

SAS/STAT® 12.1 User's Guide The CANDISC Procedure (Chapter)



This document is an individual chapter from SAS/STAT® 12.1 User's Guide.

The correct bibliographic citation for the complete manual is as follows: SAS Institute Inc. 2012. SAS/STAT® 12.1 User's Guide. Cary, NC: SAS Institute Inc.

Copyright © 2012, SAS Institute Inc., Cary, NC, USA

All rights reserved. Produced in the United States of America.

For a Web download or e-book: Your use of this publication shall be governed by the terms established by the vendor at the time you acquire this publication.

The scanning, uploading, and distribution of this book via the Internet or any other means without the permission of the publisher is illegal and punishable by law. Please purchase only authorized electronic editions and do not participate in or encourage electronic piracy of copyrighted materials. Your support of others' rights is appreciated.

U.S. Government Restricted Rights Notice: Use, duplication, or disclosure of this 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.

Electronic book 1, August 2012

SAS® Publishing provides a complete selection of books and electronic products to help customers use SAS software to its fullest potential. For more information about our e-books, e-learning products, CDs, and hard-copy books, visit the SAS Publishing Web site at **support.sas.com/publishing** or call 1-800-727-3228.

 $SAS^{@}$ 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.

Chapter 29

The CANDISC Procedure

Contents

Overview: CANDISC Procedure	1745
Getting Started: CANDISC Procedure	1747
Syntax: CANDISC Procedure	1751
PROC CANDISC Statement	1752
BY Statement	1755
CLASS Statement	1756
FREQ Statement	1756
VAR Statement	1756
WEIGHT Statement	1756
Details: CANDISC Procedure	1757
Missing Values	1757
Computational Details	1757
Input Data Set	1758
Output Data Sets	1759
Computational Resources	1761
Displayed Output	1762
ODS Table Names	1764
Example: CANDISC Procedure	1765
Example 29.1: Analysis of Iris Data With PROC CANDISC	1765
References	1771

Overview: CANDISC Procedure

Canonical discriminant analysis is a dimension-reduction technique related to principal component analysis and canonical correlation. The methodology used in deriving the canonical coefficients parallels that of a one-way MANOVA. MANOVA tests for equality of the mean vector across class levels. Canonical discriminant analysis finds linear combinations of the quantitative variables that provide maximal separation between classes or groups. Given a classification variable and several quantitative variables, the CANDISC procedure derives *canonical variables*, linear combinations of the quantitative variables that summarize between-class variation in much the same way that principal components summarize total variation.

The CANDISC procedure performs a canonical discriminant analysis, computes squared Mahalanobis distances between class means, and performs both univariate and multivariate one-way analyses of variance. Two output data sets can be produced: one containing the canonical coefficients and another containing, among other things, scored canonical variables. The canonical coefficients output data set can be rotated

by the FACTOR procedure. It is customary to standardize the canonical coefficients so that the canonical variables have means that are equal to zero and pooled within-class variances that are equal to one. PROC CANDISC displays both standardized and unstandardized canonical coefficients. Correlations between the canonical variables and the original variables as well as the class means for the canonical variables are also displayed; these correlations, sometimes known as loadings, are called canonical structures. To aid the visual interpretation of group differences, you can use ODS Graphics to display graphs of pairs of canonical variables from the scored canonical variables output data set.

Given two or more groups of observations with measurements on several quantitative variables, canonical discriminant analysis derives a linear combination of the variables that has the highest possible multiple correlation with the groups. This maximal multiple correlation is called the *first canonical correlation*. The coefficients of the linear combination are the *canonical coefficients* or *canonical weights*. The variable defined by the linear combination is the *first canonical variable* or *canonical component*. The second canonical correlation is obtained by finding the linear combination uncorrelated with the first canonical variable that has the highest possible multiple correlation with the groups. The process of extracting canonical variables can be repeated until the number of canonical variables equals the number of original variables or the number of classes minus one, whichever is smaller.

The first canonical correlation is at least as large as the multiple correlation between the groups and any of the original variables. If the original variables have high within-group correlations, the first canonical correlation can be large even if all the multiple correlations are small. In other words, the first canonical variable can show substantial differences between the classes, even if none of the original variables do. Canonical variables are sometimes called *discriminant functions*, but this usage is ambiguous because the DISCRIM procedure produces very different functions for classification that are also called discriminant functions.

For each canonical correlation, PROC CANDISC tests the hypothesis that it and all smaller canonical correlations are zero in the population. An *F* approximation (Rao 1973; Kshirsagar 1972) is used that gives better small-sample results than the usual chi-square approximation. The variables should have an approximate multivariate normal distribution within each class, with a common covariance matrix in order for the probability levels to be valid.

Canonical discriminant analysis is equivalent to canonical correlation analysis between the quantitative variables and a set of dummy variables coded from the class variable. Canonical discriminant analysis is also equivalent to performing the following steps:

- 1. Transform the variables so that the pooled within-class covariance matrix is an identity matrix.
- 2. Compute class means on the transformed variables.
- 3. Perform a principal component analysis on the means, weighting each mean by the number of observations in the class. The eigenvalues are equal to the ratio of between-class variation to within-class variation in the direction of each principal component.
- 4. Back-transform the principal components into the space of the original variables, obtaining the canonical variables.

An interesting property of the canonical variables is that they are uncorrelated whether the correlation is calculated from the total sample or from the pooled within-class correlations. The canonical coefficients are not orthogonal, however, so the canonical variables do not represent perpendicular directions through the space of the original variables.

Getting Started: CANDISC Procedure

The data in this example are measurements of 159 fish caught in Finland's lake Laengelmavesi; this data set is available from the Puranen. For each of the For each of the seven species (bream, roach, whitefish, parkki, perch, pike, and smelt) the weight, length, height, and width of each fish are tallied. Three different length measurements are recorded: from the nose of the fish to the beginning of its tail, from the nose to the notch of its tail, and from the nose to the end of its tail. The height and width are recorded as percentages of the third length variable. The fish data set is available from the Sashelp library.

The following step uses PROC CANDISC to find the three canonical variables that best separate the species of fish in the Sashelp. Fish data and create the output data set outcan. With the NCAN=3 option, only the first three canonical variables are displayed. The ODS EXCLUDE statement is specified to exclude the canonical structure tables and most of the canonical coefficient tables in order to obtain a more compact set of results. The TEMPLATE and SGRENDER procedures are used to create a plot of the first two canonical variables. The following statements produce Figure 29.1 through Figure 29.6:

```
title 'Fish Measurement Data';
proc candisc data=sashelp.fish ncan=3 out=outcan;
   ods exclude tstruc bstruc pstruc tcoef pcoef;
   class Species;
   var Weight Length1 Length2 Length3 Height Width;
run;
proc template;
   define statgraph scatter;
      begingraph;
         entrytitle 'Fish Measurement Data';
         layout overlayequated / equatetype=fit
            xaxisopts=(label='Canonical Variable 1')
            yaxisopts=(label='Canonical Variable 2');
            scatterplot x=Can1 y=Can2 / group=species name='fish';
            layout gridded / autoalign=(topright);
               discretelegend 'fish' / border=false opaque=false;
            endlayout;
         endlayout;
      endgraph;
   end;
run;
proc sgrender data=outcan template=scatter;
```

PROC CANDISC begins by displaying summary information about the variables in the analysis. This information includes the number of observations, the number of quantitative variables in the analysis (specified with the VAR statement), and the number of classes in the classification variable (specified with the CLASS statement). The frequency of each class is also displayed.

