Clarifying the relationship between average excesses and average effects of allele substitutions

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INTRODUCTION

Since Fisher (1918), partitioning of the genotypic values at a locus into additive and dominance effects has been used for conventional quantitative genetic analyses and recently for mapping quantitative trait loci (QTL; see, e.g., Lynch and Walsh, 1998). Numerous statistical models have been proposed for such partitioning. Some of them are restricted to populations under Hardy–Weinberg equilibrium (HWE; see, e.g., Falconer and MacKay, 1996), including a special case of gene frequency being one half (Mather and Jinks, 1982). Others also adequately account for Hardy–Weinberg disequilibrium (HWD; e.g., Cockerham, 1954; Li, 1976; Zeng et al., 2005). Such contrasts enter the regression model as:

$$G_{ij} = \mu + \tilde{\alpha}w_j + \tilde{\delta}v_j,$$  \hspace{1cm} (1)

where $G_{ij}$ are the genotypic values, $\mu$ is the population mean, $\tilde{\alpha}$ and $\tilde{\delta}$ are the additive and dominance genetic effects, and $w_j$ and $v_j$ are, respectively, the coefficients for the additive and dominance contrasts.

In this context, the values 0 and 1 can naturally be used to indicate the presence of alleles $A_1$ and $A_2$ in the genotypes, leading to the genotype indicator variable $z_{ij}$ taking the values $z_{11} = 0$, $z_{12} = 1$, and $z_{22} = 2$ for $A_1A_1$, $A_1A_2$, and $A_2A_2$, respectively, and to the coefficients for the additive effects through $w_{ij} = z_{ij} - E(z)$, where $E(z)$ is the expectation of $z$ (see, e.g., Zeng et al., 2005). This indicator variable has thus a clear biological meaning – the gene content of one of the alleles, $A_2$. When using this indicator variable, the additive parameter is the average effect, i.e., $\tilde{\alpha} = \alpha$, and the dominance parameter is the dominance genetic effect $\tilde{\delta} = \delta$.

On the other hand, the average excesses of alleles in a population under HWD were proffered to further entail the effects of alleles due to correlations with other alleles in that population (Fisher, 1941). Aiming to allow for such correlations in our derivations, we here consider more general indexes. In particular, we introduce a constant $c$ as the ratio of the average effect over the average excess (cf. Eq. 3 of Fisher, 1941). Multiplying $z_{ij}$ by this constant leads to a new genotype indicator variable with $z_{11} = 0$, $z_{12} = c$, $z_{22} = 2c$, and

$$G_{ij} = c\mu + c\tilde{\alpha}w_j + 2c\tilde{\delta}v_j,$$  \hspace{1cm} (2)

For Fisher’s concepts of average effects and average excesses are at the core of the quantitative genetics theory. Their meaning and relationship have regularly been discussed and clarified. Here we develop a generalized set of one locus two-allele orthogonal contrasts for average excesses and average effects, based on the concept of the effective gene content of alleles. Our developments help understand the average excesses of alleles for the biallelic case. We dissect how average excesses relate to the average effects and to the decomposition of the genetic variance.

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and $z_{22} = 2c$. This new genotype indicator variable will serve to indicate the effective content of allele $A_2$ in the three genotypes, as it will be further illustrated below.

The use of effective gene contents for obtaining orthogonal contrasts under HWD is summarized in Table 1. Obtaining the coefficients for the orthogonal additive contrast, $w_{ij}$, as $z_{ij} - E(z)$, warrants that $\Sigma p_{ij}w_{ij} = 0$, where $p_{ij}, i = 11, 12, 22$, are the genotypic frequencies of the population (see, e.g., Cockerham, 1954). The coefficients for the orthogonal dominance contrasts, $w_{ij}$, are obtained to fulfill $\Sigma p_{ij}w_{ij} = 0$ and $\Sigma p_{ij}w_{ij} = 0$ (Álvarez-Castro and Carlborg, 2007). These are the deviations of the observed genotypic values from the expected values as predicted from the regression of the genotypic values on the effective gene contents.

Additive and dominance contrasts (e.g., the ones expressed in Table 1) can be conveniently expressed in matrix notation. This allows for a straightforward extension of the one locus model to and arbitrary number of loci with arbitrary epistasis under linkage equilibrium (LE; Tiwari and Elston, 1997). It has also been shown that the matrix notation enables straightforward transformations between parameters that have previously been expressed using appropriate contrasts (Álvarez-Castro and Carlborg, 2007).

