## ANALYSIS OF COVARIANCE

Experimental error is due to variability among experimental units. To increase precision, minimize experimental error.

Plan A: Blocking

Maximize differences among blocks and minimize differences within blocks
Does not work if heterogeneity does not follow a pattern, eg variation in plant stand, spotty soil heterogeneity
Blocking isolates variation known before the experiment is installed. Does not work if variability occurs after blocking, eg insect or disease damage

Plan B: Covariance analysis
Covariance isolates variation that occurs after the experiment is installed or that could not be isolated by blocking.
Corrects for variability in observations Y related to measured variability in factor $X$ - the covariate
Combines analysis of variance and regression analysis
Y can be adjusted linearly based on size of X for that observation
Can adjust for sources of bias in observational studies
Covariance is not interchangeable with blocking.
Covariance can be combined with blocking.

## Covariance

Measurements are made on the observed response Y and also on an additional variable X , called the covariate.

Covariate, X
Must be quantitative
Measures differences among experimental units
Can be measured before or during experiment
Assumed to be linearly related to $\mathrm{Y}(\mathrm{Y}=\mathrm{a}+\mathrm{bX})$
Analysis of covariance adjusts the values of Y for variation in the covariate X .
Example: An experiment measuring the yield of rice in high and low fertility treatments was affected by insect damage. Initial analysis showed a mean of 46.3 for the low fertility treatment and a mean of 38.4 for the high fertility treatment. Regression was done to determine the effect of insect damage on yield. After correcting for the effect of insect damage on yield, the high fertility treatment was shown to produce higher yields.

Yield of Two Fertility Treatments


Yield vs Score of Insect Damage
Score 1= little damage, 9 =most damage


- Low $=$ High


## Relationship of Covariate to Treatment

Covariate can be independent of treatment, eg initial weight of animals, insect infestation

Adjust for effects of covariate to reduce error
Covariate can be affected by treatment
Use to investigate mechanism or nature of treatment effects
eg water management of rice depth of water affects both grain yield and weed population is effect on grain yield primarily due to effect on weed population? is there a significant difference in yield after correcting for weed population?
eg how much of the increase in crop yield following manure application is due to nematode control?
eg physiological mechanisms: can the effect of growth hormone be explained by the increase in IGF-1?

Covariance should be used whenever there is an important independent variable that can be measured but can not be controlled by blocking. For example:

Plant stand
Severity of pest damage
Initial size of plants
Stage of development
Soil properties
Initial weight of animals
Food intake
Environmental effects
Uses of Covariance:

1. To assist in the interpretation of data.
2. To control error and increase precision.
3. To adjust treatment means of the dependent variable for differences in values of the corresponding independent variable.
4. To partition a total covariance or sum of cross products into component parts.
5. To estimate missing data.

## Mathematical Model

The mathematical model for covariance is that for the analysis of variance plus an additional term for the independent variable.

For the Randomized Complete Block Design

$$
Y_{i j}=\bar{Y}_{. .}+T_{i}+B_{j}+b\left(X_{i j}-\bar{X}_{. .}\right)+e_{i j}
$$

Where $Y_{\overline{\mathrm{Y}}}$ is the observed dependent variable
$\bar{Y} . . \quad$ is the experiment mean of the dependent variable
$T_{i}$ is the treatment effect
$B_{j} \quad$ is the block effect
b is the regression coefficient of the relationship between the dependent variable, Y and the independent variable, X
$\mathrm{X}_{\mathrm{ii}} \quad$ is the observed independent variable
$\bar{X}$.. is the experiment mean of the independent variable
$\mathrm{e}_{\mathrm{ij}} \quad$ is the residual variance or experimental error

## Assumptions in Covariance

1. The X's are fixed and measured without error. (Not always true)
2. The regression of $Y$ on $X$ after removal of block and treatment differences is linear and independent of treatments and blocks.
3. The residuals are normally and independently distributed with zero mean and a common variance.