Figure 29.1 Summary Information

	Fis	h Measurement I	ata	
	The	CANDISC Proces	lure	
Total Sampl	le Size	158 DI	' Total	157
Variables		6 DI	' Within Classe	s 151
Classes		7 DI	Between Class	es 6
N	Number of Obs	ervations Read	159	
_		ervations Used		
	Clas	s Level Informa	tion	
		s Level Informa	tion	
Species	Clas Variable Name	s Level Informa		Proportion
Species Bream	Variable			Proportion 0.215190
-	Variable Name	Frequency	Weight :	0.215190
Bream	Variable Name Bream	Frequency	Weight : 34.0000 11.0000	0.215190 0.069620
Bream Parkki	Variable Name Bream Parkki	Frequency 34 11	Weight 34.0000 11.0000 56.0000	0.215190 0.069620 0.354430
Bream Parkki Perch	Variable Name Bream Parkki Perch	Frequency 34 11 56	Weight 34.0000 11.0000 56.0000 17.0000	0.215190 0.069620 0.354430 0.107595
Bream Parkki Perch Pike	Variable Name Bream Parkki Perch Pike	Frequency 34 11 56 17	Weight 34.0000 11.0000 56.0000 17.0000	0.215190 0.069620 0.354430 0.107595 0.126582

PROC CANDISC performs a multivariate one-way analysis of variance (one-way MANOVA) and provides four multivariate tests of the hypothesis that the class mean vectors are equal. These tests, shown in Figure 29.2, indicate that not all of the mean vectors are equal (p < .0001).

Figure 29.2 MANOVA and Multivariate Tests

Fish Measurement Data					
	The CANDISC	Procedure			
Multivari	ate Statistics	and F Appro	ximations		
	S=6 M=-0	.5 N=72			
Statistic	Value	F Value	Num DF	Den DF	Pr > F
Wilks' Lambda	0.00036325	90.71	36	643.89	<.0001
Pillai's Trace	3.10465132	26.99	36	906	<.0001
Hotelling-Lawley Trace	52.05799676	209.24	36	413.64	<.0001
Roy's Greatest Root	39.13499776	984.90	6	151	<.0001
NOTE: F Statistic	for Roy's Gre	atest Root i	s an uppe	r bound.	

The first canonical correlation is the greatest possible multiple correlation with the classes that can be achieved by using a linear combination of the quantitative variables. The first canonical correlation, displayed in Figure 29.3, is 0.987463. A likelihood ratio test is displayed of the hypothesis that the current canonical correlation and all smaller ones are zero. The first line is equivalent to Wilks' lambda multivariate test.

Figure 29.3 Canonical Correlations

	Fi	sh Measurement	Data		
	The	e CANDISC Proce	edure		
		Adjusted	Approxim	ate	Squared
	Canonical	Canonical	Stand	ard	Canonical
	Correlation	Correlation	Er	ror C	orrelation
1	0.987463	0.986671	0.001	989	0.975084
2	0.952349	0.950095	0.007	425	0.906969
3	0.838637	0.832518	0.023	678	0.703313
4	0.633094	0.623649	0.047	821	0.400809
5	0.344157	0.334170	0.070	356	0.118444
6	0.005701		0.079	806	0.000033
		Eigenvalues	of Inv(E)	*H	
		= CanRsq/	(1-CanRsq)		
	Eigenvalue	Difference	Proporti	on Cu	mulative
1	39.1350	29.3859	0.75	18	0.7518
2	9.7491	7.3786	0.18	73	0.9390
3	2.3706	1.7016	0.04	55	0.9846
4	0.6689	0.5346	0.01	28	0.9974
5	0.1344	0.1343	0.00	26	1.0000
6	0.0000		0.00	00	1.0000
	Test of H0: T	he canonical co	rrelations	in the	
	current row	and all that f	follow are	zero	
	Likelihood	Approximate			
	Ratio	F Value	Num DF	Den DF	Pr > F
1	0.00036325	90.71	36	643.89	<.0001
2	0.01457896	46.46	25	547.58	<.0001
3	0.15671134	23.61	16	452.79	<.0001
4	0.52820347	12.09	9	362.78	<.0001
5	0.88152702	4.88	4	300	0.0008
6	0.99996749	0.00	1	151	0.9442

The first canonical variable, Can1, shows that the linear combination of the centered variables Can1 = $-0.0006 \times \text{Weight} - 0.33 \times \text{Length1} \ 2.49 \times \text{Length2} + 2.60 \times \text{Length3} + 1.12 \times \text{Height} - 1.45 \times \text{Width}$ separates the species most effectively (see Figure 29.4).

Figure 29.4 Raw Canonical Coefficients

	Fish Measurement Data				
	The CANDISC Procedure				
	Raw Canonic	cal Coefficients			
Variable	e Can1	Can2	Can3		
Weight	-0.000648508	-0.005231659	-0.005596192		
Length1	-0.329435762	-0.626598051	-2.934324102		
Length2	-2.486133674	-0.690253987	4.045038893		
Length3	2.595648437	1.803175454	-1.139264914		
Height	1.121983854	-0.714749340	0.283202557		
Width	-1.446386704	-0.907025481	0.741486686		

PROC CANDISC computes the means of the canonical variables for each class. The first canonical variables able is the linear combination of the variables Weight, Length1, Length2, Length3, Height, and Width that provides the greatest difference (in terms of a univariate F test) between the class means. The second canonical variable provides the greatest difference between class means while being uncorrelated with the first canonical variable.

Figure 29.5 Class Means for Canonical Variables

	Class Means on Ca	anonical Variables	
Species	Can1	Can2	Can3
Bream	10.94142464	0.52078394	0.23496708
Parkki	2.58903743	-2.54722416	-0.49326158
Perch	-4.47181389	-1.70822715	1.29281314
Pike	-4.89689441	8.22140791	-0.16469132
Roach	-0.35837149	0.08733611	-1.10056438
Smelt	-4.09136653	-2.35805841	-4.03836098
Whitefish	-0.39541755	-0.42071778	1.06459242

A plot of the first two canonical variables (Figure 29.6) shows that Can1 discriminates between three groups: 1) bream; 2) whitefish, roach, and parkki; and 3) smelt, pike, and perch. Can2 best discriminates between pike and the other species.

