A Quantitative-Trait Genome-Wide Association Study of Alcoholism Risk in the Community: Findings and Implications

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Background: Given moderately strong genetic contributions to variation in alcoholism and heaviness of drinking (50% to 60% heritability) with high correlation of genetic influences, we have conducted a quantitative trait genome-wide association study (GWAS) for phenotypes related to alcohol use and dependence.

Methods: Diagnostic interview and blood/buccal samples were obtained from sibships ascertained through the Australian Twin Registry. Genome-wide single nucleotide polymorphism (SNP) genotyping was performed with 8754 individuals (2062 alcoholdependent cases) selected for informativeness for alcohol use disorder and associated quantitative traits. Family-based association tests were performed for alcohol dependence, dependence factor score, and heaviness of drinking factor score, with confirmatory case-population control comparisons using an unassessed population control series of 3393 Australians with genome-wide SNP data.

Results: No findings reached genome-wide significance ($p = 8.4 \times 10^{-8}$ for this study), with lowest p value for primary phenotypes of 1.2 \times 10⁻⁷. Convergent findings for quantitative consumption and diagnostic and quantitative dependence measures suggest possible roles for a transmembrane protein gene (*TMEM108*) and for *ANKS1A*. The major finding, however, was small effect sizes estimated for individual SNPs, suggesting that hundreds of genetic variants make modest contributions (1/4% of variance or less) to alcohol dependence risk.

Conclusions: We conclude that 1) meta-analyses of consumption data may contribute usefully to gene discovery; 2) translation of human alcoholism GWAS results to drug discovery or clinically useful prediction of risk will be challenging; and 3) through accumulation across studies, GWAS data may become valuable for improved genetic risk differentiation in research in biological psychiatry (e.g., prospective high-risk or resilience studies).

Key Words: Alcoholism, genome-wide association, nonreplication, quantitative trait

vidence for important genetic contributions to the intergenerational transmission of alcohol use disorder (AUD) has accumulated over many years. This began with the observation of pedigrees with multiple alcoholic family members (1), continuing with adoption study data suggesting that adopted-away offspring risk correlates with biological parent alcoholism (2,3), and twin studies suggesting that, across a range of phenotype definitions ranging from severe to broad, genetic factors may explain as much as 60% of the variance in risk of alcoholism (4,5). Genetic linkage studies (6–8) and, most recently, the first generation of genome-wide association studies (GWAS) of alcoholic case-control series (9–11) have sought to identify genes contributing to risk of alcohol use disorder. In parallel, a second literature has developed, mostly based on twin studies, showing comparably high genetic variance in alcohol

consumption patterns (12) and noting genetic overlap between alcohol dependence (AD) and heaviness of consumption (13). This overlap persists even when consumption data from alcohol dependent individuals are omitted from the analysis to avoid an artifactual correlation because of escalating consumption secondary to the onset of alcoholism (14). Studies of alcohol metabolism genes (ALDH2, ADH gene family), initially in Asian samples and then in those of Jewish, European, or African ancestry, have shown how genetic variants can make important contributions to differences in risk of alcohol dependence, consumption, or other aspects of response to alcohol (15–19). Finally, a third approach using various forms of latent structure modeling has emphasized the quasi-continuous nature of alcoholism (20–22), drawing attention to the undesirability, for some research purposes, of dichotomizing into affected and unaffected.

Published first-generation case-control genome-wide association studies of alcoholism have yielded generally disappointing findings. They have not achieved the robustly replicated associations seen using quantitative smoking phenotypes, for example, between the CHRNA3/CHRNA5/CHRNB4 gene complex and heaviness of smoking (23–25). Here, we report results of a GWAS using a different research strategy that seeks to take advantage of the quantitative information that can be obtained for heaviness of alcohol consumption (as operationalized in [14]) and for alcoholism symptom severity. This is the first study to apply a genome-wide association approach to a quantitative measure of alcoholism risk in the general community and potentially complements the clinical case-control studies.

