

1 **Genetic influences on alcohol flushing in East Asian populations**

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26 **ABSTRACT**

27 **Background:** Although it is known that variation in the *aldehyde dehydrogenase 2 (ALDH2)*
28 gene family influences the East Asian alcohol flushing response, knowledge about other
29 genetic variants that affect flushing symptoms is limited. **Methods:** We performed a genome-
30 wide association study meta-analysis and heritability analysis of alcohol flushing in 15,105
31 males of East Asian ancestry (Koreans and Chinese) to identify genetic associations with
32 alcohol flushing. We also evaluated whether self-reported flushing can be used as an
33 instrumental variable for alcohol intake. **Results:** We identified variants in the region of
34 *ALDH2* strongly associated with alcohol flushing, replicating previous studies conducted in
35 East Asian populations. Additionally, we identified variants in the alcohol dehydrogenase 1B
36 (*ADH1B*) gene region associated with alcohol flushing. Several novel variants were identified
37 after adjustment for the lead variants (*ALDH2*-rs671 and *ADH1B*-rs1229984), which need to
38 be confirmed in larger studies. The estimated SNP-heritability on the liability scale was 13%
39 (S.E. = 4%) for flushing, but the heritability estimate decreased to 6% (S.E. = 4%) when the
40 effects of the lead variants were controlled for. Genetic instrumentation of higher alcohol
41 intake using these variants recapitulated known associations of alcohol intake with
42 hypertension. Using self-reported alcohol flushing as an instrument gave a similar association
43 pattern of higher alcohol intake and cardiovascular disease-related traits (e.g. stroke).
44 **Conclusion:** This study confirms that *ALDH2*-rs671 and *ADH1B*-rs1229984 are associated
45 with alcohol flushing in East Asian populations. Our findings also suggest that self-reported
46 alcohol flushing can be used as an instrumental variable in future studies of alcohol
47 consumption.
48
49 **Keywords:** GWAS, alcohol, alcohol flushing, *ALDH2*, *ADH1B*, heritability, Mendelian
50 randomization

51 **Background**

52 Alcohol flushing is a heritable condition in which a person develops flushes on the face or
53 skin after drinking alcohol. Whilst pronounced alcohol flushing is rarely observed in
54 Europeans, approximately 36% of East Asians experience alcohol flushing as well as other
55 unpleasant symptoms (e.g. nausea and tachycardia) [1]. Previous genome-wide association
56 studies (GWAS) identified two key genes associated with alcohol flushing, *alcohol*
57 *dehydrogenase 2 (ALDH2)* and *aldehyde dehydrogenase 1B (ADH1B)* [2-4]. These genes
58 encode enzymes that metabolize alcohol into acetaldehyde (*ADH1B*) and acetaldehyde into
59 acetate (*ALDH2*). Genetic variants in *ALDH2* and *ADH1B* alter alcohol metabolism leading
60 to prolonged, elevated levels of acetaldehyde. The excess acetaldehyde leads to physiological
61 responses to alcohol consumption, including erythema on the face, nausea, and rapid heart
62 rate [5, 6].

63

64 Most previous GWAS have focused on genetic associations with alcohol drinking status,
65 rather than alcohol-induced responses, such as alcohol flushing [7, 8]. Candidate gene
66 association studies have provided evidence for the association of *ALDH2* or *ADH1B* with
67 alcohol flushing [9], but it is unclear whether there are loci other than *ALDH2* or *ADH1B* at
68 which genetic variation appreciably influences flushing symptoms. Furthermore,
69 investigations of putative causal genes for alcohol-related physiological responses have been
70 conducted almost exclusively in individuals of European ancestry to date [7, 10], which risks
71 missing variants with very low frequencies in European populations. Genetic biobanks from
72 East Asian populations are growing in number, and with alcohol flushing highly prevalent
73 amongst those participants there is an opportunity to improve our understanding of the
74 relevant risk variants for the condition.

