

# Neurotransmitter and Neuromodulator Genes Associated With a History of Depressive Symptoms in Individuals With Alcohol Dependence

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**Background:** Depressive symptoms are common among individuals with alcohol use disorders and impact treatment outcome. Substantial overlap exists among the neurobiological systems proposed in the pathophysiology of depressive and alcohol use disorders; however, specific genetic effects contributing to risk for depressive comorbidity remain poorly understood.

**Methods:** This study examines the association of depressive symptom scores for lifetime depression (the sum of DSM-IV major depression co-endorsed criteria for lifetime depression) with markers in 120 candidate genes in 554 alcohol-dependent individuals. The candidate genes code for molecules involved in dopamine, serotonin, glutamate, gamma-aminobutyric acid (GABA), and opioid neurotransmission, cell signaling, pharmacokinetics, stress biology, and behavioral control. Analyses were conducted at the single marker level with experimentwise permutation to control for multiple testing.

**Results:** Results revealed nominal associations for markers in 20 genes. Following experimentwise permutation, markers in the corticotropin-releasing hormone-binding protein (*CRHBP*) the  $\mu$ -opioid receptor (*OPRM1*) and the  $\beta 1$  subunit of GABA A ( $GABA_A$ ) receptors (*GABRB1*) met or exceeded the significance threshold. None of the markers associated with depressive symptom scores were significantly associated with alcohol dependence symptom scores.

**Conclusion:** These findings suggest potential risk genes for depressive symptoms in alcohol-dependent individuals.

**Key Words:** Major Depression, Alcohol Dependence, Genetic Association.

VULNERABILITY TO STRESS-RELATED psychiatric disorders such as alcohol dependence (AD) is influenced by environmental factors; however, family, twin, and adoption studies consistently demonstrate that AD aggregates in families due partly to genetic risks (McGue, 1999). Biological vulnerability to addiction is supported by the observation that disturbances in neuroendocrine activity occur among offspring of AD individuals, which in turn predicts future experimentation with illicit substances (Lovallo, 2006). Despite the strong familial component to AD, complex interactions among neurobiological systems involved in addiction and high rates

of comorbid clinical diagnoses complicate efforts to identify genetic vulnerabilities.

Genomewide association studies (GWAS) have had modest success in identifying genes associated with AD (e.g., Edenberg and Foroud, 2006). Because they are empirically rather than hypothesis driven, GWA studies require conservative statistical correction that renders detection of common vulnerability alleles difficult. An alternate strategy is to select candidate genes in a focused array based on the neurobiology of addiction. Whereas a neurobiology-guided approach may have its own limitations, this strategy has identified some important AD-linked genes, such as the alcohol dehydrogenase cluster (Edenberg, 2007). In addition to ethanol metabolism, however, a strong motivational component involving multiple neurobiological systems is also important for the development and maintenance of addiction.

An influential conceptual framework provided by Koob and colleagues conceptualizes AD, and addiction generally, as a cycle of increasing dysregulation of brain reward/antireward mechanisms, resulting in a negative emotional state contributing to the compulsive use of drugs. According to this model, the reward mechanisms disturbed in AD are mediated by activation of dopamine, serotonin, opioid peptides, and  $\gamma$ -aminobutyric acid (GABA) systems, as well as receptor transduction mechanisms such as adenylate cyclase, protein kinase A, cAMP response element-binding (CREB),

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and deltaFosB. Activation of brain stress systems, most notably corticotropin-releasing hormone (CRH) but also noradrenaline, dynorphin, and the antistress neuropeptide Y system in the extended amygdala is considered the anti-reward mechanisms associated with the emotional dysregulation observed in addiction (Koob and Le Moal, 2008). Koob argues that CRH and other brain stress systems play a role in the transition to dependence, its maintenance, and relapse during abstinence in the face of life stress by contributing to a “residual negative emotional state” in the cycle of addiction (Koob and Le Moal, 2008). The negative affective component in the cycle of AD consists of motivational changes including chronic irritability, emotional pain, malaise, dysphoria, alexithemia, sensitivity to stress and pain, sleep disturbance, and loss of interest in natural rewards (Valdez and Koob, 2004).

Although negative affect that occurs in the context of heavy drinking does not necessarily warrant an independent diagnosis of major depression (MD), clinical data show a large proportion of AD individuals meet criteria for concurrent or lifetime history of MD. Moreover, disturbances in the CRH system, the brain stress system central to Koob’s model, are believed to play a central role in the pathogenesis of MD via similar feedback/feedforward processes (DeRijk and de Kloet, 2005). Epidemiological studies of AD and MD comorbidity show there is up to a 4-fold increased lifetime risk for developing one disorder, if the other disorder has occurred (Hasin and Grant, 2002; Lynskey, 1998). Comorbidity rates are even higher among treatment-seeking populations (Grant et al., 2004). Research conducted by trained clinicians using standard diagnostic interviews has shown that over two-thirds of severely affected (i.e., hospitalized) AD patients meet criteria for lifetime history of MD (Grant et al., 1989). Among individuals with AD, symptoms of MD predict treatment outcomes, in particular relapse to drinking (Greenfield et al., 1998). These clinical implications further underscore the importance of understanding factors contributing to depressive symptoms in individuals with AD.