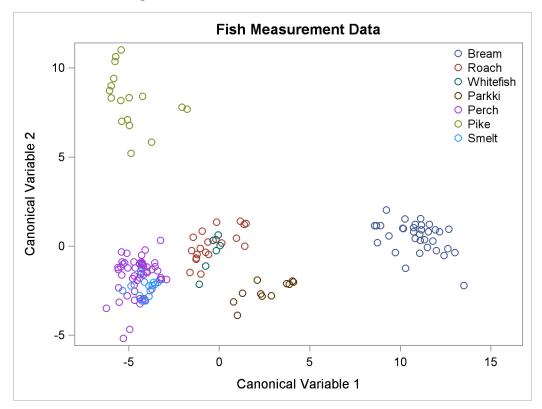


Figure 29.6 Plot of First Two Canonical Variables

Syntax: CANDISC Procedure

The following statements are available in the CANDISC procedure:

```
PROC CANDISC < options>;
CLASS variable;
BY variables;
FREQ variable;
VAR variables;
WEIGHT variable;
```

The BY, CLASS, FREQ, VAR, and WEIGHT statements are described after the PROC CANDISC statement.

PROC CANDISC Statement

PROC CANDISC < options> ;

The PROC CANDISC statement invokes the CANDISC procedure. Table 29.1 summarizes the options available in the PROC CANDISC statement.

Table 29.1 CANDISC Procedure Options

Option	Description
Input Data Set	t .
DATA=	Specifies input SAS data set
Output Data S	ets
OUT=	Specifies output data set with canonical scores
OUTSTAT=	Specifies output statistics data set
Method Detail	s
NCAN=	Specifies the number of canonical variables
PREFIX=	Specifies a prefix for naming the canonical variables
SINGULAR=	Specifies the singularity criterion
Control Displa	yed Output
ALL	Displays all output
ANOVA	Displays univariate statistics
BCORR	Displays between correlations
BCOV	Displays between covariances
BSSCP	Displays between SSCPs
DISTANCE	Displays squared Mahalanobis distances
NOPRINT	Suppresses all displayed output
PCORR	Displays pooled correlations
PCOV	Displays pooled covariances
PSSCP	Displays pooled SSCPs
SHORT	Suppresses some displayed output
SIMPLE	Displays simple descriptive statistics
STDMEAN	Displays standardized class means
TCORR	Displays total correlations
TCOV	Displays total covariances
TSSCP	Displays total SSCPs
WCORR	Displays within correlations
WCOV	Displays within covariances
WSSCP	Displays within SSCPs

ALL

activates all of the display options.

ANOVA

displays univariate statistics for testing the hypothesis that the class means are equal in the population for each variable.

BCORR

displays between-class correlations.

BCOV

displays between-class covariances. The between-class covariance matrix equals the between-class SSCP matrix divided by n(c-1)/c, where n is the number of observations and c is the number of classes. The between-class covariances should be interpreted in comparison with the total-sample and within-class covariances, not as formal estimates of population parameters.

BSSCP

displays the between-class SSCP matrix.

DATA=SAS-data-set

specifies the data set to be analyzed. The data set can be an ordinary SAS data set or one of several specially structured data sets created by SAS statistical procedures. These specially structured data sets include TYPE=CORR, COV, CSSCP, and SSCP. If you omit the DATA= option, the procedure uses the most recently created SAS data set.

DISTANCE

MAHALANOBIS

displays squared Mahalanobis distances between the group means, *F* statistics, and the corresponding probabilities of greater squared Mahalanobis distances between the group means.

NCAN=n

specifies the number of canonical variables to be computed. The value of n must be less than or equal to the number of variables. If you specify NCAN=0, the procedure displays the canonical correlations, but not the canonical coefficients, structures, or means. A negative value suppresses the canonical analysis entirely. Let v be the number of variables in the VAR statement, and let c be the number of classes. If you omit the NCAN= option, only $\min(v, c-1)$ canonical variables are generated; if you also specify an OUT= output data set, v canonical variables are generated, and the last v - (c-1) canonical variables have missing values.

NOPRINT

suppresses the normal display of results. Note that this option temporarily disables the Output Delivery System (ODS); see Chapter 20, "Using the Output Delivery System," for more information.

OUT=SAS-data-set

creates an output SAS data set containing the original data and the canonical variable scores. If you want to create a SAS data set in a permanent library, you must specify a two-level name. For more information about permanent libraries and SAS data sets, see SAS Language Reference: Concepts.

OUTSTAT=SAS-data-set

creates a TYPE=CORR output SAS data set that contains various statistics, including class means, standard deviations, correlations, canonical correlations, canonical structures, canonical coefficients, and means of canonical variables for each class. If you want to create a SAS data set in a permanent library, you must specify a two-level name. For more information about permanent libraries and SAS data sets, see SAS Language Reference: Concepts.

PCORR

displays pooled within-class correlations (partial correlations based on the pooled within-class covariances).

PCOV

displays pooled within-class covariances.

PREFIX=name

specifies a prefix for naming the canonical variables. By default the names are Can1, Can2, Can3, and so forth. If you specify PREFIX=Abc, the components are named Abc1, Abc2, and so on. The number of characters in the prefix plus the number of digits required to designate the canonical variables should not exceed 32. The prefix is truncated if the combined length exceeds 32.

PSSCP

displays the pooled within-class corrected SSCP matrix.

SHORT

suppresses the display of canonical structures, canonical coefficients, and class means on canonical variables; only tables of canonical correlations and multivariate test statistics are displayed.

SIMPLE

displays simple descriptive statistics for the total sample and within each class.

SINGULAR=p

specifies the criterion for determining the singularity of the total-sample correlation matrix and the pooled within-class covariance matrix, where 0 . The default is SINGULAR=1E-8.

Let S be the total-sample correlation matrix. If the R square for predicting a quantitative variable in the VAR statement from the variables preceding it exceeds 1 - p, then S is considered singular. If S is singular, the probability levels for the multivariate test statistics and canonical correlations are adjusted for the number of variables with R square exceeding 1 - p.

If S is considered singular and the inverse of S (squared Mahalanobis distances) is required, a quasi inverse is used instead. For details see the section "Quasi-inverse" on page 2083 in Chapter 33, "The DISCRIM Procedure."

STDMEAN

displays total-sample and pooled within-class standardized class means.

TCORR

displays total-sample correlations.

TCOV

displays total-sample covariances.

TSSCP

displays the total-sample corrected SSCP matrix.

WCORR

displays within-class correlations for each class level.

WCOV

displays within-class covariances for each class level.

WSSCP

displays the within-class corrected SSCP matrix for each class level.

BY Statement

BY variables;

You can specify a BY statement with PROC CANDISC to obtain separate analyses of observations in groups that are defined by the BY variables. When a BY statement appears, the procedure expects the input data set to be sorted in order of the BY variables. If you specify more than one BY statement, only the last one specified is used.