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Methods and Materials

Samples

Samples were ascertained from a pool of approximately 11,700 Australian families identified through diagnostic interview surveys of two cohorts of like-sex and unlike-sex twin pairs from a volunteer Australian twin panel (cohort 1, born 1895–1964, n = 5995 interviewed twins [4] but for the purposes of this study mostly born 1940 – 1964, and cohort 2, born 1964 – 1971, n = 6257 twins [26], as well as through an interview survey of the spouses/partners of the former cohort, n = 3846 [27]). Index cases from these families, their full siblings, and parents were recruited for three coordinated studies: 1) the Nicotine Addiction Genetics (NAG) Study (28), which ascertained heavy smoking index cases; 2) the Australian Alcohol Extreme Discordant and Concordant Sibship (OZALC-EDAC) study, which ascertained index cases with a history of alcohol dependence or scoring above the 85th percentile for heaviness of drinking factor score (operationalized as in [14]); and 3) the Australian Alcohol Large Sibship (OZALC-BIGSIB) study, which ascertained large sibships (4-14 full siblings), regardless of sibling phenotypic values. Additional cases and control subjects were recruited from cohort 1 participants and additional cohort 2 participants who did not complete the new interview protocol but had comparable alcohol use/ dependence assessments. Finally, an unassessed population control series is included in some analyses, comprising twins and their families with GWAS data who had participated in an adolescent twin study (29). Further details of the sampling design are given in the Methods and Materials in Supplement 1. All projects underwent Institutional Review Board review at the participating institutions. Genome-wide association study genotyping, using the Illumina 370K array (Illumina Inc., San Diego, California), was performed on a total of 6852 individuals selected from the BIGSIB and EDAC series (including n = 336 parents), from a subsample of the NAG families that had previously been selected for 10 cM microsatellite scans (28), and a smaller number of additional alcohol-dependent cases and control subjects from cohorts 1 and 2. Additional GWAS data are included here for sample members who had been genotyped for other projects (29). Tables S1 and S2 in Supplement 1 summarize numbers of participants and distribution of sibship size for individuals with GWAS genotyping.

Assessment

The diagnostic interview assessment was modified from the Semi-Structured Assessment for the Genetics of Alcoholism (30,31) for telephone administration, with deletion of certain diagnostic sections and elimination of non-DSM items. Assessment of history of alcohol abuse and dependence was supplemented with detailed questions about alcohol consumption (frequency of use, frequency of heavy drinking [using five or more drinks in a day], frequency of drinking to intoxication, drinks per typical drinking day). These were coded categorically, with 10 response categories and a wide range of values (e.g., 1–2 to 31 or more drinks in a typical drinking day) used to encourage more accurate reports, and were asked of both heaviest drinking period of at least 12 months duration and of past 12 months (if not included as the heaviest period). Two additional open-ended items coded maximum drinks (max drinks) in a 24hour period, lifetime, and in the past 12 months. All questions used standard Australian drinks (10 g of alcohol). Diagnostic sections on smoking, anxiety, depression, and conduct disorder were included based on their relevance for understanding alcoholism genetics. Given the potential for inaccurate reporting by interviewed parents (mean age 66 years, range 49-91 years), parent reports were limited to smoking history. In some cases, interview data were only

available from previous diagnostic assessments, with slight variations in assessment protocol, so that not all consumption measures were available for all individuals in the GWAS sample (Table S3 in Supplement 1).