This study examines genetic association of symptom scores from a lifetime history of depression in AD individuals. We conducted a candidate-gene association study using a single-nucleotide polymorphism (SNP) array designed by the group of D. Goldman at the National Institute on Alcohol Abuse and Alcoholism (NIAAA). The rationale for using this array was twofold: first, genes on the array are selected on the basis of their roles in functional domains important in the related phenotypes of addiction, depression, and anxiety (Goldman et al., 2005). In some cases, specific genes and molecules have been identified by GWAS or candidate gene focused linkage results; other genes are selected solely on the basis of animal studies or for their role in relevant biological pathways (Goldman et al., 2005). Second, along with high clinical comorbidity, there is considerable overlap among neurobiological pathways hypothesized to be involved in pathogenesis of AD, mood, and anxiety disorders including signaling networks, neuroendocrine stress response, and neurotransmitter systems

(Belmaker and Agam, 2008; Pacher and Keckskemeti, 2004; Vengeliene et al., 2008). Neurotransmitter/neuromodulator systems with potential overlap in AD and MD include serotonin, acetylcholine, dopamine, GABA, opioid, glutamate, CRH, and neuropeptide Y systems. With the exception of genes directly involved in drug metabolism, whose role in AD is well documented, we opted to conduct association analysis on the entire array. This includes glutamate, GABA, dopamine, serotonin, and opioid neural system genes, signaling genes, pharmacokinetic genes, and genes modulating stress biology and behavioral control.

## MATERIALS AND METHODS

### *Participants and Phenotypes*

Participants in this study were recruited in Ireland and Northern Ireland. A detailed account of the study design, sample ascertainment and consent procedure, is described elsewhere (Prescott et al., 2006). All procedures were approved by the Institutional Review Board at Virginia Commonwealth University, the ethics committees at the Irish Health Research Board and Belfast Health and Social Care Trust, and all the ethics boards at facilities referring participants. Briefly, ascertainment of probands was conducted primarily in public and private hospitals and community alcoholism treatment facilities. Probands were eligible for study inclusion, if they met DSM-IV (American Psychiatric Association, 1994) criteria for AD and if all 4 grandparents had been born in Ireland, Northern Ireland, Scotland, Wales or England. After a prospective family was identified through probands, parents and potentially affected siblings whom the probands provided permission to contact were recruited.

Interviews were conducted by clinically trained research interviewers. The assessment included demographic and ancestral information as well as related traits and lifetime history of symptoms of AD and MD. Depressive symptoms were assessed using the MD module from the SCID-R. Participants were asked about lifetime history of depressed mood or anhedonia lasting at least 2 weeks. Individuals responding positively to either item were assessed for the presence of remaining 7 criteria for a DSM-IV-defined MD episode during that period (or during their worst episode if there were multiple periods of 2 weeks of symptoms; Spitzer and Williams, 1985). Co-occurrence of symptoms was required to be considered a positive endorsement, as per standard diagnostic interviews. To maximize comparability to prior literature on depressive symptoms, a depressive symptom score was created by summing the 9 co-occurring DSM-IV symptom criteria. Prior work comparing a continuum of depressive symptoms and MD diagnosis indicates that a continuum of depressive symptoms has greater predictive power (Kendler and Gardner, 1998). Lifetime AD was assessed using the modified Semi-Structured Assessment for the Genetics of Alcoholism (SSAGA) interview and included items to assess the 7 DSM-IV AD criteria [version 11 (Bucholz et al., 1994)]. As with the depressive symptom score, an AD symptom score was created by summing the number of co-occurring DSM-IV AD symptom areas endorsed. Participants were also asked whether their alcohol problems occurred always, sometimes, or never in conjunction with their depressive symptoms. Of 425 individuals with clinically significant levels of depressive symptoms (those meeting criteria for MD diagnosis), few participants reported either complete separation (3.5%) or complete overlap (4.9%) with regards to co-occurrence of depressive and AD symptoms.

All participants provided informed consent. There were 1,238 individuals who completed the SSAGA interview and met criteria for DSM-IV AD diagnosis, including 591 probands and 647 affected siblings. In families where more than 1 AD individual provided a DNA sample, 1 participant was randomly drawn for inclusion in the

study. Genotyping was performed on 575 independent cases; of which, 13 were dropped for poor DNA quality (see Data QC) and 8 for insufficient information regarding depression, yielding a final sample of 554 (354 men, 200 women). Demographics, drinking history, and treatment use are shown in Table 1. This was a severely affected AD sample, with 84% of the sample endorsing at least 6 of 7 possible AD criteria, 73% of the sample having undergone 2+ different types of treatment (e.g., self-help group, inpatient, outpatient, hospital, other), with the median number of maximum drinks consumed in a 24-hour period equaling 36 drinks.

