

ANNOTATED SAS OUTPUT (ASO)

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Introduction

The present form of the Annotated SAS Output (ASO) has evolved from the original project aimed at illustrating common statistical methods using SAS (BU-664-M, BU-705-M). The primary goal of these annotated outputs has been to provide students with direction in using a statistical package to analyze data at the level of Statistics 602. This expanded version of the ASO should be of use to students as well as others outside the classroom.

Over the past five years there have been many people who have contributed to the ASO. Most of these people have been either students, lab instructors or undergraduate assistants involved with the conduct of Statistics 602. This list includes Anna Angelos, Suzanne Aref, Valerie Arneson, Calvin Berry, Foster Cady, Margaret Cecce, Patricia Firey, Laura Gnazzo, Jon Maatta, Charles McCulloch, Patricia Nolan, Norma Phalen, and Walter Piegorsch.

Description

A total of 15 data sets are employed to exemplify the application of SAS in the analysis of data. The data sets are derived from actual designed or observational experiments.

The name of the data set, which appears in the Table of Contents, and a description of the type of analysis that it serves to illustrate are provided below. The first 10 data sets are from Analyzing Experimental Data by Regression (Allen and Cady, 1982), while references are given for the sources of the remaining five.

- 1) Arsenic Data illustrate straight-line regression.
- 2) Firefly Data illustrate multiple linear regression with two predictor variables.

- 3) Soy milk Data illustrate polynomial regression and lack-of-fit from a completely randomized design.
- 4) Electricity Load Data illustrate a model sequence when several straight lines may need to be fitted.
- 5) Potato Leafhopper Data illustrate an analysis of treatment means via a complete set of orthogonal contrasts. The experiment design is a one-way classification in a completely randomized design with equal replication.
- 6) Lymphocyte Data illustrate an analysis of a 2×2 factorial experiment laid out in a completely randomized design with equal replication.
- 7) Fat Digestibility Data illustrate an analysis of a 2×2 factorial experiment laid out in a randomized complete block design.
- 8) Protein Nutrition Data illustrate an analysis of a one-way completely randomized design with unequal replication. A set of nonorthogonal contrasts are presented as well as a complete orthogonal set.
- 9) Swamp pH Data illustrate the analysis of cell means in a 2×3 factorial with unequal replication. The first analysis follows that presented in Allen and Cady (1982). The second is an analysis of unweighted means which is discussed in Snedecor and Cochran (1980). The Least Square Means of SAS are illustrated.
- 10) Soybean Physiological Data illustrate a covariate analysis. Several ways are shown to estimate treatment means adjusted and unadjusted for the covariate. The "classical" ANCOVA table is given as is a test for homogeneity of slopes.
- 11) Potato Scab Data illustrate an analysis of a 2×3 factorial experiment where one treatment factor is qualitative with two levels and the other treatment factor is quantitative with three unequally spaced levels. Two equivalent analyses are presented: the first is a model sequence approach (see Allen and Cady, Unit 19) comparing quadratic regression curves; the second is an

analysis of cell means wherein appropriate single degree-of-freedom orthogonal polynomial contrasts and interaction contrasts are estimated. These data are taken from a larger set presented in Cochran and Cox (1957, p. 97).

- 12) Alcohol-Drug Data illustrate the analysis of a split-unit experiment. The cell means model is fitted as described in Allen and Cady (1982, p. 280). The approach presented allows the simple effects to be estimated more easily than in a default model specification. The whole unit analysis proceeds by analyzing the sums and the split-unit analysis is accomplished after removing the whole-unit variability. Such an approach can result in substantial savings of computing dollars and a strengthening of the understanding of such experiments. The usual approach is presented for completeness.
- 13) Urea Synthesis Data illustrate the analysis of a repeated measures experiment with two treatment groups and two repeated measurements upon each experimental unit. The pertinent hypothesis tests may be reduced to three independent two-sample t-tests based upon the within-subject sums or differences. Such an approach tends to render the analysis quite transparent. The usual ANOVA table is also presented for completeness. These data are taken from Brogan and Kutner (1980).
- 14) Hemoglobin Data illustrate the analysis of a two-period crossover experiment. The calculations for this design are exactly the same as those of the preceding Urea Synthesis Data. The distinction between the two designs lies in the method of treatment allocation. These data are taken from Grizzle (1965).
- 15) Milk Yield Data illustrate the analysis of a three-period three-treatment crossover design which is balanced for first order carry-over effects. The layout is in repeated latin squares. Treatment means adjusted and unadjusted for carry-over effects are given as are their standard errors. These data are taken from Cochran and Cox (1957, p. 135).

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* See the description for references.

```
TITLE ARSENIC DATA; ← prints a heading on each page
DATA ARSENIC;
INPUT DIST ARSENIC;
DEV= DIST-16.1;
XO= 1;
CARDS;
```

```
2 3.19
4 3.26
8 1.82
10 1.02
12 1.85
15 2.05
21 1.34
23 0.79
30 0.66
36 0.30
```

These statements create a data set called ARSENIC, which has variables DIST, ARSENIC, DEV, and XO.
 ARSENIC is the observed arsenic level,
 DIST is the distance from the arsenic source,
 DEV is the difference between DIST and the mean distance, 16.1,
 XO is a column of ones.

(A) PROC PRINT;
 VAR XO DEV XO DIST; PROC PRINT prints the X matrices for parts (B) and (C) below.

```
PROC REG SIMPLE USCOR;
(B) MODEL ARSENIC= XO DEV/NOINT P SEGP SS1 SS2 CLM;
   OUTPUT OUT= NEW1 PREDICTED= YHAT1 RESIDUAL= RESID1
   LPSM= LOWER USPM= UPPER;
(C) MODEL ARSENIC= XO DIST/NOINT P SEGE SS1 SS2;
(D) MODEL ARSENIC= DIST/P SEGE SS1 SS2;
(E) MODEL ARSENIC= DIST/NOINT P SEGP SS1 SS2;
(F) MODEL ARSENIC= XO/NOINT;
   OUTPUT OUT= NEW2 PREDICTED= YBAR;
```

PROC REG is a regression procedure.
 Various models and options are illustrated in parts (B) through (F).
 The OUTPUT statements save the predicted and residual values
 in data sets called NEW1 and NEW2.

(G) PROC UNIVARIATE DATA= NEW1 PLOT FREQ NORMAL; PROC UNIVARIATE gives descriptive statistics for the residuals obtained in (B). DATA=NEW1 specifies
 VAR RESID1; which data set the variable to be used is contained in. If the data set is not specified SAS uses the
 last data set created.

```
PROC PLOT DATA= NEW1;
```

```
(H) PLOT RESID1*YHAT1/VREF= 0; ← VREF=0 draws a horizontal line at RESID=0.
(I) PLOT ARSENIC*DIST= '*' YHAT*DIST= 'F'
   UPPER*DIST= 'U' LOWER*DIST= 'L' /OVERRIDE;
```

Plot (H) is a residual plot for models (B), (C) and (D).
 Plot (I) prints observed, predicted, and confidence band values
 on the same set of axes using the symbols *, P, U, and L.

```
(J) DATA DECOMP; MERGE NEW1 NEW2;
DATA DECOMP; SET DECOMP;
RES1= YHAT1-YBAR;
```

DATA DECOMP is a new data set which contains the mean of the ARSENIC values,
 YBAR (from part (F)); the residuals left when the mean is subtracted from the
 predicted values, RES1; the ARSENIC values, and the residuals from (B). The
 MERGE statement combines data sets NEW1 and NEW2, and the SET statement
 allows the creation of the new variable RES1.

```
PROC PRINT DATA= DECOMP; ← PROC PRINT prints the data decomposition.
VAR ARSENIC YBAR RES1 RESID1;
TITLE DATA DECOMPOSITION FOR ARSENIC DATA;
```

OBS	The X matrix for the model ARSENIC = X0 DEV in part (B)		The X matrix for the model ARSENIC = X0 DIST in part (C)	
	X0	DEV	X0	DIST
1	1	-14.1	1	2
2	1	-12.1	1	4
3	1	-8.1	1	8
4	1	-6.1	1	10
5	1	-4.1	1	12
6	1	-1.1	1	15
7	1	4.9	1	21
8	1	6.9	1	23
9	1	13.9	1	30
10	1	19.9	1	36

The options SIMPLE and USSCP (Uncorrected Sums of Squares and Cross Products) in the PROC REG statement produce this output:

ARSENIC DATA

DESCRIPTIVE STATISTICS

	SUM	MEAN	UNCORRECTED	VARIAN	STD DEVIATI
ARSENIC	16.29000000	1.62900000	35.7268000	1.02477333	1.01231089
X0	10.00000000	1.00000000	10.0000000	0.00000000	0.00000000
DEV	0.00000000	0.00000000	1126.90000000	125.21111111	11.18977708
DIST	161.00000000	16.10000000	3719.00000000	125.21111111	11.18977708

SUMS OF SQUARES AND CROSSPRODUCTS

SSCP	ARSENIC	X0	DEV	DIST	INTERCEP
ARSENIC	35.7268	16.29	-88.068	174.04	16.29
X0	16.29	10	2.131635-14	161	10
DEV	-88.068	2.131635-14	1126.9	1126.9	2.131635-14
DIST	174.04	161	1126.9	3719	161
INTERCEP	16.29	10	2.131635-14	161	10

The circled numbers should be exactly zero.

SAS supplies an intercept which can produce the same sums of squares and cross products as X0 which was created as part of the data set.

MEAN AND SLOPE MODEL

ⓑ MODEL ARSENIC = XO DEV/NOINT P SEQB SS1 SS2 CIM;

SEQUENTIAL PARAMETER ESTIMATES ← The option SEQB produces the sequential b's.

XO 1.628 ← Coefficient for the mean model (see p.35).
 DEV 1.628 -0.078151 ← Coefficients for the mean and slope model
 (see pp. 39 and 45).

[The option NOINT instructs SAS not to supply an intercept since XO has been included in the model. When XO is in the model and the NOINT option is used, the model SS includes the SS for the mean.]

DEP VARIABLE: ARSENIC

ANOVA:	SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE	PROB > F
	MODEL	2	33.386414	16.693207	57.061	0.0001
	ERROR	8	2.340386	0.292548 = s ²		
	U TOTAL	10	35.726800	— U TOTAL stands for <u>UNCORRECTED</u> TOTAL. ROOT MSE is the standard deviation.		

ROOT MSE 0.540877 = s
 DEP MEAN 1.628000 = \bar{y}
 C.V. 33.22342
 (Coefficient of variation)

R-SQUARE 0.9345
 ADJ R-SQ 0.9263

← when NOINT is used, the reported R² is incorrect. See ASO, p.5 for the correct R².

NOTE: NO INTERCEPT TERM IS USED. R-SQUARE IS REDEFINED.

VARIABLE	DF	PARTIAL PARAMETER ESTIMATE	STANDARD ERROR	T FOR H0: PARAMETER=0	PROB > T	TYPE I SS
XO	1	1.628000 = \bar{y}	0.171040	9.518	0.0001	26.503840 = R(0)
DEV	1	-0.078151 = slope	0.016112	-4.850	0.0013	6.882574 = R(DEV 0)

The options SS1 and SS2 in the MODEL statement produce the TYPE I SS (the sequential SS) and the TYPE II SS (the partial SS).

VARIABLE	DF	TYPE II SS
XO	1	26.503840 = R(0 DEV)
DEV	1	6.882574 = R(DEV 0)

OBS	ACTUAL	PREDICT VALUE	STD ERR PREDICT	LOWER 95% MEAN	UPPER 95% MEAN	RESIDUAL
1	36190	2.730	0.284371	2.074	3.386	0.460075
2	36260	2.574	0.259352	1.976	3.172	0.686377
3	16820	2.261	0.215145	1.765	2.757	-0.441020
4	14020	2.105	0.197269	1.650	2.560	-1.085
5	16850	1.948	0.183354	1.526	2.371	-0.98418
6	26050	1.714	0.171956	1.317	2.111	0.336034
7	16340	1.245	0.188382	0.810647	1.679	0.094938
8	0.790000	1.029	0.203997	0.618339	1.559	-0.298760
9	0.660000	0.541706	0.281803	-0.108140	1.192	0.118294
10	0.300000	0.072801	0.363402	-0.765212	0.910615	0.227199

The P option in the MODEL statement prints each observed value, predicted value, and residual.

SUM OF RESIDUALS 7.67767E-15
 SUM OF SQUARED RESIDUALS 2.340386

Should be exactly zero.

The CIM option prints the upper and lower confidence limits for a 95% confidence interval on μ at each observed X.

Note that using DEV instead of DIST as the X variable produces sequential b's that are the same as the partials. See Ⓒ.

INTERCEPT AND SLOPE MODEL

(C) MODEL ARSENIC = X0 DIST/NOINT P SEQB SS1 SS2;

SEQUENTIAL PARAMETER ESTIMATES

X0 1.628 ← Coefficients of the mean and slope model.
 DIST 2.68623 -.078151 ← Coefficients of the intercept and slope model (see p.42).

The sequential coefficients are the diagonal elements of the SEQB output.

DEP VARIABLE: ARSENIC

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE	PROB > F
MODEL	2	33.366414	16.683207	57.061	0.0001
ERROR	8	2.340386	0.292548 = s ²		
U TOTAL	10	35.726800			
ROOT MSE		S=0.540677	R-SQUARE	0.9345	
DEP MEAN		Y=1.628000	ADJ R-SQ	0.9263	
C.V.		33.22342			

} The ANOVA is the same as in (B) MODEL ARSENIC = X0 DEV.

NOTE: NO INTERCEPT TERM IS USED. R-SQUARE IS REDEFINED. ← Same problem with R² as in (B). See ASO, p.5.

VARIABLE	DF	PARTIAL PARAMETER ESTIMATE	STANDARD ERROR	T FOR H0: PARAMETER=0	PROB > T	(SEQUENTIAL) TYPE I SS
X0	1	2.886226 ^{intercept}	0.310720	9.289	0.0001	25.503840 = R(0)
DIST	1	-0.078151 _{slope}	0.016112	-4.850	0.0013	6.882574 = R(DIST 0)

printed by SS1 option

VARIABLE	DF	(PARTIAL) TYPE II SS
X0	1	25.241773 = R(0 DIST)
DIST	1	6.882574 = R(DIST 0)

printed by SS2 option

OBS	ACTUAL	PREDICTED VALUE	RESIDUAL
1	3.190	2.730	0.460075
2	3.260	2.574	0.686377
3	1.820	2.261	-0.441020
4	1.020	2.105	-1.085
5	1.850	1.948	-0.098418
6	2.050	1.714	0.336034
7	1.340	1.245	0.094938
8	0.790000	1.088	-0.298760
9	0.660000	0.541706	0.118294
10	0.300000	0.072801	0.227199

Printed by P option, same values as in (B).

The parameter estimates are different from those in (B), but the TYPE I SS are the same.

PROC REG does not compute F tests for the TYPE I or TYPE II SS. The t tests of the parameter estimates are equivalent to F tests of the TYPE II SS. To perform the independent F tests of the TYPE I SS, form the ratio

SUM OF RESIDUALS 6.40994E-15 ← should be exactly zero
 SUM OF SQUARED RESIDUALS 2.340386 ← Error SS, see ANOVA above

$$\frac{(\text{TYPE I SS})/\text{df}}{(\text{Residual SS})/(\text{Residual df})} = F_{\text{Resid df}}^{\text{df}}$$

INTERCEPT AND SLOPE MODEL
 (D) MODEL ARSENIC = DIST/P SEQB SS1 SS2;

SEQUENTIAL PARAMETER ESTIMATES

INTERCEP 1.628
 DIST 2.88623 -0.078151 } same sequentials as (C)

[The NOINT option is not used, so SAS supplies an intercept. The model SS does not include the SS for the mean ($n\bar{y}^2$).

DEP VARIABLE: ARSENIC

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE	PROB > F
MODEL	1	6.882574	6.882574	23.526	0.0004
ERROR	8	2.340366	0.292548 = s^2 , same as (B) and (C)		
(C) TOTAL	9	9.222940			

Corrected total SS
 ROOT MSE $s = 0.540877$
 DEP MEAN $\bar{y} = 1.628000$
 C.V. 33.22342

R-SQUARE 0.7462 ← Correct R^2 . Compare with calculated R^2 below.
 ADJ R-SQ 0.7145 ← The adjusted R^2 is "adjusted" for the number of variables in the model.

VARIABLE	DF	PARTIAL PARAMETER ESTIMATE	STANDARD ERROR	T FOR H0: PARAMETER=0	PROB > T	(SEQUENTIAL) TYPE I SS
INTERCEP	1	2.886226	0.310720	9.289	0.0001	25.503840
DIST	1	-0.078151	0.016112	-4.850	0.0013	6.882574

↑ same as (C)

VARIABLE	DF	(PARTIAL) TYPE II SS
INTERCEP	1	25.241773
DIST	1	6.882574

↑ same as (C)

OBS	ACTUAL	PREDICT VALUE	RESIDUAL
1	34190	2.730	0.460075
2	34260	2.574	0.686377
3	16820	2.261	-0.441020
4	16020	2.105	-1.085
5	14850	1.948	-0.098418
6	24050	1.714	0.336034
7	14340	1.245	0.094936
8	0.790000	1.089	-0.298760
9	0.660000	0.541706	0.118294
10	0.300000	0.072801	0.227199

same as (B) and (C)

SUM OF RESIDUALS 8.408945-15
 SUM OF SQUARED RESIDUALS 2.340366

When NOINT is used, the reported R^2 is incorrect. $R^2 = \frac{\text{Corrected Model SS}}{\text{Corrected Total SS}}$ where both model and total SS have been corrected for the mean. NOINT causes the SS for the mean to be included in both numerator and denominator. To calculate the correct R^2 , subtract the SS for the mean ($n\bar{y}^2$) from the MODEL SS and TOTAL SS.

Ex.: Model (C) $n\bar{y}^2 = 26.503840$
 $R^2 = \frac{33.386414 - n\bar{y}^2}{35.726800 - n\bar{y}^2} = 0.7462$
 (compare to above R^2)

$R^2 = \frac{\text{Corrected Model SS}}{\text{Corrected Total SS}} = 1 - \frac{\text{RESIDUAL SS}}{\text{Corrected Total SS}}$

$R^2_{\text{adj}} = 1 - \frac{(\text{Resid SS}) * \left(\frac{n-1}{n-p}\right)}{\text{Corrected Total SS}}$ where n is the number of observations and p is the number of variables fitted, including the intercept or X0.

LINE THROUGH ORIGIN

SEQUENTIAL PARAMETER ESTIMATES

DIST .0467975 = slope of line through origin (see p. 53)

DEP VARIABLE: ARSENIC

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE	PROB > F
MODEL	1	8.144641	8.144641	2.658	0.1375
ERROR	9	27.582159	3.064684 = s ² , compare to previous models		
U TOTAL	10	35.726800			
ROOT MSE		1.750624	R-SQUARE	0.2280	
DEP MEAN		1.628000	ADJ R-SQ	0.2280	
C.V.		107.5322			

includes SS for the mean

NOTE: NO INTERCEPT TERM IS USED. R-SQUARE IS REDEFINED.

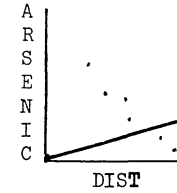
VARIABLE	DF	PARAMETER ESTIMATE	STANDARD ERROR	T FOR H0: PARAMETER=0	PROB > T	(SEQUENTIAL) TYPE I SS
DIST	1	0.046798	0.028706	1.630	0.1375	8.144641

VARIABLE	DF	(PARTIAL) TYPE II SS
DIST	1	8.144641

OBS	ACTUAL	PREDICT VALUE	RESIDUAL
1	3.190	0.093525	3.096
2	3.260	0.187190	3.073
3	1.820	0.374380	1.446
4	1.020	0.467975	0.552025
5	1.850	0.561570	1.288
6	2.050	0.701963	1.348
7	1.340	0.982748	0.357252
8	0.790000	1.075	-0.285000
9	0.660000	1.404	-0.743926
10	0.300000	1.685	-1.385

SUM OF RESIDUALS 8.745599
 SUM OF SQUARED RESIDUALS 27.58216

Ⓔ ARSENIC = DIST/NOINT;
 NOINT is used, so SAS does not supply the intercept, and X0 is not specified in the model. Result: fitting a straight line through the origin.



Such a model does not make much sense with the ARSENIC data set, although regression through the origin is useful in other circumstances.

RESIDUAL ANALYSIS

UNIVARIATE

VARIABLE=RESID1 RESIDUALS

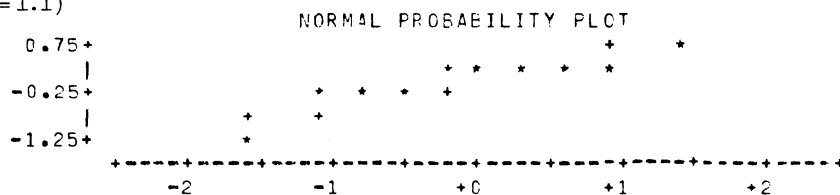
MOMENTS				QUANTILES (DEF=4)			
N	10	SUM WGT	10	100% MAX	0.686377	95%	0.686377
MEAN	7.078E-16	SUM	7.078E-15	75% Q3	0.367045	95%	0.686377
STD DEV	0.509944	VARIANCE	0.260043	50% MED	0.106616	90%	0.663747
SKEWNESS	-0.960192	KURTOSIS	1.18871	25% Q1	-0.334325	10%	-1.02035
USS	2.34039	CSS	2.34039	0% MIN	-1.08472	5%	-1.08472
CV	7.205E+16	STD MEAN	0.161258			1%	-1.08472
T:MEAN=0	4.389E-15	PROB> T	1	RANGE	1.7711		
W:NORMAL	0.948196	PROB<W	0.619	Q3-Q1	0.70137		
				MODE	-1.08472		

EXTREMES

LOWEST	HIGHEST
-1.08472	0.118294
-0.44102	0.227198
-0.29876	0.336034
-0.0984178	0.460075
0.0949383	0.686377

STEM LEAF	#	BOXPLOT
0 57	2	! } 25% of residuals
0 1123	4	-----
-0 431	3	+-----+
-0		
-1 1	1	! } 25% of residuals

(All numbers are in tenths,
i.e., -1 1=1.1)



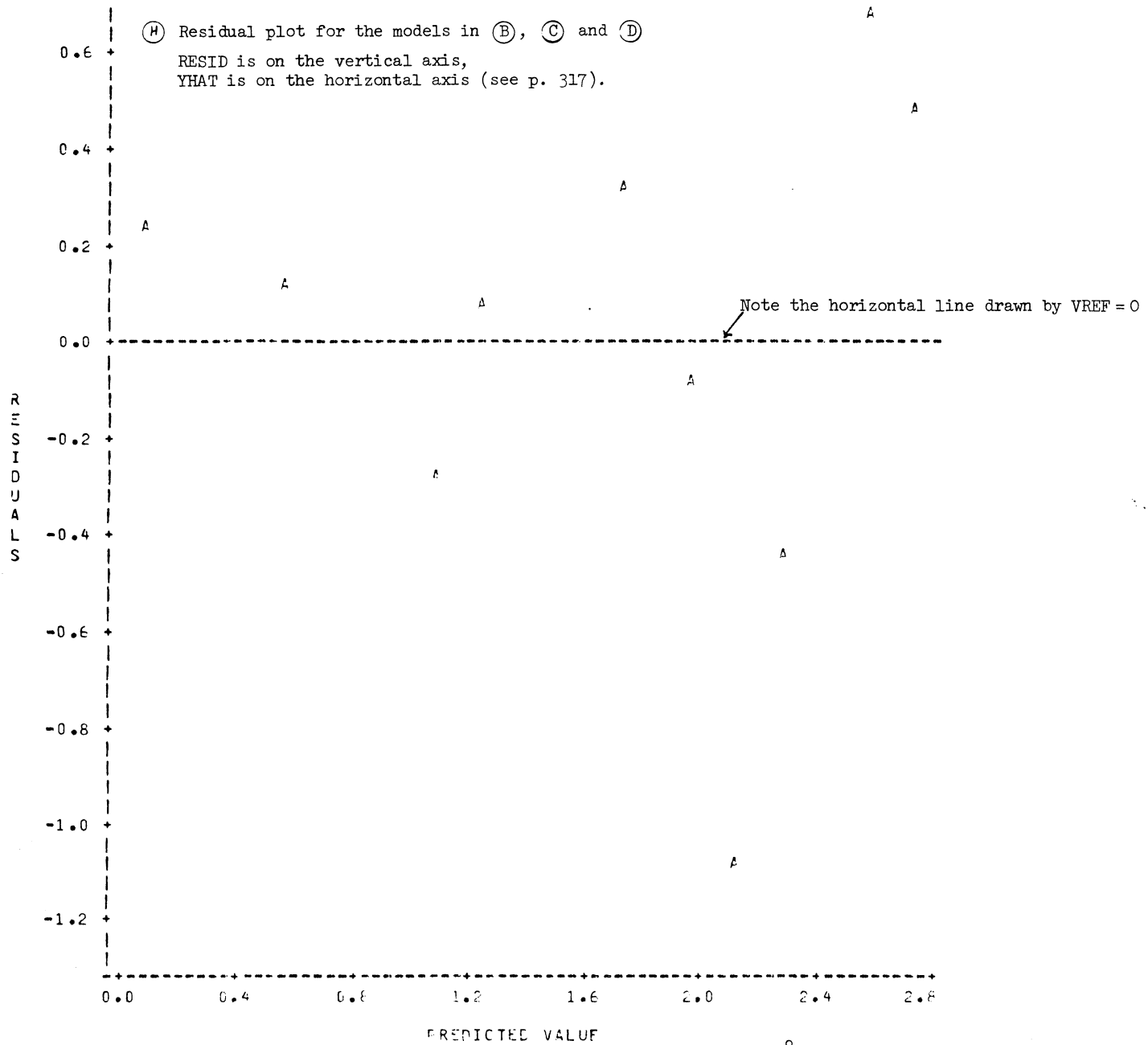
← should be ≈ a straight line if residuals are normally distributed.

FREQUENCY TABLE

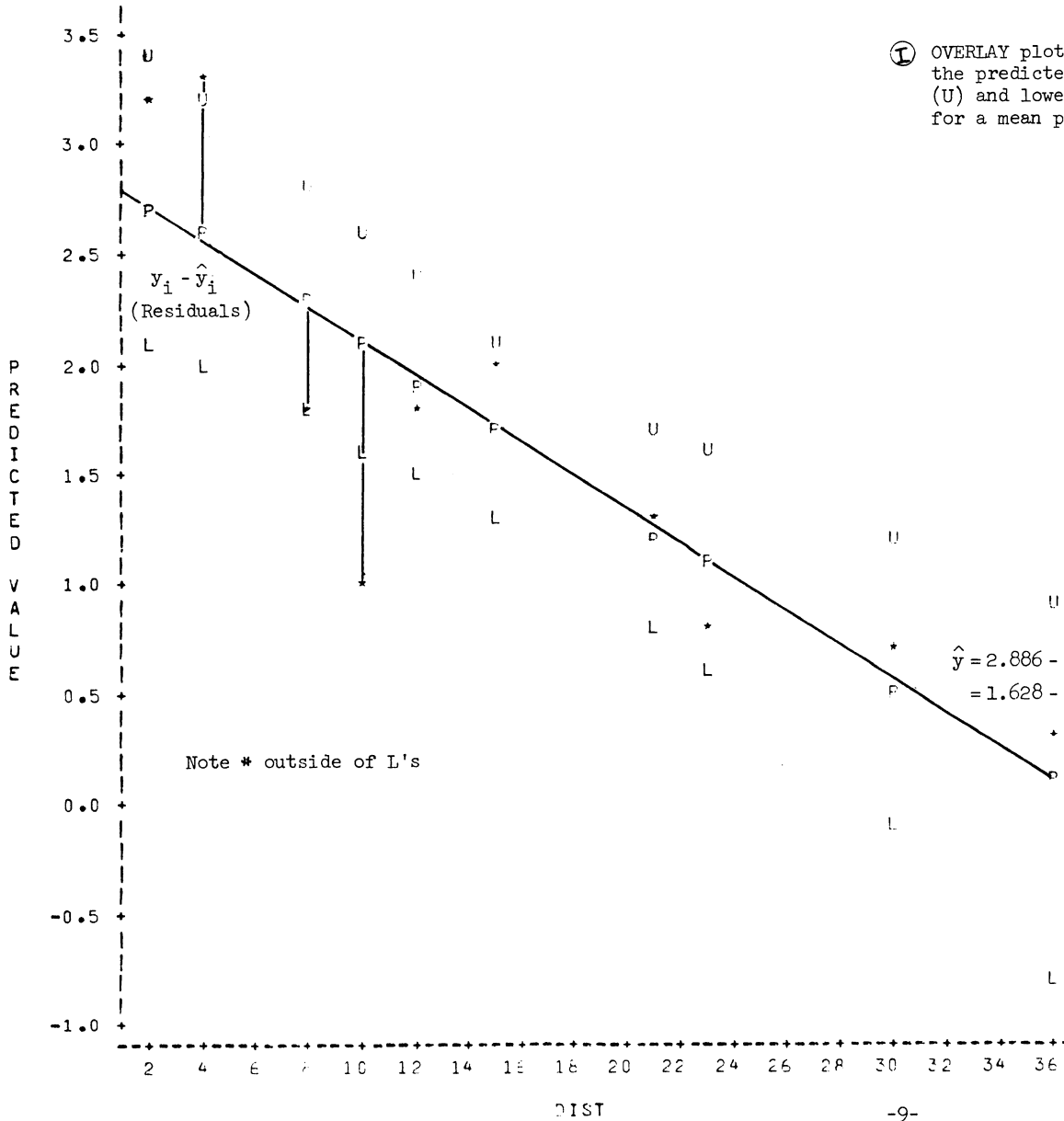
PERCENTS				PERCENTS			
VALUE	COUNT	CELL	CUM	VALUE	COUNT	CELL	CUM
-1.08472	1	10.0	10.0	0.118294	1	10.0	60.0
-0.44102	1	10.0	20.0	0.227198	1	10.0	70.0
-0.29876	1	10.0	30.0	0.336034	1	10.0	80.0
-0.098418	1	10.0	40.0	0.460075	1	10.0	90.0
0.0949383	1	10.0	50.0	0.686377	1	10.0	100.0

RESIDUAL ANALYSIS

PLOT OF RESID1*YHAT1 LEGEND: A = 1 OBS, B = 2 OBS, ETC.



