

# Computational procedure for a weighted diallel analysis\*

D. A. Schaff<sup>1</sup>\*\*, G. A. Milliken<sup>2</sup> and C. D. Clayberg<sup>1</sup>

<sup>1</sup> Department of Horticulture and

<sup>2</sup> Department of Statistics; Kansas State University, Manhattan, KS 66506, USA

Received March 25, 1987; Accepted April 2, 1987 Communicated by A. R. Hallauer

**Summary.** A diallel analysis is described for the case in which the values of the characteristic used in an inheritance study have unequal variances. Such a characteristic can be a mean, a slope of a regression line, or the estimates of some parameter of a linear or nonlinear model. Computational formulae are presented which incorporate the necessary weighting along with the statistics of the Hayman-Jinks method for diallel analysis. The method described can also be used to perform a weighted diallel analysis based on means when there are unequal numbers of replications per cross. A simple example demonstrates the computations necessary to complete a weighted diallel analysis.

Key words: Heat tolerance – Killing temperature – Inheritance

#### Introduction

The diallel analysis has been used to study genetic relationships for variables which can be measured directly (e.g., yield, earliness, or flowering) utilizing the mean values of the parents and their crosses. This method may also be used to study the inheritance of phenotypic characteristics which cannot be measured directly, such as the slopes of regression models or other parameters from various models. The estimates of the parameters are obtained by fitting a model to the data from a parent or cross. One problem with the procedure is that the variances of the estimates of the parameters are not necessarily equal. Inequality of variances can occur when there are unequal numbers of observations per cross, say  $n_{ij}$  observations in  $cross_{ij}$ , and thus the variance of the mean response is  $\hat{\sigma}^2/n_{ij}$ . A second case occurs when different types of parameters are used in the diallel analysis, such as slopes of regression lines or other parameters from linear or non-linear models used to describe a particular relationship.

The objective of this paper is to modify the equations of Hayman (1954) and Jinks (1954) to conduct a diallel analysis when the estimates of the parameters have unequal variances. The application we use for illustration is a study of the inheritance of heat tolerance as measured by the killing temperature (Schaff 1984).

## Materials and methods

To estimate the killing temperature, defined as the temperature at which 50% injury occurs, leaf samples from acclimated bean *Phaseolus vulgaris* L. plants were subjected to heat treatments of 44 °C to 52 °C (Schaff 1984), with incubation after heat treatments changed to 4 h of shaking at 10 °C to reduce the time needed to run the test. The degree of injury, evaluated as electrolyte leakage, was modeled as a function of temperature (Schaff 1984) by the sigmoidal equation:

$$Injury = 1/(1 + e^{-B(temperature - U)}) + \varepsilon.$$
(1)

The parameter U is the temperature at which 50% injury occurs. The parameters of the model, estimated from the data (Schaff 1984), were the  $U_{ij}$ , having variance  $\hat{\sigma}^2 K_{ij}^2$ . The value of  $K_{ij}$  depends on the model and the temperatures at which the data were collected. The estimate of  $\hat{\sigma}^2$  is obtained from the plant within

<sup>\*</sup> Journal paper no. 87-117-J of the Kansas Agricultural Experiment Station. This study was conducted by the senior author in partial fulfillment for the PhD degree in genetics

<sup>\*\*</sup> Current address: USDA, ARS, PGGI, Germplasm Quality and Enhancement Laboratory, BARC-West, Beltsville, MD 20705, USA

cultivar (ERROR A) mean square (Schaff 1984). Using the estimate of  $U_{ij}$ ,  $K_{ij}^2$ , and  $\hat{\sigma}^2$ , we can construct a weighted diallel analysis to study the inheritance of heat tolerance as measured by the killing temperature.

Let  $U_{ij}$  denote the estimate of the parameter from a model fit to data from the  $ij^{th}$  cross. Then the entry in the diallel table for the  $ij^{th}$  cross is  $U_{ij}$ , where the variance of  $U_{ij}$  in general is  $\hat{\sigma}^2 K_{ij}^2$ . The term  $K_{ij}^2$  is a constant associated with a particular model and depends on the values of independent variables. If  $U_{ij}$  is a mean then  $K_{ij}^2$  equals  $1/n_{ij}$ .

### Weighting procedure

The average weight  $(K_{ij}^2)$  value is Koo =  $\sum\limits_i \sum\limits_j K_{ij}^2/p^2$ 

where p = number of parents. As in the notation of Hayman (1954) and Jinks (1954), the variances and covariances of the diallel are:

Vr	= variance of one array,
Wr	= covariance between the parents and their
	offspring in one array,
V <sub>ILI</sub>	= mean variance of the arrays,
WOLOI	= mean covariance between the parents
0201	and the arrays.
V <sub>OLO</sub>	= variance of the parents,
Voli	= the variance of the means of the arrays,
0.2.	and
$(M_{LI} - M_{LO})$	= difference between the mean of the
	parents and the mean of their p <sup>2</sup> progeny.