### Single-Nucleotide Polymorphisms

A detailed description of the NIAAA “addictions array” has been published previously (Hodgkinson et al., 2008). Briefly, candidate genes were selected based on molecular pathways or prior evidence from human or animal studies suggesting involvement in the related phenotypes of addiction, depression, and anxiety. Genes assessed in this study are indicated in Table 2. Of particular interest for an examination of depressive symptoms among alcohol-dependent individuals are genes in systems potentially implicated in studies of both AD and MD, including serotonin, acetylcholine, dopamine, GABA, opioid, N-methyl-D-aspartic acid (NMDA) (glutamate), and stress [hypothalamic-pituitary-adrenocortical (HPA) axis and neuropeptide Y]. Genomic regions containing 5 kb upstream and 1 kb downstream of each candidate gene were screened for haplotype tagging SNPs selected to reconstruct all haplotypes with frequency greater than 0.5% observed in HAPMAP data from Africans, the most genetically diverse of human population groups. Performance of initially selected SNPs was validated by the manufacturer and replacements made when necessary. A total of 1,536 loci on a custom Illumina array were genotyped, including 1,350 SNPs and 186

**Table 2.** Genes Analyzed

	<i>GABA</i>	<i>Stress</i>	<i>Signal Transduction</i>	<i>Serotonin</i>
<i>Adrenergic</i>				
ADRA1A	GABRA2	CRH	ADCY7	HTR1A
ADRA2A	GABRA3	CRHBP	AVPR1A	HTR1B
ADRA2B	GABRA4	CRHR1	AVPR1B	HTR2A
ADRA2C	GABRA6	CRHR2	CDK51R	HTR2B
ADRB2	GABRB1	FKBP5	CREB1	HTR2C
ARRB2	GABRB2	GAL	CSNK1E	HTR3A
SLC6A2	GABRB3	NPY	FEV	HTR3B
DBH	GABRD	NPY1R	FOS	MAOA
	GABRE	NPY2R	FOSL1	MAOB
<i>Cholinergic</i>	GABRG2	NPY5R	FOSL2	SLC6A4
CHRM1	GABRG3	NR3C1	GSK3B	TPH1
CHRM2	GABRQ		JUN	TPH2
CHRM3	SLC6A7	<i>Dopamine</i>	MAPK1	
CHRM4	SLC6A11	COMT	MAPK3	<i>Other</i>
CHRM5	SLC6A13	DDC	MAPK14	BDNF
CHRNA4	SLC32A1	DRD1	MPDZ	CART
CHRN2	GAD1	DRD2	NGFB	CCK
	GAD2	DRD3	NTRK2	CCKAR
<i>Opioid</i>	DBI	DRD4	NTSR1	CCKBR
OPRM1		DRD5	NTSR2	CLOCK
OPRD1	<i>NMDA</i>	SLC1BA2	PPP1R1B	HCRT
OPRK1	GRIK1	SLC6A3	PRKCE	LEP
OPRL1	GRIN1	TH		OXT
PDYN	GRIN2A		<i>Glycine</i>	SLC29A1
PENK	GRIN2B	<i>Cannabinoid</i>	GLRA1	TAC1
PNOC	GRIN2C	CNR1	GLRA2	
POMC	GRM1	FAAH	GLRB	
			GPHN	

NMDA, N-methyl-D-aspartic acid; GABA, gamma-aminobutyric acid.

**Table 1.** Demographic and Clinical Characteristics of the Sample

Variable	Statistic
Age at interview M (SD)	41.8 (9.7)
Sex (% female)	36.1
Marital status	
Married (%)	39.9
Divorced/separated (%)	27.8
Widowed (%)	2.7
Never married (%)	29.1
Education	
<Secondary (%)	11.3
Some secondary (%)	34.3
Completed secondary (%)	39.0
Some postsecondary (%)	15.4
Employment	
>20 hours (%)	44.8
Unemployed (%)	36.6
Other (student, homemaker, retired) (%)	18.6
MD diagnosis (%)	75.1
Drinking history	
Age at first drink M (SD)	15.8 (4.6)
Age at first intoxication M (SD)	17.4 (6.1)
Age began drinking regularly M (SD)	18.1 (5.4)
Age began drinking most heavily M (SD)	31.9 (9.4)
Max drinks in 24 hours M (SD)	39.3 (19.3)
Age meeting AD criteria M (SD)	26.3 (8.9)
Treatment	
Any treatment (%)	79.6
Alcoholics anonymous (%)	72.2
Inpatient alcohol treatment facility (%)	64.8
Outpatient alcohol treatment facility (%)	49.6
Medical facility (%)	21.3
Other treatment (%)	9.7

AD, alcohol dependence; MD, major depression.

ancestry informative markers (AIMs). AIMs were previously developed for and tested on the Illumina platform (Enoch et al., 2006) and were analyzed using structure 2.1 (Pritchard et al., 2000) to generate population assignments for all individuals.

An additional 10 SNPs were selected in-house to tag *FKBP5*, which was not included on the array, and *AVPR1B*, for which there were no SNPs informative for Caucasian populations on the array. These SNPs were genotyped using an automated liquid handling system EP 5075 (Eppendorf, Hamburg, Germany) and standard laboratory procedures for Taqman genotyping (Applied Biosystems by Life Technologies, Carlsbad, CA).

