

## Polygenic Risk for Alcohol Misuse is Moderated by Romantic Partnerships: Primarily in Men

| Name                  | Highest Academic degrees | Affiliations   |
|-----------------------|--------------------------|--|
| Peter B. Bar          | PhD                      | Department of Psychology, VCU  |
| Sally I-Chun Kuo      | PhD                      | Department of Psychology, VCU  |
| Fazil Aliev           | PhD                      | Department of Psychology, VCU<br>Faculty of Business, Karabuk<br>University, Turkey  |
| Antti Latvala         | PhD                      | Institute for Molecular Medicine<br>Finland, University of Helsinki<br>Department of Public Health,<br>University of Helsinki          |
| Richard Viken         | PhD                      | Department of Psychological and<br>Brain Sciences, Indiana<br>University, Bloomington  |
| Richard J. Rose       | PhD                      | Department of Psychological and<br>Brain Sciences, Indiana<br>University, Bloomington  |
| Jaakko Kaprio         | PhD, MD                  | Institute for Molecular Medicine<br>Finland, University of Helsinki<br>Department of Public Health,<br>University of Helsinki, Finland |
| Jessica E. Salvatore* | PhD                      | Department of Psychology, VCU<br>Virginia Institute for Psychiatric<br>and Behavioral Genetics, VCU                                    |
| Danielle M. Dick*     | PhD                      | Department of Psychology, VCU  |

\*joint last authors

### Corresponding Authors:

Peter Barr, PhD  
Department of Psychology, Virginia Commonwealth University  
8 North Harrison St.  
Richmond, VA. 23284  
(804) 828-1266  
pbarr2@vcu.edu

Danielle M. Dick, PhD  
Department of Psychology, Virginia Commonwealth University  
800 W. Franklin Street  
Box 842018  
Richmond, VA 23284-2018  
(804) 828-8756  
ddick@vcu.edu

Word count: 2,983

## Abstract

**Importance:** Problematic alcohol use remains a leading influence on preventable mortality and morbidity across the globe. Those in committed relationships consistently report lower levels of alcohol misuse and problems.

**Objective:** To determine 1) whether genetic risk for alcohol misuse is moderated by romantic relationships (gene-environment interaction; GxE), and 2) whether GxE results are consistent across sex.

**Design:** Data came from the young adult wave of the Finnish Twin Study (FinnTwin12), a nationally representative sample of twins. Predictors included genome-wide polygenic scores (GPS), derived from a recent genome-wide association study (GWAS) of alcohol consumption in ~1 million participants; and participant reports of relationship status.

**Setting:** Finland

**Participants:** An intensively studied subset of FinnTwin12 received a diagnostic interview during the young adult phase (1,312 of 1,347 individuals provided genotypic data). The analytic sample includes those with complete interview and genetic data (N=1,201, 54% female).

**Exposure:** Self-reported involvement in a romantic partnership.

**Main Outcomes and Measures:** Drinking frequency, intoxication frequency, and DSM-IV alcohol dependence (AD) symptoms from a diagnostic interview.

**Results:** GPS predicted drinking frequency ( $b = 0.109$ ; 95% CI = 0.051, 0.167), intoxication frequency ( $b = 0.111$ ; 95% CI = 0.054, 0.168), and AD symptoms ( $b = 0.123$ ; 95% CI = 0.064, 0.182). Relationship moderated the association between GPS and drinking frequency ( $b = -0.105$ ; 95% CI = -0.211, -0.001), intoxication frequency ( $b = -0.118$ ; 95% CI = -0.220, -0.015), and AD symptoms ( $b = -0.119$ ; 95% CI = -0.229, -0.010). The interaction for drinking frequency was not significant after correcting for covariates. There was a 3-way interaction between sex, relationship status, and GPS for intoxication frequency ( $b = 0.223$ ; 95% CI = 0.014, 0.432), with the two-way interaction of relationship status and PRS on intoxication frequency being significant only in men.

**Conclusions and Relevance:** Being in a relationship reduced the association between genetic predisposition and high risk drinking. Part of the protective effect of committed partnerships on alcohol misuse observed in epidemiological research may be in limiting genetic liability. However, this protective effect was largely limited to males, mapping onto earlier findings suggesting that males benefit more from romantic partnerships.

## Key Points

**Question:** Do romantic relationships moderate polygenic risk on alcohol misuse in young adulthood?

**Findings** Involvement in romantic relationships moderated the polygenic risk on frequency of intoxication and DSM-IV alcohol dependence symptoms, such that polygenic associations with alcohol misuse were stronger among those not in a romantic relationship. Males experienced a stronger protective effect of romantic relationship in limiting the manifestation of genetic predispositions toward intoxication frequency.

**Meaning:** The interplay between genes and environment is important in understanding etiology of problematic alcohol use, and romantic relationships appear to buffer genetic risk for alcohol misuse in young adulthood. Findings underscore how social relationships may alter the risk posed by genetic predispositions.

Alcohol use is one of the leading contributors to preventable mortality and morbidity, worldwide.<sup>1-3</sup> Twin and family studies indicate that genetic influences account for approximately 50% of the variation in the population<sup>4</sup>; however, there is strong evidence the importance of genetic influences changes across environmental contexts (gene-environment interaction, or GxE).<sup>5,6</sup> Environments that allow greater access to alcohol, or acceptance of alcohol use, may create opportunity for increased manifestation of individual predispositions toward alcohol misuse and consequently the development of problems.<sup>7-11</sup> Conversely, environments that exert more social control, such as greater parental monitoring in adolescence, appear to reduce the importance of genetic predispositions.<sup>7,12</sup> Mapping which environments reduce alcohol misuse among those at greater genetic risk will be critical for developing tailored prevention intervention strategies as we move into an era of precision medicine.

