

Supporting information 1A and 1B for
 ‘Exploiting collider bias to apply two-sample
 summary data Mendelian randomization methods
 to one-sample individual level data’

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A: Derivation of the Collider-Correction formula

The asymptotic least squares estimates of the effects of X and G on Y , without conditioning on U , are

$$\begin{bmatrix} \beta^* \\ \alpha_1^* \\ \vdots \\ \alpha_k^* \end{bmatrix} = \begin{bmatrix} \text{var}(X) & \text{cov}(X, G_1) & \cdots & \text{cov}(X, G_k) \\ \text{cov}(X, G_1) & \text{var}(G_1) & \cdots & \text{cov}(G_1, G_k) \\ \vdots & \vdots & \ddots & \vdots \\ \text{cov}(X, G_k) & \text{cov}(G_1, G_k) & \cdots & \text{var}(G_k) \end{bmatrix}^{-1} \begin{bmatrix} \text{cov}(X, Y) \\ \text{cov}(G_1, Y) \\ \vdots \\ \text{cov}(G_k, Y) \end{bmatrix}$$

Assuming no LD between SNPs, so $\text{cov}(G_i, G_j) = 0$ where $i \neq j$, the variance-covariance matrix has block form with a diagonal matrix in the lower right quadrant. Block-wise inversion gives

$$\begin{bmatrix} \text{var}(X) & \text{cov}(X, G_1) & \cdots & \text{cov}(X, G_k) \\ \text{cov}(X, G_1) & \text{var}(G_1) & \cdots & \text{cov}(G_1, G_k) \\ \vdots & \vdots & \ddots & \vdots \\ \text{cov}(X, G_k) & \text{cov}(G_1, G_k) & \cdots & \text{var}(G_k) \end{bmatrix}^{-1} =$$

$$\frac{1}{\text{var}(X) - \sum_j \frac{\text{cov}(X, G_j)^2}{\text{var}(G_j)}} \left[\begin{array}{cccc} 1 & \frac{-\text{cov}(X, G_1)}{\text{var}(G_1)} & \cdots & \frac{-\text{cov}(X, G_k)}{\text{var}(G_k)} \\ \frac{-\text{cov}(X, G_1)}{\text{var}(G_1)} & \frac{\text{cov}(X, G_1)^2}{\text{var}(G_1)^2} & \cdots & \frac{\text{cov}(X, G_1)\text{cov}(X, G_k)}{\text{var}(G_1)\text{var}(G_k)} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{-\text{cov}(X, G_k)}{\text{var}(G_k)} & \frac{\text{cov}(X, G_1)\text{cov}(X, G_k)}{\text{var}(G_1)\text{var}(G_k)} & \cdots & \frac{\text{cov}(X, G_k)^2}{\text{var}(G_k)^2} \end{array} \right] +$$

$$\begin{bmatrix} 0 & 0 & \cdots & 0 \\ 0 & \frac{1}{var(G_1)} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \frac{1}{var(G_k)} \end{bmatrix}$$

Then

$$\beta^* = \frac{cov(X, Y) - \sum_j \frac{cov(X, G_j) cov(G_j, Y)}{var(G_j)}}{var(X) - \sum_j \frac{cov(X, G_j)^2}{var(G_j)}}$$

And

$$\begin{aligned} \alpha_i^* &= \frac{\frac{-cov(X, G_j)}{var(G_j)} \left(cov(X, Y) - \sum_j \frac{cov(X, G_j) cov(G_j, Y)}{var(G_j)} \right)}{var(X) - \sum_j \frac{cov(X, G_j)^2}{var(G_j)}} + \frac{cov(G_i, Y)}{var(G_i)} \\ &= -\beta_{XG_j} \beta^* + \frac{cov(G_i, Y)}{var(G_i)} \end{aligned}$$

From equation 2, $cov(G_i, Y) = (\alpha_i + \beta \beta_{XG_j}) var(G_i)$. Therefore

$$\alpha_j^* = \alpha_j + \beta_{XG_j} (\beta - \beta^*)$$

The causal effect β is therefore the observational effect β^* , plus the slope of the regression of α_j^* on β_{XG_j} .

B: Outlier SNPs sets for the Insomnia-HB1c analysis.

SNP set detected as outliers using a Bonferroni corrected exact Q statistic in analysis (a) (23andMe + UK Biobank data)

1	rs10758593
2	rs1264419
3	rs12917449
4	rs12924275
5	rs1861412
6	rs214934
7	rs2737240
8	rs2792990
9	rs3131638
10	rs34490907
11	rs429358
12	rs4788203
13	rs6888135

SNP set detected as outliers using a Bonferroni corrected exact Q statistic
in analysis (b) (23andMe data only)

1	rs10758593
2	rs1264419
3	rs214934
4	rs2792990
5	rs4788203
6	rs647905