### Data Analysis

Association analyses were conducted at the single marker level using SAS 9.1 (SAS Institute Inc., Cary, NC). Mantel-Haenszel chi-square for linear association was computed to test the association of depressive symptom scores and genotype of each marker while controlling for potential confounds of age and sex. X-linked markers were assessed for men and women separately. As expected with a clinically ascertained sample, depressive symptom scores were high ( $M = 6.02$ ,  $SD = 3.4$ ), with wide variability. One hundred and nineteen participants did not report any lifetime history of depression. Because the scores showed a bimodal distribution, normality assumptions required for linear regression analyses would have been violated. We therefore treated the depressive symptom score as an ordinal scale. To maximize generalizability to clinical populations, we grouped individuals with scores below the diagnostic cut-off of DSM-IV criteria at 1 level (0 to 4 criteria  $N = 134$ ) and the remaining individuals in 4 additional levels reflecting increasing numbers of depression criteria above the DSM-IV threshold for caseness (5 to 6 criteria  $N = 70$ ; 7 criteria  $N = 65$ ; 8 criteria  $N = 128$ ; 9 criteria  $N = 157$ ). The ordinalized depressive symptom measure was highly correlated with the original score (raw total count of depressive symptom areas endorsed; Pearson's  $r = 0.93$ ,  $p < 0.001$ ). Power



calculations for the depressive symptom measure are reported in Table S1.

Because alcohol and depressive symptoms tend to be highly comorbid, we conducted a parallel Mantel–Haenszel chi-square analysis testing association of allelic frequencies with the alcohol symptom scores to examine whether associations were unique to endorsement of symptoms of depression or common to depression and AD. Because the sample was clinically ascertained, the total AD criteria endorsed was high ( $M = 6.42$ ,  $SD = 0.96$ ) and not normally distributed. For this reason, and to compare this scale with the depressive symptom scale, the AD symptom scale was also treated as an ordinal variable (3 to 4 criteria  $N = 36$ , 5 criteria  $N = 50$ , 6 criteria  $N = 101$ , 7 criteria  $N = 367$ ). The number of individuals at each level of the depressive symptom and alcohol symptom scales is provided in Table S2. There was a significant but modest correlation in the depressive symptom and AD scores, ( $\rho = 0.26$ ,  $p < 0.01$ ).

Determining appropriate threshold for significance was a challenge for this study. Candidate genes were selected on the basis of biological pathways purported to be involved in a limited set of related phenotypes, and thus, many tests have higher a priori probabilities of yielding significant results than random genomic regions. Here, we consider results based on experimentwise permutation analysis, a conservative approach to interpreting significant results. We use the term *observed*  $p$  values to refer to nominal significance where alpha is set at  $p < 0.05$ . *Empirical*  $p$  values refer to  $p$  values generated through experimentwise permutation analysis.

Experimentwise permutation was conducted by randomly permuting depressive symptom scores across the sample while holding genotype fixed and retaining the correlational structure of all SNPs. Observed  $p$  values were compared to permuted  $p$  value distributions under the null hypotheses to generate empirical  $p$  values. Inference was based on 1,000 permutations. To interpret the results of experimentwise permutation, we utilized the method introduced by Lander and Kruglyak (1995) used extensively in linkage analyses. Under this approach, we consider a  $p$  value threshold that occurs by chance once per experiment as significant. Experimentwise thresholds can be easily calculated, if all the tests performed are independent. In this study, many tested SNPs were in high linkage disequilibrium (LD); thus, analyses were conducted on empirical  $p$  values where LD structure was preserved. We determine experimentwise thresholds as  $N$ th position, where  $N =$  number of permutations.

## RESULTS

### *Data QC and Stratification Analysis*

Among the 1,360 genotyped SNPs, 4.7% were excluded on the basis of low genotyping rate ( $<0.5$ ) and 25.2% on the basis of minor allele frequency  $<0.05$  and/or violation of Hardy–Weinberg equilibrium ( $<0.001$ ). Because markers on the array were originally selected for tagging in a different ethnic group, many markers would have been redundant in our sample. Thus, we used genetic data from a sample of 530 Irish community controls previously collected for a study of AD (Prescott et al., 2006) to determine LD among SNPs on the addictions array. Nonindependence was defined as  $r^2$  values exceeding 0.80. Where  $r^2 > 0.80$ , one SNP was randomly selected using the program Haploview (Barrett et al., 2005) for inclusion in analysis, resulting in a final SNP set of 644 markers. Thirteen individuals were excluded on the basis of low genotyping rate ( $<50\%$ ). The average genotyping rate of the remaining individuals was 92%. Of 186 AIMs, 4.8% were excluded for low genotyping rate.

Analyses of population structure testing 2 or 3 maximum populations were consistent in showing evidence for 1 ethnic cluster. For the 2 population test, individuals' assignments to population 1 was Mean = 0.50,  $\sigma = 0.02$  (min  $p = 0.45$  max  $p = 0.55$ ). Discrepancies from the mean were very small, indicating there was only 1 population. Assessments of AIM data showed no evidence for stratification, as expected in this single-ethnicity sample.

### *Associations With Age and Sex*

A small but significant correlation was observed for number of depressive symptom areas/criteria endorsed and age at the time of interview ( $r = -0.14$ ,  $p < 0.01$ ) with older participants on average endorsing a greater number of criteria. The number of depressive symptom areas endorsed was not correlated with sex ( $M = 5.92$  in men and 6.21 in women,  $t = 0.97$ , *ns*). The lack of a sex difference in depressive symptoms is consistent with epidemiological data by Van de Velde and colleagues (2010) showing no sex difference for MD diagnoses in Ireland. That report also showed that the difference in prevalence of MD between men and women in Ireland is the smallest among 23 European countries. However, because sex differences are commonly observed in other cultures and may relate to biological or demographic factors (Van de Velde et al., 2010), both age and sex covariates were preserved for possible measured and unmeasured effects.