Genotyping and Quality Control

Genotyping and the standard quality control filters that were applied are described in greater detail in Medland et al. (29) (see especially Table 1 in that publication). All genotyping was conducted on Illumina platforms, with genotypes called using Illumina BeadStudio software (Illumina Inc.). Quality control excluded single nucleotide polymorphism (SNPs) with mean GenCall score less than 70%, with call rate less than 95%, with deviation from Hardy-Weinberg significant at $p < 10^{-6}$, or with minor allele frequency less than 1%. For the present study, Illumina CNV370-Quadv3 GWAS data (Illumina Inc.) were available on 4241 individuals (including most alcohol-dependent cases) genotyped at Center for Inherited Disease Research (Baltimore, Maryland) and an additional 2611 individuals genotyped by deCODE (Reykjavik, Iceland) for the OZALC project; Illumina 317K data (Illumina Inc.) were available for 53 individuals genotyped at the University of Helsinki Genome Center and Illumina 610 Quad data were available for the remaining individuals genotyped by deCODE. Duplicate samples allowed comparison of genotyping across platforms/locations: a single SNP was identified, genotyped using the CNV370-Quadv3, which was called very differently at Center for Inherited Disease Research versus deCODE, and therefore deleted from the data set. Checks were run on genetic relatedness, with misspecified relationships corrected before analysis.

Cohorts 1 and 2 are almost entirely of European ancestry, reflecting restrictive Australian immigration policies through 1972; however, Eigenstrat analyses (32), which included data from other Australian GWAS series, identified as outliers (operationalized by \pm 6 standard deviations) a small number of families of mixed European and Asian ancestry (principally Chinese, Burmese, Indian, or Malaysian); of middle eastern (Lebanese) ancestry; with one or more grandparents of Aboriginal, Torres Strait Islander, or Maori ancestry; or with some African heritage (including individuals of self-report Maltese ancestry, consistent with the known population genetics of the Maltese population) (29). A total of 153 individuals from 60 families were thus identified and excluded from further analyses: this included 34 alcohol-dependent and 119 unaffected individuals. In the analyses presented here, we use only data from the approximately 300,000 SNPs that passed quality control and were available on all platforms. We do not attempt imputation, given the potential for statistical problems in the analysis of imputed data (e.g., biased estimates of effect size) in family-structured data sets.

Analyses

Consistent with our previous work (14), a heaviness of drinking (HOD) factor score was derived, using four items: lifetime max drinks, three heaviest period measures of frequency of heavy drinking, frequency of drinking to intoxication, and log-transformed average weekly consumption (derived as the product of frequency of use and drinks per drinking day measures). Max drinks was log-transformed and, for male subjects only, Winsorized in the left tail at three standard deviations below the mean, based on mean and standard deviation for the BIGSIB series. For the alcohol use disorder factor score (AUD-FS), we included both DSM-IV dependence and DSM-IV abuse items (excluding recurrent legal problems), consistent with the perfect genetic correlation between abuse and dependence previously reported (14) and the anticipated operationalization of alcohol dependence in DSM-V. For comparison, an

alcohol dependence factor score (AD-FS) limited to dependence items was also estimated. Factor analyses were conducted separately for the OZALC-NAG and cohorts 1 and 2 data sets; for the former, factor scoring coefficients were generated using the BIGSIB sample, by gender, and then applied to the combined OZALC-NAG data set. Factor loadings were in the range .69 to .86 for women, .63 to .92 for men, .41 to .64 for the HOD factor, and .43 to .66 for the AUD factor. Quantitative measures were adjusted for gender, age and its quadratic and cubic terms, and age*gender interaction using linear regression. Adjusted scores were then rank normalized, separately for the BIGSIB, OZALC-NAG, EDAC, and NAG, Cohort 1 only, and Cohort 2 only subsamples and combined in a single genetic association analysis. Analysis by subsample, with results combined in a meta-analysis, yielded similar results, so only the former results are reported.

