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Genome-wide search for replicable risk gene regions in alcohol and nicotine co-dependence

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Abstract

The present study searched for replicable risk genomic regions for alcohol and nicotine codependence using a genome-wide association strategy. The data contained a total of 3,143 subjects including 818 European-American (EA) cases with alcohol and nicotine co-dependence, 1,396 EA controls, 449 African-American (AA) cases and 480 AA controls. We performed separate genome-wide association analyses in EAs and AAs and a meta-analysis to derive combined p values, and calculated the genome-wide false discovery rate (FDR) for each SNP. Regions with $p < 5 \times 10^{-7}$ together with FDR < 0.05 in the meta-analysis were examined to detect all replicable risk SNPs across EAs, AAs and meta-analysis. These SNPs were followed with a series of functional expression quantitative trait locus (eQTL) analyses. We found a unique genome-wide significant gene region - SH3BP5-NR2C2 - that was enriched with 11 replicable risk SNPs for alcohol and nicotine co-dependence. The distributions of $-\log(p)$ values for all SNP-disease associations within this region were consistent across EAs, AAs, and meta-analysis (0.315 r 0.868; 8.1×10^{-52} p 3.6×10^{-5}). In the meta-analysis, this region was the only association peak throughout chromosome 3 at p<0.0001. All replicable risk markers available for eQTL analysis had nominal *cis*- and *trans*-acting regulatory effects on gene expression. The transcript expression of the genes in this region was regulated partly by several nicotine dependence-related genes and significantly correlated with transcript expression of many alcohol and nicotine dependence-related genes. We concluded that the SH3BP5-NR2C2 region on Chromosome 3 might harbor causal loci for alcohol and nicotine co-dependence.

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Keywords

GWAS; alcohol and nicotine co-dependence

Introduction

Alcohol dependence (AD) and nicotine dependence (ND) frequently co-occur in the same individuals. Alcohol and nicotine co-dependence may represent an independent phenotype. A large number of common risk genetic loci have been reported separately for AD and ND in the dopaminergic, serotoninergic, GABAergic, glutamatergic, cholinergic, opioid, and endocannabinoid systems by candidate gene approach; several genome-wide association studies (GWASs) on AD or ND reported other risk loci (summarized by [Zuo *et al.*, 2011a; Zuo *et al.*, In revision]). Only one study employed GWAS [Lind *et al.*, 2010] to examine alcohol and nicotine co-dependence, and its results have yet to be replicated.

In the present study, we searched for replicable risk gene regions for alcohol and nicotine co-dependence in two distinct American populations using GWAS. In the association analysis, we separated European-Americans (EAs) and African-Americans (AAs) to increase population homogeneity, and controlled for admixture effects. The association findings from the EAs were replicated in the AAs and vice versa. Additionally, we used an independent sample with distinct tissues to detect expression quantitative trait locus (eQTL) signals, as a confirmation of the association findings. Furthermore, we applied a stringent definition of replication (see below). The primary target of investigation in the current study was not the top-ranked SNPs in the discovery sample as previous GWASs, but rather the replicable risk regions that might harbor the population-generalizable and functional variants. This strategy led to the discovery of novel risk loci for alcohol and nicotine co-dependence.

Materials and Methods

Subjects

The sample comprised of a total of 3,143 subjects for gene-disease association analysis, including 818 European-American (EA) cases with alcohol and nicotine co-dependence, 1,396 EA controls, 449 African-American (AA) cases and 480 AA controls. This sample was extracted from SAGE (dbGaP study accession phs000092.v1.p1)[Bierut et al., 2010]. SAGE subjects were recruited from 8 different study sites in 7 states and the District of Columbia; the majority of subjects were recruited in Missouri[Bierut et al., 2010]. All subjects were interviewed using the Semi-Structured Assessment for the Genetics of Alcoholism (SSAGA)[Bucholz et al., 1994]. Affected subjects met lifetime DSM-IV criteria[American Psychiatric Association 1994] both for alcohol dependence and nicotine dependence, and were excluded if they had schizophrenia or other psychotic illnesses. Controls were defined as individuals who had been exposed to alcohol and nicotine in sufficient amounts for a sufficient time, but had never become dependent on or abused alcohol, nicotine or other illicit substances (see Supplemental Table S1). Additionally, controls were also screened to exclude individuals with major axis I disorders, including schizophrenia, mood disorders, and anxiety disorders. More demographic data were available in the Supplemental Table S1 or elsewhere [Bierut et al., 2010; Edenberg et al., 2005; Edenberg et al., 2010; Zuo et al., 2011a; Zuo et al., 2011b]. All subjects gave written informed consent to participating in protocols approved by the relevant institutional review boards and were de-identified in this study.

