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Agricolae – a free statistical library for agricultural research

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Description

Agricolae Version 1.0-3

contains a total 68 statistical routines and 33 data sets.

Agricolae is a statistical library for agricultural research with the goal of supporting developing countries. Focuses on statistical tools used in the breeding program of the International Potato Center for its main commodity crops, potato and sweetpotato. Thus, Agricolae supports a variety of field trial designs, including incomplete block designs techniques, genetic designs, stability analysis, AMMI with biplot and triplot analysis, multiple comparisons of treatments Other functions include the construction of consensus clusters, optimal size and shape of experimental field plots.

Agricolae was developed using R and is available via the CRAN, repository at <http://www.r-project.org>.

Planning of field experiment

Randomize and field book

Alpha design, Graeco, latin square, CRD, RCBD , BIB.

The planning of field experiments is one of the main areas of Agricolae.

It supports simple lattice design ([lattice.simple](#)),

Factorial a block design ([design.ab](#)),

Alpha design ([alpha.design](#)),

Balanced Incomplete Block Design ([design.bib](#)),

Randomized complete block design ([design.rcbd](#)).

Complete randomized design ([design.crd](#)),

Graeco-latin square design ([design.graeco](#)),

Latin square design ([design.lsd](#)).

Planning of field experiment

Greaco latin

```
args: trt1, trt2, number = 1, seed = 0, kinds = "Super-Duper"
```

```
> T1<-c("a", "b", "c", "d")
```

```
> T2<-c("v", "w", "x", "y")
```

```
> Plan <- design.greaco(T1, T2, number=101)
```

Plots					Treatments				
	[,1]	[,2]	[,3]	[,4]		[,1]	[,2]	[,3]	[,4]
[1,]	101	102	103	104	[1,]	"d w"	"b v"	"a x"	"c y"
[2,]	105	106	107	108	[2,]	"b y"	"d x"	"c v"	"a w"
[3,]	109	110	111	112	[3,]	"a v"	"c w"	"d y"	"b x"
[4,]	113	114	115	116	[4,]	"c x"	"a y"	"b w"	"d v"

It's not possible to construct: 6,10 and pair ≥ 14

Planning of field experiment

Alpha design

(trt, k, r, number = 1, seed = 0, kinds = "Super-Duper")

```
> Trt <- letters[1:12]
> plan<-design.alpha(trt,k=3, r=2, number=101)
```

```
alpha design (0,1) - Serie I
```

```
Parameters Alpha design
```

```
=====
```

```
treatmeans : 12
```

```
Block size : 3
```

```
Blocks      : 4
```

```
Replication: 2
```

```
Efficiency factor
```

```
(E ) 0.6470588
```

```
<<< Book >>>
```

Field Book

```
> plan
```

	plots	cols	block	trt	replication
1	101	1	1	j	1
2	102	2	1	h	1
3	103	3	1	c	1
4	104	1	2	d	1
...					
23	123	2	8	h	2
24	124	3	8	e	2

Comparison of multiple treatments

Test:

LSD, HSD, Waller, Durbin, Kruskal Wallis, Friedman, Waerden

Test parametrics:

LSD: Least significant difference and Adjust P-values

HSD: Honestly significant difference Tukey.

Waller: Bayesian t-values for multiple comparisons

Test Non parametrics

Kruskal Wallis: Complete randomized design

Friedman: Randomized complete block design

Durbin: Balanced Incomplete Block Design

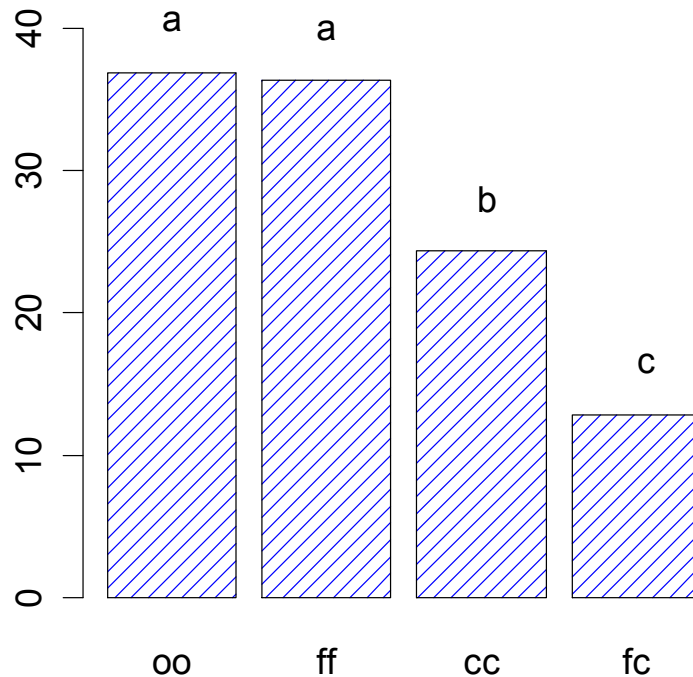
Waerden: The van der Waerden (Normal Scores)

Comparison of multiple treatments

Waller-Duncan

(y, trt, DFerror, MSerror, Fc, K = 100, group = TRUE, main = NULL)

```
> attach(sweetpotato)
> model<-aov(yield~virus)
> comparison <- waller.test(yield, virus, DFerror=8, MSerror=22.49, Fc=17.345)
> bar.group(comparison, horiz=FALSE, ylim=c(0, 45), density=10, col="blue")
```



Critical Value of Waller **2.236**
Minimum Significant Difference
8.658066

Means with the same letter are not significantly different.