Let thus $G$ be the vector of genetic effects, $E$ be the vector entailing the population mean and the additive and dominant parameters and $S$ be the genetic-effect design matrix entailing the contrasts that allow for a transformation between vectors $G$ and $E$. Then, just using the contrasts in Table 1 we obtain the matrix expression $G = S \cdot E$ as:

$$
\begin{pmatrix}
G_{11} \\
G_{12} \\
G_{22}
\end{pmatrix} =
\begin{pmatrix}
1 & -p_2c & \frac{p_1^2}{p_1} - \frac{1}{2}p_{12} \\
(p_1 - p_2)c & \frac{p_1^2}{p_1} - \frac{1}{4}p_{12} \\
2p_1c & -\frac{p_1}{p_1} + \frac{1}{2}p_{12}
\end{pmatrix}
\begin{pmatrix}
\mu \\
\delta
\end{pmatrix}
$$

where $p_i, i = 1, 2$, are the frequencies of the alleles, $p_i = p_{ii} + 1/2p_{ij}$, $j \neq i$.

A UNIFIED FRAMEWORK FOR AVERAGE EFFECTS AND AVERAGE EXCESSES

As mentioned above, the contrasts in Table 1 provide the average effects of allele substitutions when $c = 1$. It is thus not surprising that in this case Eq. 2 reduces to Álvarez-Castro and Carlborg (2007) Eq. 8 – for the average (additive and dominance) effects. For analyzing how (2) relates to the average excesses, we first recall their definition for one biallelic gene (following Fisher, 1941; Kempthorne, 1957):

$$
\begin{align*}
\alpha^*_1 & = \frac{p_{11}}{p_1} G_{11} + \frac{1}{2} p_{12} G_{12} - \mu \\
\alpha^*_2 & = \frac{1}{2} p_{12} G_{12} + \frac{p_{22}}{p_2} G_{22} - \mu
\end{align*}
$$

By inverting expression (2), it is easy to see that $\tilde{\alpha} = \alpha^*_2 - \alpha^*_1$, when $c = 1/(1 + F)$, with $F = 1 - p_{11}/2p_{12}$ being Wright’s (1965) fixation index, $F$, with the range of $-1 \leq F \leq 1$, reflects any departure from the HWE, toward either an excess or a deficiency of heterozygotes. We can thus rename $\tilde{\alpha} = \alpha^*, \tilde{\delta} = \delta^*$ when $c = 1/(1 + F)$. That is to say, Eq. 3 restores the definition of average excesses of the alleles for a biallelic locus. We will consequently refer to (2) with $c = 1/(1 + F)$ as the average-excess formulation of NOIA.

From the general expression (2), we have thus retrieved both the average effects and the average excesses as particular cases of the contrasts in Table 1, specifically with $c = 1$ and $c = 1/(1 + F)$, respectively. Therefore, by implementing the effective gene content $c$ we have actually made our model to capture the correlation between alleles that the average excesses account for. Further, using the relationship between the two values of $c$ (1 and 1/(1 + F)) we are also retrieving the relationship between average effects and average excesses reported by Kempthorne (1957), $a_i = \alpha^*_i/(1 + F)$, which actually applies to the case of multiple alleles (see also Templeton, 1987).

Evidently, the possible values of the function $1/(1 + F)$ depend on those of the fixation index, $F$. In particular, $c = 1/(1 + F)$ must always be positive and within the range $1/2 \leq c < \infty$ for the allowable values of $F$ ranging from complete homozygosity ($F = 1$) to complete heterozygosity ($F = -1$). When $F = 0$ (i.e., $c = 1$) we have the well-known case where the average effect and average excess are the same, that is under HWE. Since $c = 1/(1 + F)$ must always be positive, $\alpha$ and $\alpha^*$ will always have the same sign and will verify $|\alpha| = c|\alpha^*|$. Taking all this into account, Table 2 summarizes how the fixation index affects the relationships between average excesses and additive genetic effects under three situations: heterozygote deficiency ($F < 0$), HWE ($F = 0$) and heterozygote excess ($F > 0$). Within that table, we also stress that the mathematical relationship between average excesses and additive genetic effects does not depend upon which one(s) of all potential biological features is (are) underlying a particular set of observed genotype frequencies.

| Heterozygotes deficiency | Observed heterozygotes fit HWE | Heterozygotes excess |
|--------------------------|------------------------------|---------------------|
| 0 \leq F \leq 1          | $F = 0$                      | $-1 \leq F < 0$     |
| 1/2 \leq c < 1           | $c = 1$                      |                      |
| $|\alpha^*| > |\alpha|$  | $|\alpha^*| = \alpha$        |                      |
| Assortative mating        | Random mating and either no selection or geometric fitnesses |
| or homozygotes favored or population structure | |

The non-zero constant $c$ is introduced for accounting for effective gene contents.
PARTITIONING THE GENOTYPIC VALUES AND THE GENETIC VARIANCE

The average-excess formulation [expression (2) with \( c = 1/(1 + F) \)] comes from a linear regression (1) and it can thus be expressed by means of its intercept, \( \mu \), and its regression coefficient, \( \alpha' \), as:

\[
\hat{G}(w) = \mu + \alpha' w
\] (4)