75

76 Recently, alcohol flushing has been proposed as a phenotypic instrumental variable (IV) for
77 examining the health impacts of alcohol consumption in East Asian populations [11, 12].
78 Alcohol flushing is associated with lower levels of alcohol consumption and is assumed to be
79 independent of confounders [13]. Considering the ease of including alcohol flushing
80 questions in surveys compared with collecting genetic information, using flushing as an IV
81 may be beneficial, enabling IV analysis in a simple, cost-effective, and non-invasive manner.
82 Therefore, it would be helpful to fully understand the effects of genetic variants on alcohol
83 flushing and to further characterise its utility as an IV.

84

85 In this study, we perform the largest GWAS of alcohol flushing to date, using 15,016 male
86 individuals of East Asian ancestry from the China Kadoorie Biobank (CKB; N = 13,456) and
87 the Korean Genome and Epidemiology Study (KoGES; N = 1,560). We also estimated the
88 SNP-based heritability of alcohol flushing. Furthermore, we examined whether self-reported
89 alcohol flushing can be used as a phenotypic IV for alcohol intake, comparing estimates with
90 results from the genotypic IV (rs671 in *ALDH2*).

91

92 **METHODS**

93 **Study population**

94 This study was performed on two datasets, CKB (discovery set) and KoGES (replication
95 set). CKB is a prospective study that recruited participants between 2004 and 2008. At
96 baseline, 512,726 adults aged 30-79 years were recruited from 10 geographically defined
97 regions of China (5 urban and 5 rural areas). All participants provide a 10mL blood sample
98 which was processed into aliquots of buffy coat and plasma and stored at -70°C. Participants
99 were prospectively followed up for cause-specific morbidity and mortality through linkage to

100 death and disease registries and to the national health insurance system. Detailed information
101 on the CKB is provided elsewhere [14, 15]. For the current analyses, we excluded individuals
102 who were not genotyped or non-drinkers for whom information on alcohol flushing was not
103 collected (Figure 1). Individuals with non-local ancestry were excluded from region-stratified
104 GWAS analyses. Analyses were limited to male participants only since female participants’
105 alcohol intake is very low in China [16] and South Korea [17]. In total, 13,456 male CKB
106 participants were included in regional GWAS analyses. For the meta-analysis, data for a total
107 of 1,560 Korean men were obtained from KoGES [18]. For the IV analysis, we included
108 23,020 males from CKB who have information on alcohol flushing, alcohol intake amount
109 and the known genetic instrument for alcohol (rs671 in *ALDH2*; Figure 1). All participants
110 provided written informed consent approved by relevant local, national, and international
111 ethics committees. Detailed information on the samples is provided in Supplementary Data.

112

113 **Assessment of alcohol flushing and drinking patterns**

114 In CKB, alcohol drinking patterns were investigated using interviewer-administered
115 questionnaires. Participants were asked how often they had drunk alcohol during the previous
116 12 months (never or almost never; occasionally; only at certain seasons; every month but less
117 than weekly; usually at least once a week). Based on the questionnaire, individuals who
118 reported alcohol consumption in most weeks in the past year were identified as current
119 drinkers. Current drinkers were asked further questions including types of beverage
120 consumed, amount of alcohol drunk, and experience of flushing after drinking. Total alcohol
121 intake (g/day) was calculated using the average alcohol content of each type of alcoholic
122 beverage. Detailed information on the assessment of alcohol intake is available elsewhere
123 [16, 19]. To investigate the presence of alcohol flushing symptoms among current drinkers,
124 the following question was used: “Do you usually experience hot flushes or dizziness after

125 drinking?” Participants were offered four options: “Yes, immediately”; “Yes, after a small
126 amount of alcohol”; “Yes, but only after drinking a large amount of alcohol”, and “No”.
127 Participants who experienced flushing immediately after drinking alcohol and those who
128 flushed after a small amount of alcohol were classified as alcohol flushers. For sensitivity
129 analyses, we defined alcohol flushing using different criteria (main, relaxed, strict, and
130 continuous; see the Methods section in Supplementary Data for more details). All
131 questionnaires were provided in Mandarin. The definition of flushing for KoGES is described
132 in Supplementary Data.