## Steps in Covariance Analysis - RCBD

1. Construct ANOVA tables as RCBD for X , independent variable or covariate, and for Y , dependent variable
2. Check for treatment effect on $X$ and on $Y$ using F-test
3. Calculate sums of cross-products
4. Construct Analysis of Covariance table including sums of squares for $X$ and $Y$, and sums of cross-products. Include Trt+Err df, SSX, SP and SSY
5. Calculate SSRegr (adj for trt) and SSDev(Regr+Trt)
6. Calculate SSRegr (trt + err) and SSTrt (adj for regr)
7. Complete the Analysis of Covariance table and test MSRegr (adj for trt) and MSTrt (adj
for regr) against MSDevRegr (the remaining error)
8. Adjust treatment means

## Example Problem - Covariance in a RCBD

Initial Weights, X, and Kidney Fat, Y, for 16 steers given 4 hormone treatments in an RCBD.

| Block | Hormone |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 1 |  | 2 |  | 3 |  | 4 |  |
|  | X | Y | X | Y | X | Y | X | Y |
| 1 | 56 | 133 | 44 | 128 | 53 | 129 | 69 | 134 |
| 2 | 47 | 132 | 44 | 127 | 51 | 130 | 42 | 125 |
| 3 | 41 | 127 | 36 | 127 | 38 | 124 | 43 | 126 |
| 4 | 50 | 132 | 46 | 128 | 50 | 129 | 54 | 131 |
| Total | 194 | 524 | 170 | 510 | 192 | 512 | 208 | 516 |


| ANOVA for $X$ |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
|  | Hormone |  |  |  |  |
| Block | 1 | 2 | 3 | 4 | Total |
| 1 | 56 | 44 | 53 | 69 | 222 |
| 2 | 47 | 44 | 51 | 42 | 184 |
| 3 | 41 | 36 | 38 | 43 | 158 |
| 4 | 50 | 46 | 50 | 54 | 200 |
| Total | 194 | 170 | 192 | 208 | 764 |

ANOVA Table for X

| Source | df | SSX | MS | F | $F_{.05}$ |
| :--- | ---: | ---: | ---: | ---: | ---: |
| Total | 15 | 973 |  |  |  |
| Block | 3 | 545 | 181.67 | 6.73 | 3.86 |
| Hormone | 3 | 185 | 61.67 | 2.28 | 3.86 |
| Error | 9 | 243 | 27.00 |  |  |


| ANOVA for $Y$ |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
| Hormone |  |  |  |  |  |
| Block | 1 | 2 | 3 | 4 | Total |
| 1 | 133 | 128 | 129 | 134 | 524 |
| 2 | 132 | 127 | 130 | 125 | 514 |
| 3 | 127 | 127 | 124 | 126 | 504 |
| 4 | 132 | 128 | 129 | 131 | 520 |
| Total | 524 | 510 | 512 | 516 | 2062 |

ANOVA Table for $Y$

| Source | df | SS | MS | F | $F_{.05}$ |
| :--- | ---: | ---: | ---: | ---: | ---: |
| Total | 15 | 12 | 7.75 |  |  |
| Block | 3 | 56.75 | 18.92 | 4.03 | 3.86 |
| Hormone | 3 | 28.75 | 9.58 | 2.04 | 3.86 |
| Error | 9 | 42.25 | 4.69 |  |  |

Calculation of Sums of Cross-Products

Correction factor, cf $=\frac{\Sigma X \Sigma Y}{n}=\frac{(764)(2062)}{16}=98460.5$
SPTotal $=\Sigma X Y-c f=(56)(133)+\ldots+(54)(131)-98460.5=295.5$
SPBlock $=\frac{\Sigma X_{. j} \Sigma Y_{. j}}{r}-c f=\frac{(222)(524)+(200)(520)}{4}-98460.5=173.5$
SPHormone $=\frac{\Sigma X_{i} . \Sigma Y_{i .}}{r}-c f=\frac{(194)(524)+(208)(516)}{4}-98460.5=36.5$
SPError $=$ SPTotal - SPHormone $=295.5-173.5-36.5)=85.5$

