Supporting Information S1

Derivation of the variance of the sample allele frequency

Consider a population of M diploid individuals, from which N individuals are randomly sampled. Let the locus of interest have n > 1 alleles, denoted by A_i , $i \in \{1, 2, ..., n\}$. Also, let P_{ij} be the frequency of individuals in the population with alleles A_i and A_j , with $i, j \in \{1, 2, ..., n\}$. Thus, MP_{ij} is the number of individuals in the population with alleles A_i and A_j . The frequency of allele A_i in the population is denoted by p_i , whereas that in the sample is denoted by $p_{i,N}$ (a random variable). $p_{i,N} = Y_{i,N} / (2N)$, where $Y_{i,N}$ is the number of copies of allele i in the sample of size N. $Y_{i,N} = 2X_{ii} + \sum_{j=1,j\neq i}^{n} X_{ij}$, where X_{ij} is the number of individuals in the sample with alleles A_i and A_j . $Y_{i,N}$ can be written as $Y_{i,N} = \sum_j a_j X_{ij}$, where $a_j = 2$ for i = j and $a_j = 1$ for $i \neq j$. Using the formula for the variance of a sum of dependent variables, the variance of $Y_{i,N}$ is equal to:

$$\sigma^{2}[Y_{i,N}] = \sum_{j} a_{j}^{2} \sigma^{2}[X_{ij}] + \sum_{i \neq k} \sum_{k} a_{j} a_{k} \operatorname{Cov}[X_{ij}, X_{ik}],$$
 (S1)

where σ^2 and Cov denote the variance and covariance respectively. The X_{ij} 's, with $i, j \in \{1, 2, ..., n\}$, follow a multivariate hypergeometric distribution with parameters M, N and MP_{ij} . Thus,

$$\sigma^{2}\left[X_{ij}\right] = \left(\frac{M-N}{M-1}\right) N P_{ij} \left(1 - P_{ij}\right), \tag{S2}$$

and

$$\operatorname{Cov}\left[X_{ij}, X_{ik}\right] = -\left(\frac{M-N}{M-1}\right) N P_{ij} P_{ik}$$
 (S3)

[1]. Thus,

$$\begin{split} \sigma^{2}\left[Y_{i,N}\right] &= \left(\frac{M-N}{M-1}\right) N \left(\sum_{j} a_{j}^{2} P_{ij} \left(1-P_{ij}\right) - \sum_{k} \sum_{j \neq k} a_{j} a_{k} P_{ij} P_{ik}\right) \\ &= \left(\frac{M-N}{M-1}\right) N \left(a_{i}^{2} P_{ii} \left(1-P_{ii}\right) + \sum_{j \neq i} a_{j}^{2} P_{ij} \left(1-P_{ij}\right) - 2 \sum_{j \neq i} a_{i} a_{j} P_{ii} P_{ij} - \sum_{k \neq i} \sum_{j \neq i, k} a_{j} a_{k} P_{ij} P_{ik}\right) \\ &= \left(\frac{M-N}{M-1}\right) N \left(a_{i}^{2} P_{ii} \left(1-P_{ii}\right) + \sum_{j \neq i} a_{j}^{2} P_{ij} \left(1-P_{ij}\right) - 2 \sum_{j \neq i} a_{i} a_{j} P_{ii} P_{ij} - 2 \sum_{k \neq i} \sum_{j \neq i, k; j < k} a_{j} a_{k} P_{ij} P_{ik}\right) \end{split}$$

$$= \left(\frac{M-N}{M-1}\right) N \left(4P_{ii}\left(1-P_{ii}\right) + \sum_{j\neq i} P_{ij}\left(1-P_{ij}\right) - 2\sum_{j\neq i} (2P_{ii})P_{ij} - 2\sum_{k\neq i} \sum_{j\neq i,k;j< k} P_{ij}P_{ik}\right) \\
= \left(\frac{M-N}{M-1}\right) N \left(4P_{ii} - (2P_{ii})^{2} + \sum_{j\neq i} P_{ij} - \sum_{j\neq i} P_{ij}^{2} - 2\sum_{j\neq i} (2P_{ii})P_{ij} - 2\sum_{k\neq i} \sum_{j\neq i,k;j< k} P_{ij}P_{ik}\right) \\
= \left(\frac{M-N}{M-1}\right) N \left(\left(2P_{ii} + \sum_{j\neq i} P_{ij}\right) - \left((2P_{ii})^{2} + \sum_{j\neq i} P_{ij}^{2} + 2\sum_{j\neq i} (2P_{ii})P_{ij} + 2\sum_{k\neq i} \sum_{j\neq i,k;j< k} P_{ij}P_{ik}\right) + 2P_{ii}\right). \tag{S4}$$

 $p_i = P_{ii} + \sum_{j=1, j\neq i}^{n} (P_{ij}/2)$, and thus $2P_{ii} + \sum_{j\neq i} P_{ij} = 2p_i$. In addition, using the multinomial series [2]:

$$(2p_i)^2 = \left(2P_{ii} + \sum_{j \neq i} P_{ij}\right)^2 = (2P_{ii})^2 + \sum_{j \neq i} P_{ij}^2 + 2\sum_{j \neq i} (2P_{ii})P_{ij} + 2\sum_{k \neq i} \sum_{j \neq i, k; j < k} P_{ij}P_{ik} . \tag{S5}$$

Thus, (S4) can be simplified to

$$\sigma^{2}[Y_{i,N}] = \left(\frac{M-N}{M-1}\right)N(2p_{i}-4p_{i}^{2}+2P_{ii}).$$
 (S6)

Hence,

$$\sigma^{2}[p_{i,N}] = \frac{\sigma^{2}[Y_{i,N}]}{(2N)^{2}} = \frac{(M-N)(p_{i}-2p_{i}^{2}+P_{ii})}{2(M-1)N},$$
 (S7)

which is equivalent to equation (11).

References for Supporting Information S1

- 1. Johnson NL, Kotz S, Balakrishnan N (1997) Discrete multivariate distributions. Hoboken, USA, Wiley-Blackwell. 328 p.
- 2. Weisstein, EW. "Multinomial Series." From MathWorld--A Wolfram Web Resource. http://mathworld.wolfram.com/MultinomialSeries.html