PLOT OF ARSENIC+DIST SYMBOL USED IS *
 PLOT OF YHAT1+DIST SYMBOL USED IS P
 PLOT OF UPPER+DIST SYMBOL USED IS U
 PLOT OF LOWER+DIST SYMBOL USED IS L



(I) OVERLAY plot of the observed values (*), the predicted values (P), and the upper (U) and lower (L) 95% confidence limits for a mean predicted value.

MEAN MODEL
 (F) ARSENIC = X0/NOINT

DEP VARIABLE: ARSENIC

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE	PROB > F
MODEL	1	26.503840	26.503840	25.863	0.0007
ERROR	9	9.22860	1.024773		
U TOTAL	10	35.72640			

SS for the mean, equal to $n\bar{y}^2 = 10 * (1.628)^2$

Note that s^2 for this model is larger than for models (B), (C) and (D). Fitting the slope reduces the estimate of σ^2 .

ROOT MSE	1.012311	R-SQUARE	0.7418
DEP MEAN	1.628000	ADJ R-SQ	0.7418
C.V.	62.18126		

(NOTE: NO INTERCEPT TERM IS USED. R-SQUARE IS REDEFINED.)

VARIABLE	DF	PARAMETER ESTIMATE	STANDARD ERROR	T FOR HC: PARAMETER=0	PROB > T	TYPE I SS
X0	1	1.628000	0.320121	5.086	0.0007	0

This model was run so that the predicted values, each equal to $\bar{y}=1.628$, could be saved in the data set ONE, to be used in the data decomposition, part (J).

(J) DATA DECOMPOSITION FOR ARSENIC DATA

OBS	ARSENIC	YPAR	RES1	RESID1
1	3.19	1.628	1.1019	0.4601
2	3.26	1.628	0.7456	0.6864
3	1.82	1.628	0.6330	-0.4410
4	1.82	1.628	0.4767	-1.0847
5	1.85	1.628	0.3204	-0.0384
6	2.05	1.628	0.3860	0.3360
7	1.34	1.628	-0.3629	0.0949
8	0.79	1.628	-0.5392	-0.2988
9	0.66	1.628	-1.0863	0.1183
10	0.30	1.628	-1.5552	0.2278

$$y_1 = \bar{y} + (\hat{y}_1 - \bar{y}) + (y_1 - \hat{y}_1)$$

Each observation can be decomposed into the overall mean, the difference between the predicted values and the mean, and the difference between the observed and predicted values. See Unit 6, p. 50.

FIREFLY DATA - Units 10, 12, and 13; ACO, p. 351 (SAS)

```

TITLE FIREFLY DATA;
DATA FIREFLY;
INPUT FTIME LIGHT TEMP;
XC=1;
CARDS;
45 26 21.1
40 3 23.9
58 40 17.8
50 41 22.0
31 45 22.3
52 55 23.3
28 56 26.5
54 55 20.5
40 50 23.7
28 75 25.7
28 76 25.0
36 81 24.4
26 100 22.3
46 100 25.5
40 110 26.7
21 130 25.5
40 140 26.7
    
```

Creates the data set FIREFLY
 FTIME is the response variable,
 LIGHT and TEMP are two explanatory variables.

FIREFLY DATA				
OBS	X0	LIGHT	TEMP	FTIME
1	1	26	21.1	45
2	1	35	23.9	40
3	1	40	17.8	58
4	1	41	22.0	50
5	1	45	22.3	31
6	1	55	23.3	52
7	1	56	25.5	38
8	1	55	20.5	54
9	1	70	21.7	40
10	1	75	26.7	28
11	1	79	25.0	38
12	1	87	24.4	36
13	1	100	22.3	36
14	1	100	25.5	46
15	1	110	26.7	40
16	1	130	25.5	31
17	1	140	26.7	40

Ⓐ PROC PRINT;
 VAR XC LIGHT TEMP FTIME; Prints the X:Y matrix for Ⓒ

Ⓑ PROC PLOT;
 PLOT FTIME*LIGHT FTIME*TEMP LIGHT*TEMP; ← 3 separate plots, FTIME against each explanatory variable and LIGHT vs. TEMP.

Ⓐ X | Y matrix for the general model:
 FTIME = X0 LIGHT TEMP/NOINT in part Ⓒ.

```

PROC REG;
Ⓒ MODEL FTIME=XC LIGHT TEMP/NOINT P SS1 SS2 SECR ;
   OUTPUT OUT=NEW1 PREDICTED=YHAT1 RESIDUAL=RESID1;
Ⓓ MODEL FTIME= TEMP LIGHT/P SS1 SS2 SECR PARTIAL;
Ⓔ MODEL FTIME=XC TEMP/NOINT P SS1 SS2 SECR;
   OUTPUT OUT=NEW2 PREDICTED=YHAT2 RESIDUAL=RESID2;
    
```

} PROC REG used to fit models with both LIGHT and TEMP, and then a reduced model

PROC PLOT DATA=NEW1;
 PLOT RESID1*YHAT1/VDEF=0; } Residual plot for Ⓒ and Ⓓ

PROC PLOT DATA=NEW2;
 PLOT RESID2*TEMP/VDEF=0; } Residual plot for Ⓔ

FULL MODEL

Ⓒ FTIME = XO LIGHT TEMP/NOINT

SEQUENTIAL PARAMETER ESTIMATES ← The sequential coefficients are the diagonal elements of the SEQB output (see p. 99).

X⁰ 41.3529 = \bar{y} ← See p. 94.
 LIGHT 48.8962 -.173083 ← intercept and slope coefficients for FTIME = XO LIGHT (see p. 95).
 TEMP 91.4745 6.9E-04 -2.12753 ← intercept and two slopes which define a tilted plane (see p. 99).

DEP VARIABLE: FTIME Note that the slope in the LIGHT direction changes sign when TEMP is added to the model.

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE	PROB > F
MODEL	3	29527.857	9842.619	201.710	0.0001
ERROR	14	683.143	48.795924 = s ²		
U TOTAL	17	30211.000	← Includes SS for the mean		
ROOT MSE		6.985408 = s	R-SQUARE	0.9774	
DEP MEAN		41.352941 = \bar{y}	ADJ R-SQ	0.9742	
C.V.		16.89217			

NOTE: NO INTERCEPT TERM IS USED. R-SQUARE IS REDEFINED.

VARIABLE	DF	PARAMETER ESTIMATE	STANDARD ERROR	T FOR H ₀ : PARAMETER=0	PROB > T	TYPE I SS
Intercept						
X ⁰	1	91.474451	18.825103	4.859	0.0003	29071.118 = R(O)
LIGHT	1	0.0006919891	0.068339	0.010	0.9921	194.421 = R(L O)
TEMP	1	-2.127529	0.517509	-2.319	0.0361	262.318 = R(T L,O)

Partial coefficients

Slope of plane in LIGHT direction, see p. 100.
 Slope of plane in TEMP direction, see p. 100.

VARIABLE	DF	TYPE II SS
X ⁰	1	1152.148 = R(O T,L)
LIGHT	1	0.005003143 = R(L T,O)
TEMP	1	262.318 = R(T L,O)

when LIGHT is fitted last, it does not account for much of the remaining SS.

OBS	ACTUAL	PREDICT VALUE	RESIDUAL
1	45.000	46.672	-1.672
2	40.000	40.651	-0.650720
3	58.000	53.632	4.368
4	50.000	44.697	5.303
5	31.000	44.062	-13.062
6	52.000	41.941	10.059
7	38.000	37.261	0.738795
8	54.000	47.898	6.102
9	40.000	45.356	-5.356
10	28.000	34.721	-6.721
11	38.000	38.341	-0.340885
12	36.000	39.623	-3.623
13	36.000	44.100	-8.100
14	46.000	37.292	8.708
15	40.000	34.746	5.254

FULL MODEL

Ⓓ FTIME = TEMP LIGHT

OBS	ACTUAL	PREDICT VALUE	RESIDUAL
16	31.000	37.312	-6.312
17	40.000	34.766	5.234

SUM OF RESIDUALS 4.72511E-13 = 0
 SUM OF SQUARED RESIDUALS 683.1429

SEQUENTIAL PARAMETER ESTIMATES

INTERCEP 41.3529 = \bar{y}
 TEMP 91.3816 -2.12144 ← intercept and slope coefficients for FTIME = TEMP (no NOINT)
 LIGHT 91.4745 -2.12753 ~~6.9E-04~~ ← same as Ⓒ

Note that the sequential coefficients from ABDO are on the diagonal of the SEQB output. See p. 100.

DEP VARIABLE: FTIME

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE	PROB > F
MODEL	2	456.739	228.370	4.680	0.0182
ERROR	14	683.143	48.795924 = s^2 , same as Ⓒ		
C TOTAL	16	1139.882			

same ANOVA as Ⓒ, except that the MODEL and TOTAL SS are corrected for the mean.

ROOT MSE	DEP MEAN C.V.	R-SQUARE	ADJ R-SQ
6.985408	16.89217	0.4007	0.3151

VARIABLE	DF	PARAMETER ESTIMATE	STANDARD ERROR	T FOR H0: PARAMETER=0	PROB > T	TYPE I SS
INTERCEP	1	91.474451	18.825103	4.859	0.0003	29071.118 = R(0)
TEMP	1	-2.127529	0.917599	-2.319	0.0361	456.734 = R(T 0)
LIGHT	1	0.0006919891	0.068339	0.010	0.9921	0.005003143 = R(L T,0)

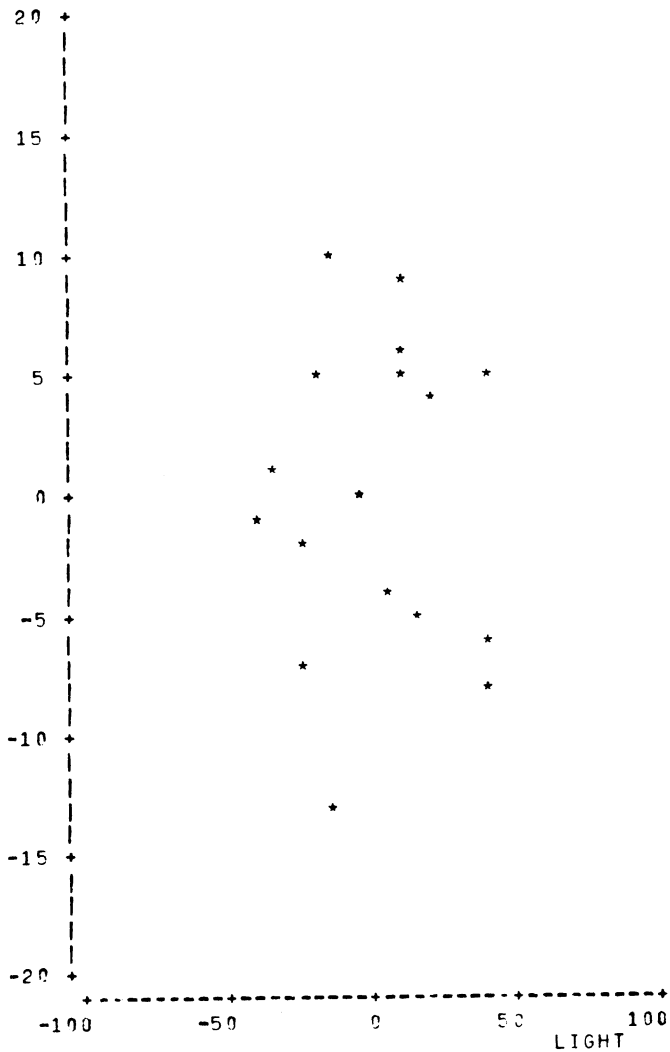
VARIABLE	DF	TYPE II SS	
INTERCEP	1	1152.148	= R(0 T,L)
TEMP	1	262.318	= R(T L,0)
LIGHT	1	0.005003143	= R(L T,0)

Again, LIGHT fitted after TEMP accounts for little of the remaining SS.

OBS	ACTUAL	PREDICT VALUE	RESIDUAL
1	45.000	46.602	-1.602
2	40.000	40.651	-.650720
3	58.000	53.632	4.368
4	50.000	44.697	5.303
5	31.000	44.062	-13.062
6	52.000	41.941	10.059
7	38.000	37.261	0.738795
8	54.000	47.898	6.102
9	40.000	45.356	-5.356
10	28.000	34.721	-6.721
11	38.000	38.341	-.340885
12	36.000	39.623	-3.623
13	36.000	44.100	-8.100
14	46.000	37.292	8.708
15	40.000	34.746	5.254
16	31.000	37.312	-6.312
17	40.000	34.766	5.234

SUM OF RESIDUALS 3.94351E-13
 SUM OF SQUARED RESIDUALS 683.1429
 Same as Ⓒ

PARTIAL REGRESSION RESIDUAL PLOTS
 \widehat{FTIME} — $FTIME$ (INTERCEPT, TEMP)

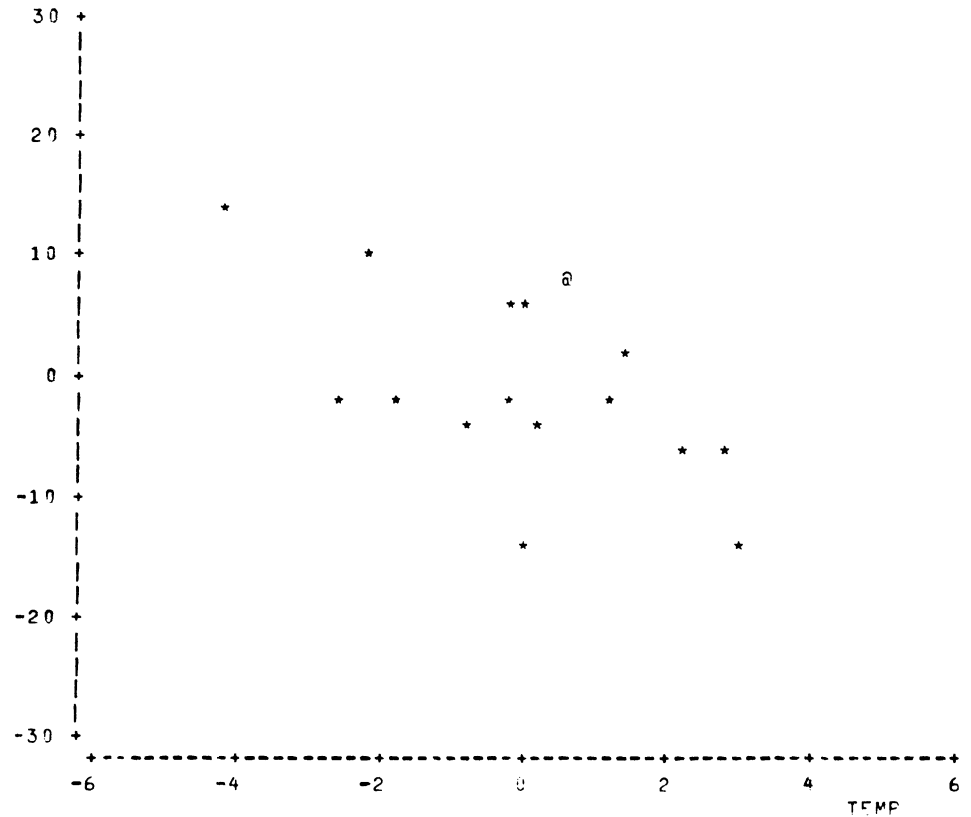


$LIGHT - \widehat{LIGHT}$ (INTERCEPT, TEMP)

There is little linear trend in this plot, so fitting LIGHT after an intercept and TEMP will account for a small portion of the remaining sum of squares, as in (C) and (D).

The option PARTIAL in the model statement from (D) produces these plots. These are similar to the plots on p. 141

PARTIAL REGRESSION RESIDUAL PLOTS
 \widehat{FTIME} — $FTIME$ (INTERCEPT, LIGHT)



$TEMP - \widehat{TEMP}$ (INTERCEPT, LIGHT)

REDUCED MODEL

(E) FTIME = X0 TEMP/NOINT

SEQUENTIAL PARAMETER ESTIMATES

X0 41.3529 = \bar{y}
 TEMP 91.3816 -2.12144 ← same estimates as second line, (D)

DEP VARIABLE: FTIME

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE	PROB > F
MODEL	2	29527.852	14763.926	324.174	0.0001
ERROR	15	683.148	45.543196 = s ²		
U TOTAL	17	30211.000			
ROOT MSE		6.748570	R-SQUARE	0.9774	
DEP MEAN		41.352941	ADJ R-SQ	0.9759	
C.V.		16.31944			

Note that s² has not increased very much from the s² in (C) and (D) where both TEMP and LIGHT were fitted.

NOTE: NO INTERCEPT TERM IS USED. R-SQUARE IS REDEFINED.

VARIABLE	DF	PARAMETER ESTIMATE	STANDARD ERROR	T FOR H0: PARAMETER=0	PROB > T	TYPE I SS
X0	1	91.381582	15.882436	5.754	0.0001	29071.118
TEMP	1	-2.121444	0.669992	-3.167	0.0064	456.734

SS for reduced model (FTIME = X0/NOINT) = SS for the mean ($n\bar{y}^2$) = 456.736

Compare with standard errors of the estimates, part (D)

VARIABLE	DF	TYPE II SS	TOLERANCE
X0	1	1507.671	1805.465876
TEMP	1	456.734	1.000000

F test for the need of the general model with TEMP and LIGHT over the reduced model with TEMP (pp. 138-140):

$$F'_{14} = \frac{[(\text{Model SS, general model}) - (\text{Model SS, reduced model})] / \text{difference in df between models}}{(\text{Residual SS, general model}) / (\text{Residual df, general model})}$$

$$= \frac{(456.739 - 456.736) / 1}{(683.143) / 14} = 0.0001 \quad \text{Reduced model is adequate .}$$

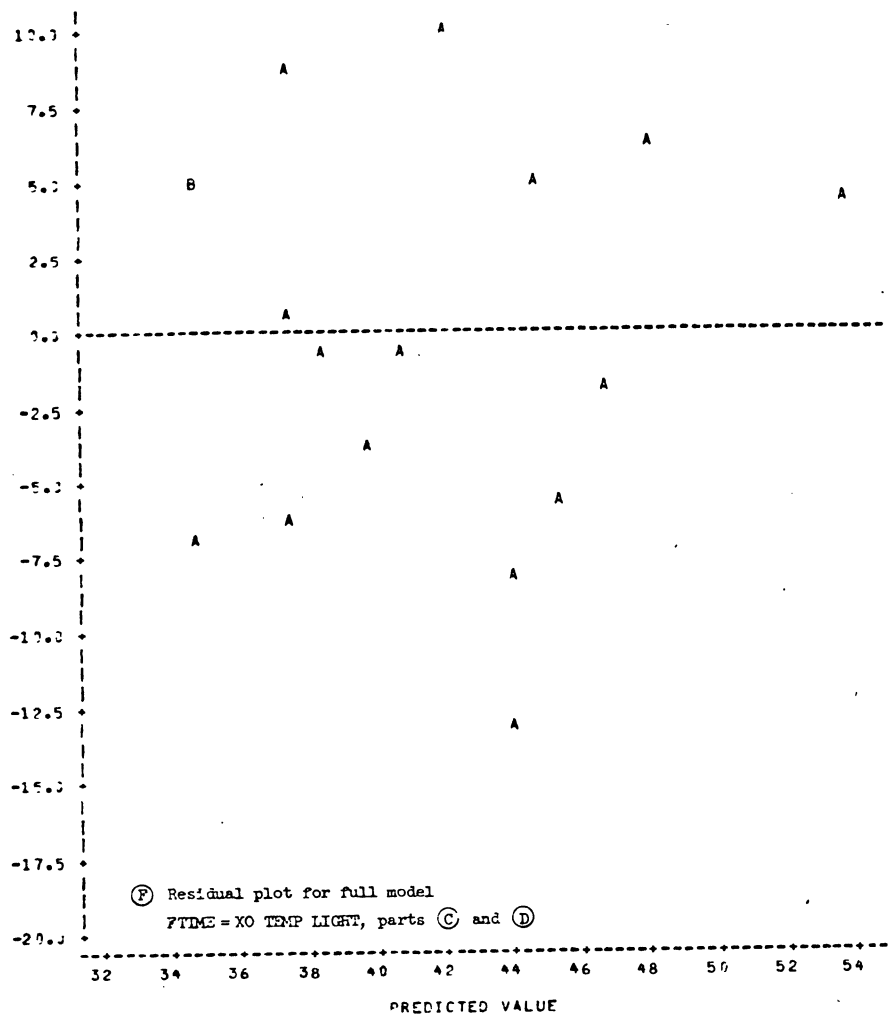
This test is equivalent to the t-test of the LIGHT parameter estimate in (C) or (D).

OBS	ACTUAL	PREDICT VALUE	RESIDUAL
1	45.000	46.619	-1.619
2	40.000	40.679	-6.79071
3	58.000	53.620	4.380
4	50.000	44.710	5.290
5	31.000	44.073	-13.073
6	52.000	41.952	10.048
7	38.000	37.285	0.715240
8	54.000	47.892	6.108
9	40.000	45.346	-5.346
10	28.000	34.739	-6.739
11	38.000	38.345	-.345482
12	36.000	39.618	-3.618
13	36.000	44.073	-8.073
14	46.000	37.285	8.715
15	40.000	34.739	5.261
16	31.000	37.285	-6.285
17	40.000	34.739	5.261

SUM OF RESIDUALS 3.65930E-13
 SUM OF SQUARED RESIDUALS 683.1479

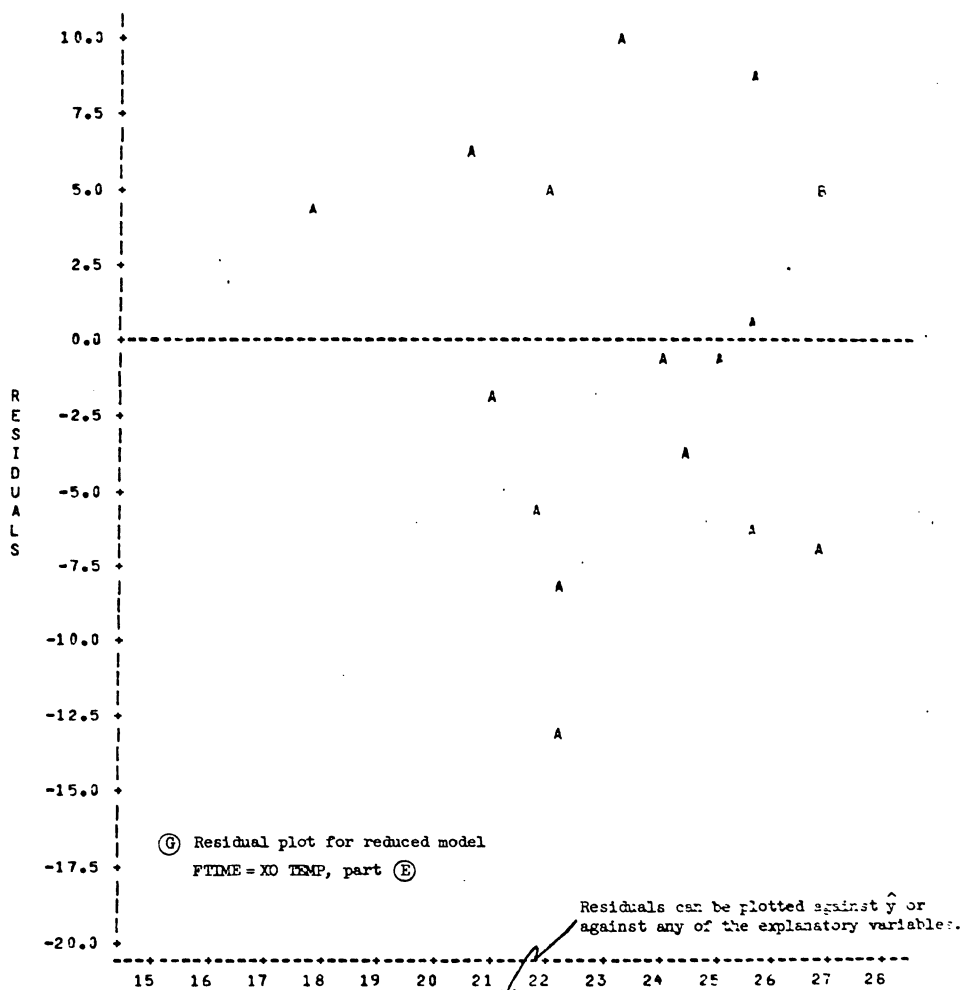
PLOT OF RESID1*YHAT1

LEGEND: A = 1 OBS, B = 2 OBS, ETC.



PLOT OF RESID2*TEMP

LEGEND: A = 1 OBS, B = 2 OBS, ETC.



```
TITLE'';
DATA SOYMILK;
INPUT TIME Y;
TIME=TIME/60.0;
TIME2=TIME*TIME;
TIME3=TIME2*TIME;
LOF=TIME;
CARDS;
0 2.74
0 2.25
0 2.34
12 3.14
12 2.68
12 2.83
30 3.44
30 3.53
30 3.63
60 3.68
60 3.75
60 3.51
```

The data set SOYMILK has variables TIME, $TIME2 = (TIME)^2$, $TIME3 = (TIME)^3$, $LOF = TIME$, and the response variable Y. The times are coded as TIME/60.0 to ensure numerical accuracy. Since there are four levels of the explanatory variable, the maximum degree polynomial that can be fitted is a cubic.

Ⓐ PROC REG;
MODEL Y=TIME TIME2 TIME3/SS1 SS2 SEQB; — Fits the cubic polynomial model

PROC SORT; BY TIME;

PROC MEANS MEAN NOPRINT;
BY TIME; VAR Y TIME;
OUTPUT OUT=NEW MEAN=YBAR X;

In order to fit the cubic polynomial to the mean response, the mean response at each level of TIME must be found. The data must be sorted BY TIME before taking MEANS BY TIME. The data set NEW contains four observations (each TIME level). The variable names are YBAR (the mean response) and X (the mean response at TIME levels).

Ⓑ PROC PRINT;
VAR X YBAR; } Prints the four observations in NEW.

DATA NEW1;
SET NEW;
X2=X*X; X3=X2*X; } Data set NEW1 will have four observations: the response variable is YBAR and the explanatory variables X, $X2 = X^2$, and $X3 = X^3$.

Ⓒ PROC REG DATA=NEW1;
MODEL YBAR=X X2 X3/SS1 SS2 SEQB; ← Fitting the cubic polynomial model to the mean response at each X level.

Ⓓ PROC GLM DATA=SOYMILK;
CLASS LOF;
MODEL Y=TIME TIME2 LOF; } Fits quadratic model. PROC GIM (General Linear Models) used to fit the full model. Unlike PROC REG, only one MODEL statement may follow each PROC GIM. If no SS type is specified GIM gives Type I (sequential) and Type IV (partials). LOF=lack of fit due to fitting a lower polynomial to the data. The CLASS statement constructs an indicator variable for each level of the LOF=TIME variable (see p. 166).

