Text S2: Relationship between BF_{mean} and Λ_{mean}

Here we develop an expression for the relationship between the Bayes factor BF_{mean} and the likelihood ratio statistic Λ_{mean} , under an additive model with d = 0 and a fixed value of the prior variance of a, σ_a . (In the main paper we use a Bayes factor that averages over several values of σ_a , and allows for some dominance, as in [1]. This is different from, though highly correlated with, the Bayes factor considered here.)

Under an additive model (d = 0) the model (12) in the main text can be written as $Y \sim N(\bar{X}\beta, 1/\tau)$, where \bar{X} is the $n \times 2$ design matrix $(\mathbf{1}, \bar{g})$ whose first column is all 1s, and second column contains the expected genotype values for each individual. Using the prior D2 from [1] the Bayes factor for the alternative $(a \neq 0)$ vs the null (a = 0) is:

$$BF_{mean} = \frac{(\alpha n)^{1/2}}{\det(\Omega)^{1/2}} \left(\frac{Y^t Y - Y^t \bar{X} \Omega^{-1} \bar{X}^t Y}{Y^t Y - n \bar{Y}^2} \right)^{-n/2}$$
(1)

where for notational convenience we have written α for σ_a^{-2} , and $\Omega = \begin{pmatrix} 0 & 0 \\ 0 & \alpha \end{pmatrix} + \bar{X}^t \bar{X}$. If we let V denote n times the sample variance of g, $V = (g^t g - n\bar{g}^2)$, then $\det(\Omega) = n(\alpha + V)$, $\det(\bar{X}^t \bar{X}) = nV$, and

$$\Omega^{-1} = \frac{1}{\det(\Omega)} \left[\det(\bar{X}^t \bar{X}) (\bar{X}^t \bar{X})^{-1} + \begin{pmatrix} \alpha & 0 \\ 0 & 0 \end{pmatrix} \right] = \frac{V}{\alpha + V} \left[(\bar{X}^t \bar{X})^{-1} + \begin{pmatrix} \alpha/nV & 0 \\ 0 & 0 \end{pmatrix} \right].$$
(2)

Now let RSS_0 and RSS_1 denote the residual sum of squares for the maximum likelihood estimates under the null and alternative hypotheses:

$$RSS_0 = Y^t Y - n\bar{Y}^2$$

$$RSS_1 = Y^t Y - Y^t \bar{X} (\bar{X}^t \bar{X})^{-1} \bar{X}^t Y.$$
(3)

Taking the log of (1), and substituting, we obtain

$$\log(\mathrm{BF}_{\mathrm{mean}}) = 0.5 \log[\alpha/(\alpha+V)] - (n/2) \log \frac{\alpha/(\alpha+V)\mathrm{RSS}_0 + V/(\alpha+V)\mathrm{RSS}_1}{\mathrm{RSS}_0} = 0.5 \log[\alpha/(\alpha+V)] - (n/2) \log\left[1 - \frac{V}{\alpha+V}\frac{\mathrm{RSS}_0 - \mathrm{RSS}_1}{\mathrm{RSS}_0}\right].$$
(4)

Note also that the log-likelihood ratio statistic is given by

$$2\log\Lambda_{\text{mean}} = n\log[\text{RSS}_0/\text{RSS}_1] = n\log[1 - (\text{RSS}_0 - \text{RSS}_1)/\text{RSS}_1].$$
(5)

Four observations come directly from these expressions for BF_{mean} and Λ_{mean} :

- For fixed V, both BF_{mean} and Λ_{mean} are monotone decreasing in $(RSS_0-RSS_1)/RSS_1$. Hence, among SNPs with the same value of V, BF_{mean} and Λ_{mean} will produce the same ranking of SNPs.
- If $\alpha = kV$ (so that for each SNP the prior variance on a is chosen to be inversely proportional to the V for that SNP) then log(BF_{mean}) and Λ_{mean} produce the same ranking of SNPs.
- For fixed V, if $(RSS_0 RSS_1)/RSS_1$ is small (which it will be in our applications, where effect sizes are small), then both $\log(BF_{mean})$ and Λ_{mean} are approximately linear in $(RSS_0 - RSS_1)/RSS_1$, because $\log(1 - x) \approx -x$ when x small. This explains the approximately linear relationships evident in Figure 3 in the main text.
- For fixed $(RSS_0 RSS_1)/RSS_0$, and fixed α , $BF_{mean} \rightarrow 1$ as $V \rightarrow 0$. Thus, compared with Λ_{mean} , BF_{mean} down-weights SNPs with small V.

References

[1] Servin, B. and Stephens, M. (2007). Efficient multipoint analysis of association studies: candidate regions and quantitative traits. *PLoS Genetics*, **3**.