Much of the foundational work on GxE in alcohol outcomes has been conducted in twin studies.<sup>6-9,12</sup> Most GxE studies to date using measured genotypes on alcohol use outcomes have focused on candidate genes or single nucleotide polymorphisms (SNP), where the effect of a specific candidate gene or single SNP varies as a function of the environment.<sup>6</sup> However, candidate gene research has generated inconsistent results, likely a reflection of being underpowered to robustly detect moderations, false positives, and publication bias.<sup>13,14</sup> Furthermore, the use of single genes in GxE studies does not align with our current molecular genetic understanding that complex behaviors, including alcohol use,<sup>15</sup> problems,<sup>16</sup> and dependence,<sup>17</sup> have a polygenic architecture, driven by many genetic variants of very small effect.<sup>18,19</sup> Large sample sizes are needed to detect robust genetic associations for complex behavioral outcomes in genome-wide association studies (GWAS), which use data from the entire genome rather than relying on predefined SNPs.<sup>20,21</sup>

To characterize individual risk across hundreds or thousands of alleles associated with an outcome in a GWAS, genome-wide polygenic risk scores (GPS) have emerged as a way to aggregate this information into a single score. As we begin to identify GPS robustly associated with substance use and dependence, one of the critical next steps toward precision medicine will be to characterize the pathways by which risk unfolds.<sup>22</sup> For alcohol-related outcomes, this will necessitate characterizing how specific environments moderate the likelihood that individuals carrying risky genetic predispositions will develop excessive use, problems, and dependence, providing important information about targeted areas for intervention.

In this study, we focused on romantic relationships, as epidemiological research has consistently shown that those in committed relationships (especially marriage) engage in fewer risky or health-deteriorating behaviors, such as alcohol misuse.<sup>23,24</sup> This reduction in risky behaviors is due in part to increased social control and monitoring associated with being in a relationship,<sup>23</sup> as well as individuals' motivation to align their behavior with the social expectations typically associated with the spousal role.<sup>25,26</sup> However, marriage-like relationships are generally beneficial for men but more or less indifferent for women,<sup>27</sup> suggesting important sex-differences in any protective effect. Finally, twin studies have found that the heritability of alcohol consumption is decreased among individuals in committed relationships,<sup>28,29</sup> suggesting that being with a partner may act as a "social control" that limits expression of genetic predispositions toward alcohol problems.

Here, we test this hypothesis using molecular genetic data in a population-based sample of young adults.<sup>30</sup> We focused on young adulthood because it is a critical period for the development of alcohol use patterns and problems,<sup>28</sup> with heavy alcohol use at its highest point<sup>31</sup> and the peak age of onset for alcohol related disorders falling during this period.<sup>32</sup> We used

results from the largest mega-analysis to date on alcohol consumption (drinks per week in ~1 million individuals),<sup>15</sup> to calculate genome-wide polygenic scores in our independent, population-based sample. We tested the association of these polygenic risk scores with alcohol use, heavy consumption, and alcohol problems, and importantly, whether being in a romantic relationship changed the association between genetic risk and alcohol outcomes. Finally, because there are sex differences in patterns of alcohol use and in the prevalence of alcohol use disorders<sup>32</sup> and heavy consumption,<sup>31</sup> we examined whether there were sex differences in GxE.<sup>33</sup>

## Methods

### *Sample*

Data come from the youngest cohort of the Finnish Twin Cohort Study (FinnTwin12). Families were identified from Finland's Population Registry, permitting comprehensive nationwide ascertainment for twins born from 1983 to 1987. Baseline collection occurred when twins were approximately 12 years old, with a sample of approximately 5600 twins (87% participation) and their families.<sup>30</sup> Follow-up surveys occurred at ages 14, 17.5, and during young adulthood (age range 20-26). Twin zygosity was determined using items developed for twin children.<sup>34</sup> Confirmation by multiple genetic markers revealed that 97% of same-sex pairs retained the original questionnaire-based zygosity classification<sup>35</sup>. The Helsinki University Central Hospital District's Ethical Committee and Indiana University's Institutional Review Board approved the FinnTwin12 study. Of those in the larger sample, a subset of intensively studied individuals also received in-depth clinical interviews (N = 1,347) and participated in DNA collection as young adults. In the present study, we limited our analyses to those who had complete information on all relevant study variables and who had initiated alcohol use (n =

1,201). The analytic subset did not differ significantly from the full sample in terms of demographic characteristics or alcohol misuse (see supplemental information for more detail).

### *Genotyping and Quality Control*

Genotyping was conducted using the Human670-QuadCustom Illumina BeadChip at the Wellcome Trust Sanger Institute.<sup>36</sup> Quality control steps included removing SNPs with minor allele frequency (MAF) <1%, genotyping success rate <95%, or Hardy-Weinberg equilibrium  $p < 1 \times 10^{-6}$ , and removing individuals with genotyping success rate <95%, a mismatch between phenotypic and genotypic gender, excess relatedness (outside of known families), and heterozygosity outliers. Genotypes were imputed to the 1,000 Genomes Phase 3 reference panel<sup>37</sup> reference panel using ShapeIT<sup>38</sup> for phasing and IMPUTE2<sup>39</sup> for imputation, resulting in 13,688,418 autosomal SNPs for analyses. Prior analyses indicated a single dimension of ancestry in the sample.<sup>40</sup> Although a single dimension of ancestry does not preclude variation along this dimension, we note that fine-scale population substructure is less of an issue for common variants (vs. rare variants), especially in the present sample given the relatively longer LD blocks that make the Finnish population more homogenous than other populations of mixed European ancestry.

### *Measures*

*Alcohol-Related Behaviors* were assessed across increasing levels of severity. Drinking frequency was measured by asking "How often do you use alcohol?" Responses included "never" (0), "once a year" (1), 2-4 times a year (2), "every other month" (3), "once a month" (4), "more than once a month" (5), "once a week" (6), "more than once a week" (7), and "daily" (8). Intoxication frequency was assessed by asking "How often do you use alcohol in such a way that you get really drunk?" Responses were the same for drinking frequency. We transformed these

ordinal measures into pseudo-continuous measures of the frequency of these behaviors in the past 30 days.<sup>41,42</sup> Finally, we included a count of lifetime DSM-IV Alcohol Dependence (AD) symptoms, assessed using the Semi-Structured Assessment for the Genetics of Alcoholism (SSAGA), a reliable and valid clinical instrument.<sup>43</sup> Each alcohol measure was log transformed (left anchored at 1) to adjust for positive skew. *Relationship Status* was measured by asking, “How long (in years) have you been together with your present partner?” Respondents that indicated they were not in a relationship were coded as 0. Those who indicated they were in a romantic relationship for any length were coded as 1. We ran sensitivity analyses with a stricter definition of relationship status (those in a relationship  $\geq 2$  years). Our results did not fundamentally differ from the more inclusive definition and we retained the original measurement of relationship status. Finally, we included age, sex, educational attainment (based on the Finnish education system: basic education; vocational training; secondary education; tertiary education), and whether or not respondents were still in school<sup>44</sup> (dichotomous yes/no) as covariates.