133

134 **DNA sampling and genotyping**

135 DNA was extracted from the buffy coat and was genotyped using the custom Affymetrix
136 Axiom arrays and Illumina Golden Gate platform at BGI (Shenzhen, China), as previously
137 described [15]. Data for a total of 100,706 individuals passed quality control criteria (call rate
138 $\geq 95\%$, no sex mismatch, heterozygosity F statistic SD score $< +3$, no XY aneuploidy, no non-
139 East Asian ancestry). Following variant QC (call rate > 0.98 , no batch or plate effect, Hardy-
140 Weinberg equilibrium $P > 10^{-6}$), imputation was performed using SHAPEITv3/IMPUTEv4 and
141 the 1000 Genomes Project Phase 3 reference panel. After imputation, SNPs were removed if
142 the MAF was low (< 0.01) or INFO was < 0.3 . After QC, 8,001,732 autosomal SNPs were
143 used for association testing. Detailed information on the genotyping method and QC for
144 KoGES is provided in Supplementary Data.

145

146 **Genome-wide association analyses**

147 In CKB, genetic loci associated with flushing were investigated using BOLT-LMM v2.3.2
148 [20]. Three models were constructed. The first model was adjusted for age, age squared, the
149 first ten genetic principal components (PCs), and genotyping array version (Model 1). We

150 performed second and third GWAS analyses adjusting for the dosages of the SNPs that are
151 known to be strongly associated with alcohol metabolism – rs671 in *ALDH2* (Model 2) and
152 additionally rs1229984 in *ADH1B* (Model 3) [12]. We performed further GWAS analyses
153 using different definitions of alcohol flushing (Supplementary Data). Each of the GWA
154 analyses described above was performed separately for each geographical region (10 study
155 areas). Within each region, SNPs with a low minor allele count (MAC < 6) or with Hardy–
156 Weinberg equilibrium test values of $P < 1 \times 10^{-6}$ were excluded. Betas and standard errors
157 (S.E.) obtained from BOLT-LMM were converted to log-odds ratios (OR) using $\log(\text{OR}) = \beta /$
158 $(\mu(1-\mu))$, where μ is the case-control ratio, following which region-level association statistics
159 were combined using a fixed-effect inverse-variance-weighted meta-analysis using METAL
160 [21]. One region (region 46, Liuzhou; n = 682) was excluded from the meta-analysis since
161 the heritability estimate in this region was close to 0. We did not apply genomic control
162 correction to the meta-analysis data because there was little evidence for inflation (all $\lambda <$
163 1.02, Figure 2).

164 In KoGES, association tests were performed using PLINK 1.90 (available at
165 <https://www.cog-genomics.org/plink2>). The GWA analysis of alcohol flushing was
166 conducted using logistic regression assuming an additive genetic model using the three
167 constructed models described above (Supplementary Data). SNPs with a low minor allele
168 count (MAC < 20) were excluded.

169 For the GWAS meta-analysis of CKB and KoGES, we performed a fixed-effect inverse
170 variance-weighted meta-analysis of the GWAS summary statistics from the CKB and KoGES
171 using METAL [21].

172 For all GWAS analyses, a genome-wide significance threshold of 5.0×10^{-8} was applied. We
173 presented variants that were identified to be independent after linkage disequilibrium (LD)

174 clumping (Supplementary Data). The distributions of the observed P-values of given SNPs
175 were plotted against the theoretical distribution of expected P-values to yield a quantile–
176 quantile (QQ) plot for flushing (Figure 2).

177

178 **Single nucleotide polymorphism-heritability analysis**

179 The SNP heritability of alcohol flushing in the CKB sample was calculated using BOLT-
180 REML, which provides a fast algorithm for multi-component modelling to partition SNP-
181 heritability [22]. Heritability (h_g^2) was estimated using the restricted maximum likelihood
182 estimation method implemented in BOLT-REML. Since we defined alcohol flushing as a
183 binary trait, we transformed the heritability on the observed scale to that on the liability scale
184 (h_l^2) [23]. Analyses were adjusted for the covariates used in the GWAS analyses. SNP
185 heritability in KOGES was estimated using the bivariate restricted maximum likelihood
186 analysis implemented in GCTA [24, 25]. Detailed methods are described in the
187 Supplementary Data.

