Madden PA, Breslau N, Johnson EO, Hatsukami D, Pomerleau OF et al. Novel genes identified in a high-density genome wide association study for nicotine dependence. Hum Mol Genet 2007; 16: 24–35.
- 15 Thorgeirsson TE, Geller F, Sulem P, Rafnar T, Wiste A, Magnusson KP et al. A variant associated with nicotine dependence, lung cancer and peripheral arterial disease. Nature 2008; 452: 638–642.
- 16 Liu JZ, Tozzi F, Waterworth DM, Pillai SG, Muglia P, Middleton L et al. Meta-analysis and imputation refines the association of 15q25 with smoking quantity. Nat Genet 2010; 42: 436–440.
- 17 Tobacco and Genetics Consortium. Genome-wide meta-analyses identify multiple loci associated with smoking behavior. Nat Genet 2010; 42: 441–447.
- 18 Thorgeirsson TE, Gudbjartsson DF, Surakka I, Vink JM, Amin N, Geller F et al. Sequence variants at CHRNB3-CHRNA6 and CYP2A6 affect smoking behavior. Nat Genet 2010: 42: 448–453.
- 19 Rice JP, Hartz SM, Agrawal A, Almasy L, Bennett S, Breslau N et al. CHRNB3 is more strongly associated with Fagerstrom Test for Cigarette Dependence-based nicotine dependence than cigarettes per day: phenotype definition changes genomewide association studies results. Addiction 2012; 107: 2019–2028.

- 20 David SP, Hamidovic A, Chen GK, Bergen AW, Wessel J, Kasberger JL et al. Genome-wide meta-analyses of smoking behaviors in African Americans. *Transl Psychiatry* 2012; 2: e119.
- 21 Loukola A, Wedenoja J, Keskitalo-Vuokko K, Broms U, Korhonen T, Ripatti S et al. Genome-wide association study on detailed profiles of smoking behavior and nicotine dependence in a twin sample. Mol Psychiatry 2014; 19: 615–624.
- 22 Gelernter J, Kranzler HR, Sherva R, Almasy L, Herman AI, Koesterer R et al. Genome-wide association study of nicotine dependence in American populations: identification of novel risk loci in both African-Americans and European-Americans. Biol Psychiatry 2015; 77: 493–503.
- 23 Hancock DB, Reginsson GW, Gaddis NC, Chen X, Saccone NL, Lutz SM et al. Genome-wide meta-analysis reveals common splice site acceptor variant in CHRNA4 associated with nicotine dependence. Transl Psychiatry 2015; 5: e651.
- 24 Wain LV, Shrine N, Miller S, Jackson VE, Ntalla I, Artigas MS *et al.* Novel insights into the genetics of smoking behaviour, lung function, and chronic obstructive pulmonary disease (UK BiLEVE): a genetic association study in UK Biobank. *Lancet Respir Med* 2015; **3**: 769–781.
- 25 Baker TB, Breslau N, Covey L, Shiffman S. DSM criteria for tobacco use disorder and tobacco withdrawal: a critique and proposed revisions for DSM-5. Addiction 2012; 107: 263–275.
- 26 Heatherton TF, Kozlowski LT, Frecker RC, Fagerstrom KO. The fagerstrom test for nicotine dependence: a revision of the fagerstrom tolerance questionnaire. Br J Addict 1991; 86: 1119–1127.
- 27 Willer CJ, Li Y, Abecasis GR. METAL: fast and efficient meta-analysis of genome-wide association scans. *Bioinformatics* 2010; **26**: 2190–2191.
- 28 Pe'er I, Yelensky R, Altshuler D, Daly MJ. Estimation of the multiple testing burden for genomewide association studies of nearly all common variants. Genet Epidemiol 2008; 32: 381–385.
- 29 Fadista J, Manning AK, Florez JC, Groop L. The (in)famous GWAS P-value threshold revisited and updated for low-frequency variants. Eur J Hum Genet 2016; 24: 1202–1205.
- 30 Kanai M, Tanaka T, Okada Y. Empirical estimation of genome-wide significance thresholds based on the 1000 Genomes Project data set. J Hum Genet 2016; 61: 861–866.
- 31 deVries PS, Sabater-Lleal M, Chasman DI, Trompet S, Ahluwalia TS, Teumer A *et al.*Comparison of HapMap and 1000 genomes reference panels in a large-scale genome-wide association study. *PLoS One* 2017; **12**: e0167742.
- 32 Sveinbjornsson G, Albrechtsen A, Zink F, Gudjonsson SA, Oddson A, Masson G et al. Weighting sequence variants based on their annotation increases power of whole-genome association studies. Nat Genet 2016: 48: 314–317.
- 33 Pruim RJ, Welch RP, Sanna S, Teslovich TM, Chines PS, Gliedt TP et al. LocusZoom: regional visualization of genome-wide association scan results. *Bioinformatics* 2010; **26**: 2336–2337.
- 34 Boyles AL, Harris SF, Rooney AA, Thayer KA. Forest Plot Viewer: a new graphing tool. *Epidemiology* 2011; **22**: 746–747.