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

- Sort the data by using the SORT procedure with a similar BY statement.
- Specify the NOTSORTED or DESCENDING option in the BY statement for the CANDISC 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 by using the DATASETS procedure (in Base SAS software).

For more information about BY-group processing, see the discussion in SAS Language Reference: Concepts. For more information about the DATASETS procedure, see the discussion in the Base SAS Procedures Guide.

CLASS Statement

CLASS variable;

The values of the CLASS variable define the groups for analysis. Class levels are determined by the formatted values of the CLASS variable. The CLASS variable can be numeric or character. A CLASS statement is required.

FREQ Statement

FREQ variable;

If a variable in the data set represents the frequency of occurrence for the other values in the observation, include the name of the variable in a FREQ statement. The procedure then treats the data set as if each observation appears n times, where n is the value of the FREQ variable for the observation. The total number of observations is considered to be equal to the sum of the FREQ variable when the procedure determines degrees of freedom for significance probabilities.

If the value of the FREQ variable is missing or is less than one, the observation is not used in the analysis. If the value is not an integer, the value is truncated to an integer.

VAR Statement

VAR variables;

You specify the quantitative variables to include in the analysis by using a VAR statement. If you do not use a VAR statement, the analysis includes all numeric variables not listed in other statements.

WEIGHT Statement

WEIGHT variable;

To use relative weights for each observation in the input data set, place the weights in a variable in the data set and specify the name in a WEIGHT statement. This is often done when the variance associated with each observation is different and the values of the WEIGHT variable are proportional to the reciprocals of the variances. If the value of the WEIGHT variable is missing or is less than zero, then a value of zero for the weight is assumed.

The WEIGHT and FREQ statements have a similar effect except that the WEIGHT statement does not alter the degrees of freedom.

Details: CANDISC Procedure

Missing Values

If an observation has a missing value for any of the quantitative variables, it is omitted from the analysis. If an observation has a missing CLASS value but is otherwise complete, it is not used in computing the canonical correlations and coefficients; however, canonical variable scores are computed for that observation for the OUT= data set.

Computational Details

General Formulas

Canonical discriminant analysis is equivalent to canonical correlation analysis between the quantitative variables and a set of dummy variables coded from the class variable. In the following notation the dummy variables are denoted by \mathbf{y} and the quantitative variables by \mathbf{x} . The total sample covariance matrix for the \mathbf{x} and \mathbf{y} variables is

$$\mathbf{S} = \begin{bmatrix} \mathbf{S}_{xx} & \mathbf{S}_{xy} \\ \mathbf{S}_{yx} & \mathbf{S}_{yy} \end{bmatrix}$$

When c is the number of groups, n_t is the number of observations in group t, and S_t is the sample covariance matrix for the x variables in group t, the within-class pooled covariance matrix for the x variables is

$$\mathbf{S}_p = \frac{1}{\sum n_t - c} \sum (n_t - 1) \mathbf{S}_t$$

The canonical correlations, ρ_i , are the square roots of the eigenvalues, λ_i , of the following matrix. The corresponding eigenvectors are \mathbf{v}_i .

$$S_p^{-1/2}S_{xy}S_{yy}^{-1}S_{yx}S_p^{-1/2}$$

Let V be the matrix with the eigenvectors \mathbf{v}_i that correspond to nonzero eigenvalues as columns. The raw canonical coefficients are calculated as follows:

$$\mathbf{R} = \mathbf{S}_p^{-1/2} \mathbf{V}$$

The pooled within-class standardized canonical coefficients are

$$\mathbf{P} = \mathrm{diag}(\mathbf{S}_p)^{1/2}\mathbf{R}$$

The total sample standardized canonical coefficients are

$$\mathbf{T} = \mathrm{diag}(\mathbf{S}_{xx})^{1/2}\mathbf{R}$$

Let X_c be the matrix with the centered x variables as columns. The canonical scores can be calculated by any of the following:

$$X_c R$$

$$\mathbf{X}_c \operatorname{diag}(\mathbf{S}_p)^{-1/2} \mathbf{P}$$

$$\mathbf{X}_c \operatorname{diag}(\mathbf{S}_{xx})^{-1/2} \mathbf{T}$$

For the multivariate tests based on $E^{-1}H$,

$$\mathbf{E} = (n-1)(\mathbf{S}_{yy} - \mathbf{S}_{yx}\mathbf{S}_{xx}^{-1}\mathbf{S}_{xy})$$

$$\mathbf{H} = (n-1)\mathbf{S}_{yx}\mathbf{S}_{xx}^{-1}\mathbf{S}_{xy}$$

where *n* is the total number of observations.

Input Data Set

The input DATA= data set can be an ordinary SAS data set or one of several specially structured data sets created by statistical procedures available with SAS/STAT software. For more information about special types of data sets, see Appendix A, "Special SAS Data Sets." The BY variable in these data sets becomes the CLASS variable in PROC CANDISC. These specially structured data sets include the following:

- TYPE=CORR data sets created by PROC CORR by using a BY statement
- TYPE=COV data sets created by PROC PRINCOMP by using both the COV option and a BY statement
- TYPE=CSSCP data sets created by PROC CORR by using the CSSCP option and a BY statement, where the OUT= data set is assigned TYPE=CSSCP with the TYPE= data set option
- TYPE=SSCP data sets created by PROC REG by using both the OUTSSCP= option and a BY statement.

When the input data set is TYPE=CORR, TYPE=COV, or TYPE=CSSCP, then PROC CANDISC reads the number of observations for each class from the observations with _TYPE_='N' and the variable means in each class from the observations with _TYPE_='MEAN'. The CANDISC procedure then reads the within-class correlations from the observations with _TYPE_='CORR', the standard deviations from the observations with _TYPE_='STD' (data set TYPE=CORR), the within-class covariances from the observations with _TYPE_='COV' (data set TYPE=COV), or the within-class corrected sums of squares and crossproducts from the observations with _TYPE_='CSSCP' (data set TYPE=CSSCP).

When the data set does not include any observations with _TYPE_='CORR' (data set TYPE=CORR), _TYPE_='COV' (data set TYPE=COV), or _TYPE_='CSSCP' (data set TYPE=CSSCP) for each class, PROC CANDISC reads the pooled within-class information from the data set. In this case, PROC CANDISC reads the pooled within-class correlations from the observations with _TYPE_='PCORR', the pooled within-class standard deviations from the observations with _TYPE_='PSTD' (data set TYPE=CORR), the pooled within-class covariances from the observations with _TYPE_='PCOV' (data set TYPE=COV), or the pooled within-class corrected SSCP matrix from the observations with _TYPE_='PSSCP' (data set TYPE=CSSCP).