Groups, Treatments and means

a	oo	36.9
a	ff	36.33333
b	cc	24.4
c	fc	12.86667

Comparison of multiple treatments

LSD

(y, trt, DFerror, MSerror, alpha = 0.05, p.adj = c("none", "holm", hochberg", "bonferroni", "BH", "BY", "fdr"), group = TRUE, main = NULL))

```
> comparison <- LSD.test(yield, virus, DFerror=8, MSerror=22.49,
p.adj="bonferroni")
```

group = TRUE

```
LSD t Test for yield
P value adjustment method: bonferroni
.....
Alpha                0.050000
Error Degrees of Freedom 8.000000
Error Mean Square     22.490000
Critical Value of t   3.478879
```

Least Significant Difference 13.47065

Means with the same letter are not significantly different.

Groups, Treatments and means

a	oo	36.9
a	ff	36.33333
ab	cc	24.4
b	fc	12.86667

group = FALSE

Treatment Means

	virus	yield	std.err	replication
1	cc	24.40000	2.084067	3
2	fc	12.86667	1.246774	3
3	ff	36.33333	4.233727	3
4	oo	36.90000	2.482606	3

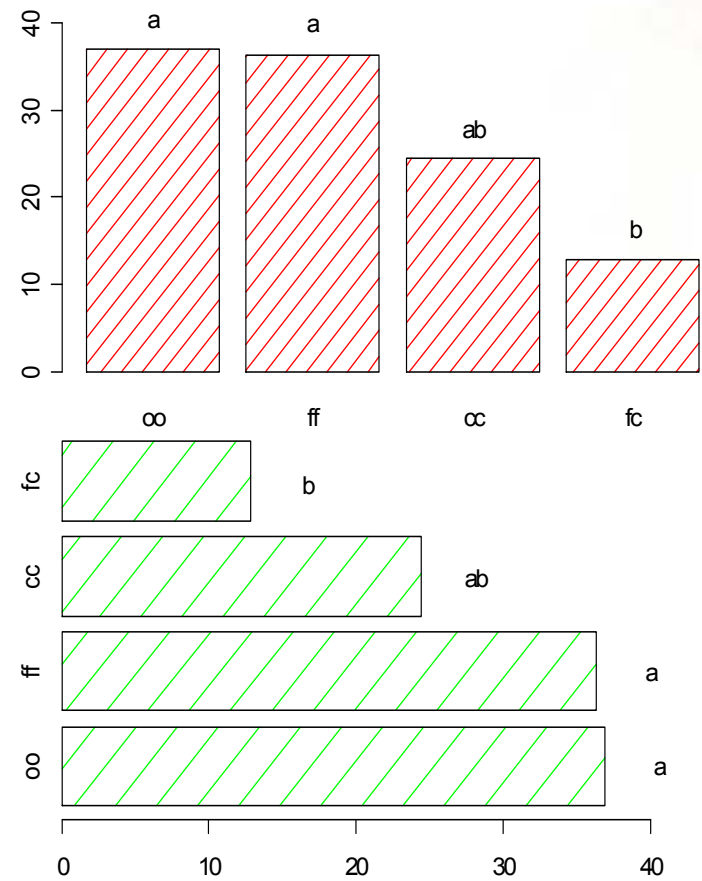
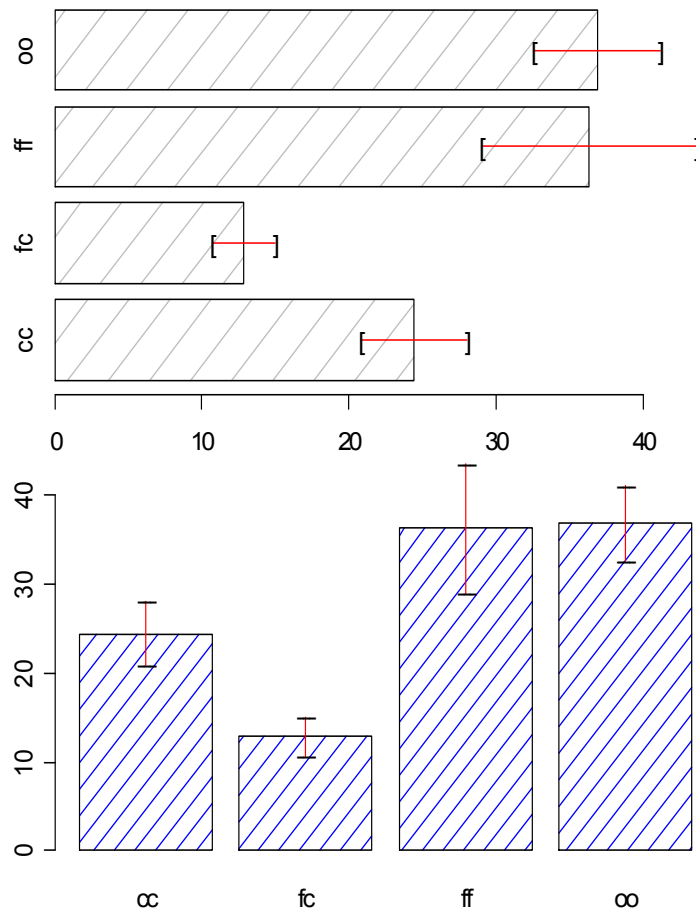
Comparison between treatments means