### *Single-Marker Association*

Results of single-marker tests are indicated in Table 3. Mantel–Haenszel chi-square for linear association was computed to examine the association of depressive symptom scores with genotype after controlling for possible confounding effects of sex and age, where age was binned into 5-year intervals to accommodate the Mantel–Haenszel chi-square test. To examine whether associations were being driven by individuals whose depressive symptoms occurred only in the context of drinking, we ran associations on a subset of individuals excluding the 21 individuals who reported symptoms of MD and AD always occurred together. None of the significant effects differed from those of the full data set, and so data from the full sample are reported. Nominal associations were observed for 31 SNPs (observed  $p$ 's  $<0.05$ ) in multiple molecular pathways and functional domains. They include the adrenergic, cholinergic, dopaminergic, opioidergic and serotonergic systems, GABA, NMDA receptors, as well as molecules involved in signal transduction and the neuroendocrine stress response.

### *Analysis of Experimentwise Permutation*

Depressive symptom scores were permuted across participants 1,000 times while retaining the correlational structure of SNPs to generate empirical  $p$  values. Using the Lander and Kruglyak (1995) approach to interpretation of

**Table 3.** Single-Marker Association With Depressive Symptom Score

Marker <sup>a</sup>	Gene	Chr.	bp <sup>a</sup>	Position <sup>a</sup>	MAF	$\chi^2$	Observed <i>p</i>
rs11688	JUN	1	59,247,993	Exonic synonymous	0.05	5.24	0.0221
rs10495447	CHRM3	1	240,051,340	Intronic	0.26	3.92	0.0478
rs4688043	GSK3B	3	119,591,248	Intronic	0.08	5.51	0.0189
rs1442060	GABRA2	4	46,366,067	Intronic	0.45	5.34	0.0208
rs2236781	GABRB1	4	47,035,087	Intronic	0.49	5.76	0.0164
rs4315750	GABRB1	4	47,051,185	Intronic	0.09	9.64	0.0019
rs971353	GABRB1	4	47,102,894	Intronic	0.28	3.95	0.0469
rs9996854	GABRB1	4	47,340,895	Intronic	0.36	5.97	0.0145
rs6350	SLC6A3	5	1,443,199	Exonic synonymous	0.08	5.63	0.0176
rs10474485	CRHBP	5	76,270,853	3'	0.16	5.83	0.0157
rs1715747	CRHBP	5	76,274,537	3'	0.31	10.61	0.0011
rs1042718	ADRB2	5	148,206,917	Exonic synonymous	0.14	4.03	0.0446
rs7724086	GABRB2	5	160,724,214	Intronic	0.09	5.82	0.0158
rs1799971	OPRM1	6	154,360,797	Exonic missense	0.10	4.80	0.0285
rs3778151	OPRM1	6	154,393,680	Intronic	0.17	5.12	0.0237
rs650245	OPRM1	6	154,428,702	Intronic <sup>b</sup>	0.10	9.86	0.0017
rs548339	OPRM1	6	154,460,799	Intronic	0.33	3.95	0.0468
rs10488599	CHRM2	7	136,589,094	Intronic	0.15	7.06	0.0079
rs10102186	ADRA1A	8	26,624,651	Intronic	0.31	4.26	0.0390
rs6356	TH	11	2,190,951	Exonic missense	0.37	5.30	0.0213
rs2587548	DRD2	11	113,292,212	Intronic	0.40	4.08	0.0434
rs1125394	DRD2	11	113,297,185	Intronic	0.16	5.55	0.0185
rs4274224	DRD2	11	113,319,452	Intronic	0.46	5.07	0.0244
rs1352252	TPH2	12	72,452,041	3'	0.37	4.43	0.0352
rs11179071	TPH2	12	72,455,185	3'	0.14	6.18	0.0129
rs6582086	TPH2	12	72,498,191	3'	0.18	5.68	0.0172
rs12324292	GABRB3	15	27,003,081	Intronic	0.31	5.27	0.0217
rs4611457	ADCY7	16	50,295,994	5'	0.47	6.74	0.0094
rs7219247	GRIN2C	17	72,847,205	Intronic	0.27	5.85	0.0156
rs2235751	PDYN	20	1,969,934	Intronic	0.20	4.10	0.0429
rs460401	GRIK1	21	31,176,765	Intronic	0.10	3.95	0.0469

<sup>a</sup>As identified in dpSNP Build 129.

<sup>b</sup>NCBI lists this marker as intronic. Recent work identifies it as exonic or in the 3'-UTR of some isoforms (Shabalina et al., 2009). MAF, minor allele frequency.