Preliminary linkage analyses were conducted with MERLIN-RE-GRESS (33), using a panel of SNPs selected for high minor allele frequency and low linkage disequilibrium (LD) ($r^2 < .02$). Familybased association analyses of quantitative phenotypes were conducted using the FASTASSOC option in MERLIN (34); family-based analyses of the categorical DSM-IV alcohol dependence diagnosis were conducted using MQLS (35) as implemented in GDT (36). Results for our primary phenotypes were compared with those from the Collaborative Study on the Genetics of Alcoholism (COGA) (10), Study of Addiction: Genetics and Environment (9), and German (11) alcoholism GWAS studies. Finally, case-population control (CPC) analyses were implemented using Huber-White adjustment for familial clustering in STATA (37). We use alpha = 1.67E-7 (Bonferroni correction for 300,000 SNPs) as the threshold for genomewide significance for a single phenotype, thus alpha = 8.35E-8 for our primary HOD and AUD-FS outcome measures, allowing for testing of two phenotypes. We also report some key findings for other consumption phenotypes that would be more readily available in other data sets. We looked for consistency of evidence of genetic effects across phenotypes—between HOD and AUD-FS for each with dependence diagnosis and with other consumption measures. Finally, we used case-population control comparisons as a further check for consistency of evidence. To provide context for our findings and guidelines for replication, we conducted power calculations for power to detect genetic association, under an additive genetic model, at alpha = 5E-8, for a SNP in complete linkage disequilibrium or with specified D', with a variant, for a range of assumed effect sizes using the Genetic Power Calculator (38).

Results

Sample Characteristics

The component subsamples shared several characteristics typical of a general community sample (Table S4 in Supplement 1): 1) most alcohol-dependent cases were mild, with 70% of

Table 1. Summary of Regression-Based Multipoint Linkage Results for Primary Quantitative Phenotype AUD Factor Score

Chromosome	AUD-Factor Score		
	Location (cM)	LOD	
1	153.06	2.30	
2	112.31	2.22	
10	152.45	2.02	

Only LOD scores > 1.5 are shown. No LOD scores ≥ 1.5 were observed for HOD factor score.

AUD, alcohol use disorder; HOD, heaviness of drinking; LOD, logarithm of odds.

Table 2. Effect Size (%h² = QTL Heritability) Range for Most Strongly Associated SNPs (p < .0001) with HOD and AUD Factor Score Measures and for Secondary Consumption and AD Factor Score Measures

	n	Effect Sizes (%h²)	Lowest p Value
AUD Factor Score ^d	7490–8209	.2235	7.2E-7
AD Factor Score	7490–8209	.21–.33	8.2E-7
Heaviest Drinking Period			
HOD factor score ^d	6194-6300	.2750	1.2E-7
Frequency of any use	6411–6481	.26–.39	2.4E-6
Frequency of heavy drinking ^{a,c}	6198–6481	.26–.47	3.7E-7
Frequency drunk ^{b,c}	6017-6098	.2740	4.7E-6
Drinks per drinking day	6198–6481	.26–.40	1.5E-6
Drinks per week ^c	6204-6481	.2639	3.7E-6
Max drinks ^c	8218-8305	.2136	5.4E-7
Past 12-Month Consumption			
Frequency of any use	7665–7947	.22–.42	1.0E-7
Drinks per drinking day	7798–7910	.22–.34	1.1E-6
Drinks per week	7819–7907	.2232	3.1E-6
Max drinks	7453–7529	.2332	4.6E-6

Also given is the most extreme *p* value observed for each measure. AD, alcohol dependence; AUD, alcohol use disorder; FS, factor score; HOD, heaviness of drinking; QTL, quantitative trait locus; SNP, single nucleotide polymorphism.

^aFrequency of drinks five or more standard drinks in a day.

^bExcludes individuals who had never been drunk; these individuals were rescored as zero on this item and included in the HOD-FS analyses.

'Component of the HOD factor score.

^dPrimary phenotype results.

those meeting alcohol dependence criteria reporting only three or four dependence symptoms and fewer than 5% reporting seven dependence symptoms (not shown); and 2) a moderately high percentage of these affected individuals denied weekly drinking to intoxication and a minority denied even weekly drinking of five or more standard drinks in a single day during their 12-month period of heaviest drinking, implying that a not insignificant number experienced an episode of less than 12 months duration. Alcohol consumption histories, stratified by alcohol dependence history and gender, were comparable across subsamples, supporting the decision to combine the subsamples in a single analysis.

