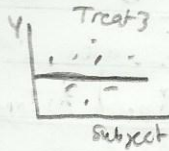
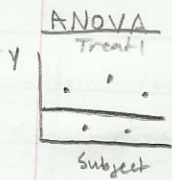
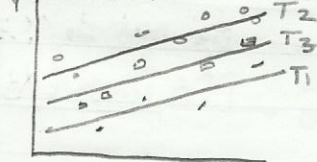


Chapter 22 ANCOVA



ANCOVA



* Reduced Variability

Concomitant variable added (unaffected by treatments) pre study

n_i = number of cases in the i^{th} level

$$n_T = \sum n_i$$

$$Y_{ij} = \mu_0 + \tau_i + \epsilon_{ij} \quad \text{ANOVA} \quad \downarrow \text{In text, we fit it differently}$$

$$Y_{ij} = \mu_0 + \tau_i + \gamma(X_{ij} - \bar{X}_{..}) + \epsilon_{ij} \quad \downarrow \quad \mu_0 + \tau_i + I_{i1} + \tau_2 I_{i2} + \gamma X_{ij} + \epsilon_{ij} \quad \text{ANCOVA}$$

μ_0 = overall mean

τ_i = fixed treatment effects $\rightarrow \sum \tau_i = 0$

γ = regression coefficient for the relation between Y and X

X_{ij} are constants (concomitant observations)

$\epsilon_{ij} \sim N(0, \sigma^2)$

$i = 1, \dots, r$; $j = 1, \dots, n_i$

$$E(Y_{ij}) = \mu_0 + \tau_i + \gamma(X_{ij} - \bar{X}_{..})$$

$$\sigma^2(Y_{ij}) = \sigma^2$$

[Note: this depends on concomitant variables so all observations of treatment won't have the same mean response]

• We can generalize this model for X 's that are random variables

• We can generalize for nonlinear Y, X relationships

ie quadratic: $Y_{ij} = \mu_0 + \tau_i + \delta_1(X_{ij1} - \bar{X}_{..1}) + \delta_2(X_{ij2} - \bar{X}_{..2}) + \epsilon_{ij}$

• We can add more concomitant observations

ie two: $Y_{ij} = \mu_0 + \tau_i + \delta_1(X_{ij1} - \bar{X}_{..1}) + \delta_2(X_{ij2} - \bar{X}_{..2}) + \epsilon_{ij}$

Assumptions ① Normality of error terms

② Constant error variance across treatments

③ Equality of slope of different treatment regression lines

④ Linearity of regression lines

⑤ Uncorrelated error terms

Hyp Test: $H_0: \tau_1 = \tau_2 = \dots = \tau_r = 0$

H_a : not H_0

$$F^* = \frac{MSR(X_2, \dots, X_{p-1} \mid X_1, \dots, X_{q-1})}{MSE}$$

Reject for large F^*

* If we fail to reject we're interested in looking at linear combinations of regression coefficients to investigate these effects.

* Look at Type III tests of fixed effects in SAS below

SAS

```
proc sgscatter; /* Gives Ancova scatterplot linear -> no interaction */  
  plot y * concomitant / group = x loess;
```

```
run;
```

```
proc glimmix; /* Get regression equation */
```

```
  class x;
```

```
  model y = x concomitant / solution; /*  $Y_{ij} = \mu_{..} + \tau_i I_{ij1} + \dots + \gamma x_{ij} + \epsilon_{ij}$  */
```

```
  lsmeans x "mu_1 - mu_2" 1 - 1 0 / adjust = t cl alpha = .05; /* treat diff? */
```

```
  lsmeans x; / gives  $\mu_1, \mu_2, \dots$  estimates
```

```
run;
```

```
proc glimmix; /* test  $x^*$  concomitant = 0  $\equiv$  parallel lines in Ancova scatter */
```

```
  class x;
```

```
  model y = x | concomitant;
```

```
run;
```

Covariance Model for Two Factors

$$Y_{ijk} = \mu_{..} + \alpha_i + \beta_j + (\alpha\beta)_{ij} - \epsilon_{ijk} \quad \text{ANOVA}$$

