

# Package ‘PaolaR6Nuevo’

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**Type** Package

**Title** Confusion Matrix

**Version** 0.1.0

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**Description** This package provides useful functions for the analysis of confusion matrices in classification problems. Includes methods to calculate overall accuracy, user accuracy, and map creator accuracy.

**Imports** R6, Rdpack

**RdMacros** Rdpack

**License** GPL

**Encoding** UTF-8

**LazyData** true

**RoxygenNote** 7.