	tr.i	tr.j	diff	pvalue
1	1	2	11.5333333	0.1056
2	1	3	11.9333333	0.0900
3	1	4	12.5000000	0.0720
4	2	3	23.4666667	0.0024
5	2	4	24.0333333	0.0012
6	3	4	0.5666667	1.0000

Comparison of multiple treatments

Graphics.

bar.err & bar.group



Stability analysis

AMMI, `stability.par`, `stability.nonpar`

AMMI: Additive Main Effects and Multiplicative Interaction models are widely used to analyze main effects and genotype by environment (GEN, ENV) interactions in multilocation variety trials. Furthermore, this function generates biplot and triplot graphs as well as principal component analysis.

`stability.par`: SHUKLA'S STABILITY VARIANCE AND KANG'S. This procedure calculates the stability variations as well as the statistics of selection for the yield and the stability

`stability.nonpar`: A method based on the statistical ranges of the study variable per environment for the stability analysis

Stability analysis

AMMI

(ENV, GEN, REP, Y, MSE=0, number=TRUE, graph="biplot", ...)

```
> model<- AMMI(ltrv[,2], ltrv[,1], ltrv[,3], ltrv[,5],  
xlim=c(-3,3),ylim=c(-4,4), graph="biplot",number=FALSE)
```

```
ANALYSIS AMMI: ltrv[, 5]
```

```
Class level information
```

```
ENV: Ayac LM-02 SR-02 Hyo-02 LM-03 SR-03
```

```
GEN: 102.18 104.22 121.31 141.28 157.26 163.9 221.19 233.11 235.6
```

```
241.2 255.7 314.12 317.6 319.20 320.16 342.15 346.2 351.26 364.21
```

```
402.7 405.2 406.12 427.7 450.3 506.2 Canchan Desiree Unica
```

```
REP: 1 2 3
```

```
Number of observations: 504
```

```
model Y: ltrv[, 5] ~ ENV + REP%in%ENV + GEN + ENV:GEN
```

```
Random effect REP%in%ENV
```

Stability analysis

AMMI

(ENV, GEN, REP, Y, MSE=0, number=TRUE, graph="biplot", ...)

Analysis of Variance Table

Response: Y

	Df	Sum Sq	Mean Sq	F value	Pr(>F)	
ENV	5	9607.4	1921.5	284.6352	4.957e-12	***
REP(ENV)	12	81.0	6.8	2.7313	0.00154	**
GEN	27	1367.4	50.6	20.4904	< 2.2e-16	***
ENV:GEN	135	1764.8	13.1	5.2891	< 2.2e-16	***
Residuals	324	800.8	2.5			

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Coeff var	Mean ltrv[, 5]
20.07525	7.831188