#### *Genome-wide Polygenic Scores*

We created polygenic scores derived from a large-scale GWAS of number of alcoholic drinks per week in approximately one million individuals<sup>15</sup> provided by the GWAS & Sequencing Consortium of Alcohol and Nicotine Use (GSCAN). As FinnTwin12 was included in the original discovery GWAS, we obtained summary statistics with all Finnish participants, including FinnTwin12 participants removed. There were 3,707,235 autosomal SNPs in common after QC. We used the well-established process of clumping and thresholding.<sup>45</sup> SNPs from the discovery GWAS were clumped based on linkage disequilibrium (LD) using the *clump* procedure in PLINK,<sup>46</sup> based on an  $R^2 = .25$ , with a 500kb window, resulting in 407,604

independent SNPs for creating scores. We then created scores based on differing thresholds of GWAS p-values ( $p < .0001$ ,  $p < .001$ ,  $p < .01$ ,  $p < .05$ ,  $p < .10$ ,  $p < .20$ ,  $p < .30$ ,  $p < .40$ ,  $p < .50$ ). We converted GPS to Z-scores for interpretation.

We note that alcohol consumption and problematic use, though highly correlated, have distinct genetic influences.<sup>47</sup> We ran a series of sensitivity analyses to determine if recent GWAS focused on alcohol problems or dependence<sup>16,17</sup> provided better assessments of genetic liability for alcohol misuse (see supplemental information). However, in each case, the scores derived from GSCAN were the most predictive.

#### *Analytic Strategy*

First, we estimated the effect of GPS across each p-value threshold to determine the most predictive score (based on model  $R^2$ ) for each alcohol phenotype. We then tested whether relationship status moderated the association of the genome-wide polygenic scores. In the instances where we found evidence for a significant interaction, we fit a more robust model for evaluating G×E,<sup>48</sup> which includes all G × covariate and E × covariate interaction terms. Finally, we tested for sex-specific G×E by including a three-way interaction term. We determined whether estimates were significant using an  $\alpha$  of  $p < .05$  (two-sided test). Because the FinnTwin12 data is a family-based data set, we evaluated all hypotheses using a linear mixed model with random intercepts for each family in the *lme4*<sup>49</sup> package in *R* 3.5.1.<sup>50</sup> We estimate effect size ( $R^2$ ) using a method designed for mixed effects models<sup>51</sup> with the *MuMIn* package.<sup>52</sup>

## **Results**

Males exhibited higher mean levels of each alcohol measure (Table 1). The alcohol phenotypes were also modestly correlated ( $r_{\text{drinking}^* \text{intox}} = .64$ ,  $r_{\text{drinking}^* \text{ADsx}} = .37$ ,  $r_{\text{intox}^* \text{ADsx}} = .43$ ),

with stronger correlations between the consumption items than with the measure of AD symptoms.<sup>47</sup>

#### *Polygenic Score Performance*

Figure 1 provides the incremental R-squared for polygenic scores at different p-value inclusion thresholds. The variance explained at each p-value threshold in GPS represents the change in R-squared from the baseline model (age and sex as covariates) after including the GPS at that p-value threshold. GPS were significantly associated with each alcohol related behavior across almost all of the p-value thresholds, with the exception of the most restrictive scores in relation to drinking frequency. GPS explained more variance as p-value thresholds became more inclusive, peaking and leveling off at thresholds between  $p < .20$  and  $p < 0.50$ . We decided to use the most liberal threshold ( $p < .50$ ) for all models going forward.

In order to ensure the GPS were predictive of alcohol problems above and beyond levels of consumption (which are genetically correlated but distinct phenotypes),<sup>47</sup> we estimated the effect of GPS while accounting for either drinking or intoxication frequency. GPS were significantly related to AD symptoms after statistically controlling for drinking frequency ( $b = 0.085$ ,  $p < .01$ ) or intoxication frequency ( $b = 0.075$ ,  $p < .01$ ; see supplemental information for full results). Finally, we estimated the polyserial correlation between GPS and relationship status ( $\rho = 0.005$ ,  $p > .05$ ) to assess the possibility of gene-environment correlation.

#### *Main Effects of Polygenic Score and Relationship Status*

Table 2 provides the estimates for the linear mixed models evaluating the joint effect of GPS and relationship status. In the model for main effects (Model 1), those currently in a relationship had lower levels of intoxication frequency, but not drinking frequency or AD symptoms. GPS remained significantly associated with each of these alcohol related behaviors.

### *Gene-Environment Interaction Models*

Model 2 (Table 2) presents the estimates for G×E. There was a significant interaction between relationship status and polygenic scores for each alcohol behavior. We refit each of these models with interactions between relationship status and each covariate and interactions between GPS and each covariate (plotted in Figure 2, see supplemental information for full results) to account for possible confounding.<sup>48</sup> P-values were attenuated, especially in the models for drinking frequency and AD symptoms, but the nature of the interactions remained unchanged for the other phenotypes. The shape of the interaction was similar across all phenotypes, but most pronounced for intoxication. In the case of intoxication frequency, there was a stronger association between genetic risk score and intoxication frequency among individuals who are not in romantic relationships, and a relatively weaker association between genetic risk score and intoxication frequency among those who were in romantic relationships.

### *Sex Differences in G×E*

Finally, we tested for sex differences in the interaction between relationship status and GPS. We found no evidence of a significant three-way interaction between sex, relationship status, and GPS for either drinking frequency or AD symptoms. However, we did find a significant three-way interaction in the models for intoxication frequency. This interaction remained significant even after adjusting for possible confounding in the G×E interactions. Figure 3 displays the predicted values from this model. For intoxication frequency, the G×E effect appears to be driven by the effect in males.

## **Discussion**

We tested whether polygenic risk scores derived from a meta-analysis of alcohol consumption were associated with alcohol outcomes in an independent, population-based young

adult sample, whether romantic relationship status moderated the association of genetic predispositions with alcohol outcomes, and whether observed effects varied between females and males.