Genotyping

All samples were genotyped on the Illumina Human 1M beadchip at the Center for Inherited Disease Research (CIDR) at Johns Hopkins University (Baltimore, MD USA). Allele cluster definitions for each marker were determined using Illumina BeadStudio Genotyping Module version 3.1.14 and the combined intensity data from the samples.

Data cleaning

Before statistical analysis, we strictly cleaned the phenotype data and then the genotype data [Zuo *et al.*, 2011a; Zuo *et al.*, In revision]. After cleaning, 805,814 markers in EAs and 895,714 markers in AAs were included for association analysis. The cleaned data had high-quality, as evidenced by the following: (1) The homogeneity of the two samples was very high; that is, EAs and AAs were well differentiated. (2) The observed and expected p-values for the associations fit very well within EAs or AAs (see QQ plots in Supplemental Figure S1). (3) We also computed from these p-values a low genomic inflation factor (GIF) of 1.04 in EAs, 1.02 in AAs and 1.04 in meta-analysis.

Data analytic procedure

We performed genome-wide association analysis separately in EAs and AAs first, and then performed a meta-analysis of EAs and AAs. Genome-wide false discovery rates (FDRs) [Benjamini and Hochberg 1995] were calculated in EAs, AAs and meta-analysis. The replicable risk regions were identified, in which (1) many markers were associated with phenotype across EAs, AAs and meta-analysis, and (2) the distributions of -log(p) values for associations for all markers were consistent across EAs, AAs and meta-analysis; that is, the number of risk markers, the effect directions, the effect sizes and the significance strengths were congruent across three groups. These distributions were compared for similarity using Pearson correlations. Among these replicable risk regions, those with $p<5\times10^{-7}$ and FDR<0.05 in the meta-analysis were selected as "genome-wide significant and replicable regions" (screening step; see correction for multiple testing section below). And then, all SNPs in those significant and replicable regions were examined to identify all replicable risk SNPs (Table 1) (testing step). Also, around these significant and replicable regions, a region spanning 5Mb to whole chromosome was carefully studied, to know the width of the selected risk region where the putative causal loci may be located.

In addition to the replication design described above, we also performed functional eQTL analysis on the replicable risk SNPs, which included (1) *cis*-eQTL analysis on exon-/ transcript-level expression changes in peripheral blood mononucleated cells (PBMCs) (n=80) and cortical brain tissues (n=93), (2) transcriptome-wide *trans*-acting eQTL analysis on transcript expression, (3) genome-wide *trans*-acting eQTL analysis on the transcript expression of the replicable risk genes, and (4) transcriptome-wide expression correlation analysis [Heinzen *et al.*, 2008; Zuo *et al.*, 2011a; Zuo *et al.*, 2011b]. Additionally, RNA secondary structure analysis [Zuker 2003; Zuo *et al.*, 2011a; Zuo *et al.*, 2011b] and a series of bioinformatic analysis (Supplemental Table S2) of the replicable risk markers were also performed.

Association analysis

a. Genome-wide association tests in EAs and AAs: The allele frequencies of the cleaned SNPs were compared between cases and controls using genome-wide logistic regression analysis implemented in the program Plink[Purcell *et al.*, 2007], separately in EAs and AAs. Diagnosis served as the dependent variable, alleles or genotype served as the independent variables, and ancestry proportions (to control for admixture effects), sex, and age served as covariates (Supplementary Figure

S2). Ancestry proportions of each individual were estimated from 3,172 completely independent ancestry-informative genetic markers [Zuo *et al.*, 2011a]. The replicable regions between EAs and AAs were identified.