188

189 **Mendelian randomisation analysis of alcohol flushing and disease outcomes**

190 The causal effect of alcohol intake on blood pressure and cardiovascular diseases and related
191 traits was evaluated using IV analyses with a two-stage least squares estimation method. A
192 total of 23,020 individuals were included in the IV analyses (Figure 1). Self-reported alcohol
193 flushing and the rs671 genotype were used as the phenotypic and genotypic instruments,
194 respectively. We used the strict definition of flushing (i.e., immediately after consuming
195 alcohol) as our IV. The magnitude of the association of alcohol intake (g/week) was scaled
196 into a 280 g/week effect, as in a previous study [16]. For binary outcomes (i.e. stroke,
197 myocardial infarction, coronary heart disease, hypertension, and diabetes), a two-stage

198 logistic model was used. In the first stage, alcohol intake was instrumented by alcohol
199 flushing or the rs671 genotype with adjustment for age, region, PCs (1-10), and genotyping
200 array, using a linear regression model. In the second stage, the effect of alcohol on the risk of
201 disease was estimated by fitting the alcohol intake value from the first stage, under a logistic
202 regression model with adjustment for the same confounders as in the first stage. For
203 continuous traits (i.e., aspartate aminotransferase [AST], gamma-glutamyl transferase [GTT],
204 cholesterol, triglycerides, blood glucose, and blood pressure), a two-stage linear model was
205 applied, similarly adjusting for confounders. Region-stratified analyses followed by meta-
206 analysis gave similar results.

207 The values were reported as ORs per 280 g/week alcohol intake with 95% CIs for the binary
208 outcomes and β -coefficients with 95% CIs for the continuous outcomes. We examined the
209 strength and validity of each instrument using the F-statistic of the association of each
210 instrument with alcohol intake (with an F-statistic >10 indicating adequate strength).

211 Statistical significance (at the 5% level) was evaluated using a P-value threshold of 0.05. The
212 difference of estimates between instruments (alcohol flushing and rs671) was assessed using
213 a difference of two means test [26] (P value threshold for significance = 0.05).

214

215 **RESULTS**

216 **General characteristics of the study population**

217 The baseline characteristics of the study subjects according to flushing status are presented in
218 Supplementary Table 1 and 2. In the CKB cohort, among 13,456 men with both alcohol
219 flushing and genotype information, 17.9% reported flushing (i.e., flushing immediately after
220 drinking alcohol or after drinking a small amount of alcohol). The mean weekly alcohol
221 intake of non-flushers was 304.5 ± 259.0 g/week (mean \pm standard deviation [SD]). Flushers

222 had a lower mean weekly alcohol intake (228.1 ± 259.0 g/week) compared to non-flushers.
223 Flushers had a higher proportion of rs671 A allele carriers (45.5 % of flushers vs 8.7 % of
224 non-flushers) as well as rs1229984 A allele carriers (90.3 % of flushers vs 87.3 % of non-
225 flushers) than non-flushers. The characteristics of 1,560 KoGES samples are described in
226 Supplementary Table 2. Similar to the CKB, flushers in KoGES had a lower proportion of
227 current drinkers who consumed relatively small amounts of alcohol compared to non-
228 flushers. Also, flushers in KoGES had a higher proportion of rs671 A allele carriers (68.4 %
229 of flushers vs 9.1 % of non-flushers) and rs1229984 A allele carriers (95.5 % of flushers vs
230 93.2 % of non-flushers) than non-flushers.

231

232 **Genome-wide association analyses of flushing**

233 In CKB, the top signal for GWAS of flushing (Model 1; See Methods) was at rs671, a
234 functional variant in *ALDH2* (Beta = 2.86, S.E. = 0.07, $P = 8.6 \times 10^{-416}$; Figure 2 and Table 1;
235 Supplementary Table 3 and 8; Supplementary Figure 6). After adjustment for rs671 (Model
236 2), the strongest signal was detected at rs1229984 in *ADH1B* (Beta = 0.24, S.E = 0.03, $P =$
237 1.1×10^{-13} ; Supplementary Table 9). Additionally, Model 2 identified a variant on
238 chromosome 3 (rs1508403 in *PTPRG*, Beta = 0.84, S.E = 0.15, $P = 3.38 \times 10^{-8}$). There were
239 no genome-wide significant SNPs after further adjustment for rs1229984 (Model 3; Figure
240 2).