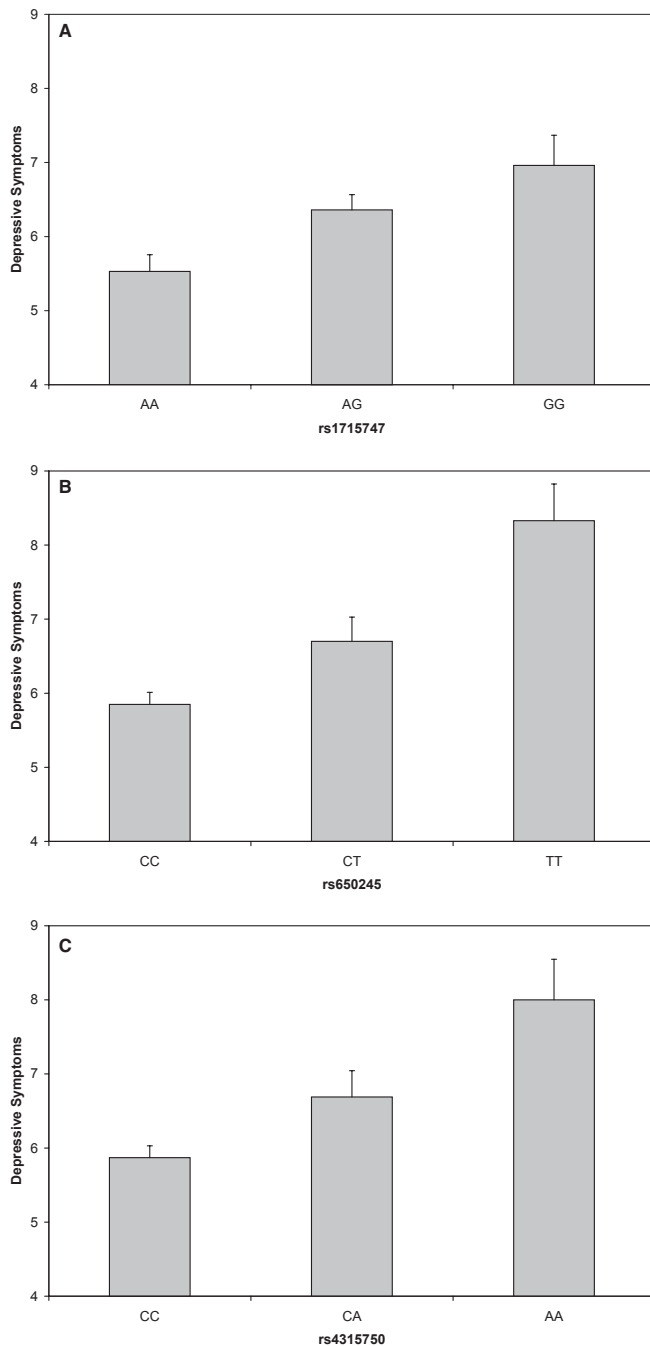
experimentwise permutation, the threshold for significance was set at  $p = 0.0017$ . Empirical  $p$  values below the threshold included the corticotropin-releasing hormone-binding protein (*CRHBP*) marker 1715747 and the *OPRM1* marker rs650245 (empirical  $p$ 's = 0.0010, 0.0014). The *GABRB1* marker rs4315750 was at the permutation threshold (empirical  $p = 0.0017$ ). Figure 1 shows the relation of the genotypes to depressive symptom scores. In each case, the minor allele was additively associated with higher depressive symptom scores. Associated allele frequencies for each depressive symptom score are indicated in Table 4. Notably, none of the markers that were significantly associated with depressive symptom scores were even nominally associated with AD symptom scores (see Table S3).

## DISCUSSION

The purpose of this study was to examine the association of potential candidate genes with lifetime history of depressive symptoms in alcohol-dependent individuals. Initial results revealed significant associations for markers in 19 genes. Because there is not yet a clear consensus as to the most appropriate method of multiple-testing correction for genetic association analysis, we used experimentwise permutation, a conservative approach to multiple testing. Of the 20 genes

reaching nominal significance, SNPs in *CRHBP*, *OPRM1*, and *GABRB1* were at or below the multiple-test correction threshold.

*CRHBP* codes for the corticotropin-releasing hormone-binding protein that regulates the bioavailability of CRH to act at its receptors, inhibits CRH activation of the HPA axis, and may be involved in HPA axis feedback and feedforward regulation via effects in limbic system structures (Behan et al., 1995; Van Den Eede et al., 2005). The CRH system is central to both the neuroendocrine hypothesis of depression (DeRijk and de Kloet, 2005) and the proposed antireward mechanism promoting the transition to, maintenance of, and relapse in addiction (Koob and Le Moal, 2008). Moreover, HPA axis feedback influences reward pathways and GABA<sub>A</sub> receptor function involved in ethanol sensitivity (Morrow et al., 2006; Oswald and Wand, 2004). Family studies show altered HPA axis activity in relatives of depressed patients and HPA response to ethanol in sons of alcoholics (DeRijk and de Kloet, 2005; Schuckit et al., 1987; Zimmerman et al., 2004). Two *CRHBP* SNPs were nominally significant in our study, with 1 passing threshold for experimentwise correction. The *CRHBP* SNP significant in our study was associated with anxiety disorder and resting electroencephalogram (EEG) activity in an independent sample of AD individuals (Enoch et al., 2008). These findings along with our results on



**Fig. 1.** Depressive symptom scores plotted by variants in *CRHBP* (A), *OPRM1* (B), and *GABRB1* (C). Analyses were conducted treating the depressive symptom score as an ordinal measure statistically correcting for age and sex; data are shown on the original scale for ease of interpretation. Error bars denote standard error of the mean. *CRHBP*, corticotropin-releasing hormone-binding protein.

depressive symptoms imply genetic variation in *CRHBP* may play a role in affective disturbance in AD individuals.