POLYNOMIAL MODEL ESTIMATION

Ⓐ Y = TIME TIME2 TIME3

SEQUENTIAL PARAMETER ESTIMATES

INTERCEP	3.12667					
TIME	2.62141	1.18884				
TIME2	2.41049	3.07182	-1.82743			
TIME3	2.44333	1.9775	1.58417	-2.35833	←	The partial coefficients for the cubic polynomial model.

The coefficients on the diagonal of the SEQB output are the sequential coefficients for the cubic polynomial model.

DEF VARIABLE: Y

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE	PROB > F
MODEL	3	2.885800	0.961933	26.116	0.0001
ERROR	8	0.294667	0.036833 = s ²		
C TOTAL	11	3.180467			
ROOT MSE		0.191920	R-SQUARE	0.9074	
DEP MEAN		3.126667	ADJ R-SQ	0.8726	
C.V.		6.13817			

s² is an estimate of "pure error" here since the maximum degree polynomial has been fitted. The error has 8 df, 2 from each level of TIME.

VARIABLE	DF	PARAMETER ESTIMATE	STANDARD ERROR	T FOR H0: PARAMETER=0	PROB > T	TYPE I SS
INTERCEP	1	2.443333	0.110805	22.051	0.0001	117.313
TIME	1	1.977500	1.485287	1.331	0.2197	2.406212
TIME2	1	1.584167	4.326547	0.366	0.7237	0.456351
TIME3	1	-2.358333	2.969196	-0.794	0.4500	0.023237

} See p. 113

VARIABLE	DF	TYPE II SS
INTERCEP	1	17.909633
TIME	1	0.065291
TIME2	1	0.004938101
TIME3	1	0.023237

The pure error can be used to test for the "lack of fit" of the quadratic and straight line models.

Lack of fit of the quadratic model:

$$\frac{SS \text{ full (cubic) model} - SS \text{ reduced (quadratic) model} / df}{\text{Pure Error SS} / \text{Pure Error df (= full model MSE)}} = \frac{(2.88580 - 2.862563) / 1}{.294667 / 8} = \frac{.023237}{.036833} = .6308$$

The quadratic model is adequate.

SOYMILK BOILING TIME DATA

DATA in NEW	OPS	X	YBAR
	1	0.0	2.44333
	2	0.2	2.88333
	3	0.5	3.53333
	4	1.0	3.64667

mean Y for each level of X. See p. 113.

POLYNOMIAL MODEL ESTIMATION USING MEANS

Ⓒ $Y = X \ X^2 \ X^3$

SEQUENTIAL PARAMETER ESTIMATES

INTERCEP 3.12667
 X 2.62141 1.18884
 X2 2.41049 3.07182 -1.82743
 X3 2.44333 1.9775 1.58417 -2.35833

Same parameter estimates as Ⓐ, so fitting the curve through the mean response is the same as using all the observations, except that with no replication, there are no df or SS for pure error.

DEP VARIABLE: YBAR

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE	PROB > F
MODEL	3	0.961933	0.320644	0.000	0.0000
ERROR	0	0	.		
C TOTAL	3	0.961933			
ROOT MSE		.		P-SQUARE	1.0000
DEP MEAN		3.126667		ADJ R-SQ	0.0000
C.V.		0			

VARIABLE	DF	PARAMETER ESTIMATE	STANDARD ERROR	T FOR H0: PARAMETER=0	PROB > T	TYPE I SS
INTERCEP	1	2.443333	0	.	1.0000	39.104178
X	1	1.977500	0	.	1.0000	0.802071
X2	1	1.584167	0	.	1.0000	0.152117
X3	1	-2.358333	0	.	1.0000	0.007745551

VARIABLE	DF	TYPE II SS
INTERCEP	1	0
X	1	0
X2	1	0
X3	1	0

QUADRATIC MODEL EVALUATION

Ⓓ Y = TIME TIME2 LOF

GENERAL LINEAR MODELS PROCEDURE

CLASS LEVEL INFORMATION

CLASS	LEVELS	VALUES
LOF	4	0 1 0.2 0.5

NUMBER OF OBSERVATIONS IN DATA SET = 12

GENERAL LINEAR MODELS PROCEDURE

DEPENDENT VARIABLE: Y

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE
MODEL	3	2.88580000	0.96193333	26.12
ERROR	8	0.29466667	0.03683333	PR > F
CORRECTED TOTAL	11	3.18046667		0.0002

R-SQUARE	C.V.	STD DEV	Y MEAN
0.707351	6.1382	0.19192012	3.12666667

(Sequentials)				
SOURCE	DF	TYPE I SS	F VALUE	PR > F
TIME	1	2.40621204	65.33	0.0001
TIME2	1	0.45635131	12.39	0.0078
LOF	1	0.02323665	0.63	0.4500

(Partials)				
SOURCE	DF	TYPE IV SS	F VALUE	PR > F
TIME	0	0.00000000	.	.
TIME2	0	0.00000000	.	.
LOF	1	0.02323665	0.63	0.4500

- Compare F=0.63 with the annotated F statistic for Model Ⓐ, Y = TIME TIME2 TIME3. In general, the lack of fit (LOF) partial sum of squares is the remaining sum of squares after the fitted model sum of squares. With t treatments (t levels of a quantitative variable for example), DF = t - p for LOF after fitting a p parameter model including one DF for the mean. In this output, DF = 4 - 3 = 1 but in general DF could be greater than 1.

ELECTRICITY LOAD DATA - Unit 13; ACO, p. 363 (EMDP)

```

OPTIONS LS=80 NODATE;
TITLE ELECTRICITY LOAD DATA;
DATA FLELOAD; /* used since DAY is a character variable
INPUT DATE DAY $ TEMP Y;
X3=0; X4=0; X5=0; X6=0; X1=1; X2=TEMP;
IF DAY='SU' OR DATE=5 THEN LINK SUNDAY;
IF DAY='SA' THEN LINK SATURDAY;
RETURN;
SUNDAY: X5=1; X6=TEMP; X1=0; X2=0; RETURN;
SATURDAY: X3=1; X4=TEMP; X1=0; X2=0; DAY='Q'; RETURN;
CARDS:

```

← The link command causes the appropriately labeled series of statements to be executed. For example, if DAY = 'SA' then the five statements between SATURDAY: and RETURN are executed.

← 'SA' is changed to 'Q' for labeling purposes in (G)

See (A) and (B) for the X variables created by the above statements.

(A) PROC PRINT;
VAR X1 X2 X3 X4 X5 X6; Prints the X matrix for the full model: 3 intercepts and 3 slopes.

(B) PROC PRINT;
VAR X1 X3 X5 TEMP; Prints the X matrix for the reduced model: 3 intercepts and 1 slope.

(C) PROC GLM;
MODEL Y=X1 X2 X3 X4 X5 X6/NOINT P;
OUTPUT OUT=NEW1 PREDICTED=YHAT1 RESIDUAL=RESID1;

(D) PROC PLOT DATA=NEW1;
PLOT RESID1*YHAT1/VR=FF=0; } Residual plot for the full model.

(E) PROC GLM;
MODEL Y=X1 X3 X5 TEMP/NOINT;
OUTPUT OUT=NEW2 PREDICTED=YHAT2 RESIDUAL=RESID2; } Fits the reduced model.

PROC PLOT DATA=NEW2;
(F) PLOT RESID2*YHAT2/VR=FF=0; The residual plot for the reduced model.
(G) PLOT Y*TEMP=DAY; Y vs. TEMP, and the character used to plot each observation (Y) is the data stored in DAY which corresponds to that observation (i.e., the first letter of the day in which the observation was taken).

(A) The X matrix for the full model in (C), p. 145

ELECTRICITY LOAD DATA

RES	Y1	Y2	X3	X4	X5	X6
1	1	77	0	0	0	0
2	1	75	0	0	0	0
3	0	0	1	74	0	0
4	0	0	0	0	1	78
5	0	0	0	0	1	77
6	1	72	0	0	0	0
7	1	79	0	0	0	0
8	1	81	0	0	0	0
9	1	85	0	0	0	0
10	0	0	1	83	0	0
11	0	0	0	0	1	82
12	1	75	0	0	0	0
13	1	77	0	0	0	0
14	1	75	0	0	0	0
15	1	77	0	0	0	0
16	1	77	0	0	0	0
17	0	0	1	82	0	0
18	0	0	0	0	1	74
19	1	72	0	0	0	0
20	1	71	0	0	0	0
21	1	73	0	0	0	0
22	1	78	0	0	0	0
23	1	78	0	0	0	0
24	0	0	1	74	0	0
25	0	0	0	0	1	77
26	1	79	0	0	0	0
27	1	73	0	0	0	0
28	1	75	0	0	0	0
29	1	69	0	0	0	0
30	1	71	0	0	0	0
31	0	0	1	68	0	0

X1 = 1 if weekday, 0 otherwise
 X3 = 1 if Saturday, 0 otherwise
 X5 = 1 if Sunday or holiday, 0 otherwise
 X2 = temp if weekday, 0 otherwise
 X4 = temp if Saturday, 0 otherwise
 X6 = temp if Sunday or holiday, 0 otherwise

(B) The X matrix for the reduced model in (E)

ELECTRICITY LOAD DATA

RES	X1	X3	X5	TEMP
1	1	0	0	77
2	1	0	0	75
3	0	1	0	74
4	0	0	1	78
5	0	0	1	77
6	1	0	0	72
7	1	0	0	79
8	1	0	0	81
9	1	0	0	85
10	0	1	0	83
11	0	0	1	82
12	1	0	0	75
13	1	0	0	77
14	1	0	0	75
15	1	0	0	77
16	1	0	0	77
17	0	1	0	82
18	0	0	1	74
19	1	0	0	72
20	1	0	0	71
21	1	0	0	73
22	1	0	0	78
23	1	0	0	78
24	0	1	0	74
25	0	0	1	77
26	1	0	0	79
27	1	0	0	73
28	1	0	0	75
29	1	0	0	69
30	1	0	0	71
31	0	1	0	68

Separate Intercepts

Common Slope

© FULL MODEL, 3 INTERCEPTS AND 3 SLOPES
 © Y = X1 X2 X3 X4 X5 X6/NOINT

DEPENDENT VARIABLE: Y

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE
MODEL	6	34740917.09721010	5790152.84953502	5432.09
ERROR	25	26647.90278990	1065.91611160	PR > F
UNCORRECTED TOTAL	31	34767565.00000000		0.0001

Like PROC REG, PROC GLM computes an incorrect R² when NOINT is used.

R-SQUARE	C.V.	STD DEV	Y MEAN
0.799234	3.1015	32.64837073	1052.67741935

SOURCE	DF	(sequentials)		
		TYPE I SS	F VALUE	PR > F
X1	1	26243932.19047613	24621.01	0.0001
X2	1	50779.42480620	47.64	0.0001
X3	1	4723920.00000000	4431.79	0.0001
X4	1	28880.00000000	27.09	0.0001
X5	1	3692841.80000000	3464.48	0.0001
X6	1	563.68192771	0.53	0.4739

SOURCE	DF	(partials)		
		TYPE IV SS	F VALUE	PR > F
X1	1	613.59455571	0.58	0.4551
X2	1	50779.42480620	47.64	0.0001
X3	1	103.72400562	0.10	0.7577
X4	1	28880.00000000	27.09	0.0001
X5	1	1603.84116067	1.50	0.2314
X6	1	563.68192771	0.53	0.4739

PARAMETER	PARTIAL ESTIMATE	T FOR H0: PARAMETER=0	PR > T	STD ERROR OF ESTIMATE
X1 weekday intercept=	110.83421927	0.76	0.4551	146.08130199
X2 weekday slope =	13.30930233	6.90	0.0001	1.92829251
X3 Satur. intercept=	-62.14285714	-0.31	0.7577	199.21029667
X4 Satur. slope =	13.57142857	5.21	0.0001	2.60728487
X5 Sunday intercept=	539.65060241	1.23	0.2314	439.54005229
X6 Sunday slope =	4.12048193	0.73	0.4739	5.66620746

PROC GLM has no option which gives estimates of the sequential b's.

(E) REDUCED MODEL, 3 INTERCEPTS AND A COMMON SLOPE
 (E) Y = X1 X3 X5 TEMP/NOINT

GENERAL LINEAR MODELS PROCEDURE

DEPENDENT VARIABLE: Y

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE
MODEL	4	34738249.28002397	8684562.32000599	7996.55
ERROR	27	29315.71997603	1085.76740652	PR > F
UNCORRECTED TOTAL	31	34767565.00000000		0.0001

R-SQUARE	C.V.	STD DEV	Y MEAN
0.999157	3.1302	32.95098491	1052.67741935

SOURCE	DF	Sequential TYPE I SS	F VALUE	PR > F
X1	1	R(1) = 26243932.19047619	24170.86	0.0001
X3 } intercepts	1	R(3 1) = 4723920.00000000	4350.77	0.0001
X5 } intercepts	1	R(5 3,1) = 3692841.80000000	3401.14	0.0001
TEMP ← slope	1	R(T 5,3,1) = 77555.28954779	71.43	0.0001

SOURCE	DF	Partial TYPE IV SS	F VALUE	PR > F
X1	1	R(1 3,5,T) = 1934.52471220	1.78	0.1931
X3	1	R(3 1,5,T) = 0.00006809	0.00	0.9998
X5	1	R(5 1,3,T) = 1325.61070453	1.22	0.2789
TEMP	1	R(T 1,3,5) = 77555.28954779	71.43	0.0001

PARAMETER	ESTIMATE	T FOR H0: PARAMETER=0	PR > T	STD ERROR OF ESTIMATE
X1 weekday intercept	152.73674326	1.33	0.1931	114.42601814
X3 Satur. intercept	0.02903497	0.00	0.9998	115.94506590
X5 Sunday intercept	-130.42865930	-1.10	0.2789	118.04118545
TEMP common slope	12.75552448	8.45	0.0001	1.50924939

Prediction equation given
on p. 146

To evaluate the adequacy of the reduced model, form the F statistic

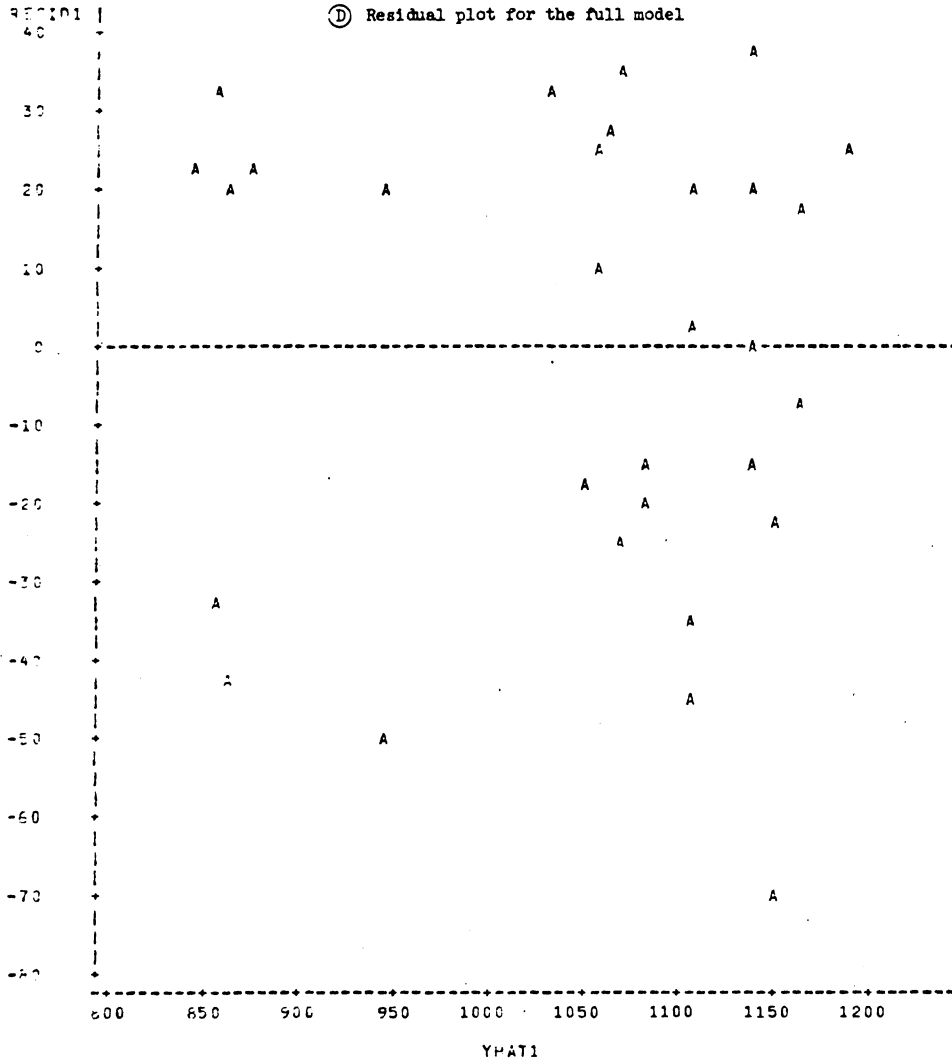
$$\frac{[(\text{Model SS, full model}) - (\text{Model SS, reduced model})] / \text{difference in df between the 2 models}}{(\text{Residual SS, full model}) / (\text{Residual df, full model})}$$

$$= \frac{[34,740,917 - 34,738,249] / 2}{1065.916} = 1.25 = F_{25}^2$$

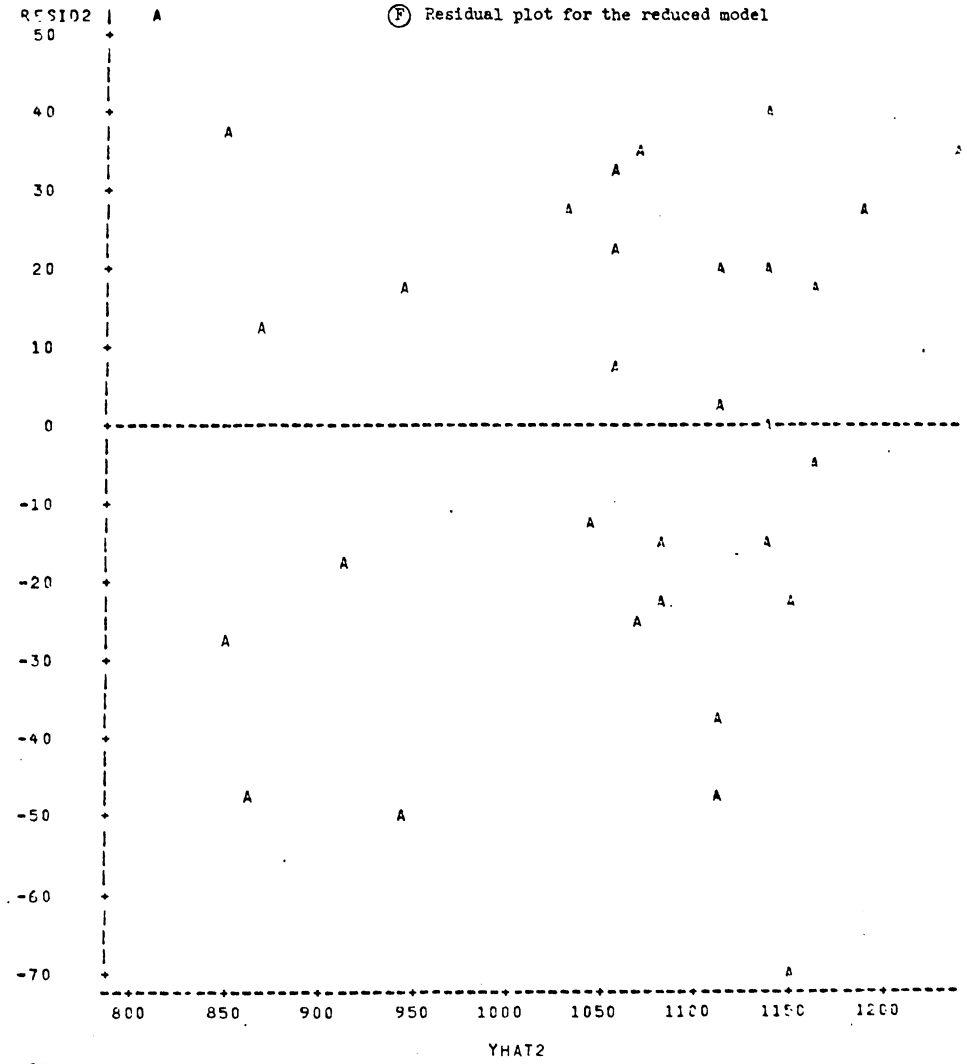
Based on the relatively small F value, the reduced model is judged to be sufficient and there is no need to fit separate slopes.

RESIDUAL ANALYSIS

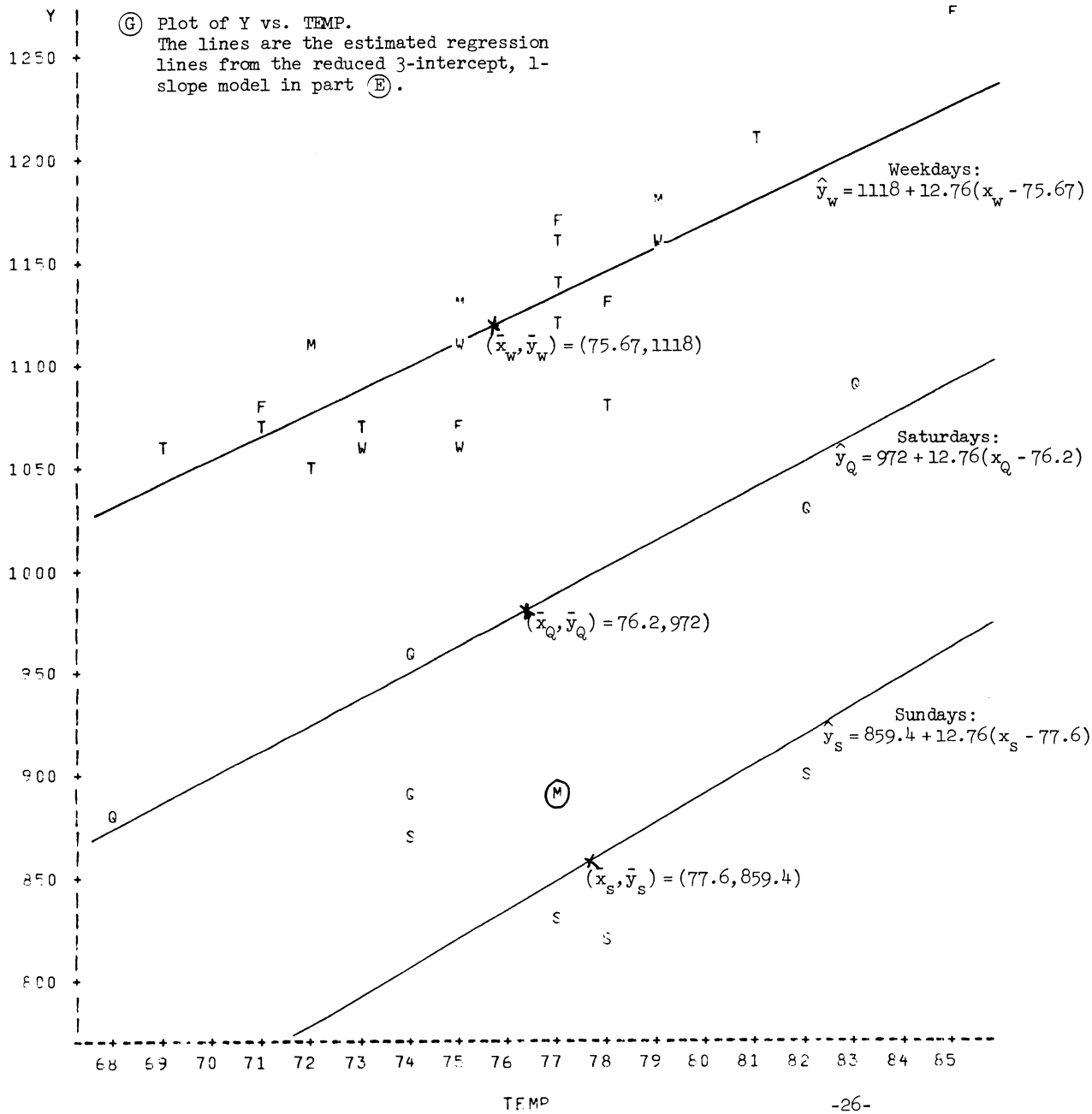
PLOT OF RESID1-YHAT1 LEGEND: A = 1 OBS, B = 2 OBS, ETC.



PLOT OF RESID2-YHAT2 LEGEND: A = 1 OBS, B = 2 OBS, ETC.



PLOT OF Y+TEMP SYMBOL IS VALUE OF DAY



```
TITLE LEAFHOPPER DATA;
DATA LHOPPER;
INPUT TRTS $ DAYS;
```

```
CARDS;
CONTROL 2.3
CONTROL 1.7
SUCROSE 3.6
SUCROSE 4.0
GLUCOSE 3.0
GLUCOSE 2.8
FRUCTOSE 2.1
FRUCTOSE 2.3
```

DAYS is the response variable

\$ indicates that TRTS is a non-numerical variable

PROC GLM; The CLASS statement constructs the treatment indicator variables. CLASS is not available in PROC REG. Treatment indicators would have to be set up in the input statement as in the Electricity Load Data or the Soybean Physiological Data.

```
(A) CLASS TRTS;
MODEL DAYS=TRTS/NOINT SOLUTION P XPX SSI; This model, following a CLASS statement, is the equivalent of a general means model. A SOLUTION option is needed after a CLASS statement so that the parameter estimates will be printed. XPX prints the X'X matrix (see pp. 179-180).
OUTPUT OUT=NEW PREDICTED=YHAT RESIDUAL=RESID;
ESTIMATE 'CONTROL VS SUGARS'
  TRTS 3 -1 -1 -1 /DIVISOR=3 E;
ESTIMATE '6-CARBONS VS SUCROSE'
  TRTS 0 -.5 -.5 1/E;
ESTIMATE 'FRUCTOSE VS GLUCOSE'
  TRTS 0 -1 1 0/E;
MEANS TRTS; ← Calculates TRT means.
```

ESTIMATE contrasts, p. 181

SAS orders levels of a classed variable alphabetically, or numerically, so the coefficients must be ordered: Control, Fructose, Glucose, Sucrose.

```
PROC PLOT;
(B) PLOT RESID*YHAT/VREF=0;
```

Residual plot for the general means model in (A)

```
PROC GLM;
CLASS TRTS;
MODEL DAYS=TRTS/P XPX SSI;
ESTIMATE 'CONTROL VS SUGARS'
  TRTS 3 -1 -1 -1 /DIVISOR=3 ;
ESTIMATE '6-CARBONS VS SUCROSE'
  TRTS 0 -.5 -.5 1;
ESTIMATE 'FRUCTOSE VS GLUCOSE'
  TRTS 0 -1 1 0;
```

LEAFHOPPER DATA

The CLASS statement produces this output.

GENERAL LINEAR MODELS PROCEDURE

CLASS LEVEL INFORMATION

CLASS	LEVELS	VALUES
TRTS	4	<u>C</u> ONTROL <u>F</u> RUCTOSE <u>G</u> LUCCOSE <u>S</u> UCROSE

Note alphabetical ordering

NUMBER OF OBSERVATIONS IN DATA SET = 8

With a CLASS statement, SAS creates indicator (dummy) variables and actually forms an X'X matrix as if indicator variables had been set up in the input statements.

LEAFHOPPER DATA

GENERAL LINEAR MODELS PROCEDURE
MATRIX ELEMENT REPRESENTATION

DEPENDENT VARIABLE: DAYS

EFFECT		REPRESENTATION
TRTS	CONTROL	DUMMY001
	FRUCTOSE	DUMMY002
	GLUCOSE	DUMMY003
	SUCROSE	DUMMY004

LEAFHOPPER DATA

GENERAL LINEAR MODELS PROCEDURE

THE X'X MATRIX

DEPENDENT VARIABLE: DAYS

	DUMMY001	DUMMY002	DUMMY003	DUMMY004
DUMMY001	2	0	0	0
DUMMY002	0	2	0	0
DUMMY003	0	0	2	0
DUMMY004	0	0	0	2

ESTIMABLE FUNCTIONS FOR CONTROL VS SUGARS

EFFECT		COEFFICIENTS	
TRTS	CONTROL	1	} Note the function of the DIVISOR=3 option.
	FRUCTOSE	0.333333	
	GLUCOSE	0.333333	
	SUCROSE	-0.333333	

The E option for each ESTIMATE statement prints the contrast vector (the c vector of cβ).