- 35 Huedo-Medina TB, Sanchez-Meca J, Marin-Martinez F, Botella J. Assessing heterogeneity in meta-analysis: Q statistic or I2 index? *Psychol Methods* 2006; 11: 193–206.
- 36 Hannon E, Spiers H, Viana J, Pidsley R, Burrage J, Murphy TM et al. Methylation QTLs in the developing brain and their enrichment in schizophrenia risk loci. Nat Neurosci 2016: 19: 48–54.
- 37 GTEx Consortium. Human genomics. The Genotype-Tissue Expression (GTEx) pilot analysis: multitissue gene regulation in humans. Science 2015; 348: 648–660.
- 38 Ramasamy A, Trabzuni D, Guelfi S, Varghese V, Smith C, Walker R *et al.* Genetic variability in the regulation of gene expression in ten regions of the human brain. *Nat Neurosci* 2014; **17**: 1418–1428.
- 39 McKay JD, Hung RJ, Han Y, Zong X, Carreras-Torres R, Christiani DC et al. Large scale genetic analysis identifies novel loci and histological variability in susceptibility to lung cancer. Nat Genet 2017; 49: 1126–1132.
- 40 Fagerstrom KO. Measuring degree of physical dependence to tobacco smoking with reference to individualization of treatment. Addict Behav 1978; 3: 235–241.
- 41 Gibbs JR, van der Brug MP, Hernandez DG, Traynor BJ, Nalls MA, Lai SL *et al.*Abundant quantitative trait loci exist for DNA methylation and gene expression in human brain. *PLoS Genet* 2010; **6**: e1000952.
- 42 Hancock DB, Wang JC, Gaddis NC, Levy JL, Saccone NL, Stitzel JA *et al.* A multiancestry study identifies novel genetic associations with CHRNA5 methylation in human brain and risk of nicotine dependence. *Hum Mol Genet* 2015; **24**: 5940–5944.
- 43 Wang JC, Spiegel N, Bertelsen S, Le N, McKenna N, Budde JP et al. Cis-regulatory variants affect CHRNA5 mRNA expression in populations of African and European ancestry. PLoS One 2013; 8: e80204.
- 44 Wang JC, Cruchaga C, Saccone NL, Bertelsen S, Liu P, Budde JP *et al.* Risk for nicotine dependence and lung cancer is conferred by mRNA expression levels and amino acid change in CHRNA5. *Hum Mol Genet* 2009; **18**: 3125–3135.

- 45 Goldstein RZ, Volkow ND. Dysfunction of the prefrontal cortex in addiction: neuroi-maging findings and clinical implications. Nat Rev Neurosci 2011; 12: 652–669.
- 46 Miquel M, Vazquez-Sanroman D, Carbo-Gas M, Gil-Miravet I, Sanchis-Segura C, Carulli D *et al.* Have we been ignoring the elephant in the room? Seven arguments for considering the cerebellum as part of addiction circuitry. *Neurosci Biobehav Rev* 2016; **60**: 1–11.
- 47 Moulton EA, Elman I, Becerra LR, Goldstein RZ, Borsook D. The cerebellum and addiction: insights gained from neuroimaging research. *Addict Biol* 2014; **19**: 317–331.
- 48 Strick PL, Dum RP, Fiez JA. Cerebellum and nonmotor function. *Annu Rev Neurosci* 2009; **32**: 413–434.
- 49 Weissman J, Naidu S, Bjornsson HT. Abnormalities of the DNA methylation mark and its machinery: an emerging cause of neurologic dysfunction. Semin Neurol 2014; 34: 249–257.
- 50 Okano M, Bell DW, Haber DA, Li E. DNA methyltransferases Dnmt3a and Dnmt3b are essential for de novo methylation and mammalian development. *Cell* 1999; 99: 247–257
- 51 Kaneda M, Okano M, Hata K, Sado T, Tsujimoto N, Li E et al. Essential role for de novo DNA methyltransferase Dnmt3a in paternal and maternal imprinting. Nature 2004: 429: 900–903.
- 52 Jin B, Tao Q, Peng J, Soo HM, Wu W, Ying J *et al.* DNA methyltransferase 3B (DNMT3B) mutations in ICF syndrome lead to altered epigenetic modifications and aberrant expression of genes regulating development, neurogenesis and immune function. *Hum Mol Genet* 2008; **17**: 690–709.
- 53 Teneng I, Tellez CS, Picchi MA, Klinge DM, Yingling CM, Snider AM *et al.* Global identification of genes targeted by DNMT3b for epigenetic silencing in lung cancer. *Oncogene* 2015; **34**: 621–630.
- 54 Oka M, Meacham AM, Hamazaki T, Rodic N, Chang LJ, Terada N. De novo DNA methyltransferases Dnmt3a and Dnmt3b primarily mediate the cytotoxic effect of 5-aza-2'-deoxycytidine. *Oncogene* 2005; **24**: 3091–3099.
- 55 Gnyszka A, Jastrzebski Z, Flis S. DNA methyltransferase inhibitors and their emerging role in epigenetic therapy of cancer. *Anticancer Res* 2013; **33**: 2989–2996.