- **b.** Meta-analysis of EAs and AAs: A meta-analysis of both populations was conducted to derive the effect directions of minor alleles, the Stouffer' Z scores, and the combined p values. The associations with opposite effects between EAs and AAs could become weaker in meta-analysis, and those with same direction of effects could become stronger in meta-analysis. Thus, to define replicable associations, we did not only require them to be positive in both EAs and AAs, but also required them to be stronger in meta-analysis, which increased the possibility that the replicable risk variants were causal. The associations in the genome-wide significant and replicable regions that were replicated across EAs, AAs and meta-analysis are shown in Table 1.
- c. *Cis*-eQTL analysis: To examine relationships between genetic variants and local gene expression levels, we performed *cis*-eQTL analysis in the PBMC and brain tissues described above. Each of these associations was analyzed using a linear regression model by correcting for age, sex, source of tissues, and principle component scores. The rare homozygotes have been merged into the heterozygotes in the association analyses. P-values less than 0.05 were listed in Table 1 and the relationships between the exon expression levels (Y-axis) and the genotypes (X-axis) of the replicable risk markers were plotted in Figure S3. The genotype frequency distributions, the strength of associations (i.e., beta values from linear regression analysis) and the exon probe ID numbers are also shown on the plots and in their legends.

Correction for multiple testing on association and cis-eQTL analysis and power analysis

To mitigate false positive rates, genome-wide associations in the screening stage need to be corrected for multiple testing. Apparently, Bonferroni correction ($\alpha = 5 \times 10^{-8}$) is overly conservative because it treats all of the one million markers in the genome as independent (which is impossible). Alternatively, a WTCCC-defined a $(=5 \times 10^{-7})$ might be more appropriate to the present study[The Wellcome Trust Case Control Consortium 2007]. As a complementary approach, we also corrected the findings in screening stage by genome-wide false discovery rate (FDR)[Benjamini and Hochberg 1995], replicated the findings, and confirmed them by functional studies. Only when a region containing at least one association in meta-analysis that survived WTCCC-defined genome-wide correction $(p<5\times10^{-7})$ together with FDR<0.05, and was replicable across EAs, AAs, and metaanalysis and confirmed by functional studies, should it be taken as a "significant" region, which was conservative enough for statistical significance. In the testing step, (1) two independent samples were used to replicate each other, which significantly reduced the chance of false positive findings (i.e., false discovery rate). (2) We aimed to detect replicable regions, not individual markers. Thus, more than one risk marker were detected in the risk regions, which reduced the chance of false positive associations too. (3) Functional analysis in distinct tissues as confirmation of association finding further reduced the chance of false positive findings (including co-localization of association signals and eQTL signals randomly), although using different independent samples in one study might increase the false negative rates due to sample heterogeneity. (4) -log(P) value distributions across EAs, AAs and meta-analysis were compared using Pearson correlation analysis. The consistency between them would significantly reduce the chance of false positive findings. Therefore, in the testing step, when an association was replicable across EAs, AAs and meta-analysis, a could be set at 0.05 (except for the exon-level *cis*-eQTL findings that needed to be corrected for the number of exons and the types of tissues). Accordingly, the power of the discovery

 $(\alpha=5\times10^{-7})$ and replication $(\alpha=0.05)$ samples to detect the significant genetic effects was analyzed using the power analysis package in R.

Results

There were a total of 29 SNPs in 15 genes in EAs, 9 SNPs in 9 genes in AAs and 22 SNPs in 16 genes in meta-analysis that were marginally to significantly $(p<10^{-5})$ associated with alcohol and nicotine co-dependence (data available on request). The p values for the 10 topranked SNPs in EAs, AAs, and meta-analysis, respectively, are listed in Supplementary Table S3, and the p values in meta-analysis for the 10 top-ranked replicable SNPs are listed in Supplementary Table S4. After correction, 4 SNPs in EAs including rs7445832, rs4700575 and rs2169520 in *IPO11-HTR1A* region (7.0×10⁻⁹ p 3.0×10⁻⁷ and 6.3×10^{-4} FDR 0.031) and rs17427389 in *PLEKHG1* (p= 4.3×10^{-7} and FDR=0.019), rs4610908 in FAM47B in AAs (p=3.2×10⁻⁷ and FDR=0.032), and 2 SNPs in meta-analysis including rs9636470 in *PLGLB2* (p=3.1×10⁻⁸ and FDR=0.003) and rs1318937 in *SH3BP5* $(p=4.1\times10^{-7} \text{ and FDR}=0.041)$ remained significant $(p<5\times10^{-7} \text{ together with FDR}<0.05)$. Among these significant SNPs, only rs9636470 in *PLGLB2* ($p=2.4\times10^{-6}$ in EAs and 0.004 in AAs) and rs1318937 in SH3BP5 ($p=2.5\times10^{-5}$ in EAs and 0.005 in AAs) were replicable between two populations. However, rs9636470 in PLGLB2 was the only replicable significant SNP in a 3Mb-wide region around this gene, and thus, this significant association could occur by chance. In contrast, the region around SH3BP5 was enriched with replicable risk SNPs and thus was the focus of interest in the present study.