All *CRHBP* markers tested were tagSNPs. Thus, the significant SNP does not have any putative function. HapMap data show LD decays rapidly both proximally and distally from *CRHBP*, suggesting the observed associations cannot be explained by effects in surrounding genes. *CRHBP* encodes a

322 amino acid protein with 5 disulfide bonds essential for CRH binding. Single amino acid changes or intronic variants affecting exon content of *CRHBP* isoforms identified in brain might directly affect peptide sequence and alter folding, stability, and/or CRH binding affinity. *CRHBP* is highly conserved across vertebrates including regions of high conservation in the 3'-UTR (possibly affecting mRNA turnover) and 3'-region introns (possibly influencing alternative exon usage).

A significant association was also observed for *OPRM1*. The mu-opioid receptor, which binds  $\beta$ -endorphin and the enkephalins, is typically activated in response to chronic or unpredictable stressful or noxious stimuli (Oswald and Wand, 2004). The opioidergic system is extensively studied in addiction and to a lesser extent in depression. Naltrexone, an opioid antagonist, is used in the treatment of AD and reduces relapse. Depressed patients demonstrate antidepressant effects of opioid peptides and enkephalinase inhibitors (Broom et al., 2002; Jutkiewicz et al., 2006). Differences in mu-opioid receptor availability are reported among depressed patients compared to controls (Kennedy et al., 2006). One mechanism of action consistent with both the neuroendocrine hypothesis of depression and Koob's model of negative affect in the cycle of addiction may be the inhibitory role of opioid neurons in facilitating the termination of the biological stress response (Drolet et al., 2001; Kreek, 1996). Notably, a well-studied *OPRM1* missense mutation with known consequences for receptor structure (rs1799971) has been associated with HPA axis activity (Chong et al., 2006; Wand et al., 2002), suicide (Hishimoto et al., 2008), alcohol sensitivity (Ray and Hutchison, 2004), and alcohol use disorders in adolescents and adults (Bart et al., 2005; Miranda et al., 2010), although a meta-analysis revealed several null associations with AD diagnosis (Arias et al., 2006). Genetic variation in *OPRM1* may have functional consequences at the interface of disturbed HPA axis activity, depression, and addiction.

In our study, rs1799971 and 3 other *OPRM1* markers reached nominal significance. One marker, rs650245, was strongly associated with depressive symptoms and survived a stringent test of multiple test correction. Early reports of the genetic architecture of *OPRM1* identified this SNP as intronic. Recent work identifies this marker as exonic or in the 3'-UTR of some *OPRM1* isoforms (Shabalina et al., 2009). Moreover, rs650245 is in high LD with and lies in the same haplotype block as known functional SNPs and a newly identified, potentially functional SNP (Shabalina et al., 2009). Markers in this haplotype block are associated with highly related phenotypes, including treatment response in MD individuals, pain threshold, negative mood, and substance dependence (Garriock et al., 2010; Shabalina et al., 2009). To our knowledge, this is the first report of an *OPRM1* marker associated with depressive symptoms. Taken together, these findings strongly suggest variation within the *OPRM1* gene may contribute to risk of depressive and addictive disorders.

*GABRB1* codes for the  $\beta 1$  subunit of GABA<sub>A</sub> (GABA<sub>A</sub>) receptors. The GABA<sub>A</sub> receptor is a heteropentameric

**Table 4.** Associated Allele Frequency for Each Level of Depressive Symptoms Scale

Gene	Marker	0–4 symptoms <sup>a</sup>	5–6 symptoms	7 symptoms	8 symptoms	9 symptoms
CRHBP	rs1715747	0.233	0.286	0.315	0.372	0.334
GABRB1	rs4315750	0.059	0.057	0.069	0.098	0.127
OPRM1	rs650245	0.059	0.071	0.069	0.118	0.127

<sup>a</sup>Individuals below clinical threshold for major depression diagnosis.

chloride channel that mediates inhibitory synaptic transmission via binding of GABA. There are 19 mammalian subunits; most GABA receptors contain  $\alpha$ ,  $\beta$ , and  $\gamma$  subunits. The majority of genes encoding human GABA<sub>A</sub> receptor subunits are organized in clusters, which may have functional significance. There is some evidence that altering the structure of one GABA subunit gene affects expression of the other genes in the same cluster (Uusi-Oukari et al., 2000).

Extensive research in rodents indicates variations in the GABA receptor genes contribute to differences in risk of alcoholism. Family-based studies show modest association with AD for *GABRB1* (Reck et al., 2005; Song et al., 2003). A microsatellite polymorphism in *GABRB1* has also been linked with AD (Parsian and Zhang, 1999). Association of *GABRA2*, which is clustered with *GABRB1* but was only nominally associated with depressive symptoms in our study, has in other reports been associated with AD and resting EEG activity (e.g., Edenberg et al., 2004). The significant *GABRB1* SNP in our study is intronic and in low LD with surrounding SNPs; thus, the relation of our findings to previous reports or in terms of potential functional consequences is difficult to determine. To our knowledge, there are no published reports examining association of *GABRB1* with MD, although this gene has been investigated in relation to bipolar disorder (Coon et al., 1994; Crowe et al., 1997). Centrally involved in the neurobiology of addiction, GABA mediates the behavioral effects of ethanol. It is perhaps noteworthy that the  $\beta 1$  subunit is preferentially expressed in the hippocampus (Michels and Moss, 2007), which provides the major GABAergic inhibitory input to CRH neurons initiating the HPA axis and for which reduced volumes have been reported in depressed patients (Herman et al., 2005; Sheline et al., 1996). This finding is also consistent with Koob's model of the role of limbic system feedback of the HPA axis in the negative affect component of the addiction cycle.