Alcohol Factor Measures

Neither of our primary measures gave genome-wide significant evidence for linkage (Table 1). Table 2 summarizes genetic association results for primary HOD and AUD-FS measures and for AD-FS and individual consumption measures, tabulating lowest observed p values and effect sizes (genetic variance explained by each SNP under an additive model, for SNPs with nominal associations at p < .0001 or less). No SNPs reached genome-wide significance, with lowest p values for our primary measures being 1.2E-7 for HOD and 7.2E-7 for AUD. Effect sizes were consistently small, half a percent or less. Out of the top 400 SNPs, ranked by p value for association with HOD, 65% had effect sizes of less than .25% of the variance (but greater or equal to .20%), and only 7.5% had effect sizes greater than .3%, with a median effect size of .23% (not shown), while for AUD-FS, the median effect size among the top 400 SNPs was .18%

(range .16% to 35%), with 94% of these top SNPs having effect sizes less than .25%.

Single nucleotide polymorphisms showing strongest association with HOD and AUD-FS measures are summarized in Tables S5 and S6 in Supplement 1. Additional Tables S7 and S8 in Supplement 1 show AD-FS and 12-month weekly consumption results. The small number of SNPs showing convergent evidence across phenotypes (at nominal p < .0001 for the primary phenotype and p < .005 for the confirmatory phenotype) are shown in Table S9 in Supplement 1. A chromosome 3 SNP, rs2369955, which was the third most highly associated SNP for HOD (p = 1.6E-6), also showed associations with AUD-FS (p = 1.3E-4), AD-FS (p = 7.3E-5), heaviest period frequency of heavy drinking (p = 4.7E-6), frequency drunk (p =3.9E-5), frequency of any use (p = 4.9E-5), and weekly alcohol consumption (p = 2.7E-5) measures and was modestly associated with DSM-IV alcohol dependence diagnosis (p = 1.5E-3). This SNP is intergenic but in moderate linkage disequilibrium with an intronic SNP in *TMEM108* (rs10935045, base pair 134282836: $r^2 = .49$, D' = .73), which is the fourth most highly associated SNP for HOD factor score (p = 1.7E-6) and also weakly associated with AUD-FS (p =6.4E-4). Both SNPs were only weakly associated with alcohol dependence diagnosis in case-population control comparisons (odds ration [OR] = 1.18, 95% confidence interval [CI] 1.02–1.37, p = .02; OR = 1.17, 95% CI 1.02-1.31, p = .03).

Of the remaining SNPs nominally associated with HOD at p <.0001, two intronic SNPs in the ANKS1A gene were associated with HOD, AUD-FS, AD-FS, and with AD diagnosis in CPC comparisons (rs1737727: p = 5.5E-5, 7.4E-6, 6.6E-6, and 3.4E-4, respectively; rs2140418: p = 4.4E-5, 4.3E-6, 4.1E-6, and 1.6E-4) (Table S9 in Supplement 1), with the former also associated with heaviest period weekly consumption (p = 6.3E-4), frequency of heavy drinking (p =3.2E-4), and max drinks (p = 4.6E-3). Neither was associated with AD diagnosis in family-based analyses. Additional ANKS1A SNPs show association with AUD-FS at p < .0001, as well as with HOD at p < .0001.005 (rs847851, rs847848, rs2273006 — the latter a TAF11 SNP but in complete LD with an ANKS1A SNP). An intronic SNP in CNGB3 (rs4961216) was associated with HOD (p = 2.8E-5), AUD-FS (p = 2.8E-5) 2.4E-4), binary AD in family-based analyses (p = 3.0E-4), and AD in CPC comparisons (p = 1.3E-4). Additional SNPs in USH2A, ITIH5, SHANK2, and C15orf32 showed nominal association with HOD and modest association with AUD-FS but without confirmatory association with AD diagnosis (Table S9 in Supplement 1), as did a number of nongenic SNPs. Overall, 22 of 64 SNPs nominally associated with HOD also showed association with AUD-FS at p < .005, compared with only 4 of 51 SNPs associated with 12-month weekly consumption that were also associated with AUD-FS (not shown).

