

## Hierarchical Cluster Analyses

To demonstrate the Hierarchical Clustering program, the data to be analyzed is the one labeled cansas.LAZ. You will see the form below with specifications for the grouping:

**Hierarchical Cluster Analysis**

Variables Available for Selection:

Variables Selected for Analysis:

weight  
waist  
pulse  
chins  
situps  
jumps

Analysis Options:

☒ Standardize Variables  
☐ Replace Grid Values  
☒ Descriptive statistics  
☒ Groups vs Errors Plot

Maximum No. of Groups: 10

Reset Cancel OK

### Specifications for the Hierarchical Cluster Analysis

Results for the hierarchical analysis that you would obtain after clicking the Compute button are presented below:

Hierarchical Cluster Analysis

Number of object to cluster = 20 on 6 variables.

Variable Means

Variables	weight	waist	pulse	chins	situps	jumps
	178.600	35.400	56.100	9.450	145.550	70.300

Variable Variances

Variables	weight	waist	pulse	chins	situps	jumps
	609.621	10.253	51.989	27.945	3914.576	2629.379

Variable Standard Deviations

Variables	weight	waist	pulse	chins	situps	jumps
	24.691	3.202	7.210	5.286	62.567	51.277

19 groups after combining group 1 (n = 1 ) and group 5 (n = 1) error = 0.386  
18 groups after combining group 17 (n = 1 ) and group 18 (n = 1) error = 0.387  
17 groups after combining group 11 (n = 1 ) and group 17 (n = 2) error = 0.556  
16 groups after combining group 1 (n = 2 ) and group 16 (n = 1) error = 0.663  
15 groups after combining group 3 (n = 1 ) and group 7 (n = 1) error = 0.805  
14 groups after combining group 4 (n = 1 ) and group 10 (n = 1) error = 1.050

13 groups after combining group 2 (n = 1 ) and group 6 (n = 1) error = 1.345  
12 groups after combining group 1 (n = 3 ) and group 14 (n = 1) error = 1.402  
11 groups after combining group 0 (n = 1 ) and group 1 (n = 4) error = 1.489  
10 groups after combining group 11 (n = 3 ) and group 12 (n = 1) error = 2.128

Group 1 (n= 5)  
Object = CASE 1  
Object = CASE 2  
Object = CASE 6  
Object = CASE 15  
Object = CASE 17

Group 3 (n= 2)  
Object = CASE 3  
Object = CASE 7

Group 4 (n= 2)  
Object = CASE 4  
Object = CASE 8

Group 5 (n= 2)  
Object = CASE 5  
Object = CASE 11

Group 9 (n= 1)  
Object = CASE 9

Group 10 (n= 1)  
Object = CASE 10

Group 12 (n= 4)  
Object = CASE 12  
Object = CASE 13  
Object = CASE 18  
Object = CASE 19

Group 14 (n= 1)  
Object = CASE 14

Group 16 (n= 1)  
Object = CASE 16

Group 20 (n= 1)  
Object = CASE 20

(... for 9 groups, 8 groups, etc. down to 2 groups)

4 groups after combining group 4 (n = 6 ) and group 9 (n = 1) error = 11.027

Group 1 (n= 8)  
Object = CASE 1  
Object = CASE 2  
Object = CASE 3  
Object = CASE 6  
Object = CASE 7  
Object = CASE 15  
Object = CASE 16  
Object = CASE 17

Group 4 (n= 4)  
Object = CASE 4  
Object = CASE 8  
Object = CASE 9  
Object = CASE 20

Group 5 (n= 7)  
Object = CASE 5  
Object = CASE 10  
Object = CASE 11  
Object = CASE 12  
Object = CASE 13  
Object = CASE 18  
Object = CASE 19

Group 14 (n= 1)  
Object = CASE 14

3 groups after combining group 0 (n = 8 ) and group 13 (n = 1) error = 13.897

Group 1 (n= 9)  
Object = CASE 1  
Object = CASE 2  
Object = CASE 3  
Object = CASE 6  
Object = CASE 7  
Object = CASE 14

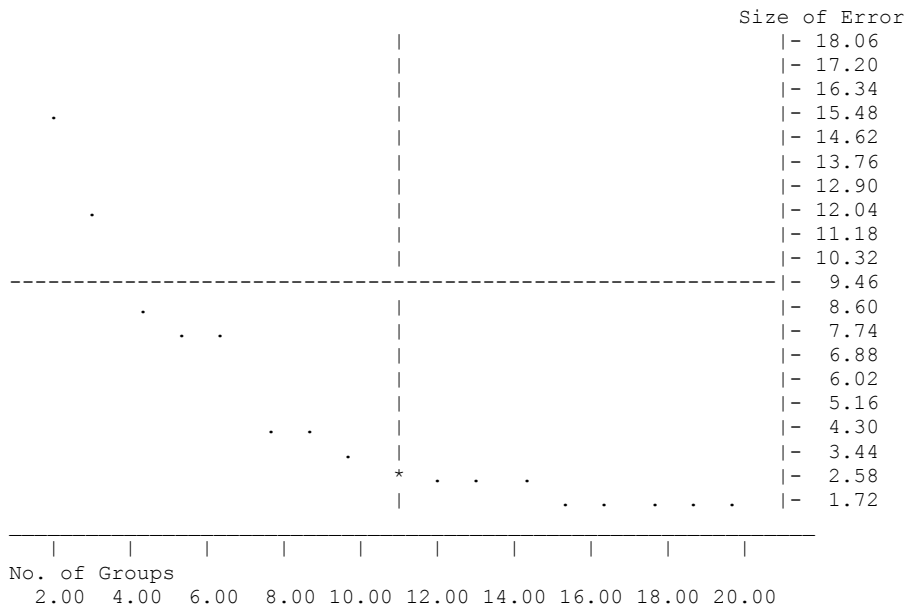
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Object = CASE 15
Object = CASE 16
Object = CASE 17
Group 4 (n= 4)
Object = CASE 4
Object = CASE 8
Object = CASE 9
Object = CASE 20
Group 5 (n= 7)
Object = CASE 5
Object = CASE 10
Object = CASE 11
Object = CASE 12
Object = CASE 13
Object = CASE 18
Object = CASE 19

2 groups after combining group 3 (n = 4 ) and group 4 (n = 7) error = 17.198
Group 1 (n= 9)
Object = CASE 1
Object = CASE 2
Object = CASE 3
Object = CASE 6
Object = CASE 7
Object = CASE 14
Object = CASE 15
Object = CASE 16
Object = CASE 17
Group 4 (n= 11)
Object = CASE 4
Object = CASE 5
Object = CASE 8
Object = CASE 9
Object = CASE 10
Object = CASE 11
Object = CASE 12
Object = CASE 13
Object = CASE 18
Object = CASE 19
Object = CASE 20

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SCATTERPLOT - Plot of Error vs No. of Groups



### Grouping Errors in Hierarchical Clustering

If you compare the results above with a discriminant analysis on the same data, you will see that the clustering procedure does not necessarily replicate the original groups. Clearly, “nearest neighbor” grouping in Euclidean space does not necessarily result in the same a priori groups from the discriminant analysis.

By examining the increase in error (variance of subjects within the groups) as a function of the number of groups, one can often make some decision about the number of groups one wishes to interpret. There is a large increase in error when going from 8 groups down to 7 in this analysis which suggests there are possibly 7 or 8 groups which might be examined. If we had more information on the objects of those groups, we might see a pattern or commonality shared by objects of those groups.