Chapter 44 The NESTED Procedure

Chapter Table of Contents

OVERVIEW
GETTING STARTED
SYNTAX
DETAILS2363Missing Values2363Unbalanced Data2364General Random Effects Model2364Analysis of Covariation2364Error Terms in F Tests2365Computational Method2365Displayed Output2366ODS Table Names2368
EXAMPLE
REFERENCES

Chapter 44 The NESTED Procedure

Overview

The NESTED procedure performs random effects analysis of variance for data from an experiment with a nested (hierarchical) structure.* A random effects model for data from a completely nested design with two factors has the following form:

$$y_{ijr} = \mu + \alpha_i + \beta_{ij} + \epsilon_{ijr}$$

where

y_{ijr}	is the value of the dependent variable observed at the r th replication with the first factor at its <i>i</i> th level and the second factor at its <i>j</i> th level.
μ	is the overall (fixed) mean of the sampling population.
$\alpha_i, \beta_{ij}, \epsilon_{ijr}$	are mutually uncorrelated random effects with zero means and respec- tive variances σ_{α}^2 , σ_{β}^2 , and σ_{ϵ}^2 (the variance components).

This model is appropriate for an experiment with a multi-stage nested sampling design. An example of this is given in Example 44.1 on page 2368, where four turnip plants are randomly chosen (the first factor), then three leaves are randomly chosen from each plant (the second factor nested within the first), and then two samples are taken from each leaf (the different replications at fixed levels of the two factors).

Note that PROC NESTED is appropriate for models with only classification effects; it does not handle models that contain continuous covariates. For random effects models with covariates, use either the GLM or MIXED procedure.

Contrasted with Other SAS Procedures

The NESTED procedure performs a computationally efficient analysis of variance for data with a nested design, estimating the different components of variance and also testing for their significance if the design is balanced (see the "Unbalanced Data" section on page 2364). Although other procedures (such as GLM and MIXED) provide similar analyses, PROC NESTED is both easier to use and computationally more efficient for this special type of design. This is especially true when the design involves a large number of factors, levels, or observations.

*PROC NESTED is modeled after the General Purpose Nested Analysis of Variance program of the Dairy Cattle Research Branch of the United States Department of Agriculture. That program was originally written by M.R. Swanson, Statistical Reporting Service, United States Department of Agriculture.

For example, to specify a four-factor completely nested design in the GLM procedure, you use the form

```
class a b c d;
model y=a b(a) c(a b) d(a b c);
```

However, to specify the same design in PROC NESTED, you simply use the form

class a b c d; var y;

In addition, other procedures require TEST statements to perform appropriate tests, whereas the NESTED procedure produces the appropriate tests automatically. However, PROC NESTED makes one assumption about the input data that the other procedures do not: **PROC NESTED assumes that the input data set is sorted by the classification (CLASS) variables defining the effects.** If you use PROC NESTED on data that is not sorted by the CLASS variables, then the results may not be valid.

Getting Started

Reliability of Automobile Models

A study is performed to compare the reliability of several models of automobiles. Three different automobile models (Model) from each of four domestic automobile manufacturers (Make) are tested. Three different cars of each make and model are subjected to a reliability test and given a score between 1 and 100 (Score), where higher scores indicate greater reliability.

The following statements create the SAS data set auto.

```
title 'Reliability of Automobile Models';
data auto;
   input Make $ Model Score @@;
  datalines;
a 1 62 a 2 77 a 3 59
a 1 67 a 2 73 a 3 64
a 1 60 a 2 79 a 3 60
b 1 72 b 2 58 b 3 80
b 1 75 b 2 63 b 3 84
b 1 69 b 2 57 b 3 89
c 1 94 c 2 76 c 3 81
      с 275 с 385
c 1 90
c 1 88 c 2 78 c 3 85
d 1 69 d 2 73 d 3 90
d 1 72 d 2 88 d 3 87
d 1 76 d 2 87 d 3 92
;
```

The Make variable contains the make of the automobile, represented here by 'a', 'b', 'c', or 'd', while the Model variable represents the automobile model with a '1', '2', or '3'. The Score variable contains the reliability scores given to the three sampled cars from each Make-Model group. Since the automobile models are nested within their makes, the NESTED procedure is used to analyze this data. The NESTED procedure requires the data to be sorted by Make and, within Make, by Model, so the following statements execute a PROC SORT before completing the analysis.

```
proc sort;
   by Make Model;
proc nested;
   class Make Model;
   var Score;
run;
```

The Model variable appears after the Make variable in the CLASS statement because it is nested within Make. The VAR statement specifies the response variable. The output is displayed in Figure 44.1.