When the input data set is TYPE=SSCP, then PROC CANDISC reads the number of observations for each class from the observations with _TYPE_='N', the sum of weights of observations from the variable INTERCEPT in observations with _TYPE_='SSCP' and _NAME_='INTERCEPT', the variable sums from the analysis variables in observations with _TYPE_='SSCP' and _NAME_='INTERCEPT', and the uncorrected sums of squares and crossproducts from the analysis variables in observations with _TYPE_='SSCP' and _NAME_=variablenames.

Output Data Sets

OUT= Data Set

The OUT= data set contains all the variables in the original data set plus new variables containing the canonical variable scores. You determine the number of new variables by using the NCAN= option. The names of the new variables are formed as described in the PREFIX= option. The new variables have means equal to zero and pooled within-class variances equal to one. An OUT= data set cannot be created if the DATA= data set is not an ordinary SAS data set.

OUTSTAT= Data Set

The OUTSTAT= data set is similar to the TYPE=CORR data set produced by the CORR procedure but contains many results in addition to those produced by the CORR procedure.

The OUTSTAT= data set is TYPE=CORR, and it contains the following variables:

- the BY variables, if any
- the CLASS variable
- _TYPE_, a character variable of length 8 that identifies the type of statistic
- _NAME_, a character variable of length 32 that identifies the row of the matrix or the name of the canonical variable
- the quantitative variables (those in the VAR statement, or if there is no VAR statement, all numeric variables not listed in any other statement)

The observations, as identified by the variable _TYPE_, have the following _TYPE_ values:

TYPE	Contents
N	number of observations both for the total sample (CLASS variable missing) and within each class (CLASS variable present)
SUMWGT	sum of weights both for the total sample (CLASS variable missing) and within each class (CLASS variable present) if a WEIGHT statement is specified
MEAN	means both for the total sample (CLASS variable missing) and within each class (CLASS variable present)
STDMEAN	total-standardized class means
PSTDMEAN	pooled within-class standardized class means
STD	standard deviations both for the total sample (CLASS variable missing) and within each class (CLASS variable present)
PSTD	pooled within-class standard deviations
BSTD	between-class standard deviations
RSQUARED	univariate R squares

The following kinds of observations are identified by the combination of the variables _TYPE_ and _NAME_. When the _TYPE_ variable has one of the following values, the _NAME_ variable identifies the row of the matrix:

TYPE	Contents
CSSCP	corrected SSCP matrix for the total sample (CLASS variable missing) and within each class (CLASS variable present)
PSSCP	pooled within-class corrected SSCP matrix
BSSCP	between-class SSCP matrix
COV	covariance matrix for the total sample (CLASS variable missing) and within each class (CLASS variable present)
PCOV	pooled within-class covariance matrix
BCOV	between-class covariance matrix
CORR	correlation matrix for the total sample (CLASS variable missing) and within each class (CLASS variable present)
PCORR	pooled within-class correlation matrix
BCORR	between-class correlation matrix

When the _TYPE_ variable has one of the following values, the _NAME_ variable identifies the canonical variable:

TYPE	Contents
CANCORR	canonical correlations
STRUCTUR	canonical structure

BSTRUCT between canonical structure

PSTRUCT pooled within-class canonical structure

SCORE total sample standardized canonical coefficients

PSCORE pooled within-class standardized canonical coefficients

RAWSCORE raw canonical coefficients

CANMEAN means of the canonical variables for each class

You can use this data set with PROC SCORE to get scores on the canonical variables for new data by using one of the following forms:

```
* The CLASS variable C is numeric;
proc score data=NewData score=Coef(where=(c = . )) out=Scores;
run;

* The CLASS variable C is character;
proc score data=NewData score=Coef(where=(c = ' ')) out=Scores;
run;
```

The WHERE clause is used to exclude the within-class means and standard deviations. PROC SCORE standardizes the new data by subtracting the original variable means that are stored in the _TYPE_='MEAN' observations, and dividing by the original variable standard deviations from the _TYPE_='STD' observations. Then PROC SCORE multiplies the standardized variables by the coefficients from the _TYPE_='SCORE' observations to get the canonical scores.

Computational Resources

In the following discussion, let

n = number of observations

c = number of class levels

v = number of variables in the VAR list

l = length of the CLASS variable

Memory Requirements

The amount of memory in bytes for temporary storage needed to process the data is

$$c(4v^2 + 28v + 4l + 68) + 16v^2 + 96v + 4l$$

With the ANOVA option, the temporary storage must be increased by 16v bytes. The DISTANCE option requires an additional temporary storage of $4v^2 + 4v$ bytes.

The following factors determine the time requirements of the CANDISC procedure:

- The time needed for reading the data and computing covariance matrices is proportional to nv^2 . PROC CANDISC must also look up each class level in the list. This is faster if the data are sorted by the CLASS variable. The time for looking up class levels is proportional to a value ranging from n to $n \log(c)$.
- The time for inverting a covariance matrix is proportional to v^3 .
- The time required for the canonical discriminant analysis is proportional to v^3 .

Each of the preceding factors has a different constant of proportionality.

Displayed Output

The displayed output from PROC CANDISC includes the class level information table. For each level of the classification variable, the following information is provided: the output data set variable name, frequency sum, weight sum, and the proportion of the total sample.

The optional output from PROC CANDISC includes the following:

- Within-class SSCP matrices for each group
- Pooled within-class SSCP matrix
- Between-class SSCP matrix
- Total-sample SSCP matrix
- Within-class covariance matrices for each group
- Pooled within-class covariance matrix
- Between-class covariance matrix, equal to the between-class SSCP matrix divided by n(c-1)/c, where n is the number of observations and c is the number of classes
- Total-sample covariance matrix
- Within-class correlation coefficients and Pr > |r| to test the hypothesis that the within-class population correlation coefficients are zero
- Pooled within-class correlation coefficients and Pr > |r| to test the hypothesis that the partial population correlation coefficients are zero
- Between-class correlation coefficients and $\Pr > |r|$ to test the hypothesis that the between-class population correlation coefficients are zero
- Total-sample correlation coefficients and Pr > |r| to test the hypothesis that the total population correlation coefficients are zero

- Simple statistics, including N (the number of observations), sum, mean, variance, and standard deviation both for the total sample and within each class
- Total-sample standardized class means, obtained by subtracting the grand mean from each class mean and dividing by the total sample standard deviation
- Pooled within-class standardized class means, obtained by subtracting the grand mean from each class mean and dividing by the pooled within-class standard deviation
- Pairwise squared distances between groups
- Univariate test statistics, including total-sample standard deviations, pooled within-class standard deviations, between-class standard deviations, R square, $R^2/(1-R^2)$, F, and Pr > F (univariate F values and probability levels for one-way analyses of variance)

By default, PROC CANDISC displays these statistics:

- Multivariate statistics and F approximations including Wilks' lambda, Pillai's trace, Hotelling-Lawley trace, and Roy's greatest root with F approximations, numerator and denominator degrees of freedom (Num DF and Den DF), and probability values (Pr > F). Each of these four multivariate statistics tests the hypothesis that the class means are equal in the population. See the section "Multivariate Tests" on page 92 in Chapter 4, "Introduction to Regression Procedures," for more information.
- Canonical correlations
- Adjusted canonical correlations (Lawley 1959). These are asymptotically less biased than the raw
 correlations and can be negative. The adjusted canonical correlations might not be computable and
 are displayed as missing values if two canonical correlations are nearly equal or if some are close to
 zero. A missing value is also displayed if an adjusted canonical correlation is larger than a previous
 adjusted canonical correlation.
- Approximate standard error of the canonical correlations
- Squared canonical correlations
- Eigenvalues of $E^{-1}H$. Each eigenvalue is equal to $\rho^2/(1-\rho^2)$, where ρ^2 is the corresponding squared canonical correlation and can be interpreted as the ratio of between-class variation to pooled within-class variation for the corresponding canonical variable. The table includes Eigenvalues, Differences between successive eigenvalues, the Proportion of the sum of the eigenvalues, and the Cumulative proportion.
- Likelihood ratio for the hypothesis that the current canonical correlation and all smaller ones are zero in the population. The likelihood ratio for the hypothesis that all canonical correlations equal zero is Wilks' lambda.
- Approx *F* statistic based on Rao's approximation to the distribution of the likelihood ratio (Rao 1973, p. 556; Kshirsagar 1972, p. 326)
- Numerator degrees of freedom (Num DF), denominator degrees of freedom (Den DF), and Pr > F, the probability level associated with the F statistic

The following statistics can be suppressed with the SHORT option:

- Total canonical structure, giving total-sample correlations between the canonical variables and the original variables
- Between canonical structure, giving between-class correlations between the canonical variables and the original variables
- Pooled within canonical structure, giving pooled within-class correlations between the canonical variables and the original variables
- Total-sample standardized canonical coefficients, standardized to give canonical variables with zero mean and unit pooled within-class variance when applied to the total-sample standardized variables
- Pooled within-class standardized canonical coefficients, standardized to give canonical variables with zero mean and unit pooled within-class variance when applied to the pooled within-class standardized variables
- Raw canonical coefficients, standardized to give canonical variables with zero mean and unit pooled within-class variance when applied to the centered variables
- Class means on the canonical variables

ODS Table Names

PROC CANDISC 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 Table 29.2. For more information about ODS, see Chapter 20, "Using the Output Delivery System."

Table 29.2 ODS Tables Produced by PROC CANDISC

ODS Table Name	Description	PROC CANDISC Option
ANOVA	Univariate statistics	ANOVA
AveRSquare	Average R square	ANOVA
BCorr	Between-class correlations	BCORR
BCov	Between-class covariances	BCOV
BSSCP	Between-class SSCP matrix	BSSCP
BStruc	Between canonical structure	default
CanCorr	Canonical correlations	default
CanonicalMeans	Class means on canonical variables	default
Counts	Number of observations, variables,	default
	classes, df	
CovDF	DF for covariance matrices, not printed	any *COV option
Dist	Squared distances	DISTANCE
DistFValues	F statistics based on squared distances	DISTANCE
DistProb	Probabilities for <i>F</i> statistics from	DISTANCE
	squared distances	

Table 29.2 continued

ODS Table Name	Description	PROC CANDISC Option
Levels	Class level information	default
MultStat	MANOVA	default
NObs	Number of observations	default
PCoef	Pooled standard canonical coefficients	default
PCorr	Pooled within-class correlations	PCORR
PCov	Pooled within-class covariances	PCOV
PSSCP	Pooled within-class SSCP matrix	PSSCP
PStdMeans	Pooled standardized class means	STDMEAN
PStruc	Pooled within canonical structure	default
RCoef	Raw canonical coefficients	default
SimpleStatistics	Simple statistics	SIMPLE
TCoef	Total-sample standard canonical	default
	coefficients	
TCorr	Total-sample correlations	TCORR
TCov	Total-sample covariances	TCOV
TSSCP	Total-sample SSCP matrix	TSSCP
TStdMeans	Total standardized class means	STDMEAN
TStruc	Total canonical structure	default
WCorr	Within-class correlations	WCORR
WCov	Within-class covariances	WCOV
WSSCP	Within-class SSCP matrices	WSSCP

Example: CANDISC Procedure

Example 29.1: Analysis of Iris Data With PROC CANDISC

The iris data published by Fisher (1936) have been widely used for examples in discriminant analysis and cluster analysis. The sepal length, sepal width, petal length, and petal width are measured in millimeters on 50 iris specimens from each of three species: *Iris setosa*, *I. versicolor*, and *I. virginica*. The iris data set is available from the Sashelp library.

This example is a canonical discriminant analysis that creates an output data set containing scores on the canonical variables and plots the canonical variables.

The following statements produce Output 29.1.1 through Output 29.1.6:

```
title 'Fisher (1936) Iris Data';
proc candisc data=sashelp.iris out=outcan distance anova;
  class Species;
  var SepalLength SepalWidth PetalLength PetalWidth;
run;
```

PROC CANDISC first displays information about the observations and the classes in the data set in Output 29.1.1.

Output 29.1.1 Iris Data: Summary Information

	Fisher (1936)	Iris Da	ta	
	The CANDISC	Procedur	e	
Total Sample Si	ze 150	DF T	otal	149
Variables	4	DF W	ithin Classe:	s 147
Classes	3	DF B	etween Class	es 2
	Class Level I	nformati	on	
Va	nriable			
•		iency	Weight	Proportion
Species Na		iency 50	Weight	Proportion 0.333333
Species Na	mme Frequetosa	_	-	-

The DISTANCE option in the PROC CANDISC statement displays squared Mahalanobis distances between class means. Results from the DISTANCE option are shown in Output 29.1.2.

Output 29.1.2 Iris Data: Squared Mahalanobis Distances and Distance Statistics

	Fisher (1936) Iris Data						
	The CANDISC Procedure						
	Squared Distance to Species						
From Species	Setosa	Versicolor	Virginica				
Setosa Versicolo		89.86419	179.38471 17.20107				
Virginica	179.38471	17.20107	0				

Output 29.1.2 continued

I beatibeles, RE	F=4, DDF=144 f	for Squared Dist	tance to Specie
From			
Species	Setosa	Versicolor	Virginica
Setosa	0	550.18889	1098
Versicolor	550.18889	0	105.31265
Virginica Prob > Mahalano	1098 Obis Distance f		0 tance to Specie
Virginica Prob > Mahalanc			-
Prob > Mahalano	bis Distance f		-
Prob > Mahalano	bis Distance f	For Squared Dis	tance to Specie
Prob > Mahaland From Species	bis Distance f Setosa	For Squared Disa Versicolor <.0001	tance to Specie Virginica

The ANOVA option uses univariate statistics to test the hypothesis that the class means are equal. The resulting R-square values (see Output 29.1.3) range from 0.4008 for SepalWidth to 0.9414 for PetalLength, and each variable is significant at the 0.0001 level. The multivariate test for differences between the classes (which is displayed by default) is also significant at the 0.0001 level; you would expect this from the highly significant univariate test results.