Throughout the whole chromosome 3, *SH3BP5-NR2C2* region was the only one that had gene-disease associations with $p<10^{-4}$ in meta-analysis. In EAs, within the 8Mb-range around, *SH3BP5-NR2C2* region was the only association peak with $p<10^{-4}$.

SH3BP5-NR2C2 region contains five known genes including SH3BP5, CAPN7, ZFYVE20, MRPS25 and NR2C2, all of which except MRPS25 were enriched with replicable risk SNPs. Thirty, 34 and 35 SNPs in this region were nominally associated with alcohol and nicotine co-dependence in EAs $(2.5 \times 10^{-5} \text{ p} 0.038)$, AAs $(9.3 \times 10^{-4} \text{ p} 0.046)$ and meta-analysis $(4.1 \times 10^{-7} \text{ p} 0.049)$, respectively. Among them, 11 SNPs were replicable across EAs and AAs and became more significant in meta-analysis $(4.1 \times 10^{-7} \text{ p} 4.1 \times 10^{-3})$ (Table 1). Effects of all of these 11 SNPs were in the same direction between two populations. These 11 SNPs were located in two haplotype blocks, i.e., ZFYVE20-NR2C2 (block 1) and SH3BP6-CAPN7 (block 2) (Figure 1d). Minor alleles of all replicable SNPs in block 2 increased risk for disease (OR>1 in both EAs and AAs, and Z score > 0 in meta-analysis), but minor alleles in block 1 protected against disease (OR<1 in both EAs and AAs, and Z score < 0 in meta-analysis). The -log(p) values for all available SNPs across SH3BP5-NR2C2 region were plotted in Figure 1. The distributions of -log(p) values were consistent across EAs, AAs, and meta-analysis (0.315 r 0.868; $8.1 \times 10^{-52} \text{ p} 3.6 \times 10^{-5}$; Table 2).

The LD structure (Figure 1d), the gene effect directions and the gene effect sizes shown in Table 1 indicated that *SH3BP5-NR2C2* region could be represented by two independent risk SNPs from two LD blocks. After regressing out the effects of the two top-ranked SNPs in each block (i.e., rs17040623 at *NR2C2* and rs1318937 at *SH3BP5*) by conditioning on them in the regression analysis, no other SNPs remained significantly associated with the phenotype (all p > 0.05; data not shown).

Our samples had a high power in detecting risk markers. For example, given that the risk allele of the most significant marker, i.e., rs7445832 (see Table S3), in the EA discovery sample had a frequency of 0.2892 in cases (n=818) and 0.2111 in controls (n=1396), our EA sample had a power of 90.1% to detect any risk marker that had a similar effect size to

rs7445832 (α =5×10⁻⁷). Given that the risk allele of the most significant replicable marker, i.e., rs17040623 (see Table 1), in the AA replication sample had a frequency of 0.0.059 in cases (n=449) and 0.096 in controls (n=480), our AA sample had a power of 86.4% to detect any replicable risk marker that had a similar effect size to rs17040623 (α =0.05).

Among those 11 replicable risk markers, four SNPs were available for eQTL analysis, including rs1318937 and rs8225 in *SH3BP5*, rs735659 in *CAPN7* and rs9868848 in *ZFYVE20*. eQTL analysis showed that all of them had nominal *cis*-acting regulatory effects on exon-level expression of local genes in Brain tissue and/or peripheral blood mononucleated cells (PBMCs) (0.004 p 0.041; Table 1 and Figure S3) and nominally regulated transcript expression of many genes across transcriptome $(3.3 \times 10^{-5} \text{ p } 0.05; \text{ data not shown})$. After Bonferroni correction, none of these *cis*- and *trans*-acting regulatory effects remained significant.