This regression entails a decomposition of the genotypic values in which the predictions from the regression are the additive components and the deviations of the regression – due to dominance interactions – are the dominance components. For instance, the predicted [by (4)] value for genotype A1A1 is \( G_{11} = \hat{G}(w) \). Now, both Table 1 and expression (2) show that the dominance contrasts, \( \nu_0 \), do not depend upon the scaling factor \( c \) and, hence, they are equal for the statistical and the statistical excess formulations. This implies that the dominance deviations are the same in both cases, i.e., \( \delta_{ij} = \delta'_{ij} \) and that, therefore, \( \alpha_{ij} = \alpha'_{ij} = a_i + a_j \). That is to say, both formulations lead to the same decomposition of genotypic values,

\[
G_{ij} = \mu + \alpha_{ij} + \delta_{ij} = \mu + \alpha'_{ij} + \delta'_{ij}.
\] (5)

This is illustrated in Figure 1, where we show the graphical interpretation of the decomposition of genotypic values coming from the average excesses and compare it with the classical decomposition coming from the average effects (Fisher, 1918). Note, particularly, that the decomposition of genotype A1A1 into additive and dominance parts is the same regardless of which linear regression is used. Interestingly, although for the average effects formulation (with \( c = 1 \)) the predictions of the regression can be obtained by just summing up the appropriate average effects (see, e.g., Álvarez-Castro and Carlberg, 2007), this does not hold for the average-excess formulation [with \( c = 1/(1 + F) \)], i.e., \( \alpha'_{ij} \neq \alpha^*_{ij} + \alpha^*_{ij} \), unless the genotypic frequencies are under HWE.

The reason for this is also noted in Figure 1, where it can be seen that \( \alpha^*_{ij} \) and \( \alpha^*_{ij} \) associated to different values for the regression independent variable \( (\alpha_{ij} = \hat{G}(w) = \hat{G}(c)) \) whereas \( \alpha^*_{ij} + \alpha^*_{ij} = \hat{G}(1) \). The exact relationship between these values under HWD is straightforward from \( \alpha^*_{ij} = \alpha_{ij} = \alpha_i + \alpha_j \) and \( \alpha_{ij} = c \alpha^*_{ij} \) (Kempthorne, 1957), which lead to:

\[
\alpha^*_{ij} = c \left( \alpha^*_{ij} + \alpha^*_{ij} \right).
\] (6)

Although a common way to express and compute the additive variance under HWD entails both the average (additive) effects and the average excesses [see, e.g., expression (4.23a) in Lynch and Walsh, 1998], here we have shown that either formulation alone suffices to provide the additive variance under HWD. We recall that this is true as long as the formulations are built using contrasts that are appropriate to HWD – as the ones we are providing in this communication for both the biallelic case.

EFFECTIVE GENE CONTENT

Hardy–Weinberg disequilibrium implies that alleles become (either positively or negatively) correlated in zygotes as compared to the expected genotype frequencies under HWE. A deficiency of heterozygotes, for instance, causes alleles to become positively correlated, leading to their effective additive contribution to the genotypes of a population to be more extreme (i.e., further away from their expectation) than under HWE. Fisher (1941) noted that this is accounted for by the average excesses. We note that this is not in contradiction with the interpretation of the average excesses of one allele as the conditional average genotypic deviation of the individuals that received that allele from at least one parent (see, e.g., Templeton, 2006).

For the biallelic case, we can trace Fisher’s (1941) remark in our graphical interpretation (Figure 1). We first recall that although
both the average effects and the average excesses are linear regressions of the genotypic values (the regression dependent variable) as expressed in (1), each of them is regressed on a different independent variable. The independent variable of the formulation of average effects is the actual content of allele $A_2$ (which is in Figure 1 shown as rescaled by its expectation) whereas the independent variable of the average-excess formulation is the effective content of allele $A_2$ measured by a factor $c$. This factor being greater than one in our example ($c = 5/3$) reflects an excess of heterozygotes (particularly with $F = 2/5$) and makes the slope of the regression for the average excess, $\alpha'$, to be less steep than the one on the actual gene content, $\alpha$, as noted in Table 2. Conversely, a deficiency of heterozygotes would make the slope of the average-excess regression to become steeper than the one of the regression for the average effects. Thus, the effective gene content $c$ leads to the average excesses to reflect the effective contributions of the alleles to the genotypes of a population.

CLOSING PERSPECTIVE

In conclusion, we have showed here that Fisher’s (1941) definition of average excesses can be phrased within a new regression framework that also generalizes the average effects. This has enabled us to clarify the significance of the average excesses in different ways. First, we have expressed the average excesses in terms of matrix notation within the NOIA framework, which entails the extension of that theory to multiple loci with arbitrary epistasis under LE and allows us to easily transform between average excesses and other genetic parameters. Second, we have fully integrated the average excesses into the theory for the decomposition of the genotypic values and the genetic variance into additive and dominant components. Third, we have provided a graphical interpretation of the average excesses that is analogous to the one of the average effects. Finally, we interpret the factor determining the relationship between average effects and average excesses as the effective gene content of individuals, accounting not only for the effects of their alleles but also for how pairs of alleles are correlated in a particular population.

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