241

242 GWA analyses using different criteria for defining flushing showed no difference in the top
243 signals for Models 1 and 2 across the different definitions of flushing (see Supplementary
244 Methods) although the P -values for the lead SNPs varied (Table 1; Supplementary Figure 1-
245 3; Supplementary Table 10-16); The P values for the strongest signals became less significant
246 for the relaxed flushing definition (ie., flushing after drinking any amount of alcohol) (Table

247 1; Supplementary Table 10-11). For the relaxed flushing definition, Model 2 identified
248 additional signals on chromosome 2 (rs532522882 *HPCAL1*; $P = 1.29 \times 10^{-8}$) along with the
249 signal at *ADH1B* on chromosome 4 (Table 1; Supplementary Table 11). For the strict
250 flushing definition (ie., flushing immediately after drinking alcohol), Model 3 identified a
251 few rare variants ($MAF \leq 0.01$; Table 1 and Supplementary Table 14) that reached genome-
252 wide significance including rs150099059 in *KCNH1* ($P = 9.4 \times 10^{-9}$), rs1011755 on
253 chromosome 11 ($P = 1.6 \times 10^{-8}$), and rs142761523 in *CNTN* ($P = 2.6 \times 10^{-8}$). For each
254 flushing definition, Model 3 also identified further suggestive associations marginally below
255 the genome-wide significance threshold. These include rs148407052 in *LOC105375361* ($P =$
256 5.1×10^{-7}) for the relaxed flushing definition; and rs2903308 in *SHISA9* ($P = 1.4 \times 10^{-7}$) for
257 the continuous flushing definition. However, we were not able to replicate these findings in
258 KoGES: either the association of these variants was strongly attenuated towards the null, or
259 they were not available in KoGES (Supplementary Table 6).

260

261 The GWAS results from an independent Korean cohort (KoGES) are presented in
262 Supplementary Table 3 and 4. The GWAS identified strong association signals on
263 chromosome 12 including rs671. In KoGES, *ADH1B* rs1229984 did not reach genome-wide
264 significance across models 1-2. An apparent independent association at the chromosome 12
265 locus harbouring the *ALDH2* gene was identified after adjusting for rs671 (rs2074356, $\beta =$
266 2.85 , $S.E = 0.26$, 2.7×10^{-28} ; Model 2; Supplementary Figure 4 and Supplementary Table 4),
267 or adjusting for rs12231737, which was the top signal obtained from Model 1 (rs2074356,
268 $\beta = 2.26$, $S.E = 0.28$, 2.9×10^{-16} ; Model 4; Supplementary Table 4). To explore the
269 obtained signals further, we conducted fine mapping using SuSiE which returned a single
270 credible set. The credible set suggested that the conditionally independent signals are likely

271 due to measurement error induced by relatively low imputation quality around the rs671
272 locus (data available on request).

273 A summary of the strongest association signals from the meta-analysis is presented in
274 Supplementary Table 3 and 17-19.

275

276

277 **SNP heritability for alcohol flushing in the CKB and KoGES**

278 SNP heritability of alcohol flushing among drinkers was estimated to be 12.6 % (SE = 4.0 %)
279 on the liability scale (h_l^2). It decreased to 8.4 % (S.E. = 4.2 %) when we controlled for rs671
280 in *ALDH2* (Supplementary Table 5), and decreased further when we also controlled for
281 rs1229984 in *ADH1B* (h_l^2 = 6.3 %; S.E. = 4.2 %), suggesting that rs671 and rs1229984
282 together explain half of the common variant genetic variance in alcohol flushing in Chinese
283 males. SNP heritability estimates of alcohol flushing amongst drinkers and non-drinkers in
284 the Korean population were imprecise due to the relatively small sample size but showed a
285 pattern consistent with that seen in CKB.