Genome-wide *trans*-eQTL analysis showed that transcript expression of genes in *SH3BP5*-*NR2C2* region was nominally regulated by multiple genes across the genome (data not shown). After Bonferroni correction (α =4.4×10⁻⁸), only rs7667919 in *PET112L* (Chromosome 4) showed significant regulatory effect on *CAPN7* transcript expression in PBMC (p=7.8×10⁻⁹). Furthermore, 19 SNPs in 17 genes in brain and 5 SNPs in 5 genes in PBMC had nominal replicable *trans*-acting regulatory effects on all four genes in *SH3BP5*-*NR2C2* region (5.0×10⁻⁵ p 4.5×10⁻³; data not shown).

Transcriptome-wide expression correlation analysis showed that expression of NR2C2, ZFYVE20, CAPN7 and SH3BP5 transcripts was significantly correlated with each other both in Brain and PBMC. Their expression was also significantly correlated with many alcohol or nicotine dependence-related genes (although some of these genes have not been widely replicated so far; data not shown). These genes were from the dopaminergic (DRD1, DRD2, NCAM1, TTC12, DRD3, DRD4, SLC6A3 and TH)[Batel et al., 2008; Stapleton et al., 2007], serotoninergic (HTR1B, HTR2A, HTR3B and SLC6A4)[Hasegawa et al., 2002], cholinergic (CHRM1, CHRNA1, CHRNA3, CHRNA7, CHRNB2, CHRND and CHRNG) [Lou et al., 2006; Saccone et al., 2009], GABAergic (GABARAP, GABBR1, GABRA1, GABRA2, GABRA4, GABRA6, GABRB1, GABRB2, GABRB3, GABRG1, GABRG2 and GABRG3) [Chang et al., 2002], glutamatergic (GAD1, GRIK2, GRIK3, GRIN2A, GRIN2B, GRIN2C, GRM5 and GRM7) [Edenberg et al., 2010], histaminergic (HNMT) [Oroszi et al., 2005], endocannabinoid (CNR1) [Zuo et al., 2007], opoid (OPRD1, OPRM1 and POMC) [Zhang et al., 2008], alcohol metabolic (ADH5 and ALDH3A2) [Luo et al., 2006; Uhl et al., 2008] and neuropeptide (NPY1R, NPY2R and NPY5R)[Wetherill et al., 2008] systems ($\alpha = 4.2 \times 10^{-7}$).

The main findings from the RNA secondary structure analysis included that (1) rs9868848 in *ZFYVE20* significantly altered the RNA secondary structure; and (2) rs1318937 in *SH3BP5*, rs28445844 in *NR2C2* and rs2306853 in *ZFYVE20* slightly altered the RNA secondary structures (data not shown).

Discussion

Using a replication approach, we identified a unique genome-wide significant risk region - *SH3BP5-NR2C2* - for alcohol and nicotine co-dependence. This region was enriched with replicable risk SNPs in two genetically distinct populations. Additionally, the effect directions and significance strengths of all available SNPs across this whole region matched between two populations; that is, the distributions of -log(p) values for these markers were consistent between two populations, and the associations became more significant in meta-

Multiple lines of evidence support this conclusion. First, the *SH3BP5-NR2C2* region was the only genome-wide significant (both $p<5\times10^{-7}$ and FDR<0.05) as well as replicable region across whole genome. Second, this region was the only association peak throughout Chromosome 3 at p<0.0001 in meta-analysis (in which the association signals were majorly attributable to the EA sample). It is thus highly likely that the putative causal loci for alcohol and nicotine co-dependence were located within this region. Third, RNA secondary structure may affect RNA stability, RNA 3D structure, intron splicing, exon recognition, transcription level and translation efficiency. Many replicable risk SNPs in *SH3BP5-NR2C2* had potentials to slightly (in *SH3BP5* and *NR2C2*) to significantly (in *ZFYVE20*) alter the RNA secondary structures (Supplemental Table S2), which might further influence the function of proteins and eventually affect the risk for disease, providing additional evidence in support of the hypothesis that *SH3BP5-NR2C2 per se* contributes to alcohol and nicotine co-dependence. Fourth, all SNPs in this region among those 11 replicable risk markers available for eQTL analysis had nominal *cis-* and *trans-*acting regulatory effects on gene expression.