Polygenic scores derived from variants associated with consumption are predictive of use, misuse, and problems among young adults. As hypothesized, being in a romantic relationship moderated the association between GPS and each alcohol phenotype (drinking frequency, intoxication frequency and AD symptoms). Similar to previous twin research,<sup>53,54</sup> among individuals with elevated genetic predisposition, levels of misuse were lower in those in a romantic partnership. We posit that the constraints and responsibilities placed on individuals within romantic partnerships limits their ability to express underlying predispositions towards alcohol misuse, fitting with the social control model of gene-environment interaction.<sup>23,55</sup> Additional inspection (available in supplemental information) revealed these interactions did not appear to be driven by outliers at either end of the distribution.

Finally, we examined whether there were sex differences in these G×E effects. We found no evidence of sex differences in the G×E effect for drinking frequency or AD symptoms. However, the G×E effect for intoxication frequency was driven primarily by the effect in males. Previous work in social epidemiology has documented how males tend to “over-benefit” from relationships in terms of health.<sup>27</sup> This may reflect the tendency for women in relationships to be the emotional and social support providers, of which men are the receivers.<sup>56</sup> In the current study, we see that this effect may be due in part to limiting genetic liability among a riskier drinking group (see supplemental information for additional analyses). This difference does not appear in AD symptoms, which is likely the result of these symptoms measuring aspects of both

consumption and problems. Any role relationship status has in limiting genetic liability seems limited to levels of heavy consumption.

Our findings have important practical implications for researchers and clinicians interested in those at greater risk for alcohol misuse. First, the signal for genetic associations may be drastically reduced in young adults in a committed relationship. Future research on gene identification efforts may benefit from the inclusion of important environmental information in order to increase power to detect genetic variants associated with various forms of alcohol misuse. Considering G×E in the discovery GWAS may be of even more importance in regards to alcohol use phenotypes, as there is consistent evidence of G×E from twin studies.<sup>41,53,57,58</sup> For clinicians, these analyses point to committed relationships as a malleable environmental condition that may help reduce individuals' level of misuse, in part, by limiting realization of genetic predisposition. Although gene-environment correlation is always an important consideration, we note that our GPS was uncorrelated with relationship status, consistent with previous research using more causally-identified designs.<sup>59</sup>

This research has several limitations. First, although the polygenic scores explained more variance in these outcomes than previous iterations using smaller discovery GWAS, the variance explained by the largest meta-analysis of alcohol consumption to date, compiling data from ~1 million individuals, continued to be small ( $R^2 \sim .015$ ), especially compared to other complex phenotypes with similar sample sizes, like education attainment ( $R^2 \sim .12$ )<sup>60</sup>. Discovery samples with better phenotyping will be necessary to create scores that explain the total SNP-based heritability. Second, though we found evidence of G×E, it does not rule out other confounding factors. Larger twin samples with genotypic data that allow for within-family designs will help to further account for possible environmental confounders shared across families (e.g.

neighborhood factors, religiosity, socioeconomic status; see supplemental for sensitivity analyses). Third, our measure of romantic partnerships did not examine which relationship characteristics moderate polygenic scores (e.g. relationship quality, partner's drinking, emotional support). Finally, our measure of AD symptoms was a lifetime measure. Supplemental analyses revealed similar patterns between lifetime and past 12-month symptoms.

In conclusion, polygenic scores from a large-scale GWAS of drinks per week predicted levels of alcohol use and misuse among a sample of young adults. However, the likelihood an individual carrying riskier genetic predispositions would display problematic patterns of use changed as a function of the environment. Individuals at greater genetic risk who were in romantic relationships were less likely to misuse alcohol. For drinking to intoxication, this interaction appears to occur primarily among males. This finding is consistent with previous research on social determinants of health that men tend to over-benefit from romantic partnerships.<sup>27</sup> This research underscores the importance of considering the interplay between genes and environment when considering etiology and intervention for problematic alcohol use. In order for genetic risk scores to be useful in clinical settings, we must understand how genetic risk interacts with the environment.

## Acknowledgments

Research reported in this publication was supported by the National Institute on Alcohol Abuse and Alcoholism of the National Institutes of Health under award numbers R01AA015416, K02AA018755, K01AA024152, and F32AA022269; the Academy of Finland (grants 100499, 205585, 118555, 141054, 265240, 263278, and 264146); and the Scientific and Technological Research Council of Turkey (TÜBİTAK) under award number 114C117. The content is solely the responsibility of the authors and does not necessarily represent the official views of the National Institutes of Health, the Academy of Finland, or the Scientific and Technological Research Council of Turkey. The authors have no conflict of interests to report.

## References

1. Gakidou E, Afshin A, Abajobir AA, et al. Global, regional, and national comparative risk assessment of 84 behavioural, environmental and occupational, and metabolic risks or clusters of risks, 1990–2016: a systematic analysis for the Global Burden of Disease Study 2016. *The Lancet*. 2017;390(10100):1345-1422.
2. The USBoDC. The state of us health, 1990-2016: Burden of diseases, injuries, and risk factors among us states. *JAMA*. 2018;319(14):1444-1472.
3. World Health Organization. *Global status report on alcohol and health - 2018*. Geneva, Switzerland2018.
4. Verhulst B, Neale MC, Kendler KS. The heritability of alcohol use disorders: a meta-analysis of twin and adoption studies. *Psychological Medicine*. 2015;45(5):1061-1072.
5. Shanahan MJ, Hofer SM. Social Context in Gene–Environment Interactions: Retrospect and Prospect. *The Journals of Gerontology: Series B*. 2005;60(Special\_Issue\_1):65-76.
6. Young-Wolff KC, Enoch M-A, Prescott CA. The influence of gene–environment interactions on alcohol consumption and alcohol use disorders: A comprehensive review. *Clinical Psychology Review*. 2011;31(5):800-816.
7. Dick DM, Pagan JL, Viken R, et al. Changing Environmental Influences on Substance Use Across Development. *Twin Research and Human Genetics*. 2007;10(2):315-326.
8. Harden KP, Hill JE, Turkheimer E, Emery RE. Gene-Environment Correlation and Interaction in Peer Effects on Adolescent Alcohol and Tobacco Use. *Behavior Genetics*. 2008;38(4):339-347.
9. Cooke ME, Meyers JL, Latvala A, et al. Gene–Environment Interaction Effects of Peer Deviance, Parental Knowledge and Stressful Life Events on Adolescent Alcohol Use. *Twin Research and Human Genetics*. 2015;18(5):507-517.