286

287 **Using self-reported flushing as an instrumental variable**

288 IV analyses among 23,020 men in CKB with flushing data showed that higher alcohol intake
289 (as instrumented by absence of self-reported alcohol flushing) was nominally associated with
290 a higher risk of intracerebral haemorrhage (OR per 280 g/week increase in alcohol intake =
291 3.28; 95% CI = 1.58 – 6.81), and total stroke (OR per 280 g/day increase in alcohol intake =
292 1.89; 95% CI = 1.28 – 6.81) as well as higher levels of AST, GGT, HDL cholesterol, log-
293 transformed random blood glucose, and diastolic blood pressure (DBP; beta per 280 g/day
294 increase in alcohol intake = 2.3 mm Hg; 95% CI = 0.9-3.7; Table 2). These associations were

295 generally consistent in direction and magnitude, although the estimates were more precise
296 when using the rs671 genotype as an IV, which also provided evidence that higher alcohol
297 intake caused a higher risk of hypertension and higher levels of systolic blood pressure
298 (SBP), as well as increased risk of stroke types, coronary heart disease, and diabetes.

299

300 **DISCUSSION**

301 In this study, we investigated genetic variation associated with alcohol flushing and estimated
302 the heritability of flushing in Chinese and Korean male populations. Strong signals were
303 detected in *ALDH2* (Supplementary Table 3) in both populations, supporting the previous
304 evidence [27]. The SNP-based heritability estimate on the liability scale was 13% for flushing
305 and decreased by 6% when the key variants (rs671 and rs1229984) were accounted for. The
306 decrease in heritability supports the role of *ALDH2* and *ADH1B* as major contributors to the
307 self-reported alcohol flushing response in the Chinese and Korean populations.

308

309 In both cohorts (CKB and KoGES), a small proportion of non-flushers were carriers of
310 *ALDH2*-rs671 A, whilst some flushers were not A allele carriers, suggesting that other
311 genetic variants may play a role in alcohol flushing metabolism. Therefore, we adjusted for
312 the *ALDH2* rs671 genotype to identify other variants that may influence alcohol flushing: this
313 revealed a strong association of *ADH1B* rs1229984 with alcohol flushing: this revealed a
314 strong association of *ADH1B* rs1229984 with alcohol flushing. rs1229984 is a missense
315 variant that has been extensively reported to be associated with alcohol consumption
316 phenotypes such as alcohol intake status, and alcohol use disorders, including in European
317 populations where the variant is present at low-frequency [28-30].

318

319 There has been some disagreement relating to the association of *ADH1B* with alcohol
320 flushing. A low-dose alcohol challenge followed by a metabolite screen in Han Chinese men
321 suggested that *ADH1B* did not associate with elevated blood acetaldehyde [31]. However, in
322 a candidate gene study involving *ALDH2* and *ADH1B* in a sample of Japanese individuals
323 with alcohol dependence, *ADH1B* did associate with flushing [32]. In CKB, the power to
324 detect the *ADH1B* association is improved by reducing the residual variance after
325 conditioning on rs671. However, the *ADH1B* association did not reach statistical significance
326 in the Korean population. One theoretical explanation for that result is collider bias [33], in
327 which flushing and *ADH1B* each influence alcohol dependence independently [32], and
328 amongst cases become associated. Here, the *ADH1B* association is unlikely to arise due to
329 this form of technical issue, because the association replicates in KoGES (albeit not at
330 genome-wide significance) which has no alcohol consumption-related sample selection.
331 Further GWAS in larger samples are required given the sample size of KoGES.