#### Limitations and Considerations

There are several limitations to the present results. The depressive symptom scale was created from a structured interview designed to assess clinical diagnoses. We opted to utilize the full range of scores for the number of symptom areas endorsed, rather than dividing individuals only above/below clinical threshold because of evidence suggesting the continuum of depressive symptoms has greater predictive power than standard diagnostic conventions (Kendler and Gardner, 1998). Our treatment of the data, which was non-normally

distributed, as ordinal groups with 1 group representing all individuals below the clinical threshold for MD, and other groups reflecting increasing numbers of endorsed symptom areas, maximized comparability to diagnostic studies while maintaining the validity of the statistical results. In addition, lifetime histories of depression and alcohol use problems were assessed retrospectively, which may have resulted in unmeasured recall error or bias. Replication of key findings in a prospectively studied sample is warranted.

Participants were ascertained from hospitals and treatment facilities for alcohol use disorders. It is possible that observed associations were simply caused by severe alcoholism. However, none of the markers associated with depressive symptom scores in this study were significantly associated with a comparable index of AD symptoms. Although the analysis of AD symptoms is of somewhat limited utility because of range restriction, it provides some confidence that observed associations were unique to the depressive symptoms. Moreover, the correlation among the number of depressive and AD symptoms was modest.

Our sample was severely affected with AD, with the majority of participants endorsing most or all of the AD criteria. The rate of MD comorbidity was also high; however, our comorbidity rate was similar to another study of mostly male, severely affected (hospitalized) AD patients whose symptoms were assessed via structured clinical interview (Grant et al., 1989). Thus, our results must be interpreted in light of the severely affected nature of the sample. The rate of alcohol-induced MD (Fergusson et al., 2009) or something akin to depressive-spectrum disorder (Winoker et al., 1971; see also Nurnberger et al., 2001) might be enhanced in samples of severe AD. Whether the associations observed here would be detected in AD individuals from a community-based sample or 1 in which the incidence of severe AD was lower is worthy of investigation.

Depressive symptoms may be causally involved in or result from AD in different individuals. We were unable to distinguish with confidence the temporal or etiologic relationship of alcohol and depression problems in this sample. Thus, we cannot determine whether genes associated with depressive symptom scores may be unique to depressive symptoms in AD individuals, susceptibility genes to both AD and MD, or modifier genes, that is, genes that modify susceptibility to one disorder in the presence of or genetic risk for the other. Reciprocal effects of the HPA, opioidergic, and mesolimbic reward pathways implicated in alcohol use disorders suggest that altered activity in one system, owing to either genetic or



other factors could simultaneously impact HPA axis function, mesolimbic dopamine production, and ethanol reinforcement (Oswald and Wand, 2004). Future research examining genetic contributions to depressive symptoms in the context of various developmental pathways to AD is warranted. In addition, examining these genes in subjects ascertained initially for MD will be important to establish whether associations identified with depressive symptoms are common to or distinct from primary MD. It is perhaps noteworthy that some genes previously published for association with MD such as *FKBP5* were not associated with depressive symptoms in this sample (Lekman et al., 2008). Similarly, *GABRA2* and the 1799971 SNP in *OPRM1*, both of which have been associated with AD in other samples, were only nominally associated with depressive symptoms among AD individuals in this study. There is some evidence to suggest that AD/MD comorbid individuals may manifest a different course of illness than noncomorbid individuals (Grant et al., 2004) and that comorbid AD/MD tends to aggregate in relatives of individuals with this comorbidity (Nurnberger et al., 2001). This raises the possibility of genetic vulnerability to the comorbid phenotype. Careful attention to a history of comorbid symptoms is clearly warranted in future studies.

## CONCLUSION

In sum, this study investigated the genetic contribution to depression symptoms in alcohol-dependent individuals. Nominal associations were noted for markers in 20 genes. Following experimentwise permutation, markers in 3 genes were at or below the threshold for significance. The strongest signals were in the CRH system, which is potentially involved in both the negative affect component of the addiction cycle and the pathophysiology of MD, and the opioidergic and GABAergic systems, both of which are implicated in AD and provide the major inhibitory input to the CRH system. These findings suggest that further investigation may be warranted of genes involved in regulation of stress biology as they contribute to depressive symptoms comorbid with and potentially contributing to the cycle of addiction.

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## SUPPORTING INFORMATION

Additional Supporting Information may be found in the online version of this article:

**Table S1.** Power table for the depressive symptom scale at varying allele frequencies.

**Table S2.** Number of individuals at each level of the depressive symptom and alcohol symptom scales.

**Table S3.** Associations of significant markers with alcohol symptom scores.

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