Analysis

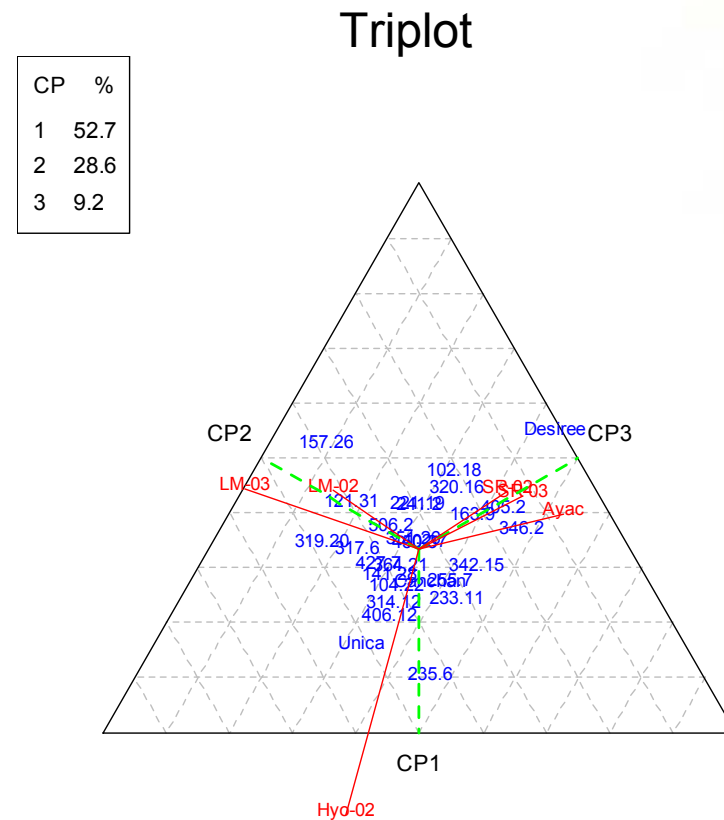
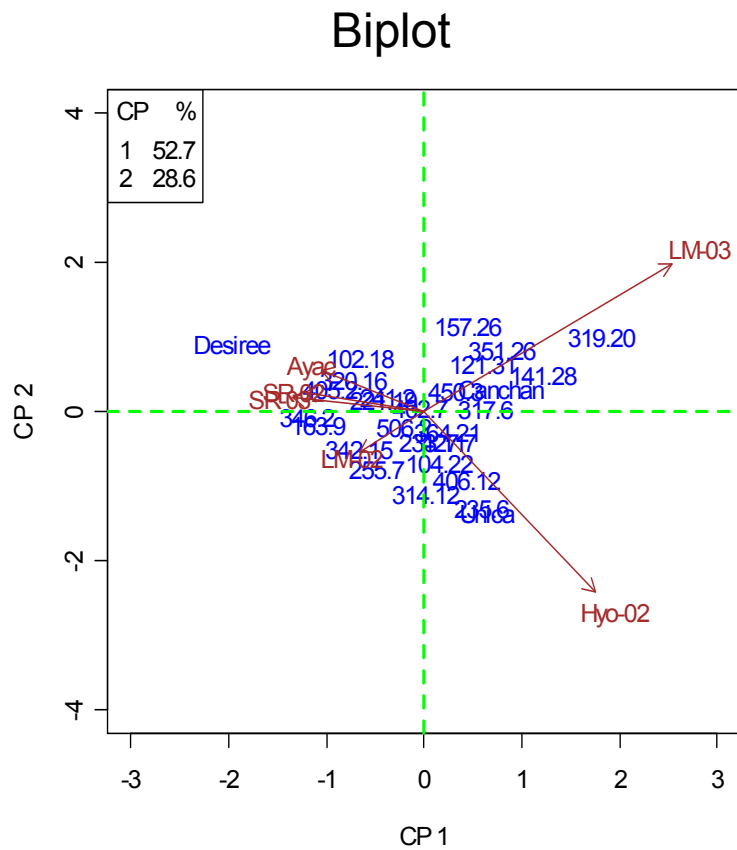
	percent	acum	Df	Sum.Sq	Mean.Sq	F.value	Pr.F
CP1	52.7	52.7	31	929.89935	29.996753	12.14	0.0000
CP2	28.6	81.3	29	503.95903	17.377898	7.03	0.0000

... More ...

Stability analysis

AMMI

(ENV, GEN, REP, Y, MSE=0, number=TRUE,graph="biplot",...)



Stability analysis

AMMI.contour

(model, distance, shape, ...)

Limit, radio: 1.645421

Genotype in: 31

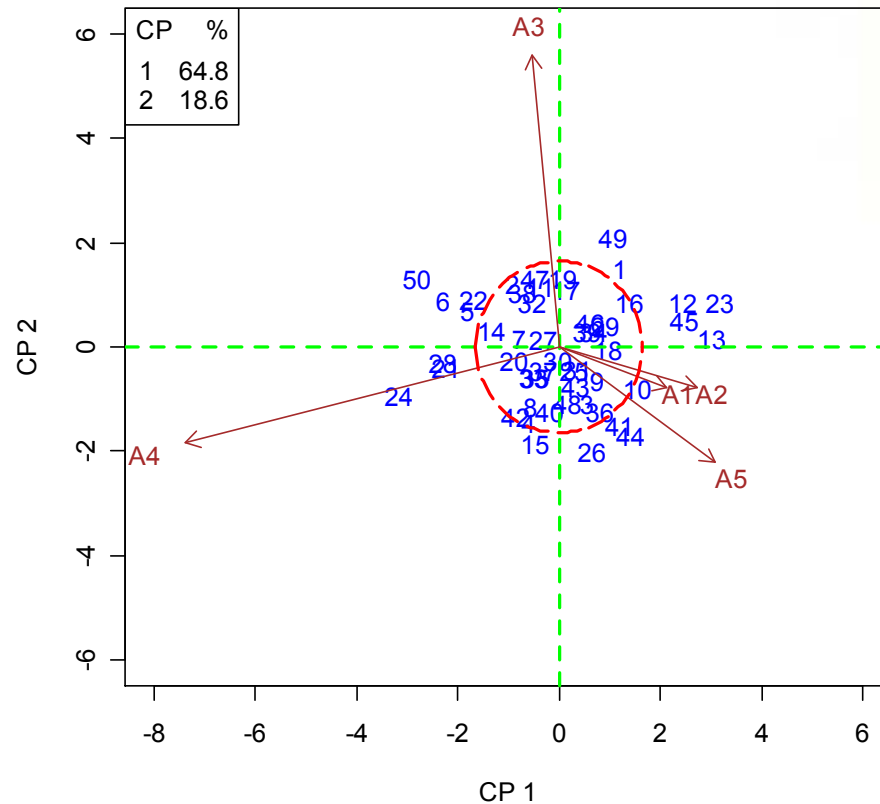
Genotype out: 19

GENOTYPE IN:

"10" "11" "12" "15" "16" "17" "19"
"21" "24" "25" "26" "27" "31" "33"
"35" [16] "36" "37" "38" "39" "4"
"40" "41" "42" "43" "44" "45" "47"
"48" "50" "6" "7"

GENOTYPE OUT:

"1" "13" "14" "18" "2" "20" "22"
"23" "28" "29" "3" "30" "32" "34"
"46" "49" "5" "8" "9"



Stability analysis

AMMI

`(ENV, GEN, REP, Y, MSE=0, number=TRUE, graph="biplot", ...)`

Input data:

a) complete or missing value. Experiments in localities under randomized complete block design.