10. Olfson E, Edenberg HJ, Nurnberger J, et al. An ADH1B Variant and Peer Drinking in Progression to Adolescent Drinking Milestones: Evidence of a Gene-by-Environment Interaction. *Alcoholism: Clinical and Experimental Research*. 2014;38(10):2541-2549.
11. Virtanen S, Kaprio J, Viken R, Rose RJ, Latvala A. Birth cohort effects on the quantity and heritability of alcohol consumption in adulthood: A Finnish longitudinal twin study. *Addiction*. 2018;0(ja).
12. Miles DR, Silberg JL, Pickens RW, Eaves LJ. Familial influences on alcohol use in adolescent female twins: testing for genetic and environmental interactions. *J Stud Alcohol*. 2005;66(4):445-451.
13. Duncan LE, Keller MC. A critical review of the first 10 years of candidate gene-by-environment interaction research in psychiatry. *Am J Psychiatry*. 2011;168.
14. Dick DM, Agrawal A, Keller MC, et al. Candidate Gene–Environment Interaction Research: Reflections and Recommendations. *Perspectives on Psychological Science*. 2015;10(1):37-59.
15. Liu M, Jiang Y, Wedow R, et al. Association studies of up to 1.2 million individuals yield new insights into the genetic etiology of tobacco and alcohol use. *Nature Genetics*. 2019.
16. Sanchez-Roige S, Palmer AA, Fontanillas P, et al. Genome-Wide Association Study Meta-Analysis of the Alcohol Use Disorders Identification Test (AUDIT) in Two Population-Based Cohorts. *American Journal of Psychiatry*. 2018:appi.ajp.2018.18040369.
17. Walters RK, Polimanti R, Johnson EC, et al. Transancestral GWAS of alcohol dependence reveals common genetic underpinnings with psychiatric disorders. *Nature Neuroscience*. 2018;21(12):1656-1669.
18. Plomin R, Haworth CMA, Davis OSP. Common disorders are quantitative traits. *Nature Reviews Genetics*. 2009;10:872.
19. Sullivan PF, Daly MJ, O'Donovan M. Genetic architectures of psychiatric disorders: the emerging picture and its implications. *Nature Reviews Genetics*. 2012;13:537.

20. Bogdan R, Baranger DAA, Agrawal A. Polygenic Risk Scores in Clinical Psychology: Bridging Genomic Risk to Individual Differences. *Annual Review of Clinical Psychology*. 2018;14(1):119-157.
21. Dudbridge F. Polygenic Epidemiology. *Genetic Epidemiology*. 2016;40(4):268-272.
22. Dick DM, Barr PB, Cho SB, et al. Post-GWAS in Psychiatric Genetics: A Developmental Perspective on the “Other” Next Steps. *Genes, Brain and Behavior*. 2018;17(3):e12447.
23. Umberson D, Crosnoe R, Reczek C. Social Relationships and Health Behavior Across Life Course. *Annual Review of Sociology*. 2010;36:139-157.
24. Staff J, Schulenberg JE, Maslowsky J, et al. Substance use changes and social role transitions: Proximal developmental effects on ongoing trajectories from late adolescence through early adulthood. *Development and Psychopathology*. 2010;22(4):917-932.
25. Horn EE, Xu Y Fau - Beam CR, Beam Cr Fau - Turkheimer E, Turkheimer E Fau - Emery RE, Emery RE. Accounting for the physical and mental health benefits of entry into marriage: a genetically informed study of selection and causation. *Journal of Family Psychology*. 2013;27(1):30-41.
26. Yamaguchi K, Kandel DB. On the Resolution of Role Incompatibility - a Life Event History Analysis of Family Roles and Marijuana Use. *American Journal of Sociology*. 1985;90(6):1284-1325.
27. Kiecolt-Glaser JK, Newton TL. Marriage and health: His and hers. *Psychological Bulletin*. 2001;127(4):472-503.
28. Barr PB, Salvatore JE, Maes HH, et al. Social Relationships Moderate Genetic Influences on Heavy Drinking in Young Adulthood. *Journal of Studies on Alcohol and Drugs*. 2017;78(6):817-826.
29. Heath AC, Jardine R, Martin NG. Interactive effects of genotype and social environment on alcohol consumption in female twins. *Journal of Studies on Alcohol*. 1989;50(1):38-48.

30. Kaprio J. The Finnish twin cohort study: an update. *Twin Research and Human Genetics*. 2013;16(01):157--162.
31. Chen P, Jacobson KC. Developmental trajectories of substance use from early adolescence to young adulthood: gender and racial/ethnic differences. *Journal of Adolescent Health*. 2012;50(2):154-163.
32. Kessler RC, Berglund P, Demler O, Jin R, Merikangas KR, Walters EE. Lifetime prevalence and age-of-onset distributions of dsm-iv disorders in the national comorbidity survey replication. *Archives of General Psychiatry*. 2005;62(6):593-602.
33. Salvatore JE, Cho SB, Dick DM. Genes, Environments, and Sex Differences in Alcohol Research. *Journal of studies on alcohol and drugs*. 2017;78(4):494-501.
34. Hill Goldsmith H. A zygosity questionnaire for young twins: A research note. *Behavior Genetics*. 1991;21(3):257-269.
35. Knaapila A, Silventoinen K, Broms U, et al. Food Neophobia in Young Adults: Genetic Architecture and Relation to Personality, Pleasantness and Use Frequency of Foods, and Body Mass Index—A Twin Study. *Behavior Genetics*. 2011;41(4):512-521.
36. Kaprio J. The Finnish Twin Cohort Study: an update. *Twin Res Hum Genet*. 2013;16(1):157-162.
37. The 1000 Genomes Project Consortium. A global reference for human genetic variation. *Nature*. 2015;526:68.
38. Delaneau O, Zagury J-F, Marchini J. Improved whole-chromosome phasing for disease and population genetic studies. *Nature Methods*. 2012;10:5.
39. Howie BN, Donnelly P, Marchini J. A Flexible and Accurate Genotype Imputation Method for the Next Generation of Genome-Wide Association Studies. *PLOS Genetics*. 2009;5(6):e1000529.
40. Meyers JL. *Elucidating genetic and environmental influences on alcohol-related phenotypes*. ProQuest Dissertations & Theses Global, Virginia Commonwealth University; 2012.