		R	eliabilit	y of Aut	omobile	Models		
			The	NESTED P	rocedure	2		
		Coef	ficients	of Expec	ted Mean	n Squares		
	Source Make			lake	Model	Error		
		Make		9	3	1		
		Mode	1	0	3	1		
		Erro	r	0	0	1		
	Nes	ted Random Ef:	fects Ana	lysis of	Varian	ce for Variable	Score	
Variance	Nes	sted Random Ef: Sum of	fects Ana	lysis of	Varian Error	ce for Variable	Score Variance	Percent
	Nes DF	Sum of	fects Ana F Value	-	Error	ce for Variable Mean Square	Variance	
Source		Sum of Squares		-	Error		Variance Component	of Total
Source	DF	Sum of Squares 4177.8888889	F Value	Pr > F	Error Term	Mean Square 119.368254	Variance Component	of Total
Source Total Make	DF 35 3	Sum of Squares 4177.8888889	F Value 2.15	Pr > F 0.1719	Error Term Model	Mean Square 119.368254 569.666667	Variance Component 131.876543 33.867284	of Total 100.0000 25.6811
Variance Source Total Make Model Error	DF 35 3 8	Sum of Squares 4177.888889 1709.000000	F Value 2.15	Pr > F 0.1719	Error Term Model	Mean Square 119.368254 569.666667	Variance Component 131.876543 33.867284 83.425926	of Total 100.0000 25.6811 63.2606
Source Total Make Model	DF 35 3 8	Sum of Squares 4177.888889 1709.000000 2118.888889	F Value 2.15 18.16	Pr > F 0.1719	Error Term Model	Mean Square 119.368254 569.666667 264.861111	Variance Component 131.876543 33.867284 83.425926	of Total 100.0000 25.6811 63.2606

Figure 44.1. Output from PROC NESTED

Figure 44.1 first displays the coefficients of the variance components that make up each of the expected mean squares, then the ANOVA table is displayed. The results do not indicate significant variation between the different automobile makes (F = 2.15, p = 0.1719). However, they do suggest that there is significant variation between the different models within the makes (F = 18.16, p < 0.0001). This is evident in the fact that the make of car accounts for only 25.7% of the total variation in the data, while the car model accounts for 63.3% (as shown in the Percent of Total column). The estimated variance components are shown in the Variance Component column.

Syntax

The following statements are available in PROC NESTED.

```
PROC NESTED < options > ;
CLASS variables ;
VAR variables ;
BY variables ;
```

The PROC NESTED and CLASS statements are required. The BY, CLASS, and VAR statements are described after the PROC NESTED statement.

PROC NESTED Statement

PROC NESTED < options > ;

The PROC NESTED statement has the following options:

AOV

displays only the analysis of variance statistics when there is more than one dependent variable. The "analysis of covariation" statistics are suppressed (see the "Analysis of Covariation" section on page 2364).

DATA=SAS-data-set

names the SAS data set to be used by PROC NESTED. By default, the procedure uses the most recently created SAS data set.

BY Statement

BY variables;

You can specify a BY statement with PROC NESTED to obtain separate analyses on observations in groups defined by the BY variables. When a BY statement appears, the procedure expects the input data set to be sorted in the order of the BY variables.

Note: When you use the NESTED procedure, your data must be sorted first by the BY variables and, within the BY variables, by the CLASS variables.

If your input data set is not sorted in ascending order, use one of the following alternatives:

- Sort the data using the SORT procedure with a similar BY statement.
- Specify the BY statement option NOTSORTED or DESCENDING in the BY statement for the NESTED procedure. The NOTSORTED option does not mean that the data are unsorted but rather that the data are arranged in groups (according to values of the BY variables) and that these groups are not necessarily in alphabetical or increasing numeric order.

• Create an index on the BY variables using the DATASETS procedure (in base SAS software).

For more information on the BY statement, refer to the discussion in *SAS Language Reference: Concepts.* For more information on the DATASETS procedure, refer to the discussion in the *SAS Procedures Guide*.

CLASS Statement

CLASS variables;

You must include a CLASS statement with PROC NESTED specifying the classification variables for the analysis.

Values of a variable in the CLASS statement denote the levels of an effect. The name of that variable is also the name of the corresponding effect. The second effect is assumed to be nested within the first effect, the third effect is assumed to be nested within the second effect, and so on.