Output 29.1.3 Iris Data: Univariate and Multivariate Statistics

Fisher (1936) Iris Data								
The CANDISC Procedure								
			THE CH	DIDC TICCO	dure			
Univariate Test Statistics								
		F Stat	istics,	Num DF=2,	Den DF	=147		
		Total	Pooled	Between				
		Standard	Standard	Standard		R-Square		
Variable	Label	Deviation	Deviation	Deviation	R-Square	/ (1-RSq)	F Value	Pr > F
_	Sepal Length (mm)	8.2807	5.1479	7.9506	0.6187	1.6226	119.26	<.0001
Sepal Width		4.3587	3.3969	3.3682	0.4008	0.6688	49.16	<.0001
Petal Length	• •		4.3033	20.9070	0.9414	16.0566	1180.16	<.0001
Petal Width		7.6224	2.0465	8.9673	0.9289	13.0613	960.01	<.0001

Output 29.1.3 continued

	Average R	-Square					
U	nweighted	0.72	24358				
W	eighted by Varia	nce 0.86	89444				
Multivariate Statistics and F Approximations S=2 M=0.5 N=71							
Statistic	Value	F Value	Num DF	Den DF	Pr > F		
Wilks' Lambda	0.02343863	199.15	8	288	<.0001		
Pillai's Trace	1.19189883	53.47	8	290	<.0001		
Hotelling-Lawley Trace	32.47732024	582.20	8	203.4	<.0001		
Roy's Greatest Root	32.19192920	1166.96	4	145	<.0001		
NOTE: F Statist	ic for Roy's Gre	atest Root	is an uppe	r bound.			

The R square between Can1 and the class variable, 0.969872, is much larger than the corresponding R square for Can2, 0.222027. This is displayed in Output 29.1.4.

Output 29.1.4 Iris Data: Canonical Correlations and Eigenvalues

	Fi	sher (1936) Iris	s Data		
	Tl	ne CANDISC Proce	edure		
		Adjusted	Approximat	te	Squared
	Canonical	Canonical	Standa	rd	Canonical
	Correlation	Correlation	Erro	or Co	rrelation
1	0.984821	0.984508	0.0024	68	0.969872
2	0.471197	0.461445	0.06373	34	0.222027
		Eigenvalues	s of Inv(E) *I	H	
		= CanRsq	(1-CanRsq)		
	Eigenvalue	Difference	Proportion	n Cum	nulative
1	. 32.1919	31.9065	0.9912	2	0.9912
2	0.2854		0.0088	3	1.0000
	Test of HO: '	The canonical co	orrelations :	in the	
	current ro	w and all that i	follow are ze	ero	
	Likelihood	Approximate			
	Ratio	F Value	Num DF I	Den DF	Pr > F
1	0.02343863	199.15	8	288	<.0001
2	0.77797337	13.79	3	145	<.0001

Output 29.1.5 Iris Data: Correlations between Canonical and Original Variables

	_,		
	The CANDISC Pr	rocedure	
	Total Canonical	Structure	
Variable	Label	Can1	Can2
SepalLength	Sepal Length (mm)	0.791888	0.217593
SepalWidth	Sepal Width (mm)	-0.530759	0.757989
PetalLength	Petal Length (mm)	0.984951	0.046037
PetalWidth	Petal Width (mm)	0.972812	0.222902
	Between Canonical	l Structure	
Variable	Label	Can1	Can2
SepalLength	Sepal Length (mm)	0.991468	0.130348
SepalWidth	Sepal Width (mm)	-0.825658	0.564171
PetalLength	Petal Length (mm)	0.999750	0.022358
PetalWidth	Petal Width (mm)	0.994044	0.108977
	Pooled Within Canoni	ical Structure	
Variable	Label	Can1	Can2
SepalLength	Sepal Length (mm)	0.222596	0.310812
SepalWidth	Sepal Width (mm)	-0.119012	0.863681
PetalLength	Petal Length (mm)	0.706065	0.167701

The raw canonical coefficients (shown in Output 29.1.6) for the first canonical variable, Can1, show that the classes differ most widely on the linear combination of the centered variables: $-0.0829378 \times \text{SepalLength} - 0.153447 \times \text{SepalWidth} + 0.220121 \times \text{PetalLength} + 0.281046 \times \text{PetalWidth}$.

Output 29.1.6 Iris Data: Canonical Coefficients

Fisher (1936) Iris Data					
The CANDISC Procedure					
Total-Sample Standardized Canonical Coefficients					
Variable	Label	Can1	Can2		
SepalLength	Sepal Length (mm)	-0.686779533	0.019958173		
SepalLength SepalWidth	Sepal Length (mm) Sepal Width (mm)	-0.686779533 -0.668825075	0.019958173 0.943441829		

Output 29.1.6 continued

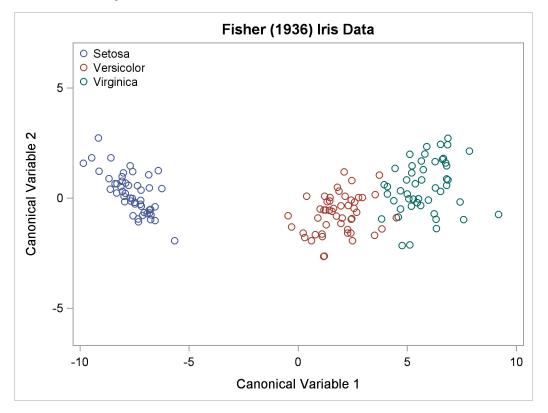
Pooled	Within-Class Standardiz	ed Canonical Coeff	icients
Variable	Label	Can1	Can2
SepalLength	Sepal Length (mm)	4269548486	0.0124075316
SepalWidth	Sepal Width (mm)	5212416758	0.7352613085
PetalLength	Petal Length (mm)	0.9472572487	4010378190
PetalWidth	Petal Width (mm)	0.5751607719	0.5810398645
	Raw Canonical C	Coefficients	
Variable	Label	Can1	Can2
SepalLength	Sepal Length (mm)	0829377642	0.0024102149
SepalWidth	Sepal Width (mm)	1534473068	0.2164521235
PetalLength	Petal Length (mm)	0.2201211656	0931921210
PetalWidth	Petal Width (mm)	0.2810460309	0.2839187853

Output 29.1.7 Iris Data: Canonical Means

Class	Means on Canonical	Variables
Species	Can1	Can2
Setosa	-7.607599927	0.215133017
Versicolor	1.825049490	-0.727899622
Virginica	5.782550437	0.512766605

The TEMPLATE and SGRENDER procedures are used to create a plot of the first two canonical variables. The following statements produce Output 29.1.8:

```
proc template;
   define statgraph scatter;
      begingraph;
         entrytitle 'Fisher (1936) Iris Data';
         layout overlayequated / equatetype=fit
            xaxisopts=(label='Canonical Variable 1')
            yaxisopts=(label='Canonical Variable 2');
            scatterplot x=Can1 y=Can2 / group=species name='iris';
            layout gridded / autoalign=(topleft);
               discretelegend 'iris' / border=false opaque=false;
            endlayout;
         endlayout;
      endgraph;
   end;
run;
proc sgrender data=outcan template=scatter;
run;
```



Output 29.1.8 Iris Data: Plot of First Two Canonical Variables

The plot of canonical variables in Output 29.1.8 shows that of the two canonical variables, Can1 has more discriminatory power.