41. Cooke ME, Meyers JL, Latvala A, et al. Gene-Environment Interaction Effects of Peer Deviance, Parental Knowledge and Stressful Life Events on Adolescent Alcohol Use. *Twin Research and Human Genetics*. 2015;18(5):507-517.
42. Barr PB, Salvatore JE, Maes HH, et al. Education and alcohol use: A study of gene-environment interaction in young adulthood. *Social Science & Medicine*. 2016;162:158-167.
43. Bucholz KK, Cadoret R, Cloninger CR, et al. A new, semi-structured psychiatric interview for use in genetic linkage studies: a report on the reliability of the SSAGA. *Journal of Studies on Alcohol*. 1994;55(2):149-158.
44. Timberlake DS, Hopfer CJ, Rhee SH, et al. College attendance and its effect on drinking behaviors in a longitudinal study of adolescents. *Alcoholism, clinical and experimental research*. 2007;31(6):1020-1030.
45. International Schizophrenia Consortium. Common polygenic variation contributes to risk of schizophrenia and bipolar disorder. *Nature*. 2009;460(7256):748-752.
46. Purcell S, Neale B, Todd-Brown K, et al. PLINK: a tool set for whole-genome association and population-based linkage analyses. *Am J Hum Genet*. 2007;81(3):559-575.
47. Dick DM, Meyers JL, Rose RJ, Kaprio J, Kendler KS. Measures of Current Alcohol Consumption and Problems: Two Independent Twin Studies Suggest a Complex Genetic Architecture. *Alcoholism: Clinical and Experimental Research*. 2011;35(12):2152-2161.
48. Keller MC. Gene x environment interaction studies have not properly controlled for potential confounders: the problem and the (simple) solution. *Biol Psychiatry*. 2014;75(1):18-24.
49. Bates D, Maechler M, Bolker B, Walker S. Fitting Linear Mixed-Effects Models Using lme4. *Journal of Statistical Software*. 2015;67(1):1-48.
50. *A language and environment for statistical computing* [computer program]. Vienna, Austria: R Foundation for Statistical Computing; 2017.

51. Nakagawa S, Schielzeth H, O'Hara RB. A general and simple method for obtaining R-squared from generalized linear mixed-effects models. *Methods in Ecology and Evolution*. 2013;4(2):133-142.
52. Bartoń K. MuMIn: Multi-Model Inference. R package version 1.40.4.; 2018.
53. Barr PB, Salvatore JE, Maes HH, et al. Social Relationships Moderate Genetic Influences on Heavy Drinking in Young Adulthood. *Journal of studies on alcohol and drugs*. 2017;78(6):817-826.
54. Heath AC, Jardine R, Martin NG. Interactive effects of genotype and social environment on alcohol consumption in female twins. *Journal of Studies on Alcohol*. 1989;50(1):38--48.
55. Shanahan MJ, Hofer S, M. Social context in gene--environment interactions: Retrospect and prospect. *The Journals of Gerontology Series B: Psychological Sciences and Social Sciences*. 2005;60(Special Issue 1):65--76.
56. Kawachi I, Berkman LF. Social ties and mental health. *Journal of Urban Health*. 2001;78(3):458-467.
57. Dick DM, Bernard M, Aliev F, et al. The role of socioregional factors in moderating genetic influences on early adolescent behavior problems and alcohol use. *Alcoholism: Clinical and Experimental Research*. 2009;33(10):1739-1748.
58. Dick DM, Viken R, Purcell S, Kaprio J, Pulkkinen L, Rose RJ. Parental monitoring moderates the importance of genetic and environmental influences on adolescent smoking. *Journal of Abnormal Psychology*. 2007;116(1):213-218.
59. Barnes JC, Beaver KM. Marriage and Desistance From Crime: A Consideration of Gene-Environment Correlation. *Journal of Marriage and Family*. 2012;74(1):19-33.
60. Lee JJ, Wedow R, Okbay A, et al. Gene discovery and polygenic prediction from a genome-wide association study of educational attainment in 1.1 million individuals. *Nature Genetics*. 2018;50(8):1112-1121.

## Figure Captions

### Figure 1: Predictive Power of GSCAN Polygenic Scores

Change in model R2 from base model (age and sex as covariates) to model including polygenic scores at various p-value inclusion thresholds (determined by p-value from discovery GWAS).

\* association  $p < .05$ .

### Figure 2: Gene-Environment Interaction across Relationship Status and Polygenic Risk

Predicted values of each alcohol phenotype (standardized) across the range of polygenic scores for those in a relationship (blue) and those not in a relationship (red). Shaded areas represent 95% pointwise confidence intervals of estimates.

### Figure 3: Sex Differences in GxE for Intoxication Frequency

Predicted values of intoxication frequency (standardized) across the range of polygenic scores and sex for those in a relationship (blue) and those not in a relationship (red). Shaded areas represent 95% pointwise confidence intervals of estimates.

## Tables

Table 1: Descriptive Statistics for Finnish Twin Study (FinnTwin12)

|                                    | Males (N = 551) |       | Females (N = 650) |       | $\chi^2$ / t-test |
|------------------------------------|-----------------|-------|-------------------|-------|-------------------|
|                                    | Mean/N          | SD/%  | Mean/N            | SD/%  |                   |
| Drinking Frequency                 | 5.10            | 4.54  | 3.43              | 3.24  | *                 |
| Intoxication Frequency             | 2.02            | 1.97  | 1.12              | 1.44  | *                 |
| DSM-IV Alcohol Dependence Symptoms | 1.54            | 1.35  | 1.29              | 1.37  | *                 |
| GPS <sup>†</sup>                   | -0.03           | 1.02  | 0.03              | 0.98  |                   |
| Age                                | 21.94           | 0.77  | 21.95             | 0.76  |                   |
| Educational Attainment             |                 |       |                   |       | *                 |
| Basic Education                    | 38              | 6.9%  | 30                | 4.6%  |                   |
| Vocational Training                | 207             | 37.6% | 157               | 24.2% |                   |
| Secondary Education                | 299             | 54.3% | 424               | 65.2% |                   |
| Tertiary Education                 | 7               | 1.3%  | 39                | 6.0%  |                   |
| Enrolled in school                 | 285             | 51.7% | 401               | 61.7% | *                 |
| In relationship                    | 269             | 48.8% | 416               | 64.0% | *                 |

