A recursion relaion for online coviarnce of haplotype counts

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In this short write-up I document a recursion relation for single-pass, or online, estimation of the unbiased covariance of haplotype counts. This is necessary because our genetic data sets are just too large to load into memory and for inefficient algoritms. Indeed, such a relationship is well established for the covariance, [1], however I was unsure how this procedure would extend to my definition of haplotype count covariance. In what follows I define the haplotype count covariance, how for short, and then show the recursion relation that I'll implement in C++.

I begin by introducing the quantities of interest. Denote the expected number of haplotype p copies, at locus m, and for sample i as $x_{mi}^{(p)} \in [0,2]$. Importantly, the $x_{mi}^{(p)}$ is a real number as it represents an expected copy number, and not the actual copy number. The number of distinct haplotypes are the number of K founder strains in our heterogenous stock rat colony making $p \in \{1,2,\ldots,K\}$. A locus $m \in \{1,2,\ldots,M\}$ is any integer from 1 to the M total number of loci. Lastly, we assume that the population consists of N animals, i.e. samples.

As we are interested in the single-pass computation of the covariance, we must first define the haplotype mean and covariance statistics.

Definition 1. The mean of expected haplotype counts for haplotype p is

$$\bar{x}_{Mi}^{(p)} = \frac{1}{M} \sum_{m=1}^{M} x_{mi}^{(p)}.$$

Next the unbiased covariance over markers and haplotypes.

Definition 2. The unbiased covariance of expected haplotype counts over p distinct haplotypes and M markers is defined as

$$cov^{(M)}(x_i, x_j) = \frac{1}{M-1} \sum_{p=1}^K \sum_{m=1}^M \left(x_{mi}^{(p)} - \bar{x}_{Mi}^{(p)} \right) \left(x_{mj}^{(p)} - \bar{x}_{Mj}^{(p)} \right)$$

for which we will use $C_{ij}^{(M)}$ as shorthand.

The aim for this analysis is to write the mean Def 1 and covariance Def 2 as a recursion relation. That is, suppose there are a total of M' loci in which we want an estimate of the mean and covariance. The statistics will be computed by a single for loop over the total number of markers M'. Meaning that for any iteration $1 \leq M \leq M'$ we must figure out how to update the mean and covariance estimates from the previous M-1 observations. That is we recursively update our estimates of the mean and covariance. When we have updated our mean and covariance estimates for all M' markers, the for loop will terminate and we'll have our desired estimates all without loading the entire data set at one time.

In presenting the recursion, let us define the quantity $\delta_{Mi}^{(p)}$

Definition 3. The value $\delta_{Mi}^{(p)}$ is a quantity for updating the mean and covariances from iteration M-1 to iteration M of sample i and is defined as follows

$$\delta_{Mi}^{(p)} = x_{Mi}^{(p)} - \bar{x}_{(M-1)i}^{(p)}.$$

Result 1.

$$\bar{x}_{Mi}^{(p)} = \bar{x}_{(M-1)i}^{(p)} + \frac{\delta_{Mi}^{(p)}}{M}$$

Result 2.

$$C_{ij}^{(M)} = \frac{M-2}{M-1}C_{ij}^{(M-1)} + \frac{1}{M}\sum_{p=1}^{K} \delta_{Mi}^{(p)} \delta_{Mj}^{(p)}$$

References

[1] Wikipedia contributors, 2025. Last access 2025-01-15.