

Supplementary Materials

A novel analytical framework to quantify co-gradient and counter-gradient variation

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Supplemental Methods. VARIANCE PARTITIONING

In the main manuscript, we focus on estimating the effect size of Cov_{GE} and $\bar{\Delta}_{GE}$ and their significance. The effect size provides information about the strength of the pattern, which is a distinct type of information from the percent of variation in the phenotype explained by the following components: (i) genetic effects on phenotype, (ii) environmental effects on phenotypes, (iii) genetic x environment interactions, (iv) covariance between genetic and environmental effects on phenotypes, and (iv) residual error. Both Falconer (1989) and Conover and Schultz (1995) have previously discussed Cov_{GE} in this more traditional sense as the percent of variation in the phenotype explained by different variance components:

$$(1) \quad V_P = V_G + V_E + V_{GE} + xCov_{GE}$$

where $x = 2$.

Here, we show how to extend sums of squares (SS) calculations from a traditional analysis of variance to incorporate $SS_{Cov_{GE}}$, which can in turn be used to understand the percent of variation in phenotypes explained by different components. These calculations assume a fully factorial reciprocal transplant design. We do not advocate that these SS be used to test the significance of the variance components with a traditional F -test, because the presence of Cov_{GE} likely violates the assumption of independence among samples and complicates calculations of degrees of freedom. It is, however, useful to compare the percent of variation explained by different components to their effect sizes, because it furthers understanding of the relative influence of genetic differentiation and plasticity on the evolved patterns in the population.

In a reciprocal transplant experiment, there are g genotypes transplanted into e environmental patches, for a total of $g * e = n_{ge}$ genotype-environment combinations. In a fully factorial reciprocal transplant experiment, $g = e$, g is the number of levels of genotypes from $i = 1, 2, \dots, g$, and e is the number of levels of environments from $j = 1, 2, \dots, e$.

Assuming the equal sample sizes r ($k = 1, 2, \dots, r$) within each genotype-environment combination, the following sums of squares can be estimated as:

$$(2) \quad V_G = SS_G = re \sum_{i=1}^g (\bar{y}_i - \bar{y})^2$$

$$(3) \quad V_E = SS_E = rg \sum_{j=1}^e (\bar{y}_j - \bar{y})^2$$

$$(4) \quad V_{GE} = SS_{GE} = r \sum_{i=1}^g \sum_{j=1}^e (\bar{y}_{ij} - \bar{y}_i - \bar{y}_j + \bar{y})^2$$

$$(5) \quad V_{Cov_{GE}} = SS_{Cov_{GE}} = |xr \sum_{i=1}^g \sum_{j=1}^e (\bar{y}_i - \bar{y})(\bar{y}_j - \bar{y})I|$$

$$(6) \quad V_{error} = SS_{Error} = \sum_{i=1}^g \sum_{j=1}^e \sum_{k=1}^r (y_{ijk} - \bar{y}_{ij})^2$$