Or

b) Only means and missing value. Estimation variance of error and replication:

MSE = variance error = Mean square error

Rep = constant = Harmonic Mean (r_1, r_2, \dots, r_k)

Stability analysis (parametric)

stability.par

```
(data, rep, MSerror, alpha = 0.1, main = NULL, cova = F,  
name.cov = NULL, file.cov = 0)
```

```
> stability.par(data, rep=4, MSerror=1.8, alpha=0.1, main="Genotype")
```

```
INTERACTIVE PROGRAM FOR CALCULATING SHUKLA'S STABILITY VARIANCE AND  
KANG'S
```

```
YIELD - STABILITY (YSi) STATISTICS
```

```
Genotype
```

```
Environmental index - covariate
```

```
Analysis of Variance
```

```
-----  
Source          d.f.    Sum of Squares  Mean Squares      F  
-----  
TOTAL           155     2121.2544  
GENOTYPES       12      101.0877         8.4240         3.31 *  
ENVIRONMENTS    11     1684.3067       153.1188       85.07 **  
INTERACTION     132     335.8600         2.5444         1.41 **  
HETEROGENEITY  12       34.7256         2.8938         1.15 ns  
RESIDUAL        120     301.1344         2.5095         1.39 *  
POOLED ERROR    432         1.8000
```

Stability analysis (parametric)

stability.par

(data, rep, MSerror, alpha = 0.1, main = NULL, cova = F, name.cov = NULL, file.cov = 0)

Simultaneous selection for yield and stability (++)

	Genotype	Yield	Rank	Adj.rank	Adjusted	Stab.var	Stab.rating	YSi	...
1	A	7.383333	11	1	12	2.134311	0	12	+
2	B	6.783333	2	-1	1	1.672824	0	1	
3	C	7.250000	9	1	10	0.805606	0	10	+
4	D	6.783333	2	-1	1	2.919766	-2	-1	
5	E	7.075000	7	-1	6	1.604036	0	6	+
6	F	6.916667	6	-1	5	3.924945	-2	3	
7	G	7.808333	12	2	14	4.043485	-2	12	+
8	H	7.908333	13	2	15	2.899022	-2	13	+
9	I	7.275000	10	1	11	4.251970	-2	9	+
10	J	7.083333	8	-1	7	1.853320	0	7	+
11	K	6.433333	1	-2	-1	2.167039	0	-1	
12	L	6.891667	5	-1	4	1.692631	0	4	
13	M	6.791667	4	-1	3	3.108168	-2	1	

Yield Mean: 7.10641

YS Mean: 5.846154

LSD (0.05): 0.4514298

- - - - -

+ selected genotype

++ Reference: Kang, M. S. 1993. Simultaneous selection for yield and stability: Consequences for growers. Agron. J. 85:754-757

Stability analysis (Non-parametric)

Haynes K G, Lambert D H, Christ B J, Weingartner D P, Douches D S, Backlund J E, Fry W and Stevenson W. 1998. Phenotypic stability of resistance to late blight in potato clones evaluated at eight sites in the United States American Journal Potato Research 75, pag 211-217.

`Stability.nonpar(data, variable=NULL, ranking = FALSE)`

```
> haynes
```

```
      clone  FL  MI  ME ...  
1  A84118-3 284 1113 1053 ...  
2  AO80432-1 254  690 1112 ...  
3  AO84275-3 395 1089 1090 ...  
4  AWN86514-2 136  296  374 ...  
5   B0692-4  87  653  412 ...  
6   B0718-3 130  126  329 ...  
... ..
```

```
stability.nonpar(haynes,"YIELD",ranking=TRUE)
```

```
Nonparametric Method for Stability Analysis
```

```
-----  
Estimation and test of nonparametric measures
```

```
Variable: YIELD
```

```
Ranking...
```

```
              FL  MI  ME  MN  ND   NY  PA  WI  
A84118-3      7 11 11 14  8 14.0 12 11  
AO80432-1      6  9 13 13 12 12.0 15 14  
AO84275-3     10 10 12  8  9  7.0 11 12  
AWN86514-2     3  3  3  1  3  3.0  2  1  
B0692-4        1  8  4  3  2  2.0  1  3  
B0718-3        2  1  2  2  4  4.0  3  4  
...
```

Stability analysis Non-parametric

Haynes K G, Lambert D H, Christ B J, Weingartner D P, Douches D S, Backlund J E, Fry W and Stevenson W. 1998. Phenotypic stability of resistance to late blight in potato clones evaluated at eight sites in the United States American Journal Potato Research 75, pag 211-217.