Alcohol Dependence Diagnosis

Of the SNPs nominally associated with alcohol dependence diagnosis at p < .0001 in family based analyses (Table S10 in Supplement 1), four also showed association in case-population control comparisons [rs1541918, chromosome 5, p = 7.4E-6; CPC OR = 1.23 (95% CI 1.06–1.42), p = .007; rs10089021, chromosome 8, p = 4.3E-5, CPC OR = .82 (.77–.94), p = .003; rs786870, chromosome 10, p = 8.4E-5, CPC OR = .81 (.71–.02), p < .001; rs2789686, chromosome 10, p = 3.8E-5, CPC OR = .76 (.64–9.90), p < .001]. Only one of the 29 SNPs associated with alcohol dependence at p < .0001 also shows association with HOD, the intergenic chromosome 12 SNP rs2463107 SNP that is in moderate LD with intronic SNPs in the *PAWR* and *SYT1* genes. This SNP also shows predicted case-population control differences (OR = .62, 95% CI .49–.79, p = 7E-5).

Replication Failures

For associations identified by the COGA project, after correction for multiple testing, we found no SNPs or tagging SNPs that confirmed associations reported for alcohol dependence diagnosis. For BBX, we found nine SNPs that show association with one or more alcohol consumption measures, particularly rs1403774 (heaviest period drinks per drinking day, p = 9.5E-5; current drinks per drinking day, p = 1.2E-4); rs2049339 (HOD, p = 4.7E-3; heaviest period drink per week, p = 3.6E-4); and rs9875732 (heaviest period drinks per week, p = 6.6E-4; drinks per drinking day, p = 4.6E-4), with only this last SNP associated with AUD-FS (p = 7.9E-4) and with AD-FS (p = 4.5E-3). One of these, rs2049339, was in weak LD with the COGA SNP (rs10511260: $r^2 = .20$, D' = .57), with the two others having r^2 < .05. No SNPs or tagging SNPs confirmed in the German Alcoholism GWAS study were replicated in our analyses, nor were any SNPs or tagging SNPs identified as the most strongly associated SNPs in the Study of Addiction: Genetics and Environment study.

Sample Size Projections

The power of our study, because of its family-structured sampling, approximates that of a study using a random sample of 7600 individuals (a conservative figure, because it ignores the oversampling of individuals from the upper tail of the distribution of alcohol consumption). The range of effect sizes that we actually observed, of the order of .15% to .25% of the variance, implies that at least several hundred genetic variants are contributing to variation, but in a discovery sample, selecting top hits will overestimate true effect sizes. Thus, in hindsight, the achieved power of our study to detect the effect of a specific variant will be low (Table S11 in Supplement 1). Given a median effect size of .0017 for SNPs associated with AUD-FS at p < .0001, replication of a true association with 90% probability at alpha = 5E-8, in the ideal case of complete linkage disequilibrium between marker and trait variant, would require approximately 27,000 unrelated individuals, with more realistic effect sizes of .001 or .00075 requiring samples of 45,000 or 61,000. Imperfect linkage disequilibrium (D' = .9) with a modest mismatch between genetic marker allele frequency and trait allele frequency (.15 vs. .05) would increase required sample sizes approximately fourfold (e.g., 188,000 for a variant accounting for .1% of the variance).

Discussion

The primary conclusion from these analyses is that, as for many other complex phenotypes (e.g., body mass index [39]), effect sizes for the contribution of individual genetic variants to differences in heaviness of alcohol consumption and alcoholism risk are small, perhaps accounting for as little as one tenth of 1% of the variance. The approximately log-normal distribution of alcohol consumption in the general population is consistent with the hypothesis that this variation is being explained by small effects of many variants acting additively, rather than a few rare family-specific variants (40). For traits such as height (41), cumulative results do appear to support an important polygenic contribution to variation, and it seems plausible, given the many central and peripheral effects of alcohol, that this will also be true for variation in alcohol consumption and