\*p < .05 for Chi-square/T-test difference between males and females

<sup>†</sup>Standardized (Z-scores) GPS including SNPs with p < .50

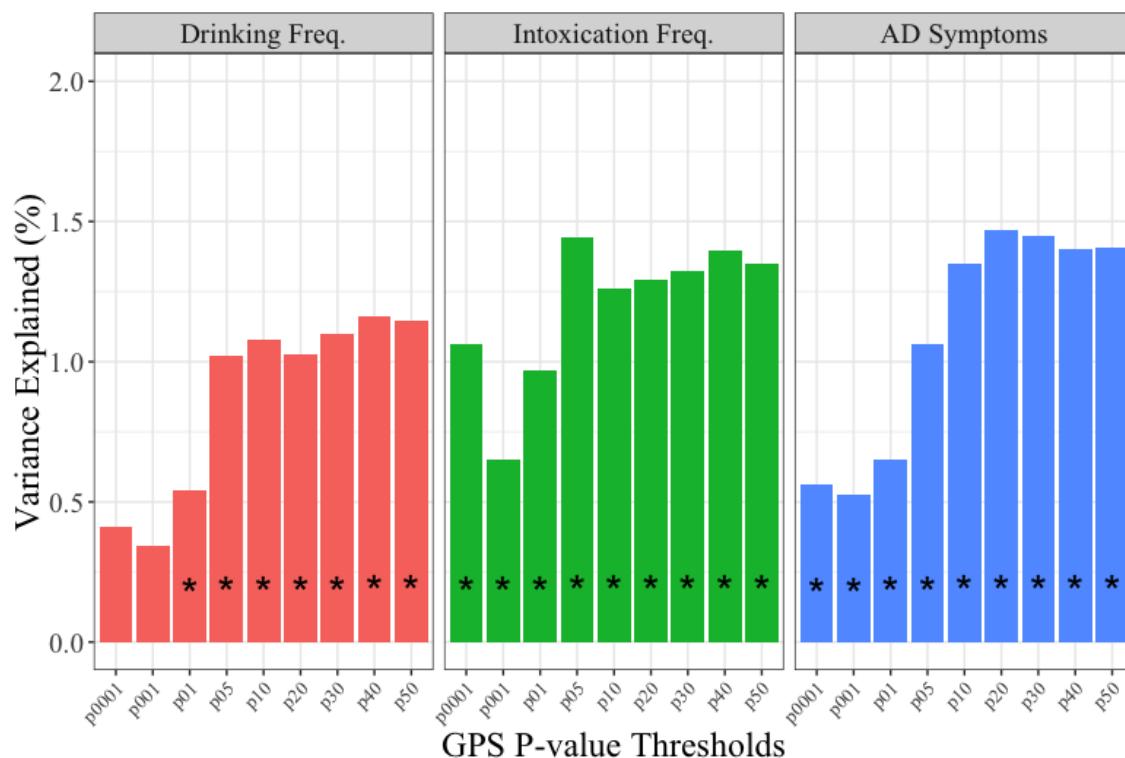
Table 2: Linear Mixed Models for Alcohol Related Behaviors (N = 1,201)

|                               | Model 1 |       | Model 2 |        | Model 3 |     |        |       |     |
|-------------------------------|---------|-------|---------|--------|---------|-----|--------|-------|-----|
|                               | B       | SE    | B       | SE     | B       | SE  |        |       |     |
| <b>Drinking Frequency</b>     |         |       |         |        |         |     |        |       |     |
| Female                        | -0.456  | 0.059 | ***     | -0.458 | 0.059   | *** | -0.437 | 0.086 | *** |
| In relationship               | -0.090  | 0.055 |         | -0.087 | 0.055   |     | -0.069 | 0.079 |     |
| GPS                           | 0.109   | 0.030 | ***     | 0.169  | 0.043   | *** | 0.193  | 0.055 | **  |
| In relationship*GPS           | -       | -     | -       | -0.105 | 0.054   | *   | -0.158 | 0.076 | *   |
| Female*GPS                    | -       | -     | -       | -      | -       | -   | -0.061 | 0.086 |     |
| Female*In relationship        | -       | -     | -       | -      | -       | -   | -0.038 | 0.110 |     |
| Female*In Relationship*GPS    | -       | -     | -       | -      | -       | -   | 0.109  | 0.110 |     |
| Pseudo-R <sup>2</sup>         | 0.073   |       | 0.076   |        | 0.077   |     |        |       |     |
| <b>Intoxication Frequency</b> |         |       |         |        |         |     |        |       |     |
| Female                        | -0.543  | 0.058 | ***     | -0.544 | 0.058   | *** | -0.535 | 0.084 | *** |
| In relationship               | -0.178  | 0.054 | **      | -0.176 | 0.054   | **  | -0.171 | 0.077 | *   |
| GPS                           | 0.111   | 0.029 | ***     | 0.179  | 0.042   | *** | 0.239  | 0.054 | *** |
| In relationship*GPS           | -       | -     | -       | -0.118 | 0.052   | *   | -0.222 | 0.073 | **  |
| Female*GPS                    | -       | -     | -       | -      | -       | -   | -0.149 | 0.084 |     |
| Female*In relationship        | -       | -     | -       | -      | -       | -   | -0.016 | 0.107 |     |
| Female*In Relationship*GPS    | -       | -     | -       | -      | -       | -   | 0.223  | 0.107 | *   |
| Pseudo-R <sup>2</sup>         | 0.110   |       | 0.114   |        | 0.117   |     |        |       |     |
| <b>AD Symptoms</b>            |         |       |         |        |         |     |        |       |     |
| Female                        | -0.197  | 0.061 | **      | -0.199 | 0.061   | **  | -0.123 | 0.089 |     |
| In relationship               | -0.097  | 0.057 |         | -0.095 | 0.057   |     | -0.028 | 0.082 |     |
| GPS                           | 0.123   | 0.030 | ***     | 0.191  | 0.044   | *** | 0.196  | 0.057 | **  |
| In relationship*GPS           | -       | -     | -       | -0.119 | 0.056   | *   | -0.154 | 0.079 |     |
| Female*GPS                    | -       | -     | -       | -      | -       | -   | -0.018 | 0.089 |     |
| Female*In relationship        | -       | -     | -       | -      | -       | -   | -0.134 | 0.115 |     |
| Female*In Relationship*GPS    | -       | -     | -       | -      | -       | -   | 0.069  | 0.114 |     |
| Pseudo-R <sup>2</sup>         | 0.029   |       | 0.032   |        | 0.034   |     |        |       |     |