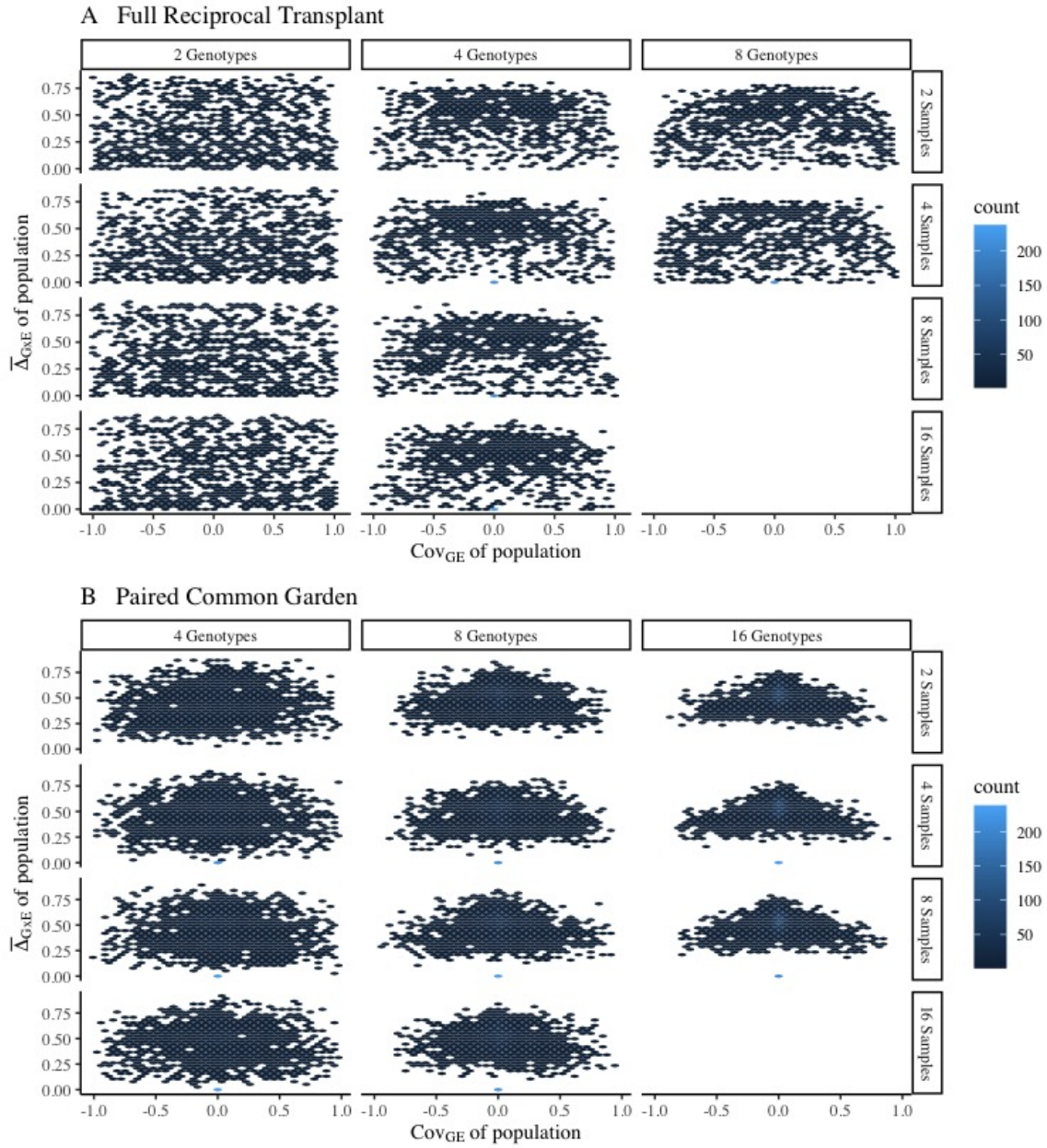
where $x = \frac{ge}{(\sum_{i=1}^g \sum_{j=1}^e I_{ij})}$, and I is an indicator variable that is 1 when the genotype i originated from environment j and 0 otherwise. In a 2x2 reciprocal transplant design, $g = 2$, $e = 2$, and $\sum_{i=1}^g \sum_{j=1}^e I_{ij} = 2$, so $x = 2$ as is assumed in Eq. S1. However in a 4x4 reciprocal transplant design, $g = 4$, $e = 4$, and $\sum_{i=1}^g \sum_{j=1}^e I_{ij} = 4$, so $x = 4$. The factor x ensures that the $SS_{Cov_{GE}}$ scales appropriately with the SS of the other components with the size of the experiment. Finally, since Cov_{GE} can be negative under countergradient scenarios, we take the absolute value for partitioning variance.