ESTIMABLE FUNCTIONS FOR 6 CARBONS VS SUCROSE

EFFECT		COEFFICIENTS
TRTS	CONTROL	0
	FRUCTOSE	-0.5
	GLUCOSE	-0.5
	SUCROSE	1

MEANS		
TRTS	N	DAYS
CONTROL	2	2.00000000
FRUCTOSE	2	2.20000000
GLUCOSE	2	2.90000000
SUCROSE	2	3.80000000

MEANS TRTS prints the treatment means. Compare to parameter estimates on next page.

ESTIMABLE FUNCTIONS FOR FRUCTOSE VS GLUCOSE

EFFECT		COEFFICIENTS
TRTS	CONTROL	0
	FRUCTOSE	-1
	GLUCOSE	1
	SUCROSE	0

GENERAL MEANS MODEL

Ⓐ Y = TRTS/NOINT

GENERAL LINEAR MODELS PROCEDURE

DEPENDENT VARIABLE: DAYS

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE
MODEL	4	63.38000000	15.84500000	211.27
ERROR	4	0.30000000	0.07500000	PR > F
UNCORRECTED TOTAL	8	63.68000000		0.0001

R-SQUARE	C.V.	STD DEV	DAYS MEAN
0.995289	10.0500	0.27386128	2.72500000

SOURCE	DF	TYPE I SS	F VALUE	PR > F
TRTS	4	63.38000000	211.27	0.0001 ← $H_0: \mu_c = \mu_f = \mu_g = \mu_s = 0$

PARAMETER	Sequentials from ABDO, p. 180	ESTIMATE	T FOR H0: PARAMETER=0	PR > T	STD ERROR OF ESTIMATE
TRTS CONTROL	$\left\{ \begin{array}{l} 2.00000000 = \bar{y}_c \\ 2.20000000 = \bar{y}_f \\ 2.90000000 = \bar{y}_g \\ 3.80000000 = \bar{y}_s \end{array} \right.$		10.33	0.0005	0.19364917
Note order FRUCTOSE			11.36	0.0003	0.19364917
GLUCOSE			14.98	0.0001	0.19364917
SUCROSE			19.62	0.0001	0.19364917
CONTROL VS SUGARS		-0.96666667	-4.32	0.0124	0.22360680
6-CARBONS VS SUCROSE		1.25000000	5.27	0.0062	0.23717082
FRUCTOSE VS GLUCOSE		0.70000000	2.56	0.0629	0.27386128

Contrasts

OBSERVATION	OBSERVED VALUE	PREDICTED VALUE	RESIDUAL
1	2.30000000	2.00000000	0.30000000
2	1.70000000	2.00000000	-0.30000000
3	3.60000000	3.80000000	-0.20000000
4	4.00000000	3.80000000	0.20000000
5	3.00000000	2.90000000	0.10000000
6	2.80000000	2.90000000	-0.10000000
7	2.10000000	2.20000000	-0.10000000
8	2.30000000	2.20000000	0.10000000

Control vs. Sugar contrast (see p. 181):

Estimate = $-0.9667 = 2.0 - .3333(2.2) - .3333(2.9) - .3333(3.8)$

Standard Error = $.2236 = \text{SQRT} \left[\frac{\sigma^2}{2} (1 + 1/9 + 1/9 + 1/9) \right]$

GENERAL MEANS MODEL

Y = TRTS

LEAFHOPPER DATA

GENERAL LINEAR MODELS PROCEDURE

DEPENDENT VARIABLE: DAYS

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE
MODEL	3	3.97500000	1.32500000	17.67
ERROR	4	0.30000000	0.07500000	PR > F
CORRECTED TOTAL	7	4.27500000		0.0000 ← $H_0: \mu_C = \mu_F = \mu_G = \mu_S = \mu$

R-SQUARE	C.V.	STD DEV	DAYS MEAN
0.929825	17.0500	0.27386128	2.72500000

SOURCE	DF	TYPE I SS	F VALUE	PR > F
TRTS	3	3.97500000	17.67	0.0000

PARAMETER	ESTIMATE	T FOR H0: PARAMETER=0	PR > T	STD ERROR OF ESTIMATE
CONTROL VS SUGARS	-0.96666667	-4.32	0.0124	0.22360680
6-CARBONS VS SUCROSE	1.25000000	5.27	0.0062	0.23717082
FRUCTOSE VS GLUCOSE	0.70000000	2.56	0.0629	0.27386128

OBSERVATION	OBSERVED VALUE	PREDICTED VALUE	RESIDUAL
1	2.30000000	2.00000000	0.30000000
2	1.70000000	2.00000000	-0.30000000
3	3.60000000	3.80000000	-0.20000000
4	4.00000000	3.80000000	0.20000000
5	3.00000000	2.90000000	0.10000000
6	2.80000000	2.90000000	-0.10000000
7	2.10000000	2.20000000	-0.10000000
8	2.30000000	2.20000000	0.10000000

SUM OF RESIDUALS	0.00000000
SUM OF SQUARED RESIDUALS	0.30000000
SUM OF SQUARED RESIDUALS + ERROR SS	0.30000000
FIRST ORDER AUTOCORRELATION	-0.20000000
DURBIN-WATSON D	2.76666667

LEAFHOPPER DATA		
GENERAL LINEAR MODELS PROCEDURE		
CLASS LEVEL INFORMATION		
CLASS	LEVELS	VALUES
TRTS	4	CONTROL FRUCTOSE GLUCOSE SUCROSE
NUMBER OF OBSERVATIONS IN DATA SET = 8		

LYMPHOCYTE DATA - Unit 17

```
TITLE LYMPHOCYTE DATA;
DATA LYMPHOCYT;
INPUT ATP IG Y;
CARDS
0 0 44
0 0 30
0 1 62.5
0 1 67.5
1 0 25.5
1 0 24.5
1 1 46.0
1 1 50.0
```

Presence of ATP or IG stimulation is denoted by a 1, otherwise a zero appears.

	no stimulation	anti-IG stimulation
no ATP	treatment (0,0)	treatment (0,1)
ATP added	treatment (1,0)	treatment (1,1)

```
PROC GLM;
  CLASSES ATP IG; ← Class on both ATP and IG
```

```
  MODEL Y= ATP IG ATP*IG/P;
  OUTPUT OUT= NEW RESIDUAL= RESID PREDICTED= YHAT; } This model fits the main effects of ATP and IG,
  and their interaction, ATP*IG.
```

```
PROC GLM;
```

```
  CLASSES ATP IG;
  MODEL Y= ATP*IG/NOINT SOLUTION SS1;
  ESTIMATE 'STIMULUS'
    ATP*IG 1 -1 1 -1/DIVISOR=2;
  ESTIMATE 'ATP PRESENCE'
    ATP*IG 1 1 -1 -1/DIVISOR=2;
  ESTIMATE 'INTERACTION'
    ATP*IG 1 -1 -1 1/DIVISOR=2;
  Use the general means model to estimate contrasts.
  Contrasts among treatment means in part (A) would be
  "non-estimable" because the main effects are fitted first.
  The CLASSES statement orders the treatments numerically:
  (0,0); (0,1); (1,0); (1,1). The contrast coefficients
  must be in this order.
```

```
PROC PLOT;
```

```
  PLOT RESID*YHAT/VREF=0; Residual plot for (A) and (B)
  ← Refers to variables in CLASSES Statement.
```

FACTORIAL ANALYSIS

(A) Y = ATP IG ATP*IG

DEPENDENT VARIABLE: Y

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE
MODEL	3	1906.00000000	635.33333333	87.63
ERROR	4	29.00000000	7.25000000	PR > F
CORRECTED TOTAL	7	1935.00000000		0.0004

R-SQUARE	C.V.	STD DEV	Y MEAN
0.985013	6.3355	2.69258240	42.50000000

SOURCE	DF	TYPE I SS	F VALUE	PR > F
ATP	1	288.00000000	39.72	0.0032
IG	1	1568.00000000	216.28	0.0001
ATP*IG	1	50.00000000	6.90	0.0584

SOURCE	DF	TYPE IV SS	F VALUE	PR > F
ATP	1	288.00000000	39.72	0.0032
IG	1	1568.00000000	216.28	0.0001
ATP*IG	1	50.00000000	6.90	0.0584

Note that TYPE I and TYPE IV SS's are the same due to orthogonality. The SSI option could have been used.

OBSERVATION	OBSERVED VALUE	PREDICTED VALUE	RESIDUAL
1	34.00000000	32.00000000	2.00000000
2	30.00000000	32.00000000	-2.00000000
3	62.50000000	65.00000000	-2.50000000
4	67.50000000	65.00000000	2.50000000
5	25.50000000	25.00000000	0.50000000
6	24.50000000	25.00000000	-0.50000000
7	46.00000000	48.00000000	-2.00000000
8	50.00000000	48.00000000	2.00000000

SUM OF RESIDUALS	0.00000000
SUM OF SQUARED RESIDUALS	29.00000000
SUM OF SQUARED RESIDUALS - ERROR SS	0.00000000
FIRST ORDER AUTOCORRELATION	0.25000000
DURBIN-WATSON D	2.22413793

GENERAL MEANS MODEL

(B) $Y = ATP * IG / NOINT$

GENERAL LINEAR MODELS PROCEDURE

DEPENDENT VARIABLE: Y

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE
MODEL	4	16356.00000000	4089.00000000	564.00
ERROR	4	29.00000000	7.25000000 = s^2	PR > F
UNCORRECTED TOTAL	8	16385.00000000		0.0001

s^2 same as for (A)

R-SQUARE	C.V.	STD DEV	Y MEAN
0.958230	6.3355	2.69258240	42.50000000

SOURCE	DF	TYPE I SS	F VALUE	PR > F
ATP*IG	4	16356.00000000	564.00	0.0001

PARAMETER	ESTIMATE	T FOR H0: PARAMETER=0	PR > T	STD ERROR OF ESTIMATE
ATP*IG 0 0	32.00000000 = \bar{y}_{00}	16.81	0.0001	1.90394328
0 1	65.00000000 = \bar{y}_{01}	34.14	0.0001	1.90394328
1 0	25.00000000 = \bar{y}_{10}	13.13	0.0002	1.90394328
1 1	48.00000000 = \bar{y}_{11}	25.21	0.0001	1.90394328
STIMULUS	-28.00000000	-14.71	0.0001	1.90394328
ATP PRESENCE	12.00000000	6.30	0.0032	1.90394328
INTERACTION	-5.00000000	-2.63	0.0584	1.90394328

Compare significance levels to those of the TYPE I SS in part (A)

In (A) the contrasts are computed as part of the model. In (B) the contrasts are specified. More than three contrasts could have been run under the model. The contrasts do not need to be orthogonal for computation purposes.

FAT DIGESTIBILITY DATA - Unit 17; ACO, p. 365 (SAS)

```
TITLE FAT DIGESTIBILITY DATA;
DATA FAT_DIG;
INPUT PERIOD FAT $ LECITHIN Y;
X0=1; X1=0; X2=0; X4=0; X5=0; X6=0;
IF PERIOD=1 THEN X1=1;
IF PERIOD=2 THEN X2=1;
X3=1-X1-X2;
IF FAT='T' AND LECITHIN=0 THEN X4=1;
IF FAT='C' AND LECITHIN=0 THEN X5=1;
IF FAT='T' AND LECITHIN=1 THEN X6=1;
X7=1-X4-X5-X6;
CARDS;
```

} Creating the indicator variables shown in (A)

PROC PRINT;

(A) VAR X0-X7 Y; Prints the $X|Y$ matrix for (B) (see p. 201).

PROC GLM;

```
(B) MODEL Y=X0 X1-X7/NOINT;
      OUTPUT OUT=NEW RESIDUAL=RESID PREDICTED=YHAT;
      ESTIMATE *W VS W0 LECITHIN*
        X4 .5 X5 .5 X6 -.5 X7 -.5;
      ESTIMATE *FAT DIFF W0 LECITHIN*
        X4 1 X5 -1;
      ESTIMATE *FAT DIFF W LECITHIN*
        X6 1 X7 -1;
```

} Model: Equal means|Period Indicators|Treatment Indicators
and contrasts as discussed in Unit 17

PROC PLOT;

(C) PLOT RESID*YHAT/VREF=0; Residual plot for (B)

OBS	Equal means	Period Indicators			Treatment Indicators				Y	(A) $X Y$ matrix for model in (B)
	X0	X1	X2	X3	X4	X5	X6	X7		
1	1	1	0	0	1	0	0	0	64.6	
2	1	0	1	0	1	0	0	0	52.4	
3	1	0	0	1	1	0	0	0	53.8	
4	1	1	0	0	0	1	0	0	66.0	
5	1	0	1	0	0	1	0	0	60.1	
6	1	0	0	1	0	1	0	0	64.4	
7	1	1	0	0	0	0	1	0	85.0	
8	1	0	1	0	0	0	1	0	68.9	
9	1	0	0	1	0	0	1	0	77.5	
10	1	1	0	0	0	0	0	1	96.0	
11	1	0	1	0	0	0	0	1	90.4	
12	1	0	0	1	0	0	0	1	98.2	

GENERAL LINEAR MODELS PROCEDURE

MODEL SEQUENCE

Ⓑ Y = Equal means | Periods | Treatments

DEPENDENT VARIABLE: Y

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE
MODEL	6	66867.48166667	11144.58027778	940.36
ERROR	6	71.10833333	11.85138889	PR > F
UNCORRECTED TOTAL	12	66938.59000000		0.0001

R-SQUARE	C.V.	STD DEV	Y MEAN
0.998938	4.7089	3.44258462	73.10833333

SOURCE	DF	TYPE I SS	F VALUE	PR > F
X0	1	$n\bar{y}^2 = 64137.94083333$	5411.85	0.0001
X1	1	Period { 137.76041667 }	11.62	0.0143
X2	1	SS { 61.05125000 }	5.15	0.0637
X3	0	0.00000000	.	.
X4	1	Treat- { 1046.52250000 }	88.30	0.0001
X5	1	ment { 1012.50000000 }	85.43	0.0001
X6	1	SS { 471.70666667 }	39.80	0.0007
X7	0	0.00000000	.	.

= 198.81 = R(1,2,3|0) Redundant

= 2530.73

= R(4,5,6|0,1,2,3)

X3 is redundant after fitting X0 X1 X2, therefore it does not account for any variability in the model.

SOURCE	DF	TYPE IV SS	F VALUE	PR > F
X0	0	0.00000000	.	.
X1	0	0.00000000	.	.
X2	0	0.00000000	.	.
X3	0	0.00000000	.	.
X4	0	0.00000000	.	.
X5	0	0.00000000	.	.
X6	0	0.00000000	.	.
X7	0	0.00000000	.	.

No matter which variable is fitted last, it will have 0 df and no sum of squares since it will be a linear combination of variables previously fitted. This redundancy causes the following message to appear.
(See also note on following page.)

PARAMETER	ESTIMATE	T FOR H0: PARAMETER=0	PR > T	STD ERROR OF ESTIMATE
X0	$(\bar{y}_4 - \bar{y}) + \text{period 3 mean} = 95.23333333$	39.12	0.0001	2.43427493
X1	period 1 mean - period 3 mean = 4.42500000	1.82	0.1190	2.43427493
X2	period 2 mean - period 3 mean = -5.52500000	-2.27	0.0637	2.43427493
X3	0.00000000	.	.	.
X4	$\bar{y}_1 - \bar{y}_4 = -37.93333333$	-13.50	0.0001	2.81085857
X5	$\bar{y}_2 - \bar{y}_4 = -31.36666667$	-11.16	0.0001	2.81085857
X6	$\bar{y}_3 - \bar{y}_4 = -17.73333333$	-6.31	0.0007	2.81085857
X7	0.00000000	.	.	.

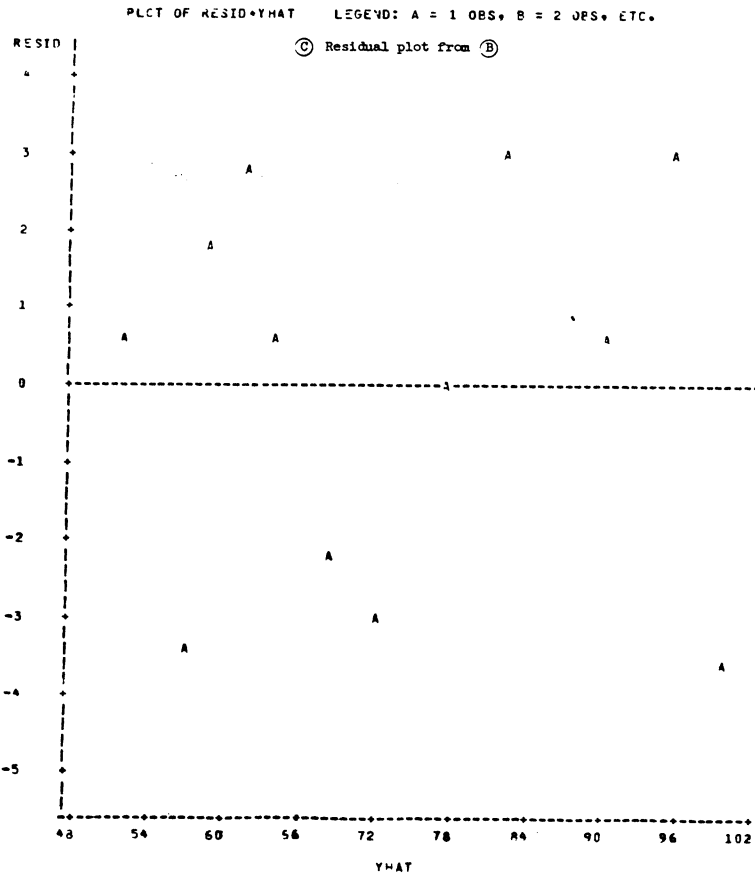
In general, the partial coefficients are not of interest but contrasts are appropriate (see next page).

GENERAL LINEAR MODELS PROCEDURE

DEPENDENT VARIABLE: Y

NOTE: THE X*X MATRIX HAS BEEN DEEMED SINGULAR AND A GENERALIZED INVERSE HAS BEEN EMPLOYED TO SOLVE THE NORMAL EQUATIONS. THE ABOVE ESTIMATES REPRESENT ONLY ONE OF MANY POSSIBLE SOLUTIONS TO THE NORMAL EQUATIONS. ESTIMATES FOLLOWED BY THE LETTER B ARE BIASED AND DO NOT ESTIMATE THE PARAMETER BUT ARE BLUE FOR SOME LINEAR COMBINATION OF PARAMETERS (OR ARE ZERO). THE EXPECTED VALUE OF THE BIASED ESTIMATORS MAY BE OBTAINED FROM THE GENERAL FORM OF ESTIMABLE FUNCTIONS. FOR THE BIASED ESTIMATORS, THE STD ERR IS THAT OF THE BIASED ESTIMATOR AND THE T VALUE TESTS $H_0: E(\text{BIASED ESTIMATOR}) = 0$. ESTIMATES NOT FOLLOWED BY THE LETTER B ARE BLUE FOR THE PARAMETER.

PARAMETER	ESTIMATE	T FOR H_0 : PARAMETER=0	PR > T	STD ERROR OF ESTIMATE
Contrasts:				
W VS WO LECITHIN	-25.78333333	-12.97	0.0001	1.98757716
FAT DIFF WO LECITHIN	-6.56666667	-2.34	0.0581	2.81085857
FAT DIFF W LECITHIN	-17.73333333	-6.31	0.0007	2.81085857



TITLE NUTRITION DATA;
DATA PROTEIN;
INPUT PROTEIN & Y;

CARDS;

H 170	}	10 horsebean observations
H 136		
H 160		
H 227		
H 217		
H 168		
H 108		
H 124		
H 143		
H 140		
L 309	}	12 linseed observations
L 229		
L 181		
L 141		
L 260		
L 203		
L 148		
L 169		
L 213		
L 257		
L 244		
L 271		
S 245	}	14 soybean observations
S 230		
S 248		
S 327		
S 329		
S 250		
S 193		
S 271		
S 316		
S 207		
S 196		
S 177		
S 158		
S 248		

(A) PROC GLM;
 CLASS PROTEIN;
 MODEL Y=PROTEIN/MOINT SOLUTION P SS1;
 OUTPUT OUT=NEW PREDICTED=YHAT RESIDUAL=RESID;
 ESTIMATE 'HORSEBEAN VS CILNEAL' See p. 217
 PROTEIN 1 -.5 -.5; Natural Contrasts
 ESTIMATE 'LINSEED VS SOYBEAN' See p. 219
 PROTEIN 0 1 -1; Ortho Contrasts
 ESTIMATE 'ORTHO H-B VS CILNEAL' See p. 219
 PROTEIN 13 -6 -7/DIVISOR=13;

Using the CLASS statement to fit the general means model, followed by natural and orthogonal contrasts, as discussed in Unit 18.

(B) PROC PLOT;
 PLOT RESID*YHAT/VREF=0; Residual plot for (A)

GENERAL MEANS MODEL

(A) Y = PROTEIN/NOINT

DEPENDENT VARIABLE: Y

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE
MODEL	3	1683997.43571429	561332.47857143	229.66
ERROR	33	80659.56428571	2444.22922078	PR > F
UNCORRECTED TOTAL	36	1764657.00000000		0.0001

R-SQUARE	C.V.	STD DEV	Y MEAN
0.954292	23.1656	49.43914664	213.41666667

SOURCE	DF	TYPE I SS	F VALUE	PR > F	Tests: $\mu_H = \mu_L = \mu_S = 0$ <u>not</u> $\mu_H = \mu_L = \mu_S = \mu$
PROTEIN	3	1683997.43571429	229.66	0.0001	

PARAMETER	ESTIMATE	T FOR H0: PARAMETER=0	PR > T	STD ERROR OF ESTIMATE
PROTEIN H	160.20000000 = \bar{y}_H	10.25	0.0001	15.63403090
L	218.75000000 = \bar{y}_L	15.33	0.0001	14.27185231
S	246.85714286 = \bar{y}_S	18.68	0.0001	13.21316773

Contrasts:

HORSEBEAN VS OILMEAL	-72.60357143	-3.94	0.0004	18.41171677
LINSEED VS SOYBEAN	-28.10714286	-1.45	0.1578	19.44925628
ORTHO H-B VS OILMEAL	-73.68461538	-4.01	0.0003	18.39651430

In this case, the natural and ORTHO contrasts give similar results.

OBSERVATION	OBSERVED VALUE	PREDICTED VALUE	RESIDUAL	OBSERVED VALUE	PREDICTED VALUE	RESIDUAL
1	179.00000000	160.20000000	18.80000000	244.00000000	218.75000000	25.25000000
2	136.00000000	160.20000000	-24.20000000	271.00000000	218.75000000	52.25000000
3	160.00000000	160.20000000	-0.20000000	243.00000000	246.85714286	-3.85714286
4	227.00000000	160.20000000	66.80000000	230.00000000	246.85714286	-16.85714286
5	217.00000000	160.20000000	56.80000000	248.00000000	246.85714286	1.14285714
6	168.00000000	160.20000000	7.80000000	327.00000000	246.85714286	80.14285714
7	108.00000000	160.20000000	-52.20000000	329.00000000	246.85714286	82.14285714
8	124.00000000	160.20000000	-36.20000000	250.00000000	246.85714286	3.14285714
9	143.00000000	160.20000000	-17.20000000	193.00000000	246.85714286	-53.85714286
10	140.00000000	160.20000000	-20.20000000	271.00000000	246.85714286	24.14285714
11	309.00000000	218.75000000	90.25000000	316.00000000	246.85714286	69.14285714
12	229.00000000	218.75000000	10.25000000	267.00000000	246.85714286	20.14285714
13	181.00000000	218.75000000	-37.75000000	199.00000000	246.85714286	-47.85714286
14	141.00000000	218.75000000	-77.75000000	177.00000000	246.85714286	-69.85714286
15	260.00000000	218.75000000	41.25000000	158.00000000	246.85714286	-88.85714286
16	203.00000000	218.75000000	-15.75000000	248.00000000	246.85714286	1.14285714
17	148.00000000	218.75000000	-70.75000000			
18	169.00000000	218.75000000	-49.75000000			
19	213.00000000	218.75000000	-5.75000000			
20	257.00000000	218.75000000	38.25000000			

SUM OF RESIDUALS	0.00000000
SUM OF SQUARED RESIDUALS	80659.56428571
SUM OF SQUARED RESIDUALS - ERROR SS	-0.00000000
FIRST ORDER AUTOCORRELATION	0.34539716
DURBIN-WATSON D	1.30480762

(B) General Means Estimation

PROC GLM; CLASSES LOC TYPE;
MODEL Y=LOC*TYPE/NOINT P CLM;

ESTIMATE *TEXT PG229 CONTRAST1*
LOC*TYPE 11 -6 -8 5 2 1 /DIVISOR=11;
ESTIMATE *TEXT PG229 CONTRAST2*
LOC*TYPE 0 1 -1 0 0 0;
ESTIMATE *TEXT PG229 CONTRAST3*
LOC*TYPE 0 0 0 14 -8 -6 /DIVISOR=14;
ESTIMATE *TEXT PG229 CONTRAST4*
LOC*TYPE 0 0 0 0 1 -1;

Interaction is judged to be important

- Objective: (1) to estimate the cell means from the general means model;
(2) to estimate column contrasts within each row (or row contrasts within each column).

DEPENDENT VARIABLE: Y

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE	PR > F	R-SQUARE	C
MODEL	6	1556.20366667	259.36727778	1334.49	0.0001	0.996391	6.6
ERROR	29	5.63633333	0.19435632		STD DEV		Y M
UNCORRECTED TOTAL	35	1561.84000000			0.44065862		6.66285

SOURCE	DF	TYPE I SS	F VALUE	PR > F	DF	TYPE IV SS	F VALUE	PR
LOC*TYPE	6	1556.20366667	1334.49	0.0001	6	1556.20366667	1334.49	0.0

PARAMETER	ESTIMATE	T FOR H0: PARAMETER=0	PR > T	STD ERROR OF ESTIMATE
TEXT PG229 CONTRAST1	0.51598909	2.52	0.0176	0.20484945
TEXT PG229 CONTRAST2	0.34656667	1.30	0.2043	0.26695315
TEXT PG229 CONTRAST3	0.08571429	0.26	0.7988	0.33325779
TEXT PG229 CONTRAST4	-0.15000000	-3.63	0.5336	0.23809087

- Interpretation: (1) Within the near location, community type North has a higher (0.5) pH than the average of the other two community types (p=.02).
(2) There is some evidence that the pH for the Mesic Community is higher than Shrub Community (p=.20).
(3) Within the away location, the pH for community types does not vary.

OBSERVATION	OBSERVED VALUE	PREDICTED VALUE	RESIDUAL	LOWER 95% CL FOR MEAN	UPPER 95% CL FOR MEAN
1	6.60000000	6.82500000	-0.22500000	6.50621858	7.14378142
2	7.20000000	6.82500000	0.37500000	6.50621858	7.14378142
3	7.20000000	6.82500000	0.37500000	6.50621858	7.14378142
4	7.00000000	6.82500000	0.17500000	6.50621858	7.14378142
5	6.80000000	6.82500000	-0.02500000	6.50621858	7.14378142
6	6.40000000	6.82500000	-0.42500000	6.50621858	7.14378142
7	7.00000000	6.82500000	0.17500000	6.50621858	7.14378142
8	6.40000000	6.82500000	-0.42500000	6.50621858	7.14378142
9	6.80000000	6.46666667	0.33333333	6.09856959	6.83476374
10	7.00000000	6.46666667	0.53333333	6.09856959	6.83476374
11	6.20000000	6.46666667	-0.26666667	6.09856959	6.83476374
12	6.20000000	6.46666667	-0.26666667	6.09856959	6.83476374
13	6.40000000	6.46666667	-0.06666667	6.09856959	6.83476374
14	6.20000000	6.46666667	-0.26666667	6.09856959	6.83476374
15	6.40000000	6.12000000	0.28000000	5.71676985	6.52323015
16	→ 5.20000000	6.12000000	-0.92000000	5.71676985	6.52323015
17	6.20000000	6.12000000	0.08000000	5.71676985	6.52323015
18	6.40000000	6.12000000	0.28000000	5.71676985	6.52323015
19	6.40000000	6.12000000	0.28000000	5.71676985	6.52323015
20	6.80000000	6.90000000	-0.10000000	6.26243716	7.53756284
21	7.00000000	6.90000000	0.10000000	6.26243716	7.53756284
22	6.20000000	6.75000000	-0.55000000	6.43121858	7.06878142
23	→ 5.60000000	6.75000000	-1.15000000	6.43121858	7.06878142
24	7.20000000	6.75000000	0.45000000	6.43121858	7.06878142
25	7.20000000	6.75000000	0.45000000	6.43121858	7.06878142
26	7.20000000	6.75000000	0.45000000	6.43121858	7.06878142
27	6.20000000	6.75000000	-0.55000000	6.43121858	7.06878142
28	7.20000000	6.75000000	0.45000000	6.43121858	7.06878142
29	7.20000000	6.75000000	0.45000000	6.43121858	7.06878142
30	7.20000000	6.90000000	0.30000000	6.53190292	7.26809708
31	6.80000000	6.90000000	-0.10000000	6.53190292	7.26809708
32	7.00000000	6.90000000	0.10000000	6.53190292	7.26809708
33	6.80000000	6.90000000	-0.10000000	6.53190292	7.26809708
34	7.00000000	6.90000000	0.10000000	6.53190292	7.26809708
35	6.60000000	6.90000000	-0.30000000	6.53190292	7.26809708

Predicted values are the cell means estimated from the general means model

LOC	TYPE		
	North	Mesic	Shrub
Near	6.825	6.467	6.12
Away	6.90	6.75	6.90

The last two columns give 95% confidence intervals on the population cell means

(C) General Means Estimation (with No Interaction)

```

PROC REG DATA=SWAMP;
MODEL Y=X1-X5/NOINT P CLM;
RESTRICT Y2-X3-X5+Y6=0;
RESTRICT X1-X3-X4+Y6=0;
NC: TEST X1-.5*Y2-.5*Y3+Y4-.5*Y5-.5*Y6=0;
OC1: TEST Y2-Y3+X5-Y6=0;
OC2: TEST .725773*X1-.395874*X2-.529897*Y3+.274227*X4-.156701*X5
      -.117526*Y6=0;
OR1: TEST .206*X1+.442*X2+.352*X3-.206*X4-.442*X5-.352*X6=0;

```

Situation: the interaction is judged not to be important and both factors are needed in the model.