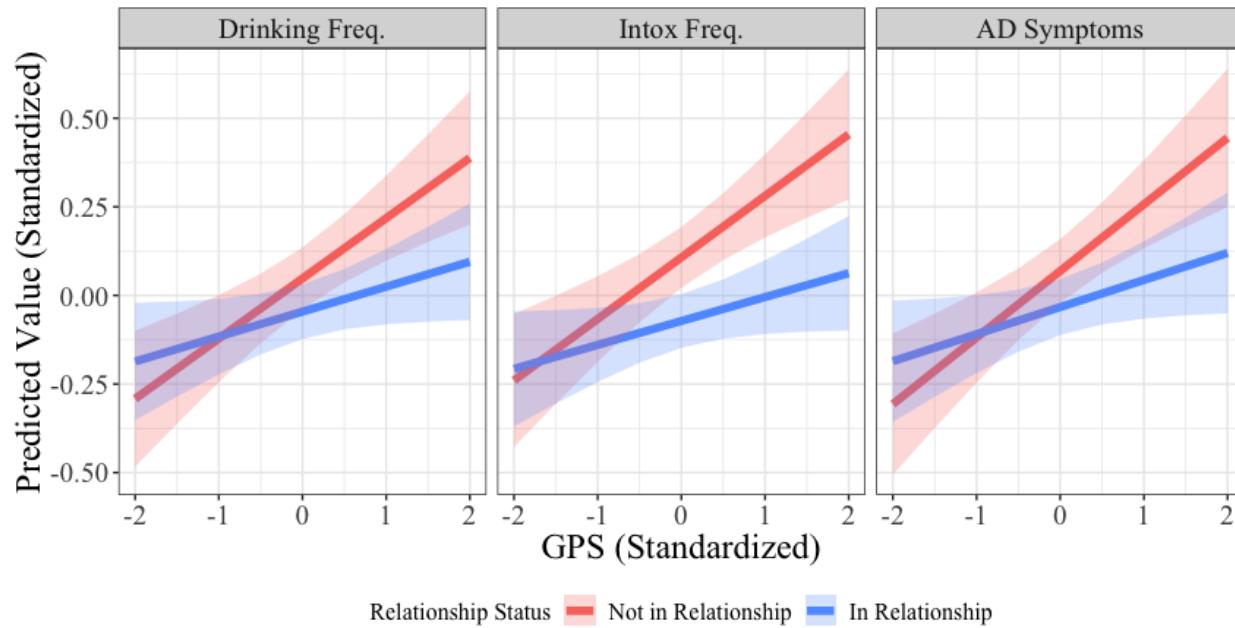
All models include age, educational attainment, and student status as covariates. Family clustering adjusted for by including random intercepts for the family level. GPS and alcohol phenotypes were standardized.

\* p < .05; \*\* p < .01; \*\*\* p < .001

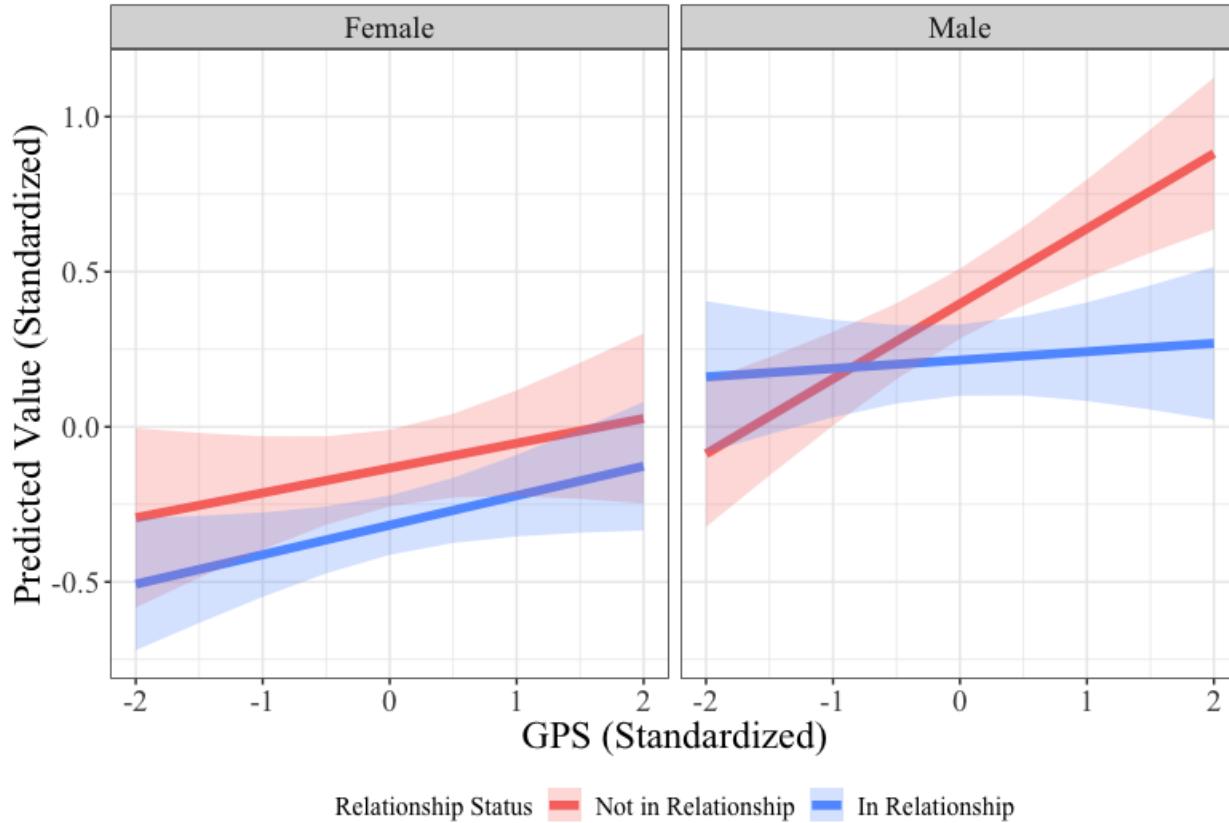
## Figures



**Figure 1: Predictive Power of GSCAN Polygenic Scores**



**Figure 2: Gene-Environment Interaction across Relationship Status and Polygenic Risk**



**Figure 3: Sex Differences in GxE for Intoxication Frequency**