Note: The data set must be sorted by the classification variables in the order that they are given in the CLASS statement. Use PROC SORT to sort the data if they are not already sorted.

VAR Statement

VAR variables;

The VAR statement lists the dependent variables for the analysis. The dependent variables must be numeric variables. If you do not specify a VAR statement, PROC NESTED performs an analysis of variance for all numeric variables in the data set, except those already specified in the CLASS statement.

Details

Missing Values

An observation with missing values for any of the variables used by PROC NESTED is omitted from the analysis. Blank values of CLASS character variables are treated as missing values.

Unbalanced Data

A completely nested design is defined to be unbalanced if the groups corresponding to the levels of some classification variable are not all of the same size. The NESTED procedure can compute unbiased estimates for the variance components in an unbalanced design, but because the sums of squares on which these estimates are based no longer have χ^2 distributions under a Gaussian model for the data, F tests for the significance of the variance components cannot be computed. PROC NESTED checks to see that the design is balanced. If it is not, a warning to that effect is placed on the log, and the columns corresponding to the F tests in the analysis of variance are left blank.

General Random Effects Model

A random effects model for data from a completely nested design with n factors has the general form

 $y_{i_1i_2\cdots i_nr} = \mu + \alpha_{i_1} + \beta_{i_1i_2} + \cdots + \epsilon_{i_1i_2\cdots i_nr}$

where

$y_{i_1i_2\cdots i_nr}$	is the value of the dependent variable observed at the <i>r</i> th replication with factor j at level i_j , for $j = 1,, n$.
μ	is the overall (fixed) mean of the sampled population.
$\alpha_{i_1}, \beta_{i_1i_2}, \ldots, \epsilon_{i_1i_2\cdots i_nr}$	are mutually uncorrelated random effects with zero means and respective variances $\sigma_{\alpha}^2, \sigma_{\beta}^2, \ldots, \sigma_{\epsilon}^2$.

Analysis of Covariation

When you specify more than one dependent variable, the NESTED procedure produces a descriptive analysis of the covariance between each pair of dependent variables in addition to a separate analysis of variance for each variable. The analysis of covariation is computed under the basic random effects model for each pair of dependent variables:

$$y_{i_1i_2\cdots i_nr} = \mu + \alpha_{i_1} + \beta_{i_1i_2} + \cdots + \epsilon_{i_1i_2\cdots i_nr}$$
$$y'_{i_1i_2\cdots i_nr} = \mu' + \alpha'_{i_1} + \beta'_{i_1i_2} + \cdots + \epsilon'_{i_1i_2\cdots i_nr}$$

where the notation is the same as that used in the preceding general random effects model.

There is an additional assumption that all the random effects in the two models are mutually uncorrelated except for corresponding effects, for which

$$\operatorname{Corr}(\alpha_{i_1}, \alpha'_{i_1}) = \rho_{\alpha}$$
$$\operatorname{Corr}(\beta_{i_1 i_2}, \beta'_{i_1 i_2}) = \rho_{\beta}$$
$$\vdots$$
$$\operatorname{Corr}(\epsilon_{i_1 i_2 \cdots i_n r}, \epsilon'_{i_1 i_2 \cdots i_n r}) = \rho_{\epsilon}$$

Error Terms in F Tests

Random effects ANOVAs are distinguished from fixed effects ANOVAs by which error mean squares are used as the denominator for F tests. Under a fixed effects model, there is only one true error term in the model, and the corresponding mean square is used as the denominator for all tests. This is how the usual analysis is computed in PROC ANOVA, for example. However, in a random effects model for a nested experiment, mean squares are compared sequentially. The correct denominator in the test for the first factor is the mean square due to the second factor; the correct denominator in the test for the second factor is the mean square due to the third factor; and so on. Only the mean square due to the last factor, the one at the bottom of the nesting order, should be compared to the error mean square.

Computational Method

The building blocks of the analysis are the sums of squares for the dependent variables for each classification variable within the factors that precede it in the model, corrected for the factors that follow it. For example, for a two-factor nested design, PROC NESTED computes the following sums of squares:

$$\begin{aligned} & \text{Total SS} \qquad \sum_{ijr} (y_{ijr} - y_{...})^2 \\ & \text{SS for Factor 1} \qquad \sum_i n_{i\cdot} \left(\frac{y_{i\cdot}}{n_{i\cdot}} - \frac{y_{...}}{n_{..}} \right)^2 \\ & \text{SS for Factor 2 within Factor 1} \qquad \sum_{ij} n_{ij} \left(\frac{y_{ij\cdot}}{n_{ij}} - \frac{y_{i\cdot}}{n_{i\cdot}} \right)^2 \\ & \text{Error SS} \qquad \sum_{ijr} \left(y_{ijr} - \frac{y_{ij\cdot}}{n_{ij}} \right)^2 \end{aligned}$$