## ANALYSIS OF COVARIANCE

Sums of Squares and Products

| Source | df | SSX | SP | SSY |
| :--- | :---: | ---: | ---: | :---: |
| Total | 15 | 973 | 295.5 | 127.75 |
| Block | 3 | 545 | 173.5 | 56.75 |
| Hormone | 3 | 185 | 36.5 | 28.75 |
| Error | 9 | 243 | 85.5 | 42.25 |
| Hormone + Error | 12 | 428 | 122.0 | 71.00 |

$$
\text { SSRegr }=r^{2} * S S Y=\frac{S P^{2}}{S S X}
$$

For Y, SSTrt+Err = 71, with 12 df . This can be divided in 2 ways:

1) SSTrt + SSRegr (adj for trt) + SSDev $=71$
2) $\operatorname{SSRegr}+$ SSTrt (adj for regr) + SSDev $=71$
1. First adjust for Trt, then subdivide Error into Regr (adj for trt) and Dev from Regr and Trt

SSTrt $=28.75$
SSError(adj Trt) $=42.25$
SSRegr(adj Trt) $=\frac{\text { SPError }^{2}}{\text { SSXError }}=\frac{85.5^{2}}{243}=30.08$
SSDev = SSError(adj Trt) - SSRegr(adj Trt) $=42.25-30.08=12.17$
2. First adjust for Regr, then subdivide Error into Trt (adj for Regr) and Dev from Regr and Trt

SSRegr $=\frac{\text { SPTrt }^{\text {Error }}{ }^{2}}{\text { SSXTrt }+ \text { Error }}=\frac{122^{2}}{428}=34.78$
SSTrt(adj Regr) = SSTrt+Error - SSRegr - SSDev

$$
=71.00-34.78-12.17=24.05
$$

Note that Trt and Regr are not orthogonal, so the SS depends on which is calculated first. Ftests are conducted on the adjusted values.

| Source | df | SS | MS | $F$ | $F_{.05}$ | $F_{.01}$ |
| :--- | ---: | ---: | ---: | ---: | ---: | ---: |
| Trt+Err | $(12)$ | 71 |  |  |  |  |
|  |  |  |  |  |  |  |
| Trt | 3 | 28.75 |  |  |  |  |
| Regr (adj Trt) | 1 | 30.08 | 30.08 | 19.79 | 5.32 | 11.26 |
| Dev | 8 | 12.17 | 1.52 |  |  |  |
|  |  |  |  |  |  |  |
| Regr | 1 | 34.78 |  |  |  |  |
| Trt (adj Regr) | 3 | 24.05 | 8.02 | 5.27 | 4.06 | 7.59 |
| Dev | 8 | 12.17 | 1.52 |  |  |  |

Adjustment of Treatment Means

Adjusted Mean, $\bar{Y}_{\mathrm{i} .}(\operatorname{adj})=\overline{\mathrm{Y}}_{\mathrm{i} .}-\mathrm{b}\left(\overline{\mathrm{X}}_{\mathrm{i} .}-\overline{\mathrm{X}}_{. .}\right)$

$$
\mathrm{b}=\frac{\text { SPErr }}{\text { SSXErr }}=\frac{85.5}{243}=0.352
$$

| Diets | $\overline{\mathrm{Y}}_{\mathrm{i} .}$ | $\overline{\mathrm{X}}_{\mathrm{i} .}$ | $\overline{\mathrm{X}}_{\mathrm{i} .}-\overline{\mathrm{X}}_{\ldots}$ | $\mathrm{b}\left(\overline{\mathrm{X}}_{\mathrm{i} .}-\overline{\mathrm{X}}_{\mathrm{X}}\right)$ | $\overline{\mathrm{Y}}_{\mathrm{i} .}(\mathrm{adj})$ |
| :--- | ---: | :---: | :---: | :---: | :---: |
| 1 | 131.0 | 48.5 | 0.75 | 0.26 | 130.7 |
| 2 | 127.5 | 42.5 | -5.25 | -1.85 | 129.3 |
| 3 | 128.0 | 48.0 | 0.25 | 0.088 | 127.9 |
| 4 | 129.0 | 52.0 | 4.25 | 1.50 | 127.5 |
|  | Summary |  |  |  |  |