332

333 Several low-frequency variants were associated with different definitions of alcohol flushing
334 in CKB (Table 1; Supplementary Tables 9-16), after controlling for the known variants
335 (*ALDH2* rs671 and *ADH1B* rs1229984). These include *PTPRG* rs1508403 (MAF = 0.013) for
336 the main flushing definition (Supplementary Table 9), *HPCAL1* rs532522882 (MAF = 0.004)
337 and rs181957632 (MAF = 0.004) for the relaxed flushing definition (Supplementary Table
338 11), and *KCNH1* rs150099059 (MAF = 0.01), and rs142761523 (MAF = 0.01) and
339 rs144350123 in *CNTN* (MAF = 0.01) for the strict flushing definition (Supplementary Table
340 13). A GWAS study in 3,838 individuals of European- and African- American ancestry
341 reported that the activities of *PTPRG* were associated with alcohol dependence [34]. A study
342 in mice reported that the expression of *HPCAL1* was associated with alcohol consumption
343 [35]. Furthermore, a study in rats reported that the *KCNH1* gene, which encodes potassium

344 voltage-gated channels, is differentially expressed in binge drinking groups [36]. The *CNTN*
345 family has been suggested to be associated with alcohol independence by GWAS studies in
346 European populations [37, 38]. Further studies with larger samples will be needed to replicate
347 these findings.

348

349 SNP-based heritability analyses estimated that around 13% of the phenotypic variation in
350 flushing is explained by common genetic variants. The heritability estimates decreased
351 substantially when *ALDH2* rs671 was controlled for illustrating the strong effect of *ALDH2*
352 on flushing in the Chinese population. These heritability estimates for flushing were much
353 lower than all previous estimates for alcohol consumption [39]. One reason could be that our
354 study only included regular drinkers. In this study, the subjects were asked about their
355 experience of flushing based on their alcohol drinking status. This can be a source of
356 selection bias where a sample can contain only those who report drinking. For example,
357 individuals from CKB who do not regularly drink due to their knowledge of flushing are
358 likely excluded from the current analysis. Also, individuals who drink regardless of their
359 flushing symptom may have developed compensatory feedback mechanisms [40], which can
360 possibly contribute to weaker flushing symptoms. Consequently, this may lead to lower
361 variance in flushing severity in the study subjects that could lead to lower heritability
362 estimates in Chinese population.

363

364 The IV results demonstrated that self-reported alcohol flushing can be used as an IV for
365 alcohol consumption levels among drinkers. The pattern of associations of alcohol and
366 disease traits was similar to a previous study in the Korean population that suggested the
367 possibility of using self-reported alcohol flushing as an IV [11, 41]. However, we observed
368 that the power to detect causal effects was generally attenuated in CKB when using self-

369 reported flushing compared with the genetic IV, whereas the previous study by Cho et al.
370 [41] demonstrated using self-reported alcohol flushing as an IV gave similar results to the use
371 of the *ALDH2* rs671 variant as an IV. One major difference between the two studies is that
372 CKB only had data available on alcohol flushing amongst individuals who self-reported
373 regular drinking. Such structured sample selection can induce collider bias [33]. Indeed, in
374 the CKB, the participants who regularly consumed alcohol had a lower prevalence of
375 hypertension and lower BP levels than non-drinkers or ex-drinkers (Supplementary Table 7).
376 This suggests that the IV analysis in CKB may have been affected by collider bias. For
377 example, if higher levels of BP and flushing are both causally related to drinking, the
378 association between alcohol intake and higher BP may be distorted (Supplementary Figure
379 7), given non-drinkers who flush were excluded from the current study. In this case, the
380 genetic instrument (e.g. rs671) for the overall population is likely to be more reliable than a
381 questionnaire as the genotypes are distributed completely randomly within the whole sample,
382 regardless of their drinking status. By contrast, the self-reported IV based on the
383 questionnaire is more likely to be subject to individuals' drinking status.

384

385 This study has several other limitations. First, despite this being the largest genome-wide
386 study of alcohol flushing to date, it is possible that there was limited statistical power to
387 detect influential loci other than *ALDH2* and *ADH1B*. Second, our analyses included flushers
388 who regularly drink, due to the design of the questionnaire used in CKB. Therefore, there is a
389 possibility that those who do not drink alcohol due to their response to alcohol were not
390 included in the current study. Nonetheless, results for our top loci are confirmed in two
391 independent samples (Chinese and Koreans) showing that the identified genetic variants are
392 likely to be strongly involved in flushing. Further GWAS and SNP heritability analyses are
393 required in other East Asian populations. Third, some variants identified in CKB were