The percent of variation explained by each component (*Comp*) can then be estimated as

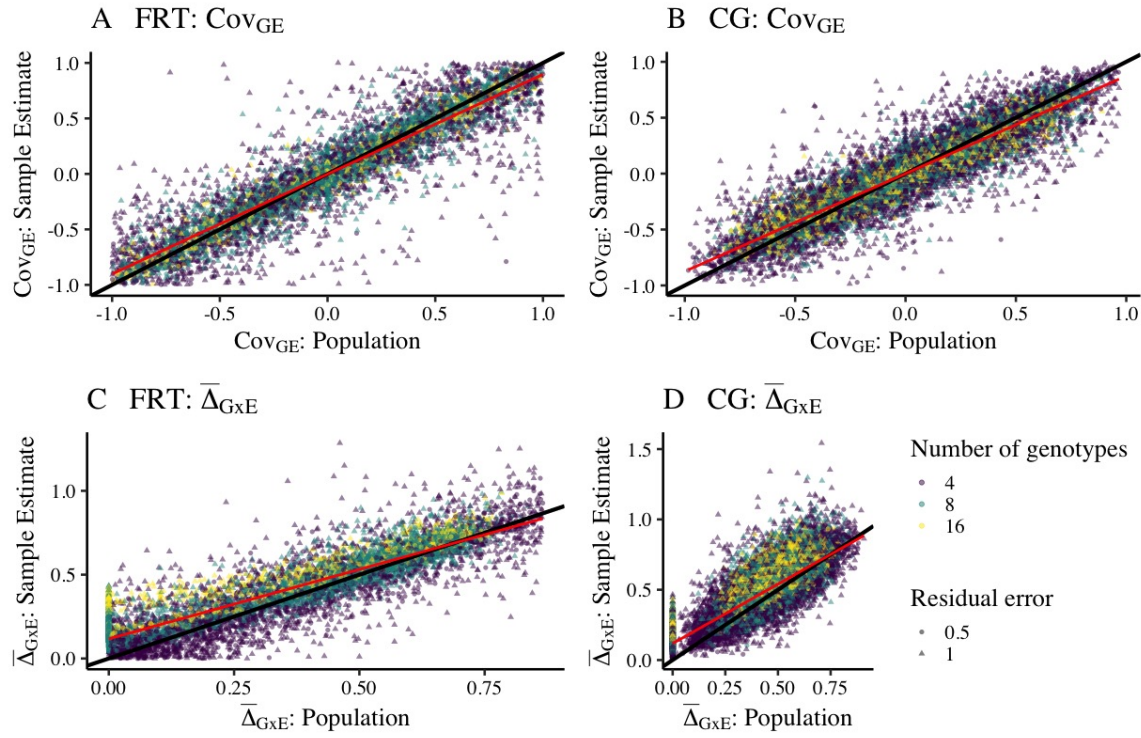
$$(7) \quad \eta_{Comp}^2 = \frac{SS_{Comp}}{SS_G + SS_E + SS_{GE} + SS_{Cov_{GE}} + SS_{Error}}$$

We recognize that this is a crude approach. However, it provides a reasonable way to compare to the percent of variation explained by different components to their effect size estimates in the main text. For example, when residual error is minimized and $|Cov_{GE}|$ is maximized, equal amounts of variance are explained by V_G , V_E , and $V_{Cov_{GE}}$ for the fully factorial reciprocal transplant experiment with an arbitrary number of populations.

Supplemental Figures. FIGURES



SUPPLEMENTARY FIGURE S1. Coverage of parameter space of Cov_{GE} and $\bar{\Delta}_{GE}$ for full reciprocal transplant (A) and paired common garden designs (B). Hexagons are colored according to the density of observations in each bin.



SUPPLEMENTARY FIGURE S2. Agreement between population measures and sample estimates of Cov_{GE} and $\bar{\Delta}_{G \times E}$ for Full Reciprocal Transplant (A, C) and paired Common Garden designs (B, D). The black line falls along a 1:1 line while the red line reflects the pattern of the data. Point colors indicate the number of genotypes, while point shapes indicate the level of residual variation. As expected, sample estimates deviate more from the population measure in situations with low sample sizes and higher residual variation.