where y_{ijr} is the *r*th replication, n_{ij} is the number of replications at level *i* of the first factor and level *j* of the second, and a dot as a subscript indicates summation over the corresponding index. If there is more than one dependent variable, PROC NESTED

also computes the corresponding sums of crossproducts for each pair. The expected value of the sum of squares for a given classification factor is a linear combination of the variance components corresponding to this factor and to the factors that are nested within it. For each factor, the coefficients of this linear combination are computed. (The efficiency of PROC NESTED is partly due to the fact that these various sums can be accumulated with just one pass through the data, assuming that the data have been sorted by the classification variables.) Finally, estimates of the variance components are derived as the solution to the set of linear equations that arise from equating the mean squares to their expected values.

Displayed Output

PROC NESTED displays the following items for each dependent variable:

• Coefficients of Expected Mean Squares, which are the coefficients of the n + 1 variance components making up the expected mean square. Denoting the element in the *i*th row and *j*th column of this matrix by C_{ij} , the expected value of the mean square due to the *i*th classification factor is

$$C_{i1}\sigma_1^2 + \dots + C_{in}\sigma_n^2 + C_{i,n+1}\sigma_\epsilon^2$$

 C_{ij} is always zero for i > j, and if the design is balanced, C_{ij} is equal to the common size of all classification groups of the *j*th factor for $i \le j$. Finally, the mean square for error is always an unbiased estimate of σ_{ϵ}^2 . In other words, $C_{n+1,n+1} = 1$.

For every dependent variable, PROC NESTED displays an analysis of variance table. Each table contains the following:

- each Variance Source in the model (the different components of variance) and the total variance
- degrees of freedom (DF) for the corresponding sum of squares
- Sum of Squares for each classification factor. The sum of squares for a given classification factor is the sum of squares in the dependent variable within the factors that precede it in the model, corrected for the factors that follow it. (See the "Computational Method" section on page 2365.)
- *F* Value for a factor, which is the ratio of its mean square to the appropriate error mean square. The next column, labeled PR > F, gives the significance levels that result from testing the hypothesis that each variance component equals zero.
- the appropriate Error Term for an *F* test, which is the mean square due to the next classification factor in the nesting order. (See the "Error Terms in F Tests" section on page 2365.)
- Mean Square due to a factor, which is the corresponding sum of squares divided by the degrees of freedom

- estimates of the Variance Components. These are computed by equating the mean squares to their expected values and solving for the variance terms. (See the "Computational Method" section on page 2365.)
- Percent of Total, the proportion of variance due to each source. For the *i*th factor, the value is

 $100 \times \frac{\text{source variance component}}{\text{total variance component}}$

• Mean, the overall average of the dependent variable. This gives an unbiased estimate of the mean of the population. Its variance is estimated by a certain linear combination of the estimated variance components, which is identical to the mean square due to the first factor in the model divided by the total number of observations when the design is balanced.

If there is more than one dependent variable, then the NESTED procedure displays an "analysis of covariation" table for each pair of dependent variables (unless the AOV option is specified in the PROC NESTED statement). See the "Analysis of Covariation" section on page 2364 for details. For each source of variation, this table includes the following:

- Degrees of Freedom
- Sum of Products
- Mean Products
- Covariance Component, the estimate of the covariance component

Items in the analysis of covariation table are computed analogously to their counterparts in the analysis of variance table. The analysis of covariation table also includes the following:

- Variance Component Correlation for a given factor. This is an estimate of the correlation between corresponding effects due to this factor. This correlation is the ratio of the covariance component for this factor to the square root of the product of the variance components for the factor for the two different dependent variables. (See the "Analysis of Covariation" section on page 2364.)
- Mean Square Correlation for a given classification factor. This is the ratio of the Mean Products for this factor to the square root of the product of the Mean Squares for the factor for the two different dependent variables.

ODS Table Names

PROC NESTED assigns a name to each table it creates. You can use these names to reference the table when using the Output Delivery System (ODS) to select tables and create output data sets. These names are listed in the following table. For more information on ODS, see Chapter 15, "Using the Output Delivery System."