394 relatively rare, and we could not test their association in KoGES, leaving the possibility that
395 these variants were detected by chance. Fourth, although the variants used for GWAS were
396 filtered to have high imputation scores (INFO \geq 0.8), imputation accuracy using the 1000
397 genomes reference panel in Korean samples as was done for KoGES may still lead to
398 measurement error. This is because, although the panel includes East Asian samples (Han
399 Chinese and Japanese), it does not include Korean samples. It has been reported that the
400 Korean population is genetically homogeneous due to geopolitical isolation, thus, Koreans
401 genetically clustered distinctly from other East Asian populations [42]. Therefore, it could be
402 speculated that while rs671 associated very strongly with flushing, it was not detected as the
403 top signal at the *ALDH2* locus due to inaccuracy in imputation. Fifth, the use of alcohol
404 flushing as an instrument may only reflect an effect of alcohol intake from a specific period
405 of the life course (e.g. in adulthood) since alcohol flushing only occurs after an individual has
406 started drinking (e.g. during adulthood).

407

408 **CONCLUSIONS**

409 Despite these limitations, the results have epidemiologic and public health implications. Our
410 findings underline the importance of additive genetic effects in modifying alcohol
411 consumption behaviour and support the use of flushing or genetic variants (e.g. rs671 in
412 *ALDH2*) as proxies for alcohol consumption in East Asian populations. To the best of our
413 knowledge, this is the first GWAS to investigate putative causal variants for alcohol flushing
414 and estimate the heritability of the condition in East-Asian populations.

415

416 **LIST OF ABBREVIATIONS**

417 ALDH2, Aldehyde dehydrogenase 2

418 ADH1B, Alcohol dehydrogenase 1B
419 GWAS, Genome-wide association studies
420 IV, Instrumental variable
421 CKB, China Kadoorie Biobank
422 KoGES, Korean Genome and Epidemiology Study
423 PCs, Principal components
424 SE, Standard errors
425 OR, Odds ratios
426 QQ, quantile-quantile
427 SD, Standard deviation
428 AST, Aspartate aminotransferase
429 GGT, Gamma-glutamyl transferase
430 BP, Blood pressure
431 DBP, Diastolic blood pressure
432 SBP, Systolic blood pressure

433

434 **DECLARATIONS**

435 **Ethics approval and consent to participate**

436 All participants for KoGES were provided written informed consent approved by relevant
437 local, and national ethics committees. The CKB complies with all the required ethical
438 standards for medical research on human subjects. Ethical approvals were granted and have
439 been maintained by the relevant institutional ethical research committees in the UK and
440 China. Informed consent was obtained from all participants included in the CKB. This study

441 was approved by the Institutional Review Board of Yonsei University (Seoul, South Korea;
442 Reference Number: 4-2015-1132).

443

444 **Consent for publication**

445 Not applicable

446

447 **Availability of data and materials**

448 The datasets supporting the conclusions of this article are not publicly available due to
449 institutional restrictions regarding accessibility, but are available from the corresponding
450 author on reasonable request and with permission of the committee of CKB and KoGES.

451

452 **Competing interests**

453 The authors declare that they have no competing interests.

454

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467

468 **Author contributions**

469 YC, IYM, GH, RGW and GDS conceptualized the project. YC, SHL, and KL performed
470 statistical analyses. YC, IYM, GH, and RGW drafted the first version of the manuscript. All
471 authors contributed to the interpretation of results and manuscript writing.

472

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476 **Figure legends**

477 **Figure 1. Flowchart of study population selection.**

478 **Figure 2. Manhattan plots and quantile-quantile for GWAS of flushing in Chinese**

479 **population.** Each plot represents the result from different models. (A) Model 1: controlling

480 for age, age squared, PCs (1-10) (B) Model 2: covariates in Model 1 plus *ALDH2* rs671 and

481 (C) Model 3: covariates in Model 2 plus *ADH1B* rs1229984. The y-axis shows the age and

482 sex-adjusted $-\log_{10}$ P values and the x-axis presents positions along the chromosome (Chr.).

483 The solid red line indicates the P value of 5×10^{-8} whereas the blue line indicates the P value

484 of 1×10^{-5} . (D-F) represent QQ plots for each model.

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