- 56 Ward LD, Kellis M. HaploReg: a resource for exploring chromatin states, conservation, and regulatory motif alterations within sets of genetically linked variants. *Nucleic Acids Res* 2012; 40: D930–D934.
- 57 Domino EF, Minoshima S, Guthrie S, Ohl L, Ni L, Koeppe RA *et al.* Nicotine effects on regional cerebral blood flow in awake, resting tobacco smokers. *Synapse* 2000; **38**: 313–321
- 58 Zubieta JK, Heitzeg MM, Xu Y, Koeppe RA, Ni L, Guthrie S *et al.* Regional cerebral blood flow responses to smoking in tobacco smokers after overnight abstinence. *Am J Psychiatry* 2005; **162**: 567–577.
- 59 Brody AL, Mandelkern MA, Jarvik ME, Lee GS, Smith EC, Huang JC *et al.* Differences between smokers and nonsmokers in regional gray matter volumes and densities. *Biol Psychiatry* 2004; **55**: 77–84.
- 60 Gallinat J, Meisenzahl E, Jacobsen LK, Kalus P, Bierbrauer J, Kienast T et al. Smoking and structural brain deficits: a volumetric MR investigation. Eur J Neurosci 2006; 24: 1744–1750.

- 61 Kuhn S, Romanowski A, Schilling C, Mobascher A, Warbrick T, Winterer G et al. Brain grey matter deficits in smokers: focus on the cerebellum. Brain Struct Funct 2012: 217: 517–522.
- 62 Sutherland MT, Riedel MC, Flannery JS, Yanes JA, Fox PT, Stein EA et al. Chronic cigarette smoking is linked with structural alterations in brain regions showing acute nicotinic drug-induced functional modulations. Behav Brain Funct 2016; 12: 16.
- 63 Wood AR, Perry JR, Tanaka T, Hernandez DG, Zheng HF, Melzer D et al. Imputation of variants from the 1000 Genomes Project modestly improves known associations and can identify low-frequency variant-phenotype associations undetected by HapMap based imputation. PLoS One 2013; 8: e64343.
- 64 Huang J, Ellinghaus D, Franke A, Howie B, Li Y. 1000 Genomes-based imputation identifies novel and refined associations for the Wellcome Trust Case Control Consortium phase 1 Data. *Eur J Hum Genet* 2012; **20**: 801–805.
- 65 Jostins L, Ripke S, Weersma RK, Duerr RH, McGovern DP, Hui KY et al. Host-microbe interactions have shaped the genetic architecture of inflammatory bowel disease. Nature 2012; 491: 119–124.
- 66 Ananthakrishnan AN. Environmental risk factors for inflammatory bowel disease. Gastroenterol Hepatol (NY) 2013; 9: 367–374.
- 67 Thorgeirsson TE, Steinberg S, Reginsson GW, Bjornsdottir G, Rafnar T, Jonsdottir I et al. A rare missense mutation in CHRNA4 associates with smoking behavior and its consequences. Mol Psychiatry 2016; 21: 594–600.
- 68 Xie P, Kranzler HR, Krauthammer M, Cosgrove KP, Oslin D, Anton RF et al. Rare nonsynonymous variants in alpha-4 nicotinic acetylcholine receptor gene protect against nicotine dependence. Biol Psychiatry 2011; 70: 528–536.
- 69 Siedlinski M, Cho MH, Bakke P, Gulsvik A, Lomas DA, Anderson W et al. Genomewide association study of smoking behaviours in patients with COPD. *Thorax* 2011; **66**: 894–902.
- 70 Hirvonen K, Korhonen T, Salomaa V, Männistö S, Kaprio J. Effect of the DBH polymorphism rs3025343 on smoking cessation in a large population-based sample. *Nic Tob Research* 2017; **19**: 1112–1115.
- 71 Yang J, Wang S, Yang Z, Hodgkinson CA, Iarikova P, Ma JZ et al. The contribution of rare and common variants in 30 genes to risk nicotine dependence. Mol Psychiatry 2015; 20: 1467–1478.
- 72 Amos Cl, Wu X, Broderick P, Gorlov IP, Gu J, Eisen T et al. Genome-wide association scan of tag SNPs identifies a susceptibility locus for lung cancer at 15q25.1. Nat Genet 2008; 40: 616–622.
- 73 Barrie ES, Weinshenker D, Verma A, Pendergrass SA, Lange LA, Ritchie MD et al. Regulatory polymorphisms in human DBH affect peripheral gene expression and sympathetic activity. Circ Res 2014; 115: 1017–1025.
- 74 Lee PN, Forey BA, Coombs KJ. Systematic review with meta-analysis of the epidemiological evidence in the 1900s relating smoking to lung cancer. BMC Cancer 2012; 12: 385.
- 75 Lee PN, Forey BA. Indirectly estimated absolute lung cancer mortality rates by smoking status and histological type based on a systematic review. BMC Cancer 2013: 13: 189.

Supplementary Information accompanies the paper on the Molecular Psychiatry website (http://www.nature.com/mp)