## $S'_B$ : a modified version of the $S_B$ statistic that is more robust to low sample sizes

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In Refoyo-Martinez et al. (2020), we defined the  $S_B$  statistic for a specific branch k of an admixture graph as:

$$S_B = \frac{((\mathbf{p} - \bar{p}\mathbf{1})^T \mathbf{b_k})^2}{\bar{p}(1 - \bar{p}) \mathbf{b_k}^T \hat{\mathbf{f}} \mathbf{b_k}}$$
(1)

Here,  $\mathbf{p}$  is the vector of sample allele frequencies across populations,  $\hat{\mathbf{F}}$  is an estimate of the genome-wide allele frequency covariance matrix, and  $\bar{p}$  is the mean allele frequency among populations. The elements of the branch vector  $\mathbf{b_k}$  are the ancestry contributions of that branch to each of the populations in the leaves of the graph.

This statistic makes use of the sample allele frequencies  $\mathbf{p}$  as an approximation the true population allele frequencies  $\mathbf{\check{p}}$ , which may be particularly poor if a SNP under study has sequence data from a few individuals in a given population.

Conditional on the population allele frequency for a population j, the sample allele frequency is binomially distributed:

$$p_i | \breve{p_i} \sim Bin(2n_i, \breve{p_i})$$
 (2)

where  $n_j$  is the number of diploid individuals for which there is reliable genotype data at a particular SNP of interest in population j. We can approximate the above equation using a Normal distribution:

$$p_j | \breve{p}_j \sim Normal(\breve{p}_j, \breve{q}_j)$$
 (3)

where  $\check{q}_j$  is equal to  $\frac{\check{p}_j(1-\check{p}_j)}{2n_j}$ . Like  $\check{p}_j$ ,  $\check{q}_j$  will also not be known, and here we approximate it as  $q_j = \frac{p_j(1-p_j)}{2n_j}$ :

$$p_j | \breve{p}_j \sim Normal(\breve{p}_j, q_j)$$
 (4)

Conditional on knowing the population allele frequencies for all populations, the sample allele frequencies for each population are independent of each other. In vector notation:

$$\mathbf{p}|\tilde{\mathbf{p}} \sim MVN(\tilde{\mathbf{p}}, diag(\mathbf{q}))$$
 (5)

The population frequencies are, in turn, assumed to depend on some population-wide ancestral allele frequency e, as in Refoyo-Martinez et al. (2020):

$$\mathbf{\breve{p}} \sim MVN(e\mathbf{1}, \mathbf{e}(\mathbf{1} - \mathbf{e})\mathbf{F}) \tag{6}$$

where **1** is a vector of ones. If we make one further approximation and treat the variance of the conditional distribution as a constant that is not dependent on the mean, we can marginalize the population allele frequencies, and obtain:

$$\mathbf{p} \sim MVN(e\mathbf{1}, diag(\mathbf{q}) + \mathbf{e}(\mathbf{1} - \mathbf{e})\mathbf{F})$$
 (7)

We then mean-center the vector  $\mathbf{p}$ :

$$\mathbf{y} = \mathbf{p} - e\mathbf{1} \sim MVN(0, diag(\mathbf{q}) + \mathbf{e}(\mathbf{1} - \mathbf{e})\mathbf{F})$$
(8)

We multiply the mean-centered vector by the branch vector  $b_k$  for a branch of interest and obtain:

$$\mathbf{y}^{T}\mathbf{b} \sim Normal(0, \mathbf{b_k}^{T} diag(\mathbf{q}) \mathbf{b_k} + e(1 - e) \mathbf{b_k}^{T} \mathbf{F} \mathbf{b_k})$$
 (9)

Finally, we derive a statistic that follows a chi-squared distribution under neutrality:

$$\frac{((\mathbf{p} - \bar{p}\mathbf{1})^T \mathbf{b_k})^2}{\mathbf{b_k}^T diag(\mathbf{q}) \mathbf{b_k} + e(1 - e) \mathbf{b_k}^T \mathbf{F} \mathbf{b_k}} \sim \chi_1^2$$
(10)

If we use the mean sample frequency across populations  $\bar{p}$  as an estimate of the ancestral frequency e, and also use the empirical covariance matrix  $\hat{\mathbf{F}}$  as an estimate of the true covariance matrix  $\mathbf{F}$ , we can obtain a statistic that penalizes sites in which the number of sampled individuals for a given branch's subtended populations is low:

$$S_B' = \frac{((\mathbf{p} - \bar{p}\mathbf{1})^T \mathbf{b_k})^2}{\mathbf{b_k}^T diag(\mathbf{q}) \mathbf{b_k} + \bar{p}(1 - \bar{p}) \mathbf{b_k}^T \hat{\mathbf{f}} \mathbf{b_k}}$$
(11)