The following computational equations are needed to evaluate these necessary statistics when the variances of the entries in the diallel table are unequal. The variance of the data from the r<sup>th</sup> parent's crosses is:

$$V_r = (1/(p-1)) \left[ \sum_{i} (U_{ir}/K_{ir})^2 - (U_r^2 \left( \sum_{i} 1/K_{ir}^2 \right)) \right] Koo$$

where  $U_r = \left(\sum_i U_{ir}/K_{ir}^2\right) / \sum_i 1/K_{ir}^2$ , which is the weighted average of the data in column r.

The covariance between the parents and their

progeny in the r<sup>th</sup> column of data is:  

$$W_r = (1/(p-1)) \left[ \sum_{i} (U_{ii} - U_p) (U_{ir} - U_r) / K_{ii} K_{ir} \right] Koo.$$

Then 
$$V_{ILI} = 1/p \sum_{r} V_r$$
 and  $W_{OLOI} = 1/p \sum_{r} W_r$ .

The variance of the parental values is:

$$V_{OLO} = (1/(p-1)) \left[ \sum_{i} (U_{ii}/K_{ii})^2 - \left( U_p^2 \left( \sum_{i} 1/K_{ii}^2 \right) \right) \right] \text{Koo},$$

where  $U_p = \left(\sum_{i} U_{ii}/K_{ii}^2\right) / \sum_{i} 1/K_{ii}^2$ , which is the weight-

ed average of the parental values.

Let  $S_r = \sum_i 1/K_{ir}^2$ . Then the variance of array means is:

$$V_{OLO} = (1/(p(p-1))) \left[ \sum_{r} U_r^2 S_r - \left( \left( \sum_{r} U_r S_r \right)^2 / \sum_{r} S_r \right) \right] Koo.$$

The square of the difference between the mean of the parents and the mean of their  $p^2$  progeny is:

$$(M_{LI} - M_{LO})^2 = \left[ \left( \left( \sum_{i} \sum_{j} U_{ij} / K_{ij}^2 \right) / \left( \sum_{i} \sum_{j} 1 / K_{ij}^2 \right) - U_p \right]^2.$$

The correlation coefficient between the parents  $(K_{ii})$  and the order of dominance  $(W_r + V_r)$  is:

$$\mathbf{r} = \sum_{\mathbf{r}} \left[ \left( (\mathbf{W}_{\mathbf{r}} + \mathbf{V}_{\mathbf{r}}) - \left( \sum_{\mathbf{r}} \left( (\mathbf{W}_{\mathbf{r}} + \mathbf{V}_{\mathbf{r}})/\mathbf{p} \right) \right) (\mathbf{U}_{\mathbf{rr}} - \mathbf{U}_{\mathbf{p}})/\mathbf{K}_{\mathbf{rr}} \right) \right] / \left( \sum_{\mathbf{r}} \left( \mathbf{W}_{\mathbf{r}}^2 + \mathbf{V}_{\mathbf{r}}^2 \right) - \left[ \sum_{\mathbf{r}} \left( \mathbf{W}_{\mathbf{r}} + \mathbf{V}_{\mathbf{r}} \right) \right]^2 / \mathbf{p} \right) \\ \cdot \left( \sum_{i} \left( \mathbf{U}_{ii} / \mathbf{K}_{ii} \right)^2 - \left( \mathbf{U}_{\mathbf{p}}^2 \left( \sum_{i} 1 / \mathbf{K}_{ii}^2 \right) \right) \right).$$

The error term is multiplied by Koo so that it is in the same units as the above statistics (V<sub>r</sub>, W<sub>r</sub>, etc.;  $\hat{E} = \text{Error A} \times \text{Koo}$ ). The remaining computations in the diallel analysis are unaffected by the unequal variances. The genetic components of the Hayman (1954) and Jinks (1954) model are:

- $\hat{D}$  = component of variation due to the additive effects of the genes =  $V_{OLO} - \hat{E}$ ,
- $\hat{F}$  = mean covariation of the additive and dominance effects over arrays

$$= 2 V_{OLO} - 4 W_{OLOI} - 2 (n-2) E/n,$$

 $\hat{H}_1$  = component of variation due to the dominance effects of the genes

$$= V_{OLO} - W_{OLOI} + 4 V_{ILI} - (3 n - 2) \ddot{E}/n,$$

- $\hat{H}_2$  = dominance indicated by asymmetry of positive and negative effects of genes =  $4 V_{ILI} - 4 V_{OLI} - 2 \hat{E}$ ,
- $\hat{h}^2 = \text{square of the dominance effects over all loci}$ in heterozygous phase in all crosses= 4 (M<sub>LI</sub> - M<sub>LO</sub>)<sup>2</sup> - 4 L E, where $L = ((\sum_i \sum_j 1/K_{ij}^2) - (\sum_i 1/K_{ii}^2))/$  $((\sum_i \sum_j 1/K_{ij}^2) (\sum_i 1/K_{ii}^2)).$

Our components are estimated as:

- $\hat{E}$  = expected environmental component of variation = Error A × Koo, and
- Koo = average value of the  $J_{ij}^2$ 's where  $\hat{\sigma}^2 K_{ij}^2$  is the variance of  $U_{ij}$ .

The expected values of the components of variation are obtained by least squares analysis and assuming that  $\hat{E} = E = E'$ . The standard errors of the genetic components are computed using the equation in Hayman (1954).

# Example

The following example uses four parents:  $P_1$  ('UI114'),  $P_2$  (PI 324607),  $P_3$  (PI 271998), and  $P_4$  ('Valley'), to

		$\mathbf{P}_1$	P <sub>2</sub>	P <sub>3</sub>	$P_4$
P <sub>1</sub>	U	47.05	46.44	47.40	47.55
	K	1.598	1.959	1.192	1.683
P <sub>2</sub>	U	46.44	46.02	48.62	47.00
	K	1.959	1.736	2.056	1.329
P <sub>3</sub>	U	47.40	48.62	49.60	46.83
	K	1.192	2.056	2.083	2.250
P <sub>4</sub>	U	47.55	47.00	46.83	46.27
	K	1.683	1.329	2.250	1.637

Table 1.  $F_1$  hybrid and parental data in a four parent diallel cross for the killing temperature, U, and its coefficient of the standard deviation K

**Table 2.** Nonlinear analysis of covariance of killing temperature for parents and  $F_1$  hybrids using conductivity test

Source	d. f.	Mean squares
Cross	18	0.2077 *
Plant (cross) – error A	58	0.0774
Error B	897	0.1109
Total	973	

\* Significant at 1% level

Table 3. Array covariances, variances, and their sums and differences

Array	Wr	Vr	$W_r + V_r$	$W_r - V_r$
1	0.3510	0.2423	0.5933	0.1087
2	1.3717	1.0773	2.4490	0.2944
3	1.0831	1.2655	2.3486	-0.1824
4	0.1123	0.3197	0.4320	- 0.2074

demonstrate the computations necessary to conduct the weighted diallel analysis. Table 1 presents the estimates of U, the killing temperature, which are the data to be analyzed, and the coefficient of its standard deviation, K. The analysis of variance, computed according to the method of Schaff et al. (1987), shows that there are significant differences between the parents and  $F_1$  hybrids for heat tolerance (Table 2). The estimates of the evironmental variance is provided by the Error A or  $\hat{\sigma}^2 = 0.0774$  and thus  $\hat{E} = 0.0774 \times 3.18$ .

The assumptions of the Hayman (1954) and Jinks (1954) diallel model are: (1) diploid segregation, (2) homozygous parents, (3) no differences between reciprocal crosses, (4) independent action of nonallelic genes, (5) no multiple allelism, and (6) genes independently distributed among parents.

The plants used in the diallel had diploid segregation and the parents were homozygous inbreds with no

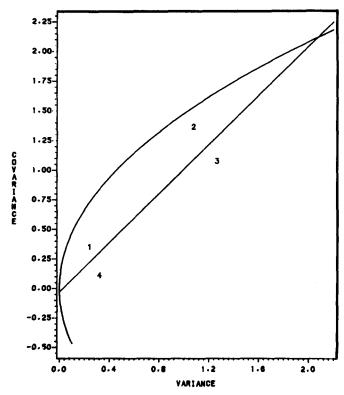


Fig. 1. Graph of variance vs covariance  $(V_r, W_r)$  of four bean cultivars for heat tolerance

differences between the reciprocal crosses. Therefore, reciprocals were combined, resulting in ten entries of four parents, and six hybrids. The remaining three assumptions will be tested with the techniques of the Hayman (1954) and Jinks (1954) diallel, weighted for unequal variances of the killing temperature. An examination of the variances and covariances of the arrays reveals that the t-test for heterogeneity of  $W_r - V_r$  is nonsignificant, with t = 0.2185 (Table 3). Also, the  $V_r$ ,  $W_r$  graph (Fig. 1) has a slope that is not significantly different from unity (slope = 1.0451 $\pm$  0.3261). Consequently, there is no information to indicate that the assumptions of independent action of nonallelic genes, no multiple allelism, and genes independently distributed between the parents were not valid.

Weighted estimates of the variances and covariances of the parents and crosses were:  $V_{OLO} = 2.2039$ ,  $W_{OLOI} = 0.7295$ ,  $V_{ILI} = 0.7262$ ,  $V_{OLI} = 0.2314$ , and  $M_{LI} - M_{LO}^2 = 0.0463$ . From these estimates the genetic components were computed (Table 4). Since the diallel analysis had less than 10 entries, these genetic estimates relate only to these parents and crosses and not to a larger population (Hayman 1963). The genetic components indicated that dominance was not significant because  $\hat{H}_1$ ,  $\hat{H}_2$ ,  $\hat{F}$ , and  $\hat{h}^2$  were not significant

**Table 4.** Genetic variance components for killing temperature of four parent common bean diallel.  $\hat{D}$  = additive effects of genes;  $\hat{H}_1$  = dominance effects of genes;  $\hat{H}_2$  = dominance indicated by asymmetry of positive and negative effects of genes;  $\hat{F}$  = covariance of dominance and additive effects;  $\hat{h}^2$  = square of the dominance effects over all loci in heterozygous phase in all crosses;  $\hat{E}$  = environmental error

Genetic components	Estimate and standard error
Ď	1.9579 ± 0.1908 *
Ê	$1.2439 \pm 0.4901$
$\hat{H}_1$ $\hat{H}_2$	$1.5757 \pm 0.5546$
$\hat{H}_2$	$1.4874 \pm 0.5119$
ĥ²	$0.0040 \pm 0.3472$
Ê	$0.2460 \pm 0.0853$

\* Significantly different from zero at 1% level

**Table 5.** Estimates of genetic parameters for killing temperature in common bean in greenhouse/growth chamber experiments.  $(H/D)^{1/2}$  = average degree of dominance;  $H_2/4 H_1$ = average frequency of positive vs negative alleles;  $K_D/K_R$ =  $((4D H_1)^{1/2} + F)/((4D H_1)^{1/2} - F)$  is the ratio of dominant to recessive alleles;  $h^2/H_2$  = the average number of factors involved; Her =  $1/4D/(1/4D - 1/4F + 1/4H_1 + E)$ 

Genetic components	Estimate and standard error		
$ \frac{(H/D)^{1/2}}{H_2/4H_1} K_D/K_R h^2/H_2 Her $	$\begin{array}{c} 0.8971 \pm 0.3555 \\ 0.2360 \pm 0.5228 \\ 2.0964 \pm 1.2923 \\ 0.0027 \pm 0.2334 \\ 0.5980 \pm 0.0866 \end{array}$		

(Tables 4 and 5), which suggests that most of the genetic variability was due to additive gene action. The coefficient of correlation between the parental degree of dominance  $(W_r + V_r)$  and the weighted parental values was nonsignificant (r = 0.3626 ± 0.6590), indicating

that there are as many positive as negative genes that show dominance. The graph (Fig. 1) shows that parents 1 and 4 had more dominant genes and parents 2 and 3 had more recessive genes.

This analysis can be used in studies where there are unequal numbers of replications per cross, such as that described by Gibori et al. (1978). Although they used an unequal number analysis of variance to obtain estimates of the variance components, they did not use the unequal numbers of replications from the means in the computation of the diallel analysis. Consequently, their analysis is inappropriate whereas the technique described here is more appropriate. A worked example for an unweighted diallel analysis is presented by Aksel and Johnson (1963).

#### References

- Aksel R, Johnson LPV (1963) Analysis of a diallel cross: a worked example. Adv Front Plant Sci 2: 37-53
- Gibori A, Hillel J, Cahaner A, Ashri A (1978) A 9×9 diallel analysis in peanuts (*A. hypogaea* L.): flowering time, tops' weight, pod yield per plant and pod weight. Theor Appl Genet 53:169-179
- Hayman BI (1954) The theory and analysis of diallel crosses. Genetics 39:789-809
- Hayman BI (1963) Notes on diallel-cross theory. In: Hanson WD, Robinson HF (eds) Statistical genetics and plant breeding. NAS-NRC 982, pp 571-578
- Jinks JL (1954) The analysis of continuous variation in a diallel cross of *Nicotiana rustica* varieties. Genetics 39: 767-788
- Schaff DA (1984) Screening for, and inheritance of, heat tolerance in common bean (*Phaseolus vulgaris* L.). PhD Dissertation, Kansas State University, Manhattan, Kansas
- Schaff DA, Milliken GA, Clayberg CD (1987) A method for comparing nonlinear models for data from a split-plot or repeated measures design. Biomet J (in press)