- Objective: (1) to estimate the cell means for the restricted general means model (the two RESTRICT statements of no interaction forces any contrast among the three columns to be the same for each row or the difference between row means will be the same for all columns).
 (2) to estimate contrasts (natural or ortho) among (i) the column cell means averaged over rows and (ii) row cell means averaged over columns.

DEP VARIABLE: Y

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE	PROB > F
MODEL	4	1555.596	388.899	1930.770	0.0001
ERROR	31	6.244072	0.201422		
U TOTAL	35	1561.840			
ROOT MSE		0.448802	R-SQUARE	0.9560	
DEP MEAN		6.662857	ADJ R-SQ	0.9556	
C.V.		6.735853			

NOTE: NO INTERCEPT TERM IS USED. R-SQUARE IS REDEFINED.

VARIABLE	DF	PARAMETER ESTIMATE	STANDARD ERROR	T FOR HC:	PROB > T
				PARAMETER=0	
X1	1	6.756998	0.145537	46.428	0.0001
X2	1	6.391427	0.151020	42.266	0.0001
X3	1	6.313082	0.161362	39.161	0.0001
X4	1	7.172003	0.191737	37.405	0.0001
X5	1	6.626430	0.138412	49.176	0.0001
X6	1	6.734092	0.153873	43.764	0.0001
RESTRICT	-1	0.451440	.	.	1.0000
RESTRICT	-1	0.544005	.	.	1.0000
TEST: NC		NUMERATOR: 1.02702	DF: 1	F VALUE: 5.0980	
		DENOMINATOR: 0.201422	DF: 31	PROB > F: 0.0311	
TEST: OC1		NUMERATOR: .0322163	DF: 1	F VALUE: 0.1599	
		DENOMINATOR: 0.201422	DF: 31	PROB > F: 0.6919	
TEST: OC2		NUMERATOR: 1.01055	DF: 1	F VALUE: 5.0171	
		DENOMINATOR: 0.201422	DF: 31	PROB > F: 0.0324	
TEST: OR1		NUMERATOR: .0322162	DF: 1	F VALUE: 0.1599	
		DENOMINATOR: 0.201422	DF: 31	PROB > F: 0.6920	
TEST: OR2		NUMERATOR: 1.33577	DF: 1	F VALUE: 6.6317	
		DENOMINATOR: 0.201422	DF: 31	PROB > F: 0.0150	
TEST: OR3		NUMERATOR: 1.33577	DF: 1	F VALUE: 6.6317	
		DENOMINATOR: 0.201422	DF: 31	PROB > F: 0.0150	

Parameter estimates are the cell means estimated from the restricted general means model

	Community		
LOC	North	Mesic	Shrub
near	6.757	6.391	6.319
away	7.172	6.806	6.734

Note that the difference between rows is the same for each column (any contrast among columns is the same for each row).

Confidence intervals on the restricted population cell means are calculated with the CLM option.

PROC GLM: CLASSES LOC TYPE:

MODEL Y=LOC TYPE/P SS1 SS2:

MEANS LOC TYPE:

LSMEANS LOC TYPE:

ESTIMATE *LOC NAT MAIN EFFECT*

LOC 1 -1:

ESTIMATE *TYPE NAT MAIN EFFECT1*

TYPE 1 -.5 -.5:

ESTIMATE *TYPE NAT MAIN EFFECT2*

TYPE 2 1 -1:

ESTIMATE *TYPE ORTHO MAIN EFFECT1*

TYPE 1 -.952577 -.447423:

① General Means Estimation (no interaction) with Reduced Model

DEPENDENT VARIABLE: Y

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE	PR > F	R-SQUARE	C
MODEL	3	1.81764195	0.60588065	3.01	0.0451	0.225466	6.7
ERROR	31	6.24407234	0.20142169		STD DEV		Y M
CORRECTED TOTAL	34	8.06171429			0.44880028		6.66285

SOURCE	DF	TYPE I SS	F VALUE	PR > F	DF	TYPE II SS	F VALUE	PR
LOC	1	0.77487218	3.85	0.0589	1	1.33577182	6.63	0.0
TYPE	2	1.04276977	2.59	0.0913	2	1.04276977	2.59	0.0

PARAMETER	ESTIMATE	T FOR HQ: PARAMETER=J	PR > T	STD ERROR OF ESTIMATE
LOC NAT MAIN EFFECT	-0.41500335	-2.58	0.0150	0.16115307
TYPE NAT MAIN EFFECT	0.40174146	2.26	0.0311	0.17791409
TYPE NAT MAIN EFFECT	0.07233758	0.40	0.6919	0.18097522
TYPE ORTHO MAIN EFFECT	0.39793817	2.24	0.0324	0.17755975

OBSERVATION	OBSERVED VALUE	PREDICTED VALUE	RESIDUAL
1	6.60000000	6.75599933	-0.15699933
2	7.20000000	6.75599933	0.44300067
3	7.20000000	6.75599933	0.44300067
4	7.00000000	6.75599933	0.24300067
5	6.80000000	6.75599933	0.04300067
6	6.40000000	6.75599933	-0.35699933
7	7.00000000	6.75599933	0.24300067
8	6.40000000	6.75599933	-0.35699933
9	6.80000000	6.39142666	0.40857334
10	7.00000000	6.39142666	0.60857334
11	6.20000000	6.39142666	-0.19142666
12	6.20000000	6.39142666	-0.19142666
13	6.40000000	6.39142666	0.00857334
14	6.20000000	6.39142666	-0.19142666
15	6.40000000	6.31998908	0.08091092
16	5.20000000	6.31998908	-1.11998908
17	6.20000000	6.31998908	-0.11998908
			18 6.40000000 6.31998908 0.08091092
			19 6.40000000 6.31998908 0.08091092
			20 6.80000000 7.17200268 -0.37200268
			21 7.00000000 7.17200268 -0.17200268
			22 6.20000000 6.80643001 -0.60643001
			23 5.60000000 6.80643001 -1.20643001
			24 7.20000000 6.80643001 0.39356999
			25 7.20000000 6.80643001 0.39356999
			26 7.20000000 6.80643001 0.39356999
			27 6.20000000 6.80643001 -0.60643001
			28 7.20000000 6.80643001 0.39356999
			29 7.20000000 6.80643001 0.39356999
			30 7.20000000 6.73409243 0.46590757
			31 6.80000000 6.73409243 0.06590757
			32 7.00000000 6.73409243 0.26590757
			33 6.80000000 6.73409243 0.06590757
			34 7.00000000 6.73409243 0.26590757
			35 6.60000000 6.73409243 -0.13409243

Note: Results from ESTIMATE statements are as in (C)

MEANS

LOC	N	Y	
1	10	6.52631579	= $\rightarrow [6.825(8) + 6.467(6) + 6.12(5)]/19$ (weighted average of observations)
2	16	6.82500000	

TYPE	N	Y	
1	10	6.84000000	= $\rightarrow [6.825(8) + 6.9(2)]/10$
2	14	6.62857143	
3	11	6.54545455	

LEAST SQUARES MEANS

LOC	Y	
	LSMEAN	
1	6.48917169	= $\rightarrow [6.757 + 6.391 + 6.319]/3$ (unweighted average of estimated cell means)
2	6.90417504	

TYPE	Y	
	LSMEAN	
1	6.96450100	= $[6.757 + 7.172]/2$
2	6.59892833	
3	6.52659076	

Note: $6.904 - 6.489 = -0.415$
 $6.965 - [6.599 + 6.527]/2 = 0.4017$
 $6.599 - 6.527 = 0.072$
 $6.965 - [(0.552577)(6.599) + (0.447423)(6.527)] = 0.398$

SWAMP DATA - Unweighted Analysis of Cell Means
see Snedecor & Cochran, 7 ed., p. 418

```
DATA SWAMP;  
INPUT LOC TYPE TMT Y;  
CARDS;
```

```
PROC ANOVA; CLASS TMT;  
MODEL Y = TMT;  
MEANS TMT / DEONLY;
```

Ⓔ The six treatment combinations are indicated in the CLASS variable TMT.
The residual MS is an estimate of σ^2 .

```
PROC SORT; BY TYPE LOC;
```

```
PROC MEANS MEAN NOPRINT; BY TYPE LOC;  
OUTPUT OUT=NEW MEAN=MY; VAR Y;
```

Ⓕ The six cell means are computed and placed
in the data set NEW.

```
PROC ANOVA; CLASS TYPE LOC;  
MODEL MY = LOC TYPE LOC*TYPE;  
MEANS LOC TYPE LOC*TYPE / DEONLY;
```

Ⓖ The main effect SS and interaction SS are determined at this step.

Ⓔ One-way ANOVA on the six treatment combinations.

ANALYSIS OF VARIANCE PROCEDURE

CLASS LEVEL INFORMATION

CLASS	LEVELS	VALUES
TMT	6	1 2 3 4 5 6

NUMBER OF OBSERVATIONS IN DATA SET = 35

DEPENDENT VARIABLE: Y

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE
MODEL	5	2.42538095	0.48507619	2.50
ERROR	29	5.63633333	0.19435632	PR > F
CORRECTED TOTAL	34	8.06171429	= s ²	0.0536

Note: s² is all that is to be used from this ANOVA table along with the associated error df = 29.

R-SQUARE	C.V.	STD DEV	Y MEAN
0.300852	6.6167	0.44085862	6.66285714

SOURCE	DF	ANOVA SS	F VALUE	PR > F
TMT	5	2.42538095	2.50	0.0536

MEANS are the cell means

TMT	N	Y
1	8	6.82500000
2	6	6.46666667
3	5	6.12000000
4	2	6.90000000
5	8	6.75000000
6	6	6.90000000

Note: Since $\frac{\max(n_{ij})}{\min(n_{ij})} = \frac{8}{2} > 2$, the analysis of unweighted cell means is of dubious worth.

Calculate: $\frac{1}{n_h} = \frac{1}{2(3)} \left(\frac{1}{8} + \frac{1}{6} + \frac{1}{5} + \frac{1}{2} + \frac{1}{8} + \frac{1}{6} \right) = 0.21389$.

Thus, $n_h = 4.675$

Ⓒ There is no output from Ⓕ

ANALYSIS OF VARIANCE PROCEDURE

DEPENDENT VARIABLE: MY

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE
MODEL	5	0.47950231	0.09590046
ERROR	0	0.00000000	0.00000000
CORRECTED TOTAL	5	0.47950231	
R-SQUARE	C.V.	STD DEV	MY MEAN
1.000000	0.0000	0.00000000	6.66027778

SOURCE	DF	ANOVA SS	F VALUE	PR > F
LOC	1	0.21596713	.	.
TYPE	2	0.13235093	.	.
TYPE*LOC	2	0.13118426	.	.

ANALYSIS OF VARIANCE PROCEDURE

CLASS LEVEL INFORMATION			Type 1 = North	Loc 1 = Near
CLASS	LEVELS	VALUES	2 = Mesic	2 = Away
TYPE	3	1 2 3	3 = Shrub	
LOC	2	1 2		

NUMBER OF OBSERVATIONS IN DATA SET = 6 = the number of cell means.

These are the desired SS which need to be multiplied by $n_h = 4.675$, the harmonic mean number of observations per cell.

ANOVA of Unweighted Cell Means

Source	df	SS	MS	F
LOC	1	1.00965	1.00965	5.19
TYPE	2	0.61874	0.30937	1.59
TYPE*LOC	2	0.61329	0.30664	1.58
ERROR	29	5.63633	0.19436	

where: LOC SS = $4.675(0.21596713) = 1.00965$
 TYPE SS = $4.675(0.13235093) = 0.61874$
 TYPE*LOC SS = $4.675(0.13118426) = 0.61329$

Error SS from part Ⓔ

$F_{1,29}^{.05} = 4.18$ $F_{2,29}^{.05} = 3.33$

$F_{1,29}^{.01} = 7.60$ $F_{2,29}^{.25} = 1.45$

ANALYSIS OF VARIANCE PROCEDURE

MEANS		
LOC	N	MY
1	3	6.47055556
2	3	6.85000000

TYPE	N	MY
1	2	6.86250000
2	2	6.60833333
3	2	6.51000000

These are unweighted means, e.g. $6.471 = \frac{1}{3}(6.825 + 6.467 + 6.120)$.

Thus, they correspond to the LSMEANS from the model which includes LOC, TYPE and LOC*TYPE.

Note: The N are not appropriate on this page.

TYPE	LOC	N	MY
1	1	1	6.82500000
1	2	1	6.90000000
2	1	1	6.46666667
2	2	1	6.75000000
3	1	1	6.12000000
3	2	1	6.90000000

These are the six cell means -- compare with part Ⓔ

The standard error of any location mean is $\sqrt{\frac{s^2}{3(n_h)}} = \sqrt{\frac{0.19436}{3(4.675)}} = 0.1177$

The standard error of any type mean is $\sqrt{\frac{0.19436}{2(4.675)}} = 0.1442$

See p. 419 of Snedecor and Cochran (7 ed.) for a discussion of calculating the correct standard errors of comparisons among row means, column means and individual cell means.

```

/*LIMITS REGION=300K
/*FORMAT A,FORMS=024,DEST=LWK2
// EXEC SAS
OPTIONS LS=80 MDATE MNOTES;
DATA SOYBEAN;
TITLE SOYBEAN DATA;
INPUT LIGHT $ HEIGHT YIELD;
DEV = HEIGHT;
X1 = (LIGHT='C'); } X1 = (LIGHT='C') is equivalent to: IF LIGHT='C' THEN X1=1;
X2 = (LIGHT='L'); } ELSE X1=0;
X3 = 1-X1-X2;

```

X1, X2, X3 are treatment indicators, HEIGHT is the covariate, YIELD is the response variable

```

CARDS;
C 43 12.2
C 52 12.4
C 42 11.9
C 35 11.3
C 40 11.9
C 48 12.1
C 60 13.1
C 61 12.7
C 50 12.4
C 33 11.4
C 48 12.3
C 51 12.2
C 56 12.6
C 65 13.2
C 51 12.3
L 63 16.6
L 50 15.8
L 63 16.5
L 33 15.0
L 38 15.4
L 45 15.6
L 50 15.8
L 46 15.8
L 50 16.0
L 49 15.8
L 35 15.0
L 50 16.2
L 62 16.7
L 49 15.8
L 52 15.9
S 52 9.5
S 54 9.5
S 53 9.6
S 45 8.8
S 57 9.5
S 62 9.6
S 52 9.1
S 47 10.3
S 55 9.5
S 40 8.5
S 41 8.5
S 67 10.4
S 55 9.4

```

- (A) PROC STANDARD MEAN=0; VAR DEV; ← Sets mean of DEV=0. It is equivalent to HEIGHT - $\overline{\text{HEIGHT}}$.
- (B) PROC PRINT; VAR X1 X2 X3 DEV; Prints the X matrix for (C).
- (C) PROC REG; MODEL YIELD=X1 X2 X3 DEV/NOINT SECF SS1 SS2; Model fitting the covariate deviations last.
 OUTPUT OUT=NEW PREDICTED=YHAT RESIDUAL=RESID;
 C_VS_TRT: TEST X1-.5*X2-.5*X3=0; Use TEST statements with PROC REG to test contrasts.
 L_VS_S: TEST X2-X3=0; Instead of specifying the coefficients as with ESTIMATE, use the equation that represents the contrasts under the null hypothesis.
- (D) PROC PLOT DATA=NEW; PLOT RESID*YHAT=LIGHT/VREF=0; Residual plot for (C).
 PLOT YIELD*HEIGHT=LIGHT; Plot of the response variable vs. the covariate.
- (E) PROC GLM; MODEL YIELD=HEIGHT X1 X2 X3; Model fitting the covariate first.
- (F) PROC GLM; CLASS LIGHT; MODEL YIELD=LIGHT HEIGHT LIGHT*HEIGHT; Model fitting separate slopes after a common slope.
- (G) PROC GLM; CLASS LIGHT; MODEL YIELD = LIGHT HEIGHT; LSMEANS LIGHT / STDERR; ← Adjusted treatment means
 MEANS LIGHT / DEONLY; ← Unadjusted treatment means
 ESTIMATE 'CONTROL VS TMT' HEIGHT 0 LIGHT 2 -1 -1 / DIVISOR=2;
 ESTIMATE 'LIGHT VS SHADE' HEIGHT 0 LIGHT 0 1 -1;
 ESTIMATE 'COMMON SLOPE' HEIGHT 1;

SOYBEAN DATA

OES	X1	X2	X3	DEV
1	1	0	0	-3.2
2	1	0	0	0.8
3	1	0	0	-9.2
4	1	0	0	-16.2
5	1	0	0	-11.2
6	1	0	0	-3.2
7	1	0	0	8.8
8	1	0	0	9.8
9	1	0	0	-1.2
10	1	0	0	-18.2
11	1	0	0	-3.2
12	1	0	0	-0.2
13	1	0	0	4.8
14	1	0	0	13.8
15	1	0	0	-0.2
16	0	1	0	11.8
17	0	1	0	-1.2
18	0	1	0	11.8
19	0	1	0	-18.2
20	0	1	0	-13.2
21	0	1	0	-6.2
22	0	1	0	-1.2
23	0	1	0	-3.2
24	0	1	0	-1.2
25	0	1	0	-2.2
26	0	1	0	-16.2
27	0	1	0	-1.2
28	0	1	0	10.8
29	0	1	0	-2.2
30	0	1	0	0.8
31	0	0	1	0.8
32	0	0	1	2.8
33	0	0	1	6.8
34	0	0	1	-6.2
35	0	0	1	5.8
36	0	0	1	10.8
37	0	0	1	0.8
38	0	0	1	15.8
39	0	0	1	3.8
40	0	0	1	-11.2
41	0	0	1	-10.2
42	0	0	1	15.8
43	0	0	1	3.8
44	0	0	1	14.8
45	0	0	1	4.8

(B) The X matrix for part (C)

Control treatment
indicator

Light treatment
indicator

Shade treatment
indicator

HEIGHT - $\overline{\text{HEIGHT}}$ (deviation of the covariate from its mean).

GENERAL MEANS MODEL

Ⓒ YIELD = Treatment indicators | covariate deviations

SEQUENTIAL PARAMETER ESTIMATES

X1 12.26 = \bar{y}_c
 X2 12.26 15.86 = \bar{y}_L
 X3 12.26 15.86 9.46667 = \bar{y}_s
 DEV 12.369 15.9806 9.23709 .0583682 = common slope ← The first three partials are the adjusted treatment means (see p. 237).

DEP VARIABLE: YIELD

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE	PROB > F
MODEL	4	7383.377	1845.844	110755.812	0.0001
ERROR	41	0.683301	0.0166666		
U TOTAL	45	7384.060			
ROOT MSE		0.129096	R-SQUARE	0.9999	
DEP MEAN		12.528889	ADJ R-SQ	0.9999	
C.V.		1.03039			

NOTE: NO INTERCEPT TERM IS USED. R-SQUARE IS REDEFINED.

VARIABLE	DF	PARAMETER ESTIMATE	STANDARD ERROR	T FOR H0: PARAMETER=0	PROB > T	TYPE I SS	
X1	1	12.368954	0.033592	368.213	0.0001	2254.614	$\bar{y}_c(\text{adj}) = 12.37$
X2	1	15.980628	0.033650	474.906	0.0001	3773.094	$\bar{y}_L(\text{adj}) = 15.98$
X3	1	9.237085	0.034469	267.984	0.0001	1344.267	$\bar{y}_s(\text{adj}) = 9.24$
DEV	1slope=0.058368		0.002231513	26.156	0.0001	11.402032	

VARIABLE	DF	TYPE II SS
X1	1	2259.578
X2	1	3758.754
X3	1	1196.866
DEV	1	11.402032

TEST: C_VS_TRT
 NUMERATOR: 0.562358 DF: 1 F VALUE: 33.7431
 DENOMINATOR: .0166659 DF: 41 PROB > F: 0.0001

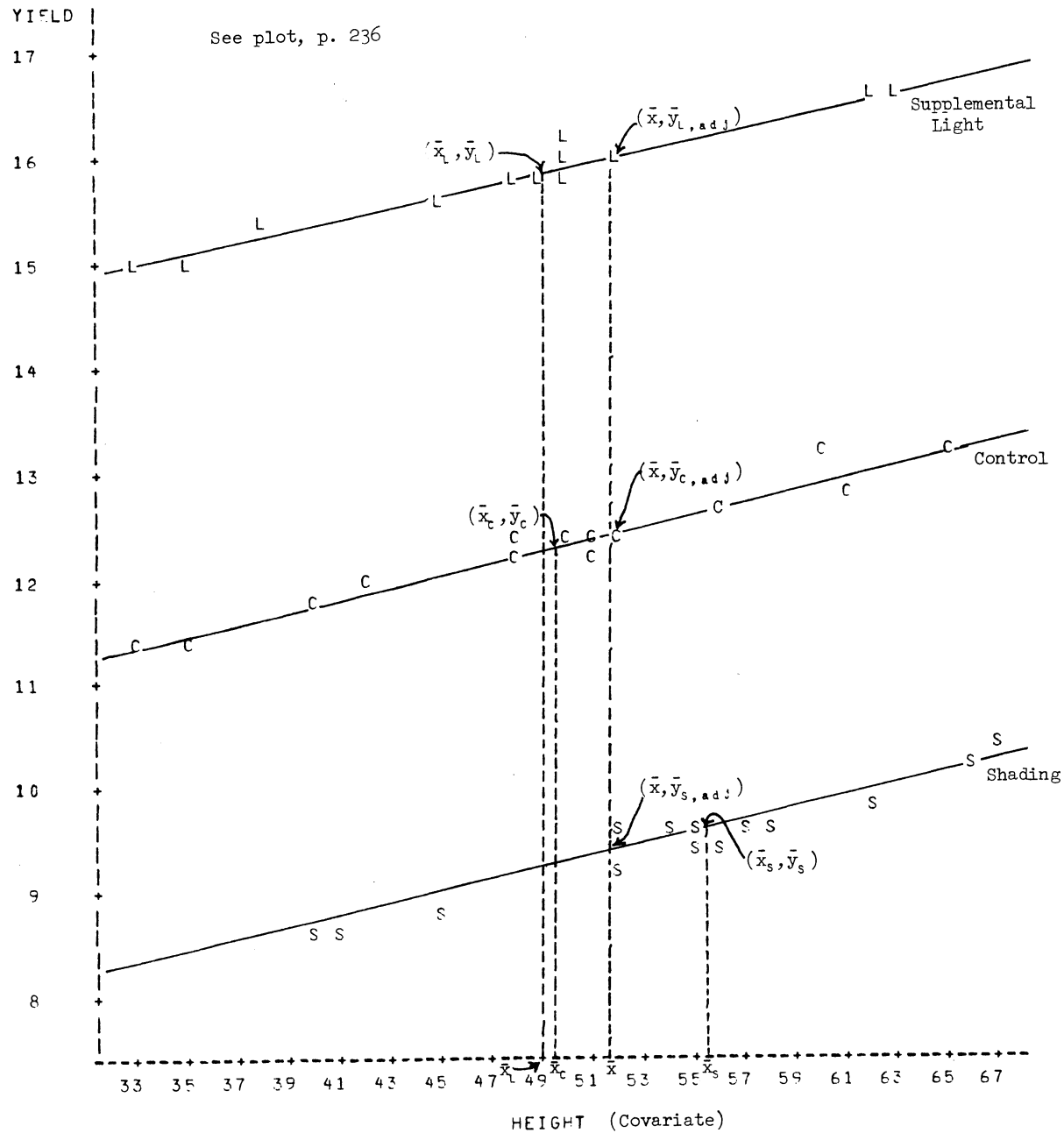
TEST: L_VS_S
 NUMERATOR: 315.604 DF: 1 F VALUE: 8937.1324
 DENOMINATOR: .0166659 DF: 41 PROB > F: 0.0001

TESTS of contrasts

These contrasts test differences between adjusted means.
 The ESTIMATE option in GIM would provide the estimate and standard error (see Ⓒ).

① PLOT OF YIELD*HEIGHT SYMBOL IS VALUE OF LIGHT

See plot, p. 236



NOTE: 5 OBS HIDDEN

GENERAL LINEAR MODELS PROCEDURE

DEPENDENT VARIABLE: YIELD

Ⓔ Covariate/treatments model

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE
MODEL	3	319.58914296	106.52971432	6392.08
ERROR	41	0.68330149	0.01666589	PR > F
CORRECTED TOTAL	44	320.27244444		0.0001

R-SQUARE	C.V.	STD DEV	YIELD MEAN
0.997866	1.0304	0.12909644	12.52888889

SOURCE	DF	TYPE I SS	F VALUE	PR > F
HEIGHT	1	1.77817807	106.70	0.0001
X1	1	2.20679982	132.41	0.0001
X2	1	315.60416506	18937.13	0.0001
X3	0	0.00000000	.	.

SOURCE	DF	TYPE IV SS	F VALUE	PR > F
HEIGHT	1	11.40203184	684.15	0.0001
X1	0	0.00000000	.	.
X2	0	0.00000000	.	.
X3	0	0.00000000	.	.

PARAMETER	ESTIMATE	T FOR H0: PARAMETER=0	PR > T	STD ERROR OF ESTIMATE
INTERCEPT	6.24863365 B	49.02	0.0001	0.12746616
HEIGHT	0.05836819	26.16	0.0001	0.00223151
X1	3.13186885 B	64.07	0.0001	0.04888388
X2	6.74354249 B	137.61	0.0001	0.04900394
X3	0.00000000 B	.	.	.

NOTE: THE X*X MATRIX HAS BEEN DEEMED SINGULAR AND A GENERALIZED INVERSE HAS BEEN EMPLOYED TO SOLVE THE NORMAL EQUATIONS. THE ABOVE ESTIMATES REPRESENT ONLY ONE OF MANY POSSIBLE SOLUTIONS TO THE NORMAL EQUATIONS. ESTIMATES FOLLOWED BY THE LETTER B ARE BIASED AND DO NOT ESTIMATE THE PARAMETER BUT ARE BLUE FOR SOME LINEAR COMBINATION OF PARAMETERS (OR ARE ZERO). THE EXPECTED VALUE OF THE BIASED ESTIMATORS MAY BE OBTAINED FROM THE GENERAL FORM OF ESTIMABLE FUNCTIONS. FOR THE BIASED ESTIMATORS, THE STD ERR IS THAT OF THE BIASED ESTIMATOR AND THE T VALUE TESTS H0: E(BIASED ESTIMATOR) = 0. ESTIMATES NOT FOLLOWED BY THE LETTER B ARE BLUE FOR THE PARAMETER.

Produces the analysis of covariance table, p. 238.