References

Fisher, R. A. (1936), "The Use of Multiple Measurements in Taxonomic Problems," *Annals of Eugenics*, 7, 179–188.

Kshirsagar, A. M. (1972), Multivariate Analysis, New York: Marcel Dekker.

Lawley, D. N. (1959), "Tests of Significance in Canonical Analysis," *Biometrika*, 46, 59–66.

Puranen, J. (1917), "Fish Catch data set (1917)," Journal of Statistics Education Data Archive, last accessed May 22, 2009.

URL http://www.amstat.org/publications/jse/datasets/fishcatch.txt

Rao, C. R. (1973), *Linear Statistical Inference and Its Applications*, Second Edition, New York: John Wiley & Sons.

Subject Index

```
analysis of variance
    multivariate (CANDISC), 1748
CANDISC procedure
    computational details, 1757
    computational resources, 1761
    input data set, 1758
    introductory example, 1747
    Mahalanobis distance, 1766
    MANOVA, 1748
    memory requirements, 1761
    missing values, 1757
    multivariate analysis of variance, 1748
    ODS table names, 1764
    output data sets, 1753, 1754, 1759
    time requirements, 1762
canonical coefficients, 1746
canonical component, 1746
canonical discriminant analysis, 1745
canonical variables, 1745
canonical weights, 1746
discriminant analysis
    canonical, 1745
discriminant functions, 1746
first canonical variable, 1746
Mahalanobis distance, 1753
    CANDISC procedure, 1766
MANOVA
    CANDISC procedure, 1748
multivariate analysis of variance
    CANDISC procedure, 1748
```

Syntax Index

ALL option	CLASS statement
PROC CANDISC statement, 1753	CANDISC procedure, 1756
ANOVA option	•
PROC CANDISC statement, 1753	DATA= option
	PROC CANDISC statement, 1753
BCORR option	DISTANCE option
PROC CANDISC statement, 1753	PROC CANDISC statement, 1753
BCOV option	
PROC CANDISC statement, 1753	FREQ statement
BSSCP option	CANDISC procedure, 1756
PROC CANDISC statement, 1753	1
BY statement	MAHALANOBIS option
	PROC CANDISC statement, 1753
CANDISC procedure, 1755	
CANDISC procedure	NCAN= option
syntax, 1751	PROC CANDISC statement, 1753
	NOPRINT option
CANDISC procedure, BY statement, 1755	PROC CANDISC statement, 1753
CANDISC procedure, CLASS statement, 1756	TROC CHI (DISC statement, 1755
CANDISC procedure, FREQ statement, 1756	OUT= option
CANDISC procedure, PROC CANDISC statement,	PROC CANDISC statement, 1753
1752	OUTSTAT= option
ALL option, 1753	
ANOVA option, 1753	PROC CANDISC statement, 1754
BCORR option, 1753	PCORR option
BCOV option, 1753	<u> </u>
BSSCP option, 1753	PROC CANDISC statement, 1754
DATA= option, 1753	PCOV option
DISTANCE option, 1753	PROC CANDISC statement, 1754
MAHALANOBIS option, 1753	PREFIX= option
NCAN= option, 1753	PROC CANDISC statement, 1754
<u> </u>	PROC CANDISC statement, see CANDISC
NOPRINT option, 1753	procedure
OUT= option, 1753	PSSCP option
OUTSTAT= option, 1754	PROC CANDISC statement, 1754
PCORR option, 1754	
PCOV option, 1754	SHORT option
PREFIX= option, 1754	PROC CANDISC statement, 1754
PSSCP option, 1754	SIMPLE option
SHORT option, 1754	PROC CANDISC statement, 1754
SIMPLE option, 1754	SINGULAR= option
SINGULAR= option, 1754	PROC CANDISC statement, 1754
STDMEAN option, 1754	STDMEAN option
TCORR option, 1755	PROC CANDISC statement, 1754
TCOV option, 1755	FROC CANDISC statement, 1734
TSSCP option, 1755	TCORR option
÷	<u>*</u>
WCOR option, 1755	PROC CANDISC statement, 1755
WCOV option, 1755	TCOV option
WSSCP option, 1755	PROC CANDISC statement, 1755
CANDISC procedure, VAR statement, 1756	TSSCP option
CANDISC procedure WEIGHT statement 1756	PROC CANDISC statement 1755

VAR statement

CANDISC procedure, 1756

WCORR option

PROC CANDISC statement, 1755

WCOV option

PROC CANDISC statement, 1755

WEIGHT statement

CANDISC procedure, 1756

WSSCP option

PROC CANDISC statement, 1755

Your Turn

We welcome your feedback.

- If you have comments about this book, please send them to yourturn@sas.com. Include the full title and page numbers (if applicable).
- If you have comments about the software, please send them to suggest@sas.com.

SAS® Publishing Delivers!

Whether you are new to the work force or an experienced professional, you need to distinguish yourself in this rapidly changing and competitive job market. SAS* Publishing provides you with a wide range of resources to help you set yourself apart. Visit us online at support.sas.com/bookstore.

SAS® Press

Need to learn the basics? Struggling with a programming problem? You'll find the expert answers that you need in example-rich books from SAS Press. Written by experienced SAS professionals from around the world, SAS Press books deliver real-world insights on a broad range of topics for all skill levels.

support.sas.com/saspress

SAS® Documentation

To successfully implement applications using SAS software, companies in every industry and on every continent all turn to the one source for accurate, timely, and reliable information: SAS documentation. We currently produce the following types of reference documentation to improve your work experience:

- Online help that is built into the software.
- Tutorials that are integrated into the product.
- Reference documentation delivered in HTML and PDF free on the Web.
- Hard-copy books.

support.sas.com/publishing

SAS® Publishing News

Subscribe to SAS Publishing News to receive up-to-date information about all new SAS titles, author podcasts, and new Web site features via e-mail. Complete instructions on how to subscribe, as well as access to past issues, are available at our Web site.

support.sas.com/spn



Sas THE POWER TO KNOW