Table 44.1. ODS Tables Produced in PROC NESTED

ODS Table Name	Description	Statement
ANCOVA	Analysis of covariance	default with more than
		one dependent variable
ANOVA	Analysis of variance	default
EMSCoef	Coefficients of expected	default
	mean squares	
Statistics	Overall statistics for fit	default

Example

Example 44.1. Variability of Calcium Concentration in Turnip Greens

In the following example from Snedecor and Cochran (1976), an experiment is conducted to study the variability of calcium concentration in turnip greens. Four plants are selected at random; then three leaves are randomly selected from each plant. Two 100-mg samples are taken from each leaf. The amount of calcium is determined by microchemical methods.

Because the data are read in sorted order, it is not necessary to use PROC SORT on the CLASS variables. Leaf is nested in Plant; Sample is nested in Leaf and is left for the residual term. All the effects are random effects. The following statements read the data and invoke PROC NESTED. These statements produce Output 44.1.1:

```
title 'Calcium Concentration in Turnip Leaves'
    '--Nested Random Model';
title2 'Snedecor and Cochran, ''Statistical Methods'''
    ', 1976, p. 286';
data Turnip;
    do Plant=1 to 4;
        do Leaf=1 to 3;
            do Sample=1 to 2;
               input Calcium @@;
               output;
              end;
              end;
              end;
              end;
```

```
datalines;
3.28 3.09 3.52 3.48 2.88 2.80
2.46 2.44 1.87 1.92 2.19 2.19
2.77 2.66 3.74 3.44 2.55 2.55
3.78 3.87 4.07 4.12 3.31 3.31
;
proc nested;
    class Plant Leaf;
    var Calcium;
run;
```

Output 44.1.1. Analysis of Calcium Concentration in Turnip Greens Using PROC NESTED

Calcium Concentrati Snedecor and Cochr		-		
	The NESTED	Procedure		
Coefficie	ents of Expe	cted Mean	Squares	
Source	Plant	Leaf	Error	
Plant	6	2	1	
Leaf	0	2	1	
Error	0	0	1	

				-		Nested Random thods', 1976, p		
			The	NESTED P	rocedur	e		
	Neste	ed Random Eff	ects Anal	ysis of	Varianc	e for Variable (Calcium	
Variance		Sum of			Error		Variance	Percent
Source	DF	Squares	F Value	Pr > F	Term	Mean Square	Component	of Total
Total	23	10.270396				0.446539	0.532938	100.0000
Plant	3	7.560346	7.67	0.0097	Leaf	2.520115	0.365223	68.5302
Leaf	8	2.630200	49.41	<.0001	Error	0.328775	0.161060	30.2212
Error	12	0.079850				0.006654	0.006654	1.2486
		Calcium M	ean			3.01208333		
		Standard	Error of	Calcium	Mean	0.32404445		

The results indicate that there is significant (nonzero) variation from plant to plant (Pr > F is 0.0097) and from leaf to leaf within a plant (Pr > F is less than 0.0001). Notice that the variance component for Plant uses the Leaf mean square as an error term in the model rather than the error mean square.

References

- Snedecor, G.W. and Cochran, W.G. (1976), *Statistical Methods*, Sixth Edition, Ames, IA: Iowa State University Press.
- Steel, R.G.D. and Torrie, J.H. (1980), *Principles and Procedures of Statistics*, New York: McGraw-Hill Book Co.

The correct bibliographic citation for this manual is as follows: SAS Institute Inc., SAS/STAT[®] User's Guide, Version 8, Cary, NC: SAS Institute Inc., 1999.

SAS/STAT[®] User's Guide, Version 8

Copyright © 1999 by SAS Institute Inc., Cary, NC, USA.

ISBN 1-58025-494-2

All rights reserved. Produced in the United States of America. No part of this publication may be reproduced, stored in a retrieval system, or transmitted, in any form or by any means, electronic, mechanical, photocopying, or otherwise, without the prior written permission of the publisher, SAS Institute Inc.

U.S. Government Restricted Rights Notice. Use, duplication, or disclosure of the software and related documentation by the U.S. government is subject to the Agreement with SAS Institute and the restrictions set forth in FAR 52.227–19 Commercial Computer Software-Restricted Rights (June 1987).

SAS Institute Inc., SAS Campus Drive, Cary, North Carolina 27513.

1st printing, October 1999

 SAS^{\circledast} and all other SAS Institute Inc. product or service names are registered trademarks or trademarks of SAS Institute Inc. in the USA and other countries.[®] indicates USA registration.

Other brand and product names are registered trademarks or trademarks of their respective companies.

The Institute is a private company devoted to the support and further development of its software and related services.