Stability.nonpar(data, variable=NULL, ranking = FALSE)

Statistics...

	Mean Rank	s1	Z1	s2	Z2	
A84118-3	741.62	13	4.82	0.22	16.70	0.34
AO80432-1	734.38	12	6.21	0.73	26.57	0.47
AO84275-3	635.88	9	6.20	0.70	28.53	0.87
...						
Sum of Z1:	20.08986					
Sum of Z2:	25.84532					

Test...

The Z-statistics are measures of stability. The test for the significance of the sum of Z1 or Z2 are compared to a Chi-Square value of chi.sum. individual Z1 or Z2 are compared to a Chi-square value of chi.ind.

MEAN	es1	es2	vs1	vs2	chi.ind	chi.sum
561.4609	5.3125	21.25	1.111905	60.75223	8.733011	26.29623

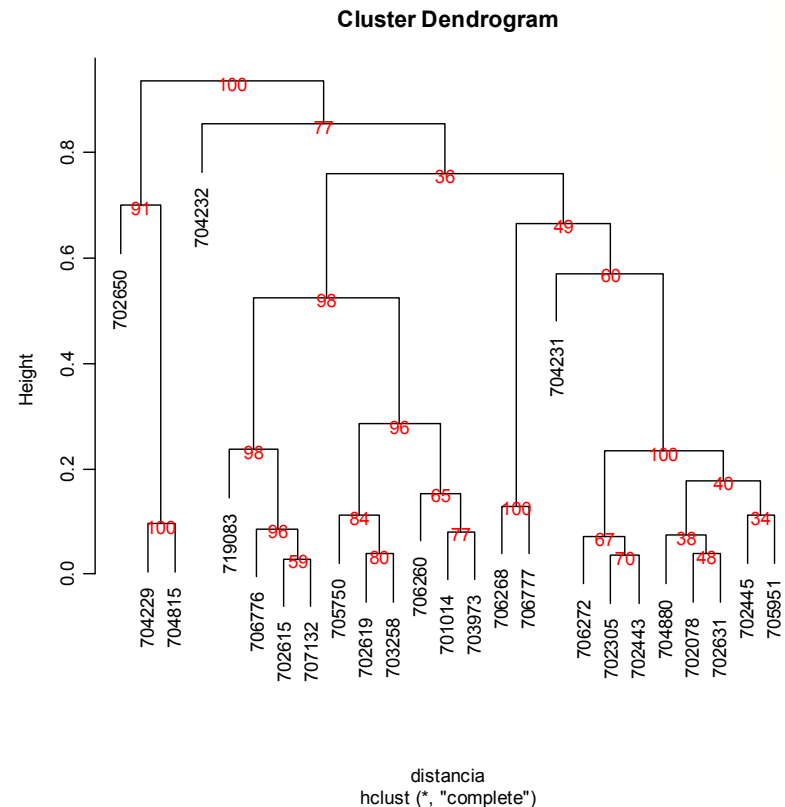
Consensus cluster

Methods distance and clustering of R, functions `dist()` and `hclust()`.

```
(data, distance = c("binary", ..), method = c("complete", ..),  
nboot = 500, duplicate = TRUE, cex.text = 1, col.text = "red",  
...)
```

```
output<-consensus( pamCIP,distance="binary", method="complete", nboot=500)
```

```
Duplicates: 18  
New data : 25 Records  
Consensus hclust  
Method distance: binary  
Method cluster : complete  
rows and cols : 25 107  
n-boostrap : 500  
  
Run time : 16.281 secs
```



Consensus cluster

Methods distance and clustering of R, functions `dist()` and `hclust()`.

```
(data, distance = c("binary", ..), method = c("complete", ..),  
nboot = 500, duplicate = TRUE, cex.text = 1, col.text = "red",  
...)
```

OUTPUT

```
> names(output)  
[1] "table.dend" "dendrogram" "duplicates"
```

to reproduce dendrogram

```
dend<-output$dendrogram  
data<-output$table.dend  
plot(dend)  
text(data[,3],data[,4],data[,5],col="blue",cex=1)
```