SOYBEAN DATA

GENERAL LINEAR MODELS PROCEDURE (F) Treatments/common slope/separate slopes model

CLASS LEVEL INFORMATION

CLASS	LEVELS	VALUES
LIGHT	3	C L S

NUMBER OF OBSERVATIONS IN DATA SET = 45

SOYBEAN DATA

GENERAL LINEAR MODELS PROCEDURE

DEPENDENT VARIABLE: YIELD

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE
MODEL	5	319.66578021	63.93315604	4110.01
ERROR	39	0.60666423	0.01555549	PR > F
CORRECTED TOTAL	44	320.27244444		0.0001

R-SQUARE	C.V.	STD DEV	YIELD MEAN
0.998106	0.9955	0.12472166	12.52888889

SOURCE	DF	TYPE I SS	F VALUE	PR > F
LIGHT	2	308.18711111	9906.05	0.0001
HEIGHT	1	11.40203184	732.99	0.0001
HEIGHT*LIGHT	2	0.07663726	2.46	0.0983

this line tests $H_0: \beta_1 = \beta_2 = \beta_3 (= \beta)$
vs. H_a : different slopes needed.

SOURCE	DF	TYPE IV SS	F VALUE	PR > F
LIGHT	2	10.53027719	338.47	0.0001
HEIGHT	1	11.47084949	737.41	0.0001
HEIGHT*LIGHT	2	0.07663726	2.46	0.0983

SOYBEAN DATA

Ⓒ Treatments/common slope model

GENERAL LINEAR MODELS PROCEDURE

DEPENDENT VARIABLE: YIELD

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE
MODEL	3	319.58914296	106.52971432	6392.00
ERROR	41	0.68330149	0.01666589	PR > F
CORRECTED TOTAL	44	320.27244444		0.0001

R-SQUARE	C.V.	STD DEV	YIELD MEAN
0.997866	1.0304	0.12909644	12.52888889

SOURCE	DF	TYPE I SS	F VALUE	PR > F
LIGHT = R(L μ)	2	308.18711111	9246.04	0.0001
HEIGHT = R(β L,μ)	1	11.40203184	684.15	0.0001

$$R(\beta|\mu) = R(L, \beta|\mu) - R(L|\beta, \mu)$$

$$= 319.58914 - 317.81096$$

$$= 1.77818 = \text{SS due to fitting a common slope before the treatments.}$$

SOURCE	DF	TYPE IV SS	F VALUE	PR > F
LIGHT = R(L β,μ)	2	317.81096489	9534.77	0.0001
HEIGHT	1	11.40203184	684.15	0.0001

PARAMETER	ESTIMATE	T FOR H0: PARAMETER=0	PR > T	STD ERROR OF ESTIMATE
CONTROL VS TMT (1)	-0.23990239	-5.81	0.0001	0.04129927
LIGHT VS SHADE (2)	6.74354249	137.61	0.0001	0.04900394
COMMON SLOPE (3)	0.05836819	26.16	0.0001	0.00223151

(1) and (2) are the contrasts among adjusted treatment means (same as the contrasts used in the TESTs of part Ⓒ).

(3) is the estimate of the common slope of the covariate.

Note that the F-value output for (2) in PROC REG is not correct.

MEANS = unadjusted treatment means

LIGHT	N	YIELD
C	15	12.2600000
L	15	15.8600000
S	15	9.4666667

LEAST SQUARES MEANS = adjusted treatment means and their standard errors.

LIGHT	YIELD LSMEAN	STD ERR LSMEAN	PROB > T H0:LSMEAN=0
C	12.3089540	0.0335918	0.0001
L	15.9806276	0.0336501	0.0001
S	9.2370851	0.0344688	0.0001

Compare the MEANS and LSMEANS with the SEQB output of PROC REG in Ⓒ.

POTATO SCAB DATA - Comparison of regression lines in a 2 x 3 factorial experiment (two qualitative levels x three quantitative levels). P. 97, Cochran and Cox, 1957.

DATA SCAB;
 INPUT YIELD TMT \$ LEVEL X0 X01 X02 X1 X11 X12 X2 X21 X22;
 CARDS;

(A) PROC PRINT N;

(B) PROC REG;
 MODEL YIELD = X0 X01 X02 X1 X11 X12 X2 X21 X22 /
 NOINT P SEQB SS1 CLM;

(B) and (C) correspond to fitting a sequence of regression models

(C) PROC REG;
 MODEL YIELD = X01 X02 X1 X2 / NOINT P SEQB SS1 CLM;
 OUTPUT OUT=NEW P=YHAT R=RES U95M=UPPER L95M=LOWER;

(D) PROC PLOT DATA=NEW;
 PLOT YIELD*LEVEL=TMT;
 PLOT RES*YHAT / VREF=0;
 PLOT YHAT*LEVEL='P' UPPER*LEVEL='U' LOWER*LEVEL='L' / OVERLAY;

A plot of the raw data as well as a residual plot and plot of predicted values for the model in part (C).

(E) PROC UNIVARIATE NORMAL PLOT DATA=NEW; VAR RES;

More analysis of residuals from the model in part (C).

(F) PROC GLM DATA=SCAB; CLASS TMT LEVEL;
 MODEL YIELD = TMT*LEVEL / NOINT P;
 ESTIMATE 'TMT' TMT*LEVEL 1 1 1 -1 -1 -1 / DIVISOR=3;
 ESTIMATE 'B LIN' TMT*LEVEL -4 -1 5 -4 -1 5 / DIVISOR=84;
 ESTIMATE 'T*B LIN' TMT*LEVEL -4 -1 5 4 1 -5 / DIVISOR=42;
 ESTIMATE 'B QUAD' TMT*LEVEL 2 -3 1 2 -3 1 / DIVISOR=108;
 ESTIMATE 'T*B QUAD' TMT*LEVEL 2 -3 1 -2 3 -1 / DIVISOR=54;

(F) Fitting the cell means model and examining single degree-of-freedom contrasts which correspond to linear, quadratic, treatment and the respective interaction terms. The results are identical to those of part (B).

Recall that $b_{2.01} = \sum l_i y_i$ where the l_i 's may be computed from, say, the ORTHO algorithm. In this example $l_1 = \frac{2}{54}$, $l_2 = \frac{-3}{54}$ and $l_3 = \frac{1}{54}$. If the levels of the quantitative factor had been equally spaced, then the l_i 's could have been obtained from a table of orthogonal polynomial coefficients.

Ⓐ The data and indicator variables

OBS	YIELD	TMT	LEVEL	X0	X01	X02	X1	X11	X12	X2	X21	X22
1	9	F	3	1	1	0	3	3	0	9	9	0
2	9	F	3	1	1	0	3	3	0	9	9	0
3	16	F	3	1	1	0	3	3	0	9	9	0
4	4	F	3	1	1	0	3	3	0	9	9	0
5	30	S	3	1	0	1	3	0	3	9	0	9
6	7	S	3	1	0	1	3	0	3	9	0	9
7	21	S	3	1	0	1	3	0	3	9	0	9
8	9	S	3	1	0	1	3	0	3	9	0	9
9	16	F	6	1	1	0	6	6	0	36	36	0
10	10	F	6	1	1	0	6	6	0	36	36	0
11	18	F	6	1	1	0	6	6	0	36	36	0
12	18	F	6	1	1	0	6	6	0	36	36	0
13	18	S	6	1	0	1	6	0	6	36	0	36
14	24	S	6	1	0	1	6	0	6	36	0	36
15	12	S	6	1	0	1	6	0	6	36	0	36
16	19	S	6	1	0	1	6	0	6	36	0	36
17	10	F	12	1	1	0	12	12	0	144	144	0
18	4	F	12	1	1	0	12	12	0	144	144	0
19	4	F	12	1	1	0	12	12	0	144	144	0
20	5	F	12	1	1	0	12	12	0	144	144	0
21	17	S	12	1	0	1	12	0	12	144	0	144
22	7	S	12	1	0	1	12	0	12	144	0	144
23	16	S	12	1	0	1	12	0	12	144	0	144
24	17	S	12	1	0	1	12	0	12	144	0	144

N=24

ⓑ Fitting the model sequence X0|X01 X02|X1|X11 X12|X2|X21 X22

SEQUENTIAL PARAMETER ESTIMATES

which is: common mean|separate means|common linear|separate linear|common quadratic|separate quadratic

X0	13.3333	= \bar{y}								
X01	16.4167	-6.16667	= $\bar{y}_F - \bar{y}_S$							
X02	16.4167	-6.16667	1.4E-14	= 0						
X1	19.6458	-6.16667	1.7E-14	-0.46131	= $b_{1,0}$	(common linear)				
X11	18.75	-4.375	1.7E-14	-0.333333	-0.255952	= $b_{F1,0} - b_{S1,0}$				
X12	19.75	-4.375	1.7E-14	-0.333333	-0.255952	1.1E-13	= 0			
X2	6.77083	-4.375	1.8E-14	3.77781	-0.255952	1.6E-13	-0.266204	= $b_{2,01}$	(common quadratic)	
X21	12.9167	-16.6667	3.3E-14	1.66667	3.95833	-2.2E-13	-0.12963	-0.273148	= $b_{F2,01} - b_{S2,01}$	
X22	12.9167	-16.6667	3.3E-14	1.66667	3.95833	-2.2E-13	-0.12963	-0.273148	1.2E-12	= 0

DEP VARIABLE: YIELD

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE	PROB>F
MODEL	6	4721.000	786.833	22.374	0.0001
ERROR	18	633.000	35.166667		
U TOTAL	24	5354.000			
ROOT MSE		5.930149			
DEP MEAN		13.333333			
C.V.		44.47612			
			R-SQUARE	0.8818	
			ADJ R-SQ	0.8489	

Standard errors and hypothesis tests are most easily determined from part ⓑ.

NOTE: NO INTERCEPT TERM IS USED. R-SQUARE IS REDEFINED.

NOTE: MODEL IS NOT FULL RANK. LEAST SQUARES SOLUTIONS FOR THE PARAMETERS ARE NOT UNIQUE. SOME STATISTICS WILL BE MISLEADING. A REPORTED DF OF 0 OR B MEANS THAT THE ESTIMATE IS BIASED. THE FOLLOWING PARAMETERS HAVE BEEN SET TO 0, SINCE THE VARIABLES ARE A LINEAR COMBINATION OF OTHER VARIABLES AS SHOWN.

X02	=+X0	-1*X01
X12	=+X1	-1*X11
X22	=+X2	-1*X21

} not of use.

VARIABLE	DF	PARAMETER ESTIMATE	STANDARD ERROR	T FOR H0: PARAMETER=0	PROB > T	TYPE I SS
X0	B	12.916667	9.932877	1.300	0.2099	4266.667
X01	B	-16.666667	14.047209	-1.186	0.2509	228.167
X02	0	0	.	.	.	0
X1	B	1.666667	3.202646	0.520	0.6091	71.502976
X11	B	3.958333	4.529226	0.874	0.3937	5.502976
X12	0	0	.	.	.	0
X2	B	-0.129630	0.205450	-0.631	0.5360	118.080
X21	B	-0.273148	0.290550	-0.940	0.3596	31.080357

(B) Results of the P and CIM option

OBS	ACTUAL	PREDICT VALUE (Cell Means)	STD ERR PREDICT	LOWER95% MEAN	UPPER95% MEAN	RESIDUAL
1	9.000	9.500	2.965	3.271	15.729	-.500000
2	9.000	9.500	2.965	3.271	15.729	-.500000
3	16.000	9.500	2.965	3.271	15.729	6.500
4	4.000	9.500	2.965	3.271	15.729	-5.500
5	20.000	16.750	2.965	10.521	22.979	13.250
6	7.000	16.750	2.965	10.521	22.979	-9.750
7	21.000	16.750	2.965	10.521	22.979	4.250
8	9.000	16.750	2.965	10.521	22.979	-7.750
9	16.000	15.500	2.965	9.271	21.729	0.500000
10	10.000	15.500	2.965	9.271	21.729	-5.500
11	18.000	15.500	2.965	9.271	21.729	2.500
12	18.000	15.500	2.965	9.271	21.729	2.500
13	18.000	18.250	2.965	12.021	24.479	-.250000
14	24.000	18.250	2.965	12.021	24.479	5.750
15	12.000	18.250	2.965	12.021	24.479	-6.250
16	19.000	18.250	2.965	12.021	24.479	0.750000
17	10.000	5.750	2.965	-.479346	11.979	4.250
18	4.000	5.750	2.965	-.479346	11.979	-1.750
19	4.000	5.750	2.965	-.479346	11.979	-1.750
20	5.000	5.750	2.965	-.479346	11.979	-.750000
21	17.000	14.250	2.965	8.021	20.479	2.750
22	7.000	14.250	2.965	8.021	20.479	-7.250
23	16.000	14.250	2.965	8.021	20.479	1.750
24	17.000	14.250	2.965	8.021	20.479	2.750

SUM OF RESIDUALS 2.34479E-13
 SUM OF SQUARED RESIDUALS 633

GENERAL LINEAR MODELS PROCEDURE

DEPENDENT VARIABLE: YIELD

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE
MODEL	6	4721.00000000	786.83333333	22.37
ERROR	18	633.00000000	35.16666667	PR > F
UNCORRECTED TOTAL	24	5354.00000000		0.0001

R-SQUARE	C.V.	STD DEV	YIELD MEAN
0.881771	44.4761	5.93014896	13.33333333

SOURCE	DF	TYPE I SS	F VALUE	PR > F
TMT*LEVEL	6	4721.00000000	22.37	0.0001

SOURCE	DF	TYPE IV SS	F VALUE	PR > F
TMT*LEVEL	6	4721.00000000	22.37	0.0001

PARAMETER	ESTIMATE	T FOR H0: PARAMETER=0	PR > T	STD ERROR OF ESTIMATE
TMT	-6.16666667	-2.55	0.0202	2.42097317
B LIN	-0.46130952	-1.43	0.1710	0.32351615
T*B LIN	-0.25595238	-0.40	0.6971	0.64703230
B QUAD	-0.26620370	-1.83	0.0835	0.14527499
T*B QUAD	-0.27314815	-0.94	0.3596	0.29054999

OBSERVATION	OBSERVED VALUE	PREDICTED VALUE (= cell means)	RESIDUAL
1	9.00000000	9.50000000	-0.50000000
2	9.00000000	9.50000000	-0.50000000
3	16.00000000	9.50000000	6.50000000
4	4.00000000	9.50000000	-5.50000000
5	30.00000000	16.75000000	13.25000000
6	7.00000000	16.75000000	-9.75000000
7	21.00000000	16.75000000	4.25000000
8	9.00000000	16.75000000	-7.75000000
9	16.00000000	15.50000000	0.50000000
10	10.00000000	15.50000000	-5.50000000
11	18.00000000	15.50000000	2.50000000
12	18.00000000	15.50000000	2.50000000

Ⓣ General Means Model

GENERAL LINEAR MODELS PROCEDURE

CLASS LEVEL INFORMATION

CLASS	LEVELS	VALUES
TMT	2	F S
LEVEL	3	3 6 12

NUMBER OF OBSERVATIONS IN DATA SET = 24

Note that the parameter estimates are the same as those given in part ⓑ under the SEQB output. Here we also have the standard errors and t-tests; however, we had to calculate the h_i 's for each contrast.

(Cell means)

		LEVEL		
		3	6	12
TMT	F	9.50	15.50	5.75
	S	16.75	18.25	14.25

The standard error of each cell mean is

$$\sqrt{\frac{35.1667}{4}} = 2.965$$

The cell means and standard errors could have been printed by using the option:

LSMEANS TMT*LEVEL/STDERR;

Ⓡ continued GENERAL LINEAR MODELS PROCEDURE

DEPENDENT VARIABLE: YIELD

OBSERVATION	OBSERVED VALUE	PREDICTED VALUE	RESIDUAL
13	18.00000000	18.25000000	-0.25000000
14	24.00000000	18.25000000	5.75000000
15	12.00000000	18.25000000	-6.25000000
16	19.00000000	18.25000000	0.75000000
17	10.00000000	5.75000000	4.25000000
18	4.00000000	5.75000000	-1.75000000
19	4.00000000	5.75000000	-1.75000000
20	5.00000000	5.75000000	-0.75000000
21	17.00000000	14.25000000	2.75000000
22	7.00000000	14.25000000	-7.25000000
23	16.00000000	14.25000000	1.75000000
24	17.00000000	14.25000000	2.75000000
SUM OF RESIDUALS			0.00000000
SUM OF SQUARED RESIDUALS			633.00000000
SUM OF SQUARED RESIDUALS - ERROR SS			0.00000000
FIRST ORDER AUTOCORRELATION			-0.63467615
DURBIN-WATSON D			3.25701027

Note: In order to get the classical ANOVA table as well as the single degree-of-freedom contrasts and means with standard errors we could have used the statements:

```
PROC GLM; CLASS TMT LEVEL;
  MODEL YIELD = TMT LEVEL TMT*LEVEL/P CLM;
  [ same ESTIMATE statements ]
  LSMEANS TMT LEVEL TMT*LEVEL/STDERR;
```

Note: To obtain the same results as in Ⓢ we would fit the model with interaction restricted to be zero That is,

```
PROC GLM; CLASS TMT LEVEL;
  MODEL YIELD = TMT LEVEL /P CLM;
```

ALCOHOL-DRUG DATA, p. 280.

Analysis of a Split-Unit Experiment

```
DATA WHOLE;
INPUT Y1 Y2 Y3;
SUBJECT = _N_;
ALCOHOL = 'YES';
IF _N_ > 6 THEN ALCOHOL='NO';
YS = (Y1+Y2+Y3)/SQRT(3);
XO = 1;
CARDS;
```

Y1 = Drug A response

Y2 = Drug B response

Y3 = Control response

$$YS = \frac{Y_1 + Y_2 + Y_3}{\sqrt{3}}$$

```
PROC SORT; BY ALCOHOL;
```

(A) PROC PRINT N; BY ALCOHOL;

(B) PROC GLM; CLASS ALCOHOL;
MODEL YS = XO ALCOHOL / NOINT;
LSMEANS ALCOHOL / STDERR;
ESTIMATE 'DIFFERENCE' ALCOHOL 1 -1;

(C) DATA SPLIT; SET WHOLE;
Y=Y1; DRUG='A'; OUTPUT;
Y=Y2; DRUG='B'; OUTPUT;
Y=Y3; DRUG='C'; OUTPUT;
DROP Y1-Y3 YS;

Rearranging the data to enable analysis
of the subplot factor.

(D) PROC PRINT N;

PROC SORT; BY ALCOHOL SUBJECT;

PROC SORT must be used on the CLASS variables
used in the ABSORB statement below. ABSORBing
the whole plot factors reduces the storage
requirements and hence the time and cost of
the analysis.

(E) PROC GLM;
ABSORB ALCOHOL SUBJECT;
CLASS DRUG ALCOHOL;
MODEL Y = DRUG ALCOHOL*DRUG; Effects model for the subplot factor and main effects contrasts.
ESTIMATE 'MAIN EFFECT: AB VS C' DRUG -1 -1 2 / DIVISOR=2;
ESTIMATE 'MAIN EFFECT: A VS B ' DRUG 1 -1 0;

(F) PROC GLM;
ABSORB ALCOHOL SUBJECT;

```

CLASSES ALCOHOL DRUG;
MODEL Y = ALCOHOL*DRUG / NOINT SS1 SS2;
ESTIMATE 'AB VS CONTROL @ NO'
        ALCOHOL*DRUG -1 -1 2 0 0 0 / DIVISOR=2 E;
ESTIMATE 'A VS B @ NO ALCOHOL'
        ALCOHOL*DRUG 1 -1 0 0 0 0;
ESTIMATE 'AB VS CONTROL @ YES'
        ALCOHOL*DRUG 0 0 0 -1 -1 2 / DIVISOR=2;
ESTIMATE 'A VS B @ YES ALCOHOL'
        ALCOHOL*DRUG 0 0 0 1 -1 0;

```

General means model and simple effects contrasts.

```

③ PROC GLM; CLASSES SUBJECT ALCOHOL DRUG;
MODEL Y = XO ALCOHOL SUBJECT(ALCOHOL)
        DRUG ALCOHOL*DRUG / SS1 SS2 NOINT;
TEST H=ALCOHOL E=SUBJECT(ALCOHOL) / HTYPE=1 ETYPE=1;
CONTRAST 'ALCOHOL DIFFERENCE' ALCOHOL 1 -1 /
        E=SUBJECT(ALCOHOL) ETYPE=1;
ESTIMATE 'AB VS CONTROL @ NO' DRUG -1 -1 2
        ALCOHOL*DRUG -1 -1 2 0 0 0 / DIVISOR=2 E;
ESTIMATE 'A VS B @ NO ALCOHOL' DRUG 1 -1 0
        ALCOHOL*DRUG 1 -1 0 0 0 0;
ESTIMATE 'AB VS CONTROL @ YES' DRUG -1 -1 2
        ALCOHOL*DRUG 0 0 0 -1 -1 2 / DIVISOR=2;
ESTIMATE 'A VS B @ YES ALCOHOL' DRUG 1 -1 0
        ALCOHOL*DRUG 0 0 0 1 -1 0;
OUTPUT OUT=NEW2 P=P R=R;

```

This analysis performs both the whole-plot and sub-plot analyses all at one time. The simple effects contrasts are more difficult to specify and the computing costs are much steeper.

```

④ PROC PLOT; PLOT R*P='*' / VREF=0;

```

Analysis of residuals

```

⑤ PROC ANOVA; CLASSES SUBJECT ALCOHOL DRUG;
MODEL Y = ALCOHOL SUBJECT(ALCOHOL) DRUG ALCOHOL*DRUG;
TEST H=ALCOHOL E=SUBJECT(ALCOHOL);
MEANS ALCOHOL SUBJECT(ALCOHOL) DRUG ALCOHOL*DRUG / DEONLY;

```

PROC ANOVA may be used since this experiment is balanced. The ESTIMATE option is not available, but the cell means and SS for the ANOVA table are computed and F-tests made.

-----ALCDHOL=NO-----

OBS	Y1	Y2	Y3	SUBJECT	YS	X0
1	2.83	2.55	2.63	7	4.62458	1
2	2.93	2.42	2.73	8	4.66499	1
3	3.58	3.99	3.38	9	6.32199	1
4	2.98	3.07	2.78	10	5.09800	1
5	2.32	2.15	2.12	11	3.80474	1
6	2.73	3.23	2.53	12	4.90170	1

N=6

(A) See p. 283 of Allen and Cady for a discussion of the assumed model.

-----ALCOHOL=YES-----

OBS	Y1	Y2	Y3	SUBJECT	YS	X0
7	3.56	4.04	3.26	1	6.27002	1
8	3.79	3.88	3.49	2	6.44323	1
9	4.09	5.32	3.79	3	7.62102	1
10	3.10	4.38	2.80	4	5.93516	1
11	3.33	3.63	3.03	5	5.76773	1
12	3.35	3.63	3.05	6	5.79082	1

N=6

GENERAL LINEAR MODELS PROCEDURE

ⓓ Analysis of sums to test for whole-plot (alcohol) differences.

CLASS LEVEL INFORMATION

CLASS LEVELS VALUES
ALCOHOL 2 NO YES

NUMBER OF OBSERVATIONS IN DATA SET = 12

DEPENDENT VARIABLE: YS

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE
MODEL	2	382.70960556	191.35480278	328.00
ERROR	10	5.83396111	0.58339611	PR > F
UNCORRECTED TOTAL	12	388.54356667		0.0001

2 = SS due to SUBJECT(ALCOHOL).

Compare SS found on this page with those in, say, analysis ⓕ.

R-SQUARE	C.V.	STD DEV	YS MEAN
0.984985	13.6304	0.76380371	5.60366549

SOURCE	DF	TYPE I SS	F VALUE	PR > F
XO	1	376.81280278	645.90	0.0001
ALCOHOL	1	5.89680278	10.11	0.0098

SOURCE	DF	TYPE IV SS	F VALUE	PR > F
XO	0	0.00000000	.	.
ALCOHOL	1	5.89680278	10.11	0.0098

PARAMETER	ESTIMATE	T FOR HO: PARAMETER=0	PR > T	STD ERROR OF ESTIMATE
DIFFERENCE	-1.40199890	-3.18	0.0098	0.44098228

LEAST SQUARES MEANS

ALCOHOL	YS LSMEAN	STD ERR LSMEAN	PROB > T HO:LSMEAN=0
NO	4.90266604	0.31182156	0.0001
YES	6.30466494	0.31182156	0.0001

ⓓ	OBS	SUBJECT	ALCOHOL	XO	Y	DRUG
	1	7	NO	1	2.83	A
	2	7	NO	1	2.55	B
	3	7	NO	1	2.63	C
	4	8	NO	1	2.93	A
	5	8	NO	1	2.42	B
	6	8	NO	1	2.73	C
	7	9	NO	1	3.58	A
	8	9	NO	1	3.99	B
	9	9	NO	1	3.38	C
	10	10	NO	1	2.98	A
	11	10	NO	1	3.07	B
	12	10	NO	1	2.78	C
	13	11	NO	1	2.32	A
	14	11	NO	1	2.15	B
	15	11	NO	1	2.12	C
	16	12	NO	1	2.73	A
	17	12	NO	1	3.23	B
	18	12	NO	1	2.53	C
	19	1	YES	1	3.56	A
	20	1	YES	1	4.04	B
	21	1	YES	1	3.26	C
	22	2	YES	1	3.79	A
	23	2	YES	1	3.88	B
	24	2	YES	1	3.49	C
	25	3	YES	1	4.09	A
	26	3	YES	1	5.32	B
	27	3	YES	1	3.79	C
	28	4	YES	1	3.10	A
	29	4	YES	1	4.38	B
	30	4	YES	1	2.80	C
	31	5	YES	1	3.33	A
	32	5	YES	1	3.63	B
	33	5	YES	1	3.03	C
	34	6	YES	1	3.35	A
	35	6	YES	1	3.63	B
	36	6	YES	1	3.05	C

N=36

Ⓔ

GENERAL LINEAR MODELS PROCEDURE

CLASS LEVEL INFORMATION

CLASS	LEVELS	VALUES
DRUG	3	A B C
ALCOHOL	2	NO YES

NUMBER OF OBSERVATIONS IN DATA SET = 36

GENERAL LINEAR MODELS PROCEDURE

DEPENDENT VARIABLE: Y

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE
MODEL	15	14.47667500	0.96511167	13.66
ERROR	20	1.41262222	0.07063111	PR > F
CORRECTED TOTAL	35	15.88929722		0.0001

R-SQUARE	C.V.	STD DEV	Y MEAN
0.911096	8.2146	0.26576514	3.23527778

SOURCE	DF	TYPE I SS	F VALUE	PR > F
ALCOHOL	1	5.89680278	83.49	0.0001
SUBJECT(ALCOHOL)	10	5.83396111	8.28	0.0001
DRUG	2	1.87722222	13.29	0.0002
DRUG*ALCOHOL	2	0.86868889	6.15	0.0083

SOURCE	DF	TYPE IV SS	F VALUE	PR > F
DRUG	2	1.87722222	13.29	0.0002
DRUG*ALCOHOL	2	0.86868889	6.15	0.0083

Note that ALCOHOL and SUBJECT(ALCOHOL) are not included in the Type IV SS.

PARAMETER	ESTIMATE	T FOR H0: PARAMETER=0	PR > T	STD ERROR OF ESTIMATE
MAIN EFFECT: AB VS C	-0.40416667	-4.30	0.0003	0.09396217
MAIN EFFECT: A VS B	-0.30833333	-2.84	0.0101	0.10849817

These contrasts are of little interest since the DRUG*ALCOHOL interaction is clearly important.

Ⓟ

GENERAL LINEAR MODELS PROCEDURE

General means model with whole-plot factor ABSORBed.

DEPENDENT VARIABLE: Y

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE
MODEL	15	14.47667500	0.96511167	13.66
ERROR	20	1.41262222	0.07063111	PR > F
CORRECTED TOTAL	35	15.88929722		0.0001

R-SQUARE	C.V.	STD DEV	Y MEAN
0.911096	8.2146	0.26576514	3.23527778

SOURCE	DF	TYPE I SS	F VALUE	PR > F
ALCOHOL	1	5.89680278	83.49	0.0001
SUBJECT(ALCOHOL)	10	5.83396111	8.26	0.0001
ALCOHOL*DRUG	4	2.74591111	9.72	0.0002

SOURCE	DF	TYPE II SS	F VALUE	PR > F
ALCOHOL*DRUG	4	2.74591111	9.72	0.0002

PARAMETER	ESTIMATE	T FOR H0: PARAMETER=0	PR > T	STD ERROR OF ESTIMATE
AB VS CONTROL @ NO	-0.20333333	-1.53	0.1416	0.13288257
A VS B @ NO ALCOHOL	-0.00666667	-0.04	0.9658	0.15343958
AB VS CONTROL @ YES	-0.60500000	-4.55	0.0002	0.13288257
A VS B @ YES ALCOHOL	-0.61000000	-3.98	0.0007	0.15343958

This is a set of simple effects contrasts
(see Table 22.3, p. 281 - Allen and Cady).