Expression of *NR2C2, ZFYVE20, CAPN7* and *SH3BP5* transcripts was significantly correlated with expression of many alcohol and nicotine dependence-related genes, including those in the dopaminergic[Batel *et al.*, 2008; Stapleton *et al.*, 2007], serotoninergic[Hasegawa *et al.*, 2002], cholinergic [Lou *et al.*, 2006; Saccone *et al.*, 2009], GABAergic [Chang *et al.*, 2002], glutamatergic [Edenberg *et al.*, 2010], histaminergic [Oroszi *et al.*, 2005], endocannabinoid [Zuo *et al.*, 2007], opoid [Zhang *et al.*, 2008], alcohol metabolic[Luo *et al.*, 2006; Uhl *et al.*, 2008] and neuropeptide[Wetherill *et al.*, 2008] systems. These findings suggested that *SH3BP5-NR2C2* might also be implicated in alcohol dependence via the classical neurotransmission systems or metabolic pathways.

Additionally, the transcript expression of the genes in this region was nominally regulated by some nicotine dependence-related genes, including *ITGA4, KCNN3, CCBE1, SGCZ, PARK2, PBK* and *CTNNA3*. All of these genes have ever been associated with nicotine dependence or related traits before[Bierut *et al.*, 2007; Uhl *et al.*, 2010; Uhl *et al.*, 2008], although some of them have not been well replicated yet. *ITGA4* and *KCNN3* regulated all four genes in *SH3BP5-NR2C2* region in brain, and *CCBE1* and *SGCZ* regulated these four genes in PBMC. *PARK2* and *PBK* regulated *SH3BP5* in PBMC and brain, respectively; and *CTNNA3* regulated *CAPN7* in brain. Although these regulatory effects were not significant after correction for multiple testing, those effects replicable across four genes might be robust. Biological functions of these nicotine dependence-related genes together with *NR2C2, ZFYVE20, CAPN7* and *SH3BP5* were summarized in Table S5. Most of these genes are integrin- or calcium-related, or encode proteases (Table S5), suggesting other possible molecular mechanisms that might be implicated in alcohol and nicotine dependence.

Taken together, these findings strongly support the hypothesis that *SH3BP5-NR2C2* harbors causal loci for alcohol and nicotine co-dependence. This region can be represented by two independent markers located in two independent LD blocks, and it might harbor two independent causal loci that are in LD with these independent blocks. The first causal locus might be rs9868848 in *ZFYVE20* in block 1 that is a non-synonymous variant (Leu591Pro) located in a coding region. It could significantly alter the RNA secondary structure. It was functional in brain and PBMC with nominal *cis-* and *trans-*acting regulatory effects (Table 1). Its minor allele protected against risk for alcohol and nicotine co-dependence. Two other SNPs (rs28445844 in *NR2C2* and rs2306853 in *ZFYVE20*) in LD with this SNP could also

slightly alter the RNA secondary structures, and rs2306853 is located in an exonic splicing silencer or enhancer (Table S2). Alternatively, *SH3BP5-CAPN7* (block 2) might possibly harbor a second putative causal locus, because the most significant SNP in this region was rs1318937 in *SH3BP5* and this SNP could also slightly alter the RNA secondary structures (Table S2). Its minor allele increased risk for alcohol and nicotine co-dependence. However, all risk markers in this region were predicted to most likely lack any phenotypic effect (by Polyphen-2 [Adzhubei *et al.*, 2010]; Table S2), so that the causal loci might not be any one of these risk markers. It is warranted in future studies to identify the causal loci by sequencing the whole *SH3BP5-NR2C2* region.

In the present study, although we studied alcohol and nicotine co-dependence, we obtained results that were similar to a previous GWAS that examined alcohol dependence using SAGE data[Bierut *et al.*, 2010]. Many top-ranked risk SNPs ($p<10^{-5}$) for alcohol dependence in that previous study (i.e., *KIAA0040, HTR1A, PKNOX2, HAO2, CTTNBP2, TMEM47, SH3BP5* and *PLEKHG1*) were also listed as top-ranked genes in the present study (Table S3). Another study found that *SH3BP5* was associated with polysubstance dependence in NIDA/MNB sample (rs9310472; p=0.008) and with cocaine dependence in SAGE sample (rs1318937; p=1.6×10⁻⁷) (both SNPs were claimed to be in *CAPN7* in that study)[Drgon *et al.*, 2010].