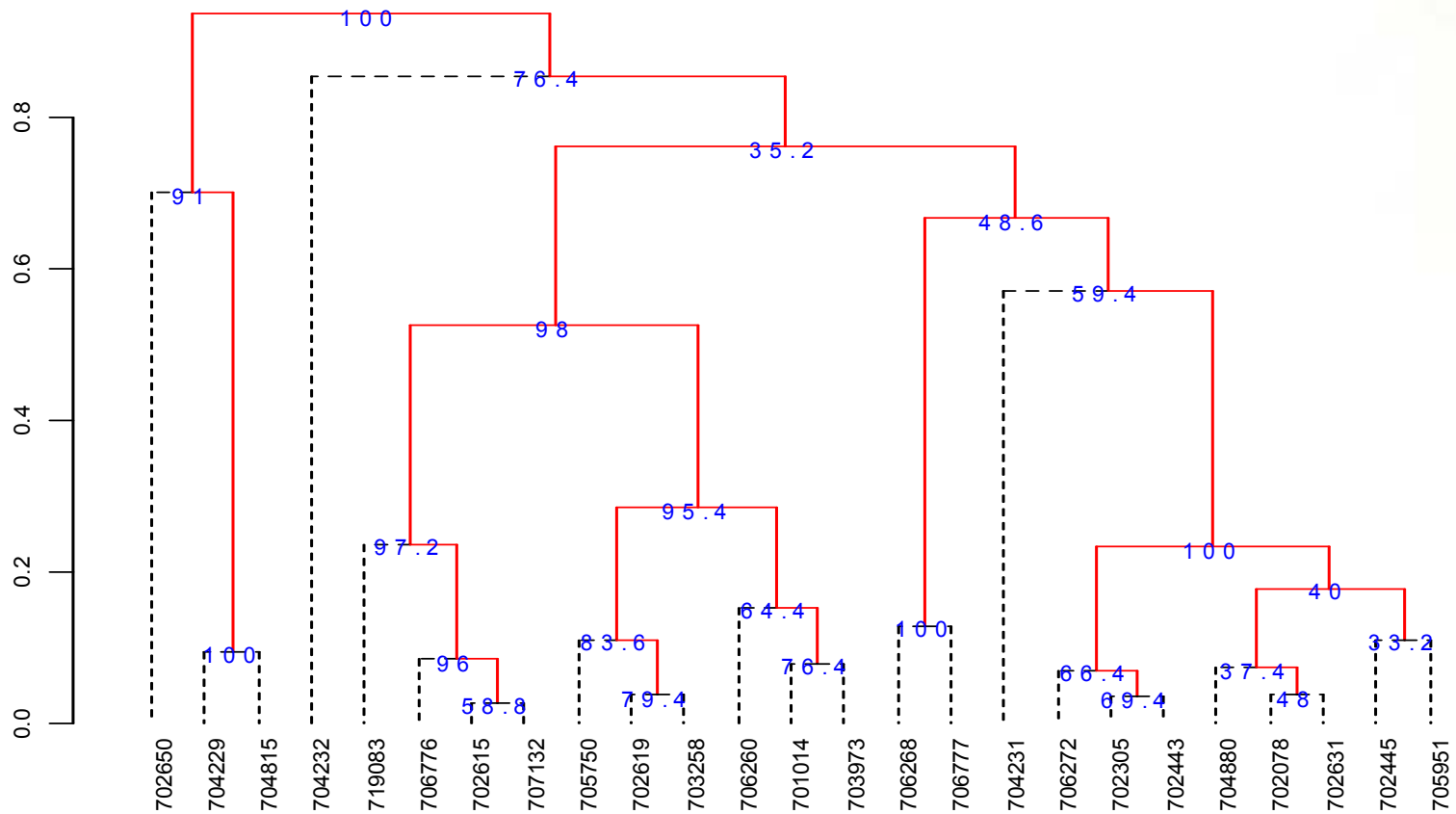
classical dendrogram

```
dend<-as.dendrogram(output$dendrogram)  
plot(dend,type="r",edgePar = list(lty=1:2, col=2:1))  
text(data[,3],data[,4],data[,5],col="blue",cex=1)
```

Consensus cluster

Methods distance and clustering of R, functions `dist()` and `hclust()`.

```
(data, distance = c("binary", ..), method = c("complete", ..),  
nboot = 500, duplicate = TRUE, cex.text = 1, col.text = "red",  
...)
```



Consensus cluster

Input: output consensus

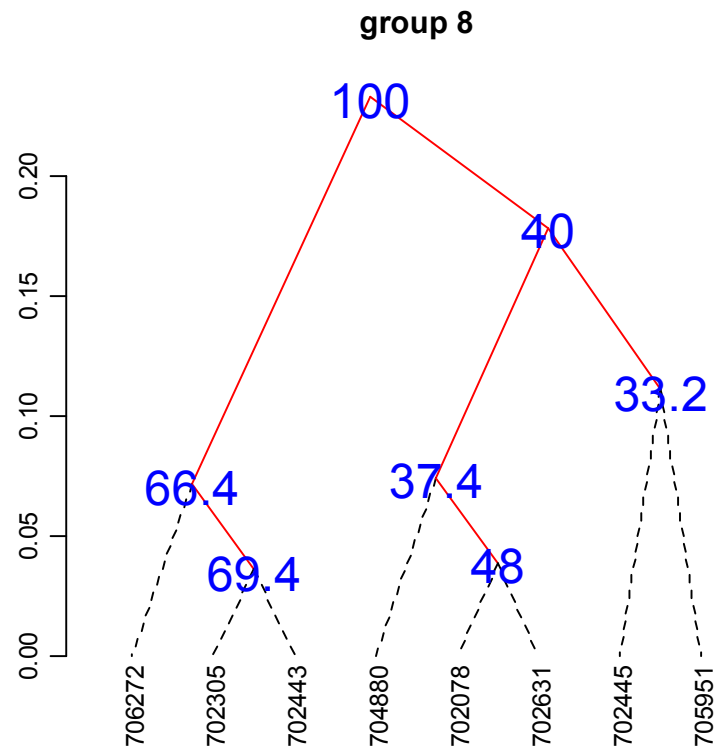
hcut()

```
(consensus, h, group, col.text = "blue", cex.text = 1, ...)
```

```
hcut(output, h=0.4, group=8, type="t", edgePar = list(lty=1:2,  
col=2:1), main="group 8", col.text="blue", cex.text=2)
```

numbers

1	1
2	2
3	1
4	4
5	6
6	2
7	1
8	8



Soil uniformity

```
Index.smith(data, ...)
```

```
table<-index.smith(rice, type="l",lty=4, lwd=3,  
main="Relationship between CV\n per unit area and plot  
size",col="red")
```