Ⓒ Combined whole-plot and split-plot analysis with simple effects contrasts.
GENERAL LINEAR MODELS PROCEDURE

DEPENDENT VARIABLE: Y

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE
MODEL	16	391.28947778	24.45559236	346.24
ERROR	20	1.41262222	0.07063111	PR > F
UNCORRECTED TOTAL	36	392.70210000		0.0001

R-SQUARE	C.V.	STD DEV	Y MEAN
0.996403	8.2146	0.26576514	3.23527778

SOURCE	DF	TYPE I SS	F VALUE	PR > F
X0	1	376.81280278	5334.94	0.0001
ALCOHOL	1	5.89680278	83.49	0.0091
SUBJECT(ALCOHOL)	10	5.83396111	8.26	0.0001
DRUG	2	1.87722222	13.29	0.0002
ALCOHOL*DRUG	2	0.86868889	6.15	0.0083

SOURCE	DF	TYPE II SS	F VALUE	PR > F
X0	0	0.00000000	.	.
ALCOHOL	1	5.89680278	83.49	0.0001
SUBJECT(ALCOHOL)	10	5.83396111	8.26	0.0001
DRUG	2	1.87722222	13.29	0.0002
ALCOHOL*DRUG	2	0.86868889	6.15	0.0083

TESTS OF HYPOTHESES USING THE TYPE I MS FOR SUBJECT(ALCOHOL) AS AN ERROR TERM

SOURCE	DF	TYPE I SS	F VALUE	PR > F
ALCOHOL	1	5.89680278	10.11	0.0098

TESTS OF HYPOTHESES USING THE TYPE I MS FOR SUBJECT(ALCOHOL) AS AN ERROR TERM

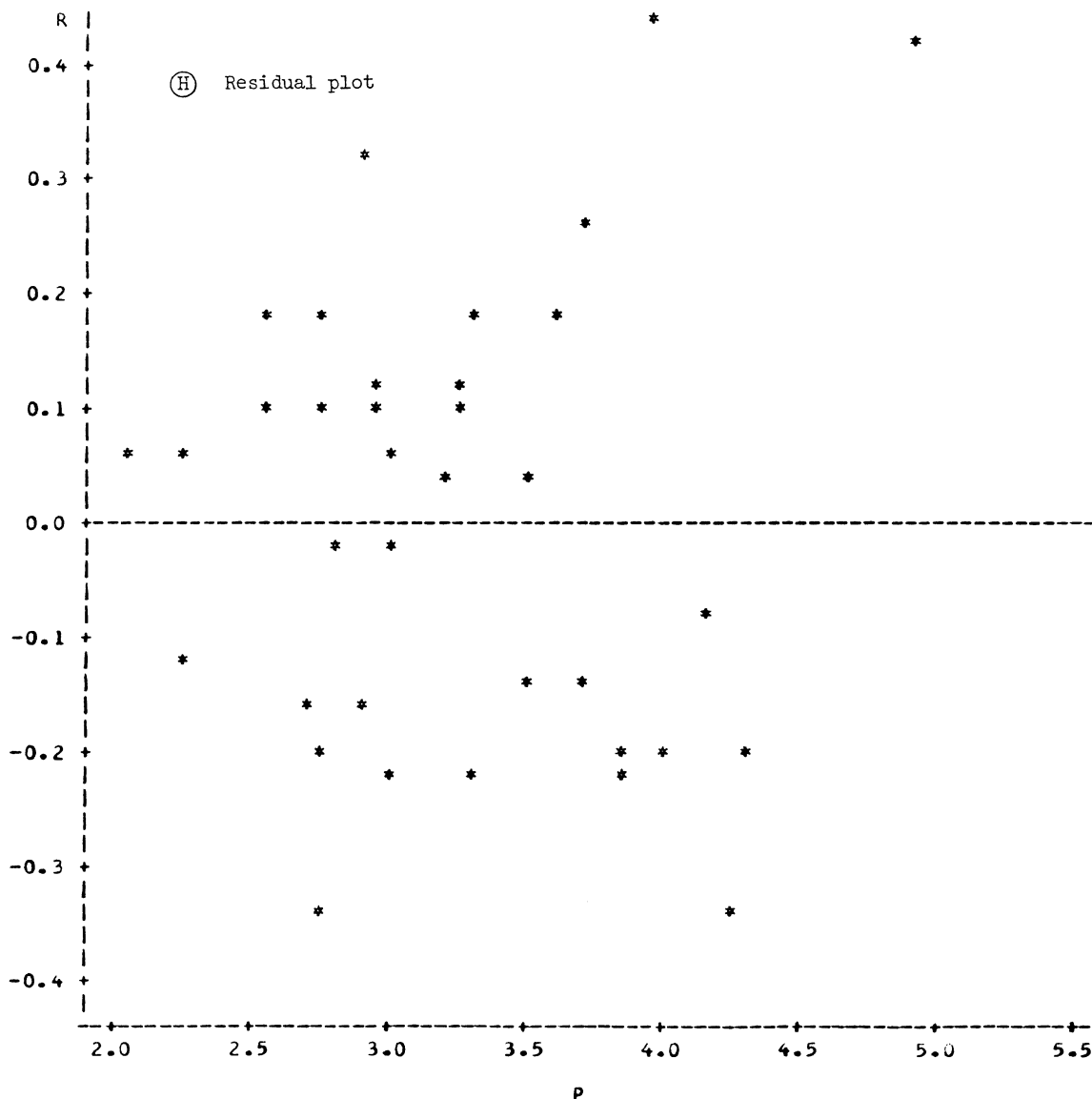
CONTRAST	DF	SS	F VALUE	PR > F
ALCOHOL DIFFERENCE	1	5.89680278	10.11	0.0098

Compare with the analysis in part Ⓓ.

PARAMETER	ESTIMATE	T FOR H0: PARAMETER=0	PR > T	STD ERROR OF ESTIMATE
AB VS CONTROL @ NO	-0.20333333	-1.53	0.1416	0.13288257
A VS B @ NO ALCOHOL	-0.00666667	-0.04	0.9658	0.15343958
AB VS CONTROL @ YES	-0.60500000	-4.55	0.0002	0.13288257
A VS B @ YES ALCOHOL	-0.61000000	-3.98	0.0007	0.15343958

Simple effects as in Ⓕ.

PLOT OF R*P SYMBOL USED IS *



① ANALYSIS OF VARIANCE PROCEDURE

MEANS		
ALCOHOL	N	Y
NO	18	2.83055556
YES	18	3.64000000

SUBJECT	ALCOHOL	N	Y
7	NO	3	2.67000000
8	NO	3	2.69333333
9	NO	3	3.65000000
10	NO	3	2.94333333
11	NO	3	2.19666667
12	NO	3	2.83000000
1	YES	3	3.62000000
2	YES	3	3.72000000
3	YES	3	4.40000000
4	YES	3	3.42666667
5	YES	3	3.33000000
6	YES	3	3.34333333

DRUG	N	Y
A	12	3.21583333
B	12	3.52416667
C	12	2.96583333

ALCOHOL	DRUG	N	Y
NO	A	6	2.89500000
NO	B	6	2.90166667
NO	C	6	2.69500000
YES	A	6	3.53666667
YES	B	6	4.14666667
YES	C	6	3.23666667

Note: Only the MEANS output is included from the PROC ANOVA output. Although all the appropriate SS have been obtained via PROC GLM, PROC ANOVA would be less costly.

UREA SYNTHESIS DATA - Analysis of a repeated measures experiment
 (Brogan & Kutner, 1980, The American Statistician 34:229-232)

```

DATA SHUNT;
INPUT PRE POST @@;
SUBJECT = _N_;
IF _N_ LE 8 THEN GROUP='NEW';
ELSE GROUP='OLD';
X0=1;
PREPOST = POST - PRE;
TMT      = POST + PRE;
TMT_X_PP= PREPOST;
IF GROUP='OLD' THEN PREPOST=-PREPOST;
CARDS;
  
```

These statements calculate the sums and differences required for the appropriate t-tests. See printout in part (A) and (B).

```
PROC SORT; BY GROUP;
```

(A) PROC PRINT; BY GROUP;

(B) PROC TTEST; CLASS GROUP;
VAR TMT PREPOST TMT_X_PP;

This gives the three appropriate t-tests for treatment effects, time effects and interaction.

(C) PROC GLM; CLASS GROUP;
MODEL TMT PREPOST TMT_X_PP = X0 GROUP / NOINT;
LSMEANS GROUP / STDERR;
ESTIMATE 'NEW VS STD' GROUP 1 -1;

Same analysis as in (B) except the analyses are performed using 1-way ANOVA. Compare with (F). Note that more than one dependent variable may appear to the left of the equal sign in the MODEL statement.

(D) DATA REPEAT; SET SHUNT;
Y= PRE; TIME=1; OUTPUT;
Y=POST; TIME=2; OUTPUT;
DROP PRE POST TMT PREPOST TMT_X_PP;

Rearrange data for a "split-plot" type analysis.

```
PROC PRINT; VAR SUBJECT X0 GROUP TIME Y;
```

(E) PROC GLM; CLASS GROUP TIME SUBJECT;
MODEL Y = X0 GROUP SUBJECT(GROUP)
TIME TIME*GROUP / NJINT SS1 SS2 SS3 SS4 P;
MEANS GROUP SUBJECT(GROUP) TIME TIME*GROUP / DEONLY;
TEST H=GROUP E=SUBJECT(GROUP) / HTYPE=1 ETYPE=1;
OUTPUT OUT=PLOT RESIDUAL=RES PREDICTED=P;

Combined ANOVA Table with two error terms.

(F) PROC PLOT; PLOT RES*P='*' / VREF=0; Residual plot.

(A)

GROUP=NEW

OBS	PRE	POST	SUBJECT	X0	PREPOST	TMT	TMT_X_PP
1	51	48	1	1	-3	99	-3
2	35	55	2	1	20	90	20
3	66	60	3	1	-6	126	-6
4	40	35	4	1	-5	75	-5
5	39	36	5	1	-3	75	-3
6	46	43	6	1	-3	89	-3
7	52	46	7	1	-6	98	-6
8	42	54	8	1	12	96	12

GROUP=OLD

OBS	PRE	POST	SUBJECT	X0	PREPOST	TMT	TMT_X_PP
9	34	16	9	1	18	50	-18
10	40	36	10	1	4	76	-4
11	34	16	11	1	18	50	-18
12	36	18	12	1	18	54	-18
13	38	32	13	1	6	70	-6
14	32	14	14	1	18	46	-18
15	44	20	15	1	24	64	-24
16	50	43	16	1	7	93	-7
17	60	45	17	1	15	105	-15
18	63	67	18	1	-4	130	4
19	50	36	19	1	14	86	-14
20	42	34	20	1	8	76	-8
21	43	32	21	1	11	75	-11

PRE = preoperative response

POST = postoperative response

		Pre	Post
Treatment	New (n=8)	μ_{11}	μ_{12}
	Old (n=13)	μ_{21}	μ_{22}

Note that: $TMT_X_PP = y_{ij2} - y_{ij1} = d_{ij}$ (= within subject differences)

$$PREPOST = \begin{cases} d_{ij} & \text{if group = NEW} \\ -d_{ij} & \text{if group = OLD} \end{cases}$$

$TMT = y_{ij1} + y_{ij2} = z_{ij}$ (= within subject sums)

The assumed model may be written as:

$$y_{ijk} = \mu_{ik} + \delta_{j(i)} + \epsilon_{ijk}$$

$$\mu_{ik} = \mu + \alpha_i + \tau_k + (\alpha\tau)_{ik}$$

$$i = 1, 2; \quad j = 1, \dots, n_i; \quad k = 1, 2$$

where:

μ_{ik} = cell mean of ith treatment, kth time combination

μ = overall mean

α_i = ith treatment effect

τ_k = kth time effect

$(\alpha\tau)_{ik}$ = treatment by time interaction

$\delta_{j(i)}$ = effect of subject j nested within ith treatment group

ϵ_{ijk} = random error

$$\epsilon_{ijk} \sim N(0, \sigma_\epsilon^2)$$

$$\delta_{j(i)} \sim N(0, \sigma_\delta^2)$$

$$y_{ijk} \sim N(\mu_{ik}, \sigma_\delta^2 + \sigma_\epsilon^2)$$

(B) TTEST PROCEDURE

VARIABLE: TMT t-test uses $\bar{z}_1 - \bar{z}_2$ "difference of sums"

GROUP	N	MEAN	STD DEV	STD ERROR	MINIMUM	MAXIMUM
NEW	8	93.50000000	16.16875293	5.71651742	75.00000000	126.00000000
OLD	13	75.00000000	24.23839929	6.72252242	46.00000000	130.00000000

VARIANCES	T	DF	PROB > T
UNEQUAL	2.0964	18.8	0.0498
EQUAL	1.9044	19.0	0.0721

FOR HO: VARIANCES ARE EQUAL, F* = 2.25 WITH 12 AND 7 DF PROB > F* = 0.2891

VARIABLE: PREPOST t-test uses $\bar{d}_1 + \bar{d}_2$ "sum of differences"

GROUP	N	MEAN	STD DEV	STD ERROR	MINIMUM	MAXIMUM
NEW	8	0.75000000	9.73579551	3.44212351	-6.00000000	20.00000000
OLD	13	12.07692308	7.63174880	2.11666628	-4.00000000	24.00000000

VARIANCES	T	DF	PROB > T
UNEQUAL	-2.8031	12.3	0.0157
EQUAL	-2.9767	19.0	0.0078

FOR HO: VARIANCES ARE EQUAL, F* = 1.63 WITH 7 AND 12 DF PROB > F* = 0.4375

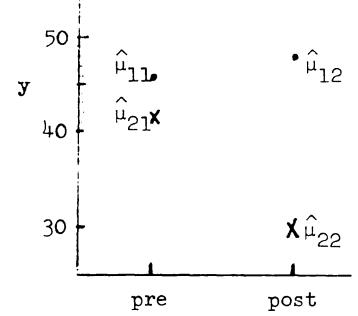
VARIABLE: TMT_X_PP t-test uses $\bar{d}_1 - \bar{d}_2$ "difference of differences"

GROUP	N	MEAN	STD DEV	STD ERROR	MINIMUM	MAXIMUM
NEW	8	0.75000000	9.73579551	3.44212351	-6.00000000	20.00000000
OLD	13	-12.07692308	7.63174880	2.11666628	-24.00000000	4.00000000

VARIANCES	T	DF	PROB > T
UNEQUAL	3.1743	12.3	0.0078
EQUAL	3.3709	19.0	0.0032

FOR HO: VARIANCES ARE EQUAL, F* = 1.63 WITH 7 AND 12 DF PROB > F* = 0.4375

Since the interaction is clearly important we need to consider simple effects



These are not helpful due to the significant interaction.

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GENERAL LINEAR MODELS PROCEDURE

CLASS LEVEL INFORMATION

CLASS	LEVELS	VALUES
GROUP	2	NEW OLD

An alternative approach to finding the same information as in parts (A) or (F).

NUMBER OF OBSERVATIONS IN DATA SET = 21

DEPENDENT VARIABLE: TMT

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE
MODEL	2	143063.00000000	71531.50000000	153.05
ERROR	19	8880.00000000	467.36842105	PR > F
UNCORRECTED TOTAL	21	151943.00000000		0.0001

R-SQUARE	C.V.	STD DEV	TMT MEAN
0.941557	26.3490	21.61870535	82.04761905

SOURCE	DF	TYPE I SS	F VALUE	PR > F
XO	1	141368.04761905	302.48	0.0001
GROUP	1	1694.95238095	3.63	0.0721

SOURCE	DF	TYPE IV SS	F VALUE	PR > F
XO	0	0.00000000	.	.
GROUP	1	1694.95238095	3.63	0.0721

PARAMETER	ESTIMATE	T FOR H0: PARAMETER=0	PR > T	STD ERROR OF ESTIMATE
NEW VS STD	18.50000000	1.90	0.0721	9.71454938

GENERAL LINEAR MODELS PROCEDURE

©

DEPENDENT VARIABLE: PREPOST

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE
MODEL	2	1900.57692308	950.28846154	13.25
ERROR	19	1362.42307692	71.70647773	PR > F
UNCORRECTED TOTAL	21	3263.00000000		0.0002

R-SQUARE	C.V.	STD DEV	PREPOST MEAN
0.582463	109.0965	8.46796775	7.76190476

SOURCE	DF	TYPE I SS	F VALUE	PR > F
XO	1	1265.19047619	17.64	0.0005
GROUP	1	635.38644689	8.86	0.0078

SOURCE	DF	TYPE IV SS	F VALUE	PR > F
XO	0	0.00000000	.	.
GROUP	1	635.38644689	8.86	0.0078

PARAMETER	ESTIMATE	T FOR HO: PARAMETER=0	PR > T	STD ERROR OF ESTIMATE
NEW VS STD	-11.32692308	-2.98	0.0078	3.80515343

DEPENDENT VARIABLE: TMT_X_PP

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE
MODEL	2	1900.57692308	950.28846154	13.25
ERROR	19	1362.42307692	71.70647773	PR > F
UNCORRECTED TOTAL	21	3263.00000000		0.0002

R-SQUARE	C.V.	STD DEV	TMT_X_PP MEAN
0.582463	117.7664	8.46796775	-7.19047619

SOURCE	DF	TYPE I SS	F VALUE	PR > F
XO	1	1085.76190476	15.14	0.0010
GROUP	1	814.81501832	11.36	0.0032

SOURCE	DF	TYPE IV SS	F VALUE	PR > F
XO	0	0.00000000	.	.
GROUP	1	814.81501832	11.36	0.0032

PARAMETER	ESTIMATE	T FOR HO: PARAMETER=0	PR > T	STD ERROR OF ESTIMATE
NEW VS STD	12.82672308	3.37	0.0032	3.80515343

LEAST SQUARES MEANS

GROUP	TMT LSMEAN	STD ERR LSMEAN	PROB > T HO:LSMEAN=0
NEW	93.5000000	7.6433666	0.0001
OLD	75.0000000	5.9959501	0.0001

GROUP	PREPOST LSMEAN	STD ERR LSMEAN	PROB > T HO:LSMEAN=0
NEW	0.7500000	2.9938787	0.8049
OLD	12.0769231	2.3485917	0.0001

GROUP	TMT_X_PP LSMEAN	STD ERR LSMEAN	PROB > T HO:LSMEAN=0
NEW	0.7500000	2.9938787	0.8049
OLD	-12.0769231	2.3485917	0.0001

Ⓓ

OBS	SUBJECT	X0	GROUP	TIME	Y
1	1	1	NEW	1	51
2	1	1	NEW	2	48
3	2	1	NEW	1	35
4	2	1	NEW	2	55
5	3	1	NEW	1	66
6	3	1	NEW	2	60
7	4	1	NEW	1	40
8	4	1	NEW	2	35
9	5	1	NEW	1	39
10	5	1	NEW	2	36
11	6	1	NEW	1	46
12	6	1	NEW	2	43
13	7	1	NEW	1	52
14	7	1	NEW	2	46
15	8	1	NEW	1	42
16	8	1	NEW	2	54
17	9	1	OLD	1	34
18	9	1	OLD	2	16
19	10	1	OLD	1	40
20	10	1	OLD	2	36
21	11	1	OLD	1	34
22	11	1	OLD	2	16
23	12	1	OLD	1	36
24	12	1	OLD	2	18
25	13	1	OLD	1	38
26	13	1	OLD	2	32
27	14	1	OLD	1	32
28	14	1	OLD	2	14
29	15	1	OLD	1	44
30	15	1	OLD	2	20
31	16	1	OLD	1	50
32	16	1	OLD	2	43
33	17	1	OLD	1	60
34	17	1	OLD	2	45
35	18	1	OLD	1	63
36	18	1	OLD	2	67
37	19	1	OLD	1	50
38	19	1	OLD	2	36
39	20	1	OLD	1	42
40	20	1	OLD	2	34
41	21	1	OLD	1	43
42	21	1	OLD	2	32

Ⓔ

GENERAL LINEAR MODELS PROCEDURE

CLASS LEVEL INFORMATION

CLASS	LEVELS	VALUES
GROUP	2	NEW OLD
TIME	2	1 2
SUBJECT	21	1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21

Ⓔ

GENERAL LINEAR MODELS PROCEDURE

DEPENDENT VARIABLE: Y

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE
MODEL	23	76921.78846154	3344.42558528	93.28
ERROR	19	681.21153846	35.85323887	PR > F
UNCORRECTED TOTAL	42	77603.00000000		0.0001

R-SQUARE	C.V.	STD DEV	Y MEAN
0.991222	14.5958	5.98775742	41.02380952

SOURCE	DF	TYPE I SS	F VALUE	PR > F
XO	1	70684.02380952	1971.48	0.0001
GROUP	1	847.47619048	23.64	0.0001
SUBJECT(GROUP)	19	4440.00000000	6.52	0.0001
TIME	1	542.88095238	15.14	0.0010
GROUP*TIME	1	407.40750916	11.36	0.0032

SOURCE	DF	TYPE II SS	F VALUE	PR > F
XO	0	0.00000000	.	.
GROUP	1	847.47619048	23.64	0.0001
SUBJECT(GROUP)	19	4440.00000000	6.52	0.0001
TIME	1	542.88095238	15.14	0.0010
GROUP*TIME	1	407.40750916	11.36	0.0032

SOURCE	DF	TYPE III SS	F VALUE	PR > F
XO	0	0.00000000	.	.
GROUP	1	847.47619048	23.64	0.0001
SUBJECT(GROUP)	19	4440.00000000	6.52	0.0001
TIME	1	317.69322344	8.86	0.0078
GROUP*TIME	1	407.40750916	11.36	0.0032

SOURCE	DF	TYPE IV SS	F VALUE	PR > F
XO	0	0.00000000	.	.
GROUP	1	847.47619048	23.64	0.0001
SUBJECT(GROUP)	19	4440.00000000	6.52	0.0001
TIME	1	317.69322344	8.86	0.0078
GROUP*TIME	1	407.40750916	11.36	0.0032

TESTS OF HYPOTHESES USING THE TYPE I MS FOR SUBJECT(GROUP) AS AN ERROR TERM

SOURCE	DF	TYPE I SS	F VALUE	PR > F
GROUP	1	847.47619048	3.63	0.0721

$$t = \frac{\left(\frac{8}{21}\right)(-0.75) + \left(\frac{13}{21}\right)(12.0769)}{\left\{2(35.85324) \left[\left(\frac{8}{21}\right)^2 \left(\frac{1}{8}\right) + \left(\frac{13}{21}\right)^2 \left(\frac{1}{13}\right) \right] \right\}^{\frac{1}{2}}}$$

$$= 3.89 \Rightarrow F = (3.89)^2 = 15.14$$

The Type I and II SS for TIME test equality of the weighted means -- weighted according to sample size. This is probably not desired and the Type III or IV should be used as they test equality of unweighted means.

Compare with the results of Ⓑ and Ⓒ.

Ⓔ

GENERAL LINEAR MODELS PROCEDURE

MEANS

GROUP	N	Y
NEW	16	46.7500000
OLD	26	37.5000000

SUBJECT	GROUP	N	Y
1	NEW	2	49.5000000
2	NEW	2	45.0000000
3	NEW	2	63.0000000
4	NEW	2	37.5000000
5	NEW	2	37.5000000
6	NEW	2	44.5000000
7	NEW	2	49.0000000
8	NEW	2	48.0000000
9	OLD	2	25.0000000
10	OLD	2	38.0000000
11	OLD	2	25.0000000
12	OLD	2	27.0000000
13	OLD	2	35.0000000
14	OLD	2	23.0000000
15	OLD	2	32.0000000
16	OLD	2	46.5000000
17	OLD	2	52.5000000
18	OLD	2	65.0000000
19	OLD	2	43.0000000
20	OLD	2	38.0000000
21	OLD	2	37.5000000

TIME	N	Y
1	21	44.6190476
2	21	37.4285714

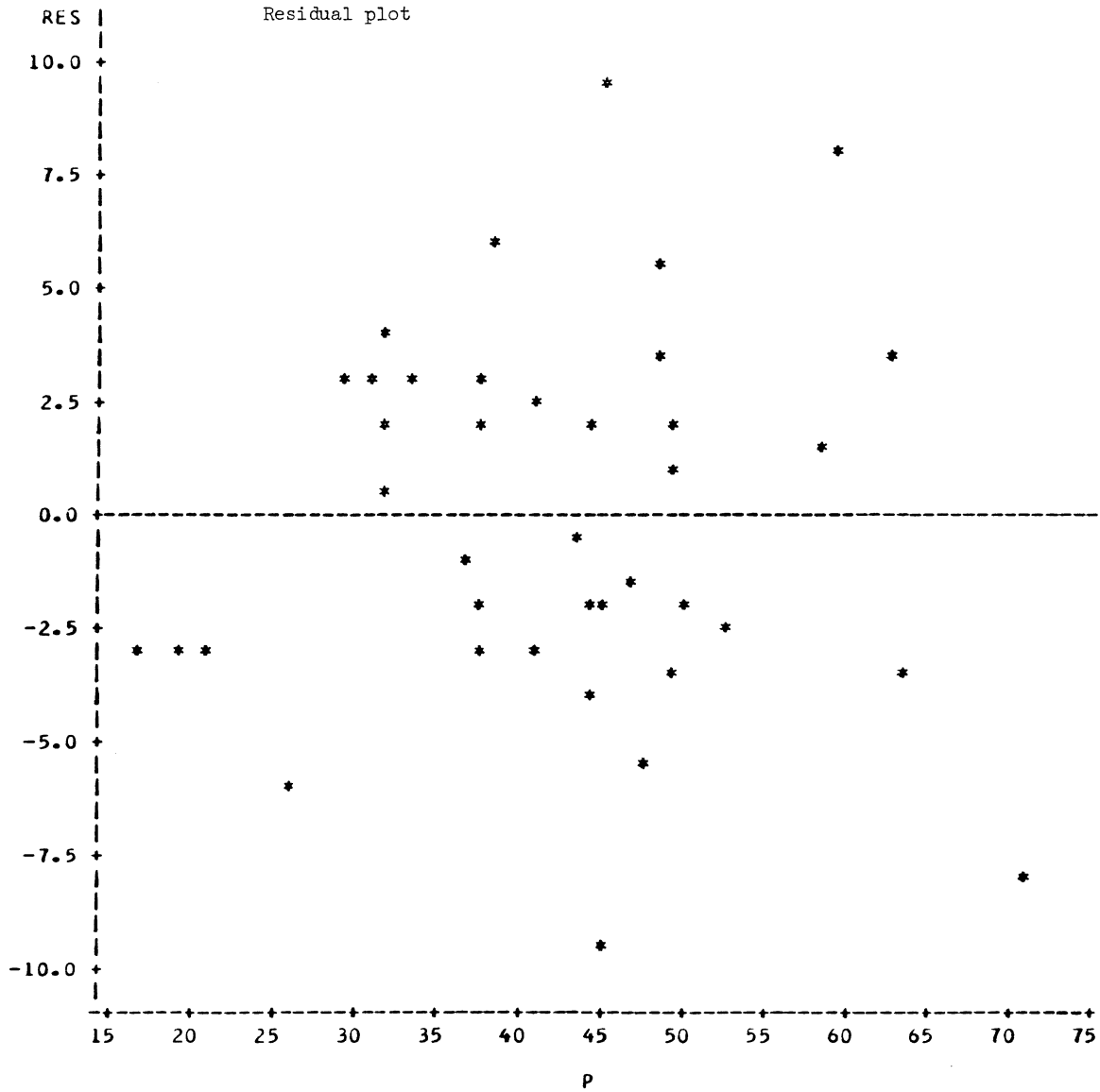
GROUP	TIME	N	Y
NEW	1	8	46.3750000
NEW	2	8	47.1250000
OLD	1	13	43.5384615
OLD	2	13	31.4615385

} These means are plotted in part Ⓐ.

Ⓟ

PLGT OF RES*P SYMBOL USED IS *

Residual plot



NOTE: 3 OBS HIDDEN

HEMOGLOBIN DATA -- Analysis of a 2-Period Cross-over Design
 (Grizzle, 1965, Biometrics 21: 467-480)

```

DATA HEMO;
INPUT Y1 Y2 @@;
SUBJECT = _N_;
IF _N_ LE 6 THEN GROUP='A1_B2';
ELSE GROUP='B1_A2';
TMT = Y2-Y1;
RESID= Y2+Y1;
TREND= TMT;
IF GROUP='B1_A2' THEN TREND=-TMT;
XJ = 1;
CARDS;

```

These statements calculate the sums and differences required for the appropriate t-tests. See printout for (A) and (B).

```
PROC SORT; BY GROUP;
```

(A) PROC PRINT N; BY GROUP;

(B) PROC TTEST; CLASS GROUP;
VAR TMT RESID TREND Y1;

PROC TTEST gives the appropriate four t-tests for treatment effects, carry-over effects, time trend and treatment differences within first period only.