Finally, three genes or gene regions, i.e., *GABRA2, CHRNA6-CHRNB3* and *CHRNA5-CHRNB4* that have previously been associated to both alcohol and nicotine dependence [Bierut *et al.*, 2007; Edenberg *et al.*, 2004; Liu *et al.*, 2010; Ray *et al.*, 2009; Saccone *et al.*, 2007; Thorgeirsson *et al.*, 2010] were also explored in the present study. They were only nominally associated with alcohol and nicotine co-dependence (Table S6); they were neither replicable between EAs and AAs, genome-wide significant, nor on the top-ranked gene list in the present study (Table S3), consistent with previous results using the same SAGE dataset [Bierut *et al.*, 2010; Wang *et al.*, 2011].

Supplementary Material

Refer to Web version on PubMed Central for supplementary material.

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						EAs	7	AAs	Meta-a	malysis	⁶ O ³	IL▲
SNP	Gene	Position (Build 36 Ref.)	Location	Minor Allele	OR	đ	OR	d	Z score	đ	p (Brain)	p (PBMC)
rs17040623	NR2C2	15003482	Intron 1	ад	0.87	0.025	0.69	3.4×10^{-3}	-3.46	5.3×10^{-4}	NA	NA
rs7635654	NR2C2	15022966	Intron 3	c	0.86	0.023	0.76	0.017	-3.20	1.4×10^{-3}	NA	NA
rs28445844	NR2C2	15049327	Intron 11	t	0.87	0.026	0.75	9.5×10^{-3}	-3.27	1.1×10^{-3}	NA	NA
rs3773478	ZFYVE20	15088046	3' UTR	а	0.85	0.033	0.72	0.036	-2.93	3.4×10^{-3}	NA	NA
rs9851219	ZFYVE20	15090727	coding	с	0.85	0.030	0.71	0.027	-3.02	2.6×10^{-3}	NA	NA
rs9868848	ZFYVE20	15090876	coding	ac	0.86	0.038	0.72	0.037	-2.87	4.1×10^{-3}	0.004	0.016
rs2306853	ZFYVE20	15092129	coding	а	0.84	0.026	0.72	0.037	-3.01	2.7×10^{-3}	NA	NA
rs735659	intergenic	15198998	24kb to 5' of CAPN7	c	1.25	$8.9{ imes}10^{-4}$	1.17	0.046	3.87	$1.1 { imes} 10^{-4}$	0.081	0.015
rs1318937	intergenic	15270368)10bp to 3' of <i>SH3BP5</i>	ac	1.33	2.5×10^{-5}	1.29	5.1×10^{-3}	5.06	4.1×10^{-7}	0.041	0.021
rs3773471	SH3BP5	15278051	Intron 6	c	1.29	1.4×10^{-4}	1.38	0.016	4.50	6.9×10^{-6}	NA	NA
rs8225	SH3BP5	15280874	Intron 5	t	1.29	1.4×10^{-4}	1.36	0.026	4.41	1.0×10^{-5}	0.040	0.103
All markers are	s in HWE. eQ	TL, expression quantitative tra	it locus analysis;									

 \mathbf{k} Exon-level expression changes in Brain and PBMC tissues (minimal p values are presented); ZFYVE20, CAPN7 and SH3BP5 have 15, 29 and 14 exons, respectively, and thus the corrected α for exonlevel eQTL analysis was 0.003, 0.002 and 0.004, respectively. NA, not available.

Table 2

Correlations of distributions of $-\log(p)$ values for gene-disease associations in *SH3BP5-NR2C2* region between different populations

	EA		AA	
	r	р	r	р
AA	0.315	3.6×10 ⁻⁵		
EA			0.315	3.6×10 ⁻⁵
Meta	0.868	8.1×10 ⁻⁵²	0.544	3.5×10 ⁻¹⁴

r, Pearson correlation coefficient; p, p-values for pairwise correlations.