Smith's index of soil heterogeneity is used primarily to derive optimum plot size. The index gives a single value as a quantitative measure of soil heterogeneity in an area. The coefficient of variance is used to determine plot size and shape

```
> table
```

```
$model
```

```
lm(formula = CV ~
```

```
I(log(x)))
```

```
Coefficients:
```

```
(Intercept)
```

```
12.4782
```

```
I(log(x))
```

```
-0.7009
```

```
$uniformity
```

```
Size Width Length plots
```

```
Vx
```

```
CV
```

```
[1,] 1 1 1 648 9044.539 13.0
```

```
[2,] 2 1 2 324 7816.068 12.1
```

```
[3,] 2 2 1 324 7831.232 12.1
```

```
[4,] 3 1 3 216 7347.975 11.7
```

```
[5,] 3 3 1 216 7355.216 11.7
```

```
...
```

```
[40,] 162 9 18 4 4009.765 8.6
```

Soil uniformity

```
Index.smith(data, ...)
```

```
table<-index.smith(rice, type="1",lty=4, lwd=3,  
main="Relationship between CV\n per unit area and plot  
size",col="red")
```

```
predict(table$model, new=data.frame(x=30))
```

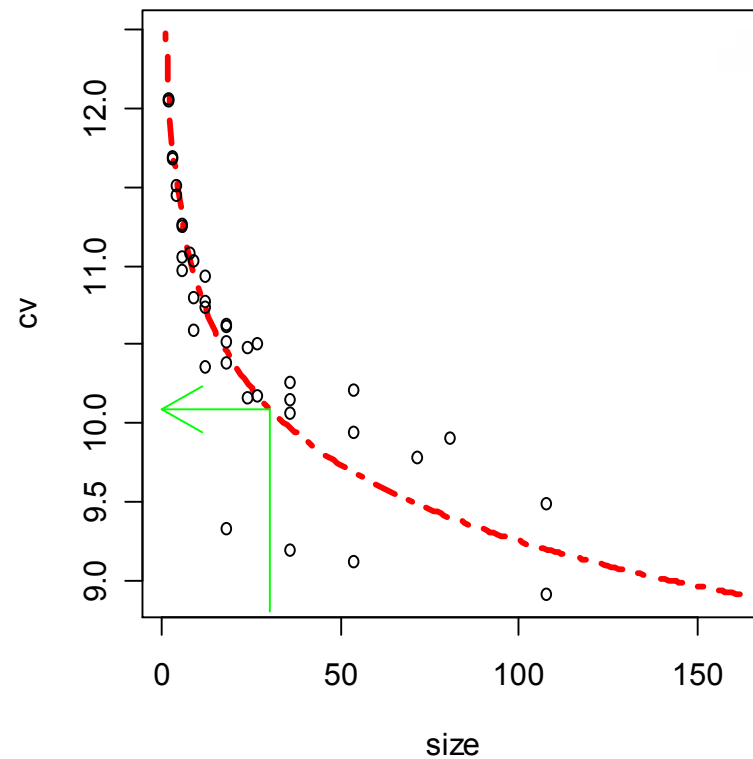
[1] 10.09436

If plot size = 30 unit ²
then CV = 10 %

rice

	V1	V2	V3	V4	V5
1	842	844	808	822	970
2	803	841	870	970	940
3	773	782	860	822	930
4	912	887	815	937	840
5	874	792	803	793	810
6	908	875	899	788	860
7	875	907	921	963	870
8	891	928	871	875	860
9	823	784	754	873	760

Relationship between CV per unit area and plot size



Other functions and data sets

Genetic design: north carolina design, line x tester.
Biodiversity index and confidence interval.
Descriptive statistical: cross tabulations, ...
Model: simulation and resampling.

Data sets main in package 'agricolae':

ComasOxapampa	Data AUDPC Comas - Oxapampa
Glycoalkaloids	Data Glycoalkaloids
RioChillon	Data and analysis Mother and baby trials
clay	Data of Ralstonia population in clay soil
disease	Data evaluation of the disease overtime
huasahuasi	Data of yield in Huasahuasi
melon	Data of yield of melon in a Latin square experiment
natives	Data of native potato
pamCIP	Data Potato Wild
paracsho	Data of Paracsho biodiversity
ralstonia	Data of population bacterial Wilt: AUDPC
soil	Data of soil analysis for 13 localities
sweetpotato	Data of sweetpotato yield
trees	Data of species trees. Pucallpa
wilt	Data of Bacterial Wilt (AUDPC) and soil

Agricolae Version 1.0-4

Please note that there is a new version of the agricolae on the link below

<http://tarwi.lamolina.edu.pe/~fmendiburu>