(C) PROC GLM; CLASS GROUP;
MODEL TMT RESID TREND Y1 = XO GROUP / NOINT;
LSMEANS GROUP / STDERR;
ESTIMATE 'CONTRAST' GROUP 1 -1;

This performs the same analyses as those found in part (B). Compare also with the results in part (E).

(D) DATA AOV; SET HEMO;
Y=Y1; PERIOD=1; IF _N_ LE 6 THEN TRT='A'; ELSE TRT='B'; OUTPUT;
Y=Y2; PERIOD=2; IF _N_ LE 6 THEN TRT='B'; ELSE TRT='A'; OUTPUT;
DROP Y1 Y2 TMT RESID TREND;

Rearranging the data so that the "usual" ANOVA table may be constructed.

```
PROC PRINT; VAR SUBJECT XO GROUP PERIOD TRT Y;
```

(E) PROC GLM; CLASS TRT PERIOD SUBJECT GROUP;
MODEL Y = XO GROUP SUBJECT(GROUP)
PERIOD TRT / NOINT P SS1 SS2 SS3 SS4;
MEANS GROUP SUBJECT(GROUP) PERIOD TRT / DEONLY;
TEST H=GROUP E=SUBJECT(GROUP) / HTYPE=1 ETYPE=1;
ESTIMATE 'TRT' TRT 1 -1;
ESTIMATE 'PERIOD' PERIOD 1 -1;

Combined ANOVA table with two error terms.

(F) PROC PLOT; PLOT RES*P='*' / VREF=0; Residual plot.

(A)

GROUP=A1_B2

UBS	Y1	Y2	SUBJECT	TMT	RESID	TREND	X0
1	0.2	1.0	1	0.8	1.2	0.8	1
2	0.0	-0.7	2	-0.7	-0.7	-0.7	1
3	-0.8	0.2	3	1.0	-0.6	1.0	1
4	0.6	1.1	4	0.5	1.7	0.5	1
5	0.3	0.4	5	0.1	0.7	0.1	1
6	1.5	1.2	6	-0.3	2.7	-0.3	1

N=6

GROUP=B1_A2

UBS	Y1	Y2	SUBJECT	TMT	RESID	TREND	X0
7	1.3	0.9	7	-0.4	2.2	0.4	1
8	-2.3	1.0	8	3.3	-1.3	-3.3	1
9	0.0	0.6	9	0.6	0.6	-0.6	1
10	-0.8	-0.3	10	0.5	-1.1	-0.5	1
11	-0.4	-1.0	11	-0.6	-1.4	0.6	1
12	-2.9	1.7	12	4.6	-1.2	-4.6	1
13	-1.9	-0.3	13	1.6	-2.2	-1.6	1
14	-2.9	0.9	14	3.8	-2.0	-3.8	1

N=8

Y1 are the responses during period 1
Y2 are the responses during period 2

Group (Sequence)

1(n=6) 2(n=8)

Period	1	2
	A μ_{11}	B μ_{21}
	B μ_{12}	A μ_{22}

TMT = $y_{ij2} - y_{ij1} = d_{ij}$ (= within subject differences)

RESID = $y_{ij2} + y_{ij1} = z_{ij}$ (= within subject sums)

TREND = $\begin{cases} d_{ij} & \text{if sequence is AB} \\ -d_{ij} & \text{if sequence is BA} \end{cases}$

The assumed model appears as:

$$y_{ijk} = \mu_{ik} + \xi_{ij} + \epsilon_{ijk}$$

where

$$\mu_{ik} = \mu + \pi_k + \phi_l + \lambda_l$$

$$j = 1, \dots, n_i; \quad i = 1, 2; \quad k = 1, 2; \quad l = 1, 2$$

and

μ = general mean,

ξ_{ij} = the effect of the j -th patient (subject) within the i -th sequence, which, for the sake of testing hypotheses, we must assume to be a normally distributed random variable with mean 0 and variance σ_{ξ}^2 ,

π_k = the effect of the k -th period,

ϕ_l = the direct effect of the l -th drug,

λ_l = the residual effect of the l -th drug, and

ϵ_{ijk} = the random fluctuation which is normally distributed with mean 0 and variance σ_{ϵ}^2 , and is independent of the ξ_{ij} .

The assumptions made about ξ_{ij} and ϵ_{ijk} imply that the variance of an observation is $\sigma_{\xi}^2 + \sigma_{\epsilon}^2$ and that two observations on an individual have covariance σ_{ξ}^2 . Observations made on different subjects are independent.

Note:

- (i) $(\bar{y}_{1.1} - \bar{y}_{1.2}) - (\bar{y}_{2.1} - \bar{y}_{2.2}) = 2(\phi_1 - \phi_2) + (\lambda_2 - \lambda_1) + (\bar{\epsilon}_{1.1} - \bar{\epsilon}_{1.2} - \bar{\epsilon}_{2.1} + \bar{\epsilon}_{2.2})$
- (ii) $(\bar{y}_{1.1} - \bar{y}_{1.2}) + (\bar{y}_{2.1} - \bar{y}_{2.2}) = 2(\pi_1 - \pi_2) - (\lambda_1 + \lambda_2) + (\bar{\epsilon}_{1.1} - \bar{\epsilon}_{1.2} + \bar{\epsilon}_{2.1} - \bar{\epsilon}_{2.2})$
- (iii) $(\bar{y}_{1.1} + \bar{y}_{1.2}) - (\bar{y}_{2.1} + \bar{y}_{2.2}) = (\lambda_2 - \lambda_1) + 2(\bar{\xi}_{1.} - \bar{\xi}_{2.}) + (\bar{\epsilon}_{1.1} + \bar{\epsilon}_{1.2} - \bar{\epsilon}_{2.1} - \bar{\epsilon}_{2.2})$
- (iv) $\bar{y}_{1.1} - \bar{y}_{2.1} = (\phi_1 - \phi_2) + (\bar{\xi}_{1.} - \bar{\xi}_{2.}) + (\bar{\epsilon}_{1.1} - \bar{\epsilon}_{2.1})$

ⓑ

TTEST PROCEDURE

VARIABLE: TMT t-test uses $\bar{d}_1 - \bar{d}_2$ "difference of differences"

GROUP	N	MEAN	STD DEV	STD ERROR	MINIMUM	MAXIMUM
A1_B2	6	0.23333333	0.65625198	0.26791375	-0.70000000	1.00000000
B1_A2	8	1.67500000	1.99051321	0.70375270	-0.60000000	4.60000000

VARIANCES T DF PROB > |T|

UNEQUAL	-1.9145	8.9	0.0882
EQUAL	-1.6915	12.0	0.1165

FOR H0: VARIANCES ARE EQUAL, F*= 9.20 WITH 7 AND 5 DF PROB > F*= 0.0205

VARIABLE: RESID t-test uses $\bar{z}_1 - \bar{z}_2$ "difference of sums"

GROUP	N	MEAN	STD DEV	STD ERROR	MINIMUM	MAXIMUM
A1_B2	6	0.83333333	1.32614730	0.54139737	-0.70000000	2.70000000
B1_A2	8	-0.80000000	1.47454593	0.52133071	-2.20000000	2.20000000

VARIANCES T DF PROB > |T|

UNEQUAL	2.1732	11.5	0.0515
EQUAL	2.1379	12.0	0.0538

There appear to be important carry-over effects. Thus, the above test for treatment effects is not free of residual effects and we must resort to the results of the first period only (variable Y1 below).

FOR H0: VARIANCES ARE EQUAL, F*= 1.24 WITH 7 AND 5 DF PROB > F*= 0.8437

VARIABLE: TREND t-test uses $\bar{d}_1 + \bar{d}_2$ "sum of differences"

GROUP	N	MEAN	STD DEV	STD ERROR	MINIMUM	MAXIMUM
A1_B2	6	0.23333333	0.65625198	0.26791375	-0.70000000	1.00000000
B1_A2	8	-1.67500000	1.99051321	0.70375270	-4.60000000	0.60000000

VARIANCES T DF PROB > |T|

UNEQUAL	2.5342	8.9	0.0323
EQUAL	2.2390	12.0	0.0449

FOR H0: VARIANCES ARE EQUAL, F*= 9.20 WITH 7 AND 5 DF PROB > F*= 0.0265

VARIABLE: Y1 t-test uses $\bar{y}_{1.1} - \bar{y}_{2.1}$ "independent two-sample t-test for the first period only"

GROUP	N	MEAN	STD DEV	STD ERROR	MINIMUM	MAXIMUM
A1_B2	6	0.30000000	0.75365775	0.30767949	-0.80000000	1.50000000
B1_A2	8	-1.23750000	1.50990775	0.53383301	-2.90000000	1.30000000

VARIANCES T DF PROB > |T|

UNEQUAL	2.4953	10.8	0.0302
EQUAL	2.2746	12.0	0.0421

There appear to be treatment effects present. Note: Had we chosen to ignore residual effects the conclusion would likely have been different.

FOR H0: VARIANCES ARE EQUAL, F*= 4.01 WITH 7 AND 5 DF PROB > F*= 0.1453

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GENERAL LINEAR MODELS PROCEDURE

DEPENDENT VARIABLE: TMT

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE
MODEL	2	22.77166667	11.38583333	4.57
ERROR	12	29.88833333	2.49069444	PR > F
UNCORRECTED TOTAL	14	52.66000000		0.0334

R-SQUARE	C.V.	STD DEV	TMT MEAN
0.432428	149.2886	1.57819341	1.05714286

SOURCE	DF	TYPE I SS	F VALUE	PR > F
XO	1	15.64571429	6.28	0.0276
GROUP	1	7.12595238	2.86	0.1165

PARAMETER	ESTIMATE	T FOR HO: PARAMETER=0	PR > T	STD ERROR OF ESTIMATE
CONTRAST	-1.44166667	-1.69	0.1165	0.85232186

DEPENDENT VARIABLE: RESID

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE
MODEL	2	9.28666667	4.64333333	2.32
ERROR	12	24.01333333	2.00111111	PR > F
UNCORRECTED TOTAL	14	33.30000000		0.1406

R-SQUARE	C.V.	STD DEV	RESID MEAN
0.278879	1414.6063	1.41460634	-0.10000000

SOURCE	DF	TYPE I SS	F VALUE	PR > F
XO	1	0.14000000	0.07	0.7959
GROUP	1	9.14666667	4.57	0.0538

PARAMETER	ESTIMATE	T FOR HO: PARAMETER=0	PR > T	STD ERROR OF ESTIMATE
CONTRAST	1.63333333	2.14	0.0538	0.76397474

LEAST SQUARES MEANS

GROUP	TMT LSMEAN	STD ERR LSMEAN	PROB > T HO:LSMEAN=0
A1_B2	0.23333333	0.64429476	0.7235
B1_A2	1.67500000	0.55797563	0.0110

GROUP	RESID LSMEAN	STD ERR LSMEAN	PROB > T HO:LSMEAN=0
A1_B2	0.83333333	0.57751062	0.1746
B1_A2	-0.80000000	0.50013887	0.1357

GROUP	TREND LSMEAN	STD ERR LSMEAN	PROB > T HO:LSMEAN=0
A1_B2	0.23333333	0.64429476	0.7235
B1_A2	-1.67500000	0.55797563	0.0110

GROUP	Y1 LSMEAN	STD ERR LSMEAN	PROB > T HO:LSMEAN=0
A1_B2	0.30000000	0.51097334	0.5680
B1_A2	-1.23750000	0.44251589	0.0161

©

GENERAL LINEAR MODELS PROCEDURE

DEPENDENT VARIABLE: TREND

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE
MODEL	2	22.77166667	11.38583333	4.57
ERROR	12	29.88833333	2.49069444	PR > F
UNCORRECTED TOTAL	14	52.66000000		0.0334

R-SQUARE	C.V.	STD DEV	TREND MEAN
0.432428	184.1226	1.57819341	-0.85714286

SOURCE	DF	TYPE I SS	F VALUE	PR > F
X0	1	10.28571429	4.13	0.0649
GROUP	1	12.48595238	5.01	0.0449

PARAMETER	ESTIMATE	T FOR H0: PARAMETER=0	PR > T	STD ERROR OF ESTIMATE
CONTRAST	1.90833333	2.24	0.0449	0.85232186

DEPENDENT VARIABLE: Y1

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE
MODEL	2	12.79125000	6.39562500	4.08
ERROR	12	18.79875000	1.56656250	PR > F
UNCORRECTED TOTAL	14	31.59000000		0.0444

R-SQUARE	C.V.	STD DEV	Y1 MEAN
0.404915	216.3301	1.25162395	-0.57857143

SOURCE	DF	TYPE I SS	F VALUE	PR > F
X0	1	4.68642857	2.99	0.1093
GROUP	1	8.10482143	5.17	0.0421

PARAMETER	ESTIMATE	T FOR H0: PARAMETER=0	PR > T	STD ERROR OF ESTIMATE
CONTRAST	1.53750000	2.27	0.0421	0.67595419

Ⓓ

OBS	SUBJECT	X0	GROUP	PERIOD	TRT	Y
1	1	1	A1_B2	1	A	0.2
2	1	1	A1_B2	2	B	1.0
3	2	1	A1_B2	1	A	0.0
4	2	1	A1_B2	2	B	-0.7
5	3	1	A1_B2	1	A	-0.8
6	3	1	A1_B2	2	B	0.2
7	4	1	A1_B2	1	A	0.6
8	4	1	A1_B2	2	B	1.1
9	5	1	A1_B2	1	A	0.3
10	5	1	A1_B2	2	B	0.4
11	6	1	A1_B2	1	A	1.5
12	6	1	A1_B2	2	B	1.2
13	7	1	B1_A2	1	B	1.3
14	7	1	B1_A2	2	A	0.9
15	8	1	B1_A2	1	B	-2.3
16	8	1	B1_A2	2	A	1.0
17	9	1	B1_A2	1	B	0.0
18	9	1	B1_A2	2	A	0.6
19	10	1	B1_A2	1	B	-0.8
20	10	1	B1_A2	2	A	-0.3
21	11	1	B1_A2	1	B	-0.4
22	11	1	B1_A2	2	A	-1.0
23	12	1	B1_A2	1	B	-2.9
24	12	1	B1_A2	2	A	1.7
25	13	1	B1_A2	1	B	-1.9
26	13	1	B1_A2	2	A	-0.3
27	14	1	B1_A2	1	B	-2.9
28	14	1	B1_A2	2	A	0.9

Ⓔ

GENERAL LINEAR MODELS PROCEDURE

CLASS LEVEL INFORMATION

CLASS	LEVELS	VALUES
TRT	2	A B
PERIOD	2	1 2
SUBJECT	14	1 2 3 4 5 6 7 8 9 10 11 12 13 14
GROUP	2	A1_B2 B1_A2

NUMBER OF OBSERVATIONS IN DATA SET = 28

DEPENDENT VARIABLE: Y

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE
MODEL	16	28.03583333	1.75223958	1.41
ERROR	12	14.94416667	1.24534722	PR > F
UNCORRECTED TOTAL	28	42.98000000		0.2780

R-SQUARE	C.V.	STD DEV	Y MEAN
0.652300	2231.9025	1.11595126	-0.05000000

SOURCE	DF	TYPE I SS	F VALUE	PR > F
X0	1	0.07000000	0.06	0.8166
GROUP	1	4.57333333	3.67	0.0794
SUBJECT(GROUP)	12	12.00666667	0.80	0.6446
PERIOD	1	7.82285714	6.28	0.0276
TRT	1	3.56297619	2.86	0.1165

The Type I SS test equality of weighted means for the PERIOD effect. This is probably not of interest.

SOURCE	DF	TYPE II SS	F VALUE	PR > F
X0	0	0.00000000	.	.
GROUP	1	4.57333333	3.67	0.0794
SUBJECT(GROUP)	12	12.00666667	0.80	0.6446
PERIOD	1	6.24297619	5.01	0.0449
TRT	1	3.56297619	2.86	0.1165

SOURCE	DF	TYPE III SS	F VALUE	PR > F
X0	0	0.00000000	.	.
GROUP	1	4.57333333	3.67	0.0794
SUBJECT(GROUP)	12	12.00666667	0.80	0.6446
PERIOD	1	6.24297619	5.01	0.0449
TRT	1	3.56297619	2.86	0.1165

The Type III SS give the appropriate tests of unweighted means for the TRT and PERIOD effects. (See Ⓐ, Ⓑ and ESTIMATE on next page.)

Ⓔ

TESTS OF HYPOTHESES USING THE TYPE I MS FOR SUBJECT(GROUP) AS AN ERROR TERM

SOURCE	DF	TYPE I SS	F VALUE	PR > F	
GROUP (Residual)	1	4.57333333	4.57	0.0538	
PARAMETER	ESTIMATE	T FOR H0: PARAMETER=0	PR > T	STD ERROR OF ESTIMATE	
TRT	0.72083333	1.69	0.1165	0.42616093	
PERIOD	-0.95416667	-2.24	0.0449	0.42616093	

Compare with parts Ⓑ and Ⓒ.

MEANS

GROUP	N	Y
A1_B2	12	0.41666667
B1_A2	16	-0.40000000

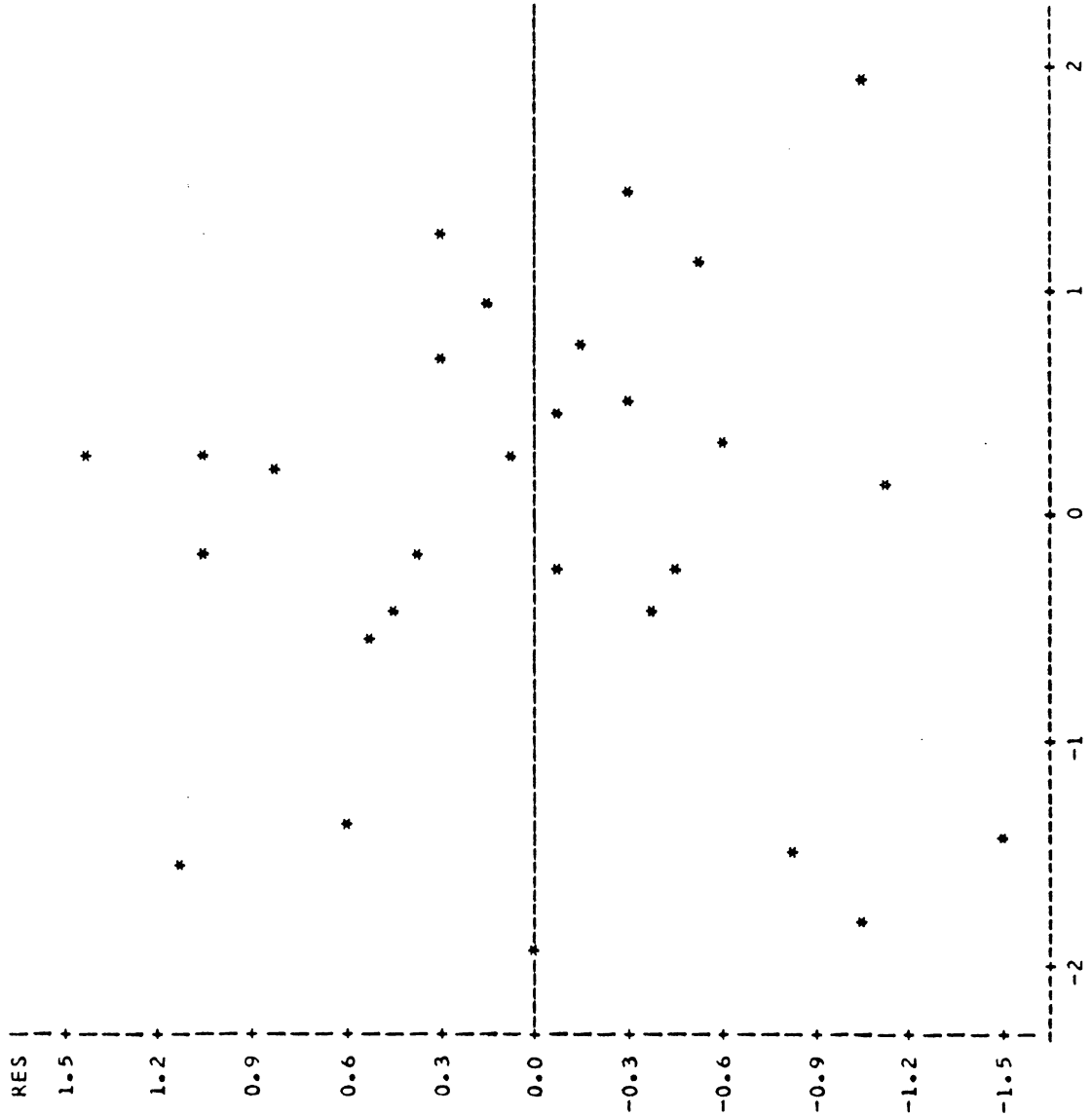
SUBJECT	GROUP	N	Y
1	A1_B2	2	0.60000000
2	A1_B2	2	-0.35000000
3	A1_B2	2	-0.30000000
4	A1_B2	2	0.85000000
5	A1_B2	2	0.35000000
6	A1_B2	2	1.35000000
7	B1_A2	2	1.10000000
8	B1_A2	2	-0.65000000
9	B1_A2	2	0.30000000
10	B1_A2	2	-0.55000000
11	B1_A2	2	-0.70000000
12	B1_A2	2	-0.60000000
13	B1_A2	2	-1.10000000
14	B1_A2	2	-1.00000000

PERIOD	N	Y
1	14	-0.57857143
2	14	0.47857143

TRT	N	Y
A	14	0.37857143
B	14	-0.47857143

(F)

PLOT OF RES*P SYMBOL USED IS *



MILK YIELD DATA - Analysis of a 3 Period - 3 Treatment Cross-over Design
Balanced for Residual Effects. (p. 134ff in Cochran and Cox, 1957)

```
DATA COW;
INPUT SQUARE COW PERIOD TRT $ YIELD RA RB RC RES;
CARDS;
```

(A) PROC PRINT;
VAR SQUARE COW PERIOD TRT RA RB RC RES YIELD;
TITLE A CROSSOVER DESIGN WHEN THERE ARE POSSIBLE RESIDUAL EFFECTS;

(B) PROC GLM;
CLASSES SQUARE COW PERIOD TRT;
MODEL YIELD = COW PERIOD(SQUARE) RA RB RC TRT / SS1;
MEANS TRT / DEONLY;
LSMEANS TRT / STDERR;

(C) PROC GLM;
CLASSES SQUARE COW PERIOD TRT;
MODEL YIELD = COW PERIOD(SQUARE) TRT RA RB RC / SS1;
ESTIMATE 'TRT A VS AVG(B+C)' TRT 1 -0.5 -0.5;
ESTIMATE 'TRT B VS TRT C' TRT 0 1 -1;
LSMEANS TRT / STDERR;
MEANS TRT / DEONLY;
OUTPUT OUT=NEW PREDICTED=P RESIDUAL=R;

(D) PROC PLOT;
PLOT R*P='*' / VREF=0;

A check on the residuals reveals no obvious violations of the assumed model.

SQUARE, COW, PERIOD, TRT and RES are CLASS variables. RA, RB and RC are 0,1 indicator variables which indicate the residual effects of each of the three treatments. YIELD is the response variable, coded milk yield.

(A) CLASS, indicator and response variables

OBS	SQUARE	COW	PERIOD	TRT	RA	RB	RC	RES	YIELD
1	1	1	1	A	0	0	0	0	38
2	1	1	2	B	1	0	0	1	25
3	1	1	3	C	0	1	0	2	15
4	1	2	1	B	0	0	0	0	109
5	1	2	2	C	0	1	0	2	86
6	1	2	3	A	0	0	1	3	39
7	1	3	1	C	0	0	0	0	124
8	1	3	2	A	0	0	1	3	72
9	1	3	3	B	1	0	0	1	27
10	2	4	1	A	0	0	0	0	86
11	2	4	2	C	1	0	0	1	76
12	2	4	3	B	0	0	1	3	46
13	2	5	1	B	0	0	0	0	75
14	2	5	2	A	0	1	0	2	35
15	2	5	3	C	1	0	0	1	34
16	2	6	1	C	0	0	0	0	101
17	2	6	2	B	0	0	1	3	63
18	2	6	3	A	0	1	0	2	1

Ⓑ The model fitting residual effects before treatment effects.

A CROSSOVER DESIGN WHEN THERE ARE POSSIBLE RESIDUAL EFFECTS

GENERAL LINEAR MODELS PROCEDURE

CLASS LEVEL INFORMATION

CLASS	LEVELS	VALUES	
SQUARE	2	1 2	
COW	6	1 2 3 4 5 6	TRT A = Roughage diet
PERIOD	3	1 2 3	B = Limited grain diet
TRT	3	A B C	C = Full grain diet

NUMBER OF OBSERVATIONS IN DATA SET = 18

DEPENDENT VARIABLE: YIELD

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE
MODEL	13	20163.19444444	1551.01495726	31.14
ERROR	4	199.25000000	49.81250000	PR > F
CORRECTED TOTAL	17	20362.44444444		0.0023

R-SQUARE	C.V.	STD DEV	YIELD MEAN
0.990215	12.0761	7.05779711	58.44444444

SOURCE	DF	TYPE I SS	F VALUE	PR > F
COW	5	5781.11111111	23.21	0.0047
PERIOD(SQUARE)	4	11489.11111111	57.66	0.0009
RA } RB } RC }	1 1 0	{ 11.75555556 26.66666667 0.00000000	{ 0.24 0.54 .	{ 0.6525 0.5049 .
TRT	2	2854.55000000	28.65	0.0043

The SS due to residual effects unadjusted for treatments is found as:

$$\begin{aligned}
 \text{SS Residual effects (unadj.)} &= \text{RA} + \text{RB} + \text{RC} \\
 &= 11.7556 + 26.6667 + 0 \\
 &= 38.4223 \text{ with 2 df.}
 \end{aligned}$$

© The model-fitting treatments before residual effects.

A CROSSOVER DESIGN WHEN THERE ARE POSSIBLE RESIDUAL EFFECTS

GENERAL LINEAR MODELS PROCEDURE

DEPENDENT VARIABLE: YIELD

SOURCE	DF	SUM OF SQUARES	MEAN SQUARE	F VALUE
MODEL	13	20163.19444444	1551.01495726	31.14
ERROR	4	199.25000000	49.81250000	PR > F
CORRECTED TOTAL	17	20362.44444444		0.0023

R-SQUARE	C.V.	STD DEV	YIELD MEAN
0.990215	12.0761	7.05779711	58.44444444

SOURCE	DF	TYPE I SS	F VALUE	PR > F
COW	5	5781.11111111	23.21	0.0047
PERIOD(SQUARE)	4	11489.11111111	57.66	0.0009
TRT	2	2276.77777778	22.85	0.0065
RA	1	258.67361111	5.19	0.0849
RB	1	357.52083333	7.18	0.0553
RC	0	0.00000000	.	.

$$\begin{aligned}
 \text{SS Residual effects (adj.)} &= \text{RA} + \text{RB} + \text{RC} \\
 &= 258.6736 + 357.5208 + 0 \\
 &= 616.1944 \text{ with 2 df.}
 \end{aligned}$$

PARAMETER	ESTIMATE	T FOR H0: PARAMETER=0	PR > T	STD ERROR OF ESTIMATE
TRT A VS AVG(B+C)	-23.93750000	-6.07	0.0037	3.94542853
TRT B VS TRT C	-20.62500000	-4.53	0.0106	4.55578844

These are contrasts among the unadjusted treatment means.

Combining the results of (B) and (C) gives the ANOVA table found on page 135 of Cochran and Cox:

Source	df	SS	MS
Cows	5	5781.1111	
Periods w/i squares	4	11489.1111	
{ Treatments (unadj.)	2	2276.7778	1138.3889
{ Residual (adj.)	2	616.1944	308.0972
{ Residual (unadj.)	2	38.4223	19.1115
{ Treatments (adj.)	2	2854.5500	1427.2750
Error	4	199.2500	49.8125
Corrected Total	17	20362.4444	

A CROSSOVER DESIGN WHEN THERE ARE POSSIBLE RESIDUAL EFFECTS

GENERAL LINEAR MODELS PROCEDURE

MEANS = The unadjusted treatment means

TRT	N	YIELD
A	6	45.1666667
B	6	57.5000000
C	6	72.6666667

A CROSSOVER DESIGN WHEN THERE ARE POSSIBLE RESIDUAL EFFECTS

GENERAL LINEAR MODELS PROCEDURE

LEAST SQUARES MEANS = The treatment means adjusted for residual effects

TRT	YIELD LSMEAN	STD ERR LSMEAN	PROB > T HO:LSMEAN=0
A	42.4861111	3.1121960	0.0002
B	56.1111111	3.1121960	0.0001
C	76.7361111	3.1121960	0.0001

Notice that the adjustments are intuitively pleasing.

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