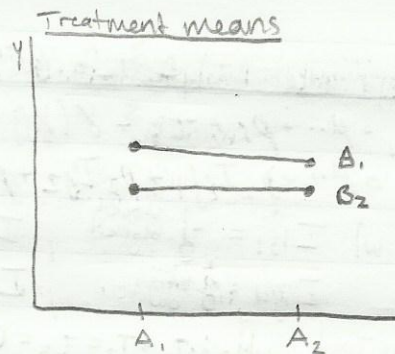
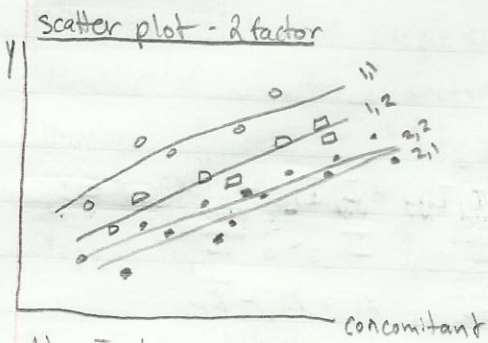
$$Y_{ijk} = \mu_{..} + \alpha_i + \beta_j + (\alpha\beta)_{ij} + \gamma(X_{ijk} - \bar{X}_{...}) + \epsilon_{ijk}$$

Intend we fit it differently

$$\rightarrow = \mu_{..} + \alpha_1 I_{ijk1} + \beta_1 I_{ijk2} + (\alpha\beta)_{11} I_{ijk1} I_{ijk2} + \gamma X_{ijk} + \epsilon_{ijk}$$

$$I_1 = \begin{cases} 1 & \text{level 1 factor A} \\ -1 & \text{level 2 factor A} \end{cases}$$

$$I_2 = \begin{cases} 1 & \text{level 1 factor B} \\ -1 & \text{level 2 factor B} \end{cases}$$



Hyp Test

H_0 : no an interaction

H_a : \exists interaction

$$F^* = \frac{SSR(I_1, I_2 | X, I_1, I_2)}{MSE}$$

Reject for large F^*

Linear Combinations

$$\left. \begin{array}{l} \alpha_1 - \alpha_2 \\ \beta_1 - \beta_2 \end{array} \right\}$$

Comparing main effects factor levels

SAS data twofactor;

set _____;

group = '11';

if $x_1 = 1$ and $x_2 = 2$ then group = '12';

if $x_1 = 2$ and $x_2 = 1$ then group = '21';

if $x_1 = 2$ and $x_2 = 2$ then group = '22';

run;

proc sgscatter data=twofactor;

plot y^* concomitant / group=group loess;

run;

proc glm; /*check type 3 test for interaction*/

class x_1 x_2 ;

model $y =$ concomitant $x_1 | x_2$ /solution;

run;

proc glimmix; class x_1 x_2 ; /*Comparing main effects*/

model $y =$ concomitant x_1 x_2 /solution;

lsestimate x_1 " $\alpha_1 - \alpha_2$ " 1 -1 / cl $\alpha = .05$;

lsestimate x_2 " $\beta_1 - \beta_2$ " 1 -1 / cl $\alpha = .05$; run;

Covariance Analysis of RCBD

$$\begin{aligned}
 Y_{ij} &= \mu_{..} + \rho_i + \tau_j + \delta(X_{ij} - \bar{X}_{.j}) + \epsilon_{ij} \\
 &= \mu_{..} + \rho_1 I_{ij1} + \rho_2 I_{ij2} + \rho_3 I_{ij3} + \tau_1 I_{ij4} + \tau_2 I_{ij5} + \delta X_{ij} + \epsilon_{ij} \\
 \text{w/ } I_{ij1} &= \begin{matrix} \text{block 1} \\ \text{block 4} \\ 0 \\ \text{ow} \end{matrix}, \quad I_{ij2} = \begin{matrix} \text{block 2} \\ \text{block 3} \\ 0 \\ \text{ow} \end{matrix}, \quad I_{ij3} = \begin{matrix} \text{block 3} \\ \text{block 4} \\ 0 \\ \text{ow} \end{matrix} \\
 I_{ij4} &= \begin{matrix} \text{treat 1} \\ \text{treat 3} \\ 0 \\ \text{ow} \end{matrix}, \quad I_{ij5} = \begin{matrix} \text{treat 2} \\ \text{treat 3} \\ 0 \\ \text{ow} \end{matrix}, \quad X_{ij} = X_{ij} - \bar{X}_{.j}
 \end{aligned}$$

Hyp Test $H_0: \tau_1 = \tau_2 = \tau_3 = 0$

$H_a: \text{not } H_0$

Use p value in type III table