Given the low power of genetic linkage analyses in general community samples relative to finding genetic association (42) and the accumulating evidence for small effect sizes, our failure to find genome-wide significant linkage signals is unsurprising. However, whereas findings for heaviness of drinking were uniformly negative (logarithm of odds [LOD] scores < 1.5), for AUD-FS, our second highest peak coincided with a location identified in previous alco-

holism linkage studies (chromosome 2, LOD 2.22 at 112 cM [8,43]), while a second peak (chromosome 10, LOD 2.02 at 152 cM) occurred within approximately 20 cM of the peak reported in (44). We did not find evidence for clustering of SNP associations in these regions. In association analyses, while in a few cases we found suggestive convergence for consumption versus dependence phenotypes or between inferences from family-based versus case-population control comparisons, never did results reach genome-wide significance. Of the genes noted as of interest, TMEM 108 codes for a transmembrane protein of unknown function but has previously been reported as associated with smoking cessation in a pooling GWAS study (45). SHANK2 is a scaffolding gene implicated in the formation of the postsynaptic density at glutamatergic synapses, and there have been reports of rare SHANK2 variants overrepresented in cases of autism spectrum disorder (46). For the remainder, no obvious link with alcoholism risk can be identified. In no case do we have confidence that a true positive association has been identified, and because of small effect sizes, confirmation by the first generation of alcohol GWAS studies is not to be expected.

For alcoholism, published GWAS studies have been seriously underpowered so that accumulation of many more alcohol dependent cases with GWAS will be necessary. Our secondary analyses of current (12-month) alcohol consumption measures, which had effect sizes and p values comparable with those for heaviest drinking period, also provide some limited grounds for hope that large-scale cross-study analyses will ultimately be successful in identifying some of the variants that contribute to consumption differences and thus indirectly to differences in dependence risk. Current, though not heaviest period, alcohol consumption measures will have been obtained in studies of many medical phenotypes as part of a dietary assessment (e.g., food frequency questionnaires [47]), albeit typically using truncated scales that do not well characterize individuals with highly elevated consumption levels, and in older age groups whose consumption may have declined substantially from their heaviest drinking period. Still, accumulating such data on 100,000 or more individuals (necessary to detect effect sizes of the order of .1% [one tenth of 1%] of the variance with reasonable power) would be feasible. The immediate clinical value of identifying a number of very weakly but significantly associated variants by such an approach may be low but identification of genes and pathways involved in individual variation in liability to alcohol use disorders could be great. Perhaps more narrowly defined consequences of alcohol effects (alcoholic liver disease, severe alcohol withdrawal) will give more hopeful outcomes. It is also possible, however, that work on genomic profiling using random effects modeling of genome-wide SNP data (48) will point to new directions in biological psychiatry, yielding greater understanding of the genetic contributions to individual phenotypes and enabling better quantification of genetic risk, thereby overcoming one of the primary challenges in prospective research on high-risk groups, and in resilience research, namely the inability to achieve sharp differentiation of genetic risk between groups.

There are several limitations of this research. The low-density coverage of the 370K array may have contributed to negative findings. While the family-structured sampling design that we have used remains powerful for quantitative phenotypes, this is achieved at the cost of a loss of power for alcohol dependence diagnosis: in a general community sample, the majority of such cases will be mild and cases and their unaffected siblings may differ only modestly in terms of symptom count. Second, while the study at the time it was implemented was powerful considering anticipated effect sizes, subsequent findings across many complex phenotypes suggests that it is seriously underpowered given effect

sizes that are likely to emerge for alcohol consumption and dependence outcomes. Third, our strategy of relying upon convergence of findings across consumption and dependence phenotypes (14) could cause us to miss associations that are specific to dependence. Finally, the cohorts that we used in this research mostly were raised at a time of restrictive Australian divorce practices, so that even in families with parental alcoholism, it was usual for that parent to remain in the family. While this might increase effect size, through gene-environment interaction effects, it also requires confirmation of generalizability to more contemporary cohorts.

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