Adjusting for covariate (initial weight)
Reduced MSError, increasing precision
Increased MSTreatment, increasing power
Compensated for the fact that treatment 2 steers had lower mean initial weight and treatment 4 steers had a higher initial weight, chance occurrences which tended to obscure hormone effects.

## Interpretation of Covariance Analysis



If the regression of $Y$ on $X$ is not significant (or if $P>.1$ ), remove the covariate from the model. If the regression is significant, examine the effects of treatment on X and Y .

Treatment effects on:

| X | Y |  |  |
| :--- | :---: | :---: | :--- |
| Before <br> Covariance | After <br> Covariance | Conclusions |  |
| ns | sig | ns | Apparent treatment effect due to variation in X. |
| ns | ns | sig | True treatment effects obscured by variation in X. |
| sig | sig | ns | Apparent treatment effect may be due to a treatment <br> effect on X, which then affects Y (the mechanism of the <br> treatment action may be via X ). |
| sig | sig | sig | Treatment had significant effect on Y beyond that due <br> to variation in X. |

Using Covariance for Calculating Missing Data
(Adapted from Steel and Torrie)
Gives an unbiased estimate of treatment and error sum of squares
Leads to an unbiased test of treatment means
Analysis is convenient and simple
Mean ascorbic acid content of three 2 g samples of turnip greens in $\mathrm{mg} / 100 \mathrm{~g}$ dry weight

| Trt | Block (Day) |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | 1 |  | 2 |  |  | 3 | 4 |  | 5 |  | Total |
|  | Y | X | Y | X | Y | X | Y | X | Y X |  |  |
| A | 0 | 1 | 887 | 0 | 897 | 0 | 850 | 0 | 975 | 0 | 3609 |
| B | 857 | 0 | 1189 | 0 | 918 | 0 | 968 | 0 | 909 | 0 | 4841 |
| C | 917 | 0 | 1072 | 0 | 975 | 0 | 930 | 0 | 954 | 0 | 4848 |
| Totals | 1774 | 1 | 3148 | 0 | 2790 | 0 | 2748 | 0 | 2838 | 0 | 13298 |

## Procedure:

1. Set $\mathrm{Y}=0$ for the missing plot
2. Define covariate as $\mathrm{X}=0$ for an observed Y , and $\mathrm{X}=+1$ (or -1 ) for $\mathrm{Y}=0$
3. Carry out the analysis of covariance to obtain the error sums of squares and products
4. Compute $\mathrm{B}=\mathrm{SP} / \mathrm{SSX}$ and change sign to estimate the missing value.

| Source | df | ANCOVA |  |  |
| :---: | :---: | :---: | :---: | :---: |
|  |  | SSX | SP | SSY |
| Total | 14 | 0.9333 | -886.53 | 945.296 |
| Block | 4 | 0.2667 | -295.20 | 359.823 |
| Treatment | 2 | 0.1333 | -174.73 | 203.533 |
| Error | 8 | 0.5333 | -426.60 | 381.940 |
| $b=S$ <br> Chan | $\begin{aligned} & X= \\ & \text { e sig } \end{aligned}$ | $\begin{aligned} & 60 / 0.533 \\ & 300(-1)= \end{aligned}$ | $799.92=$ |  |

800 is the missing value.
Redo the ANOVA with this value. Subtract 1 df from error value and total df. Can calculate several missing data by introducing a new independent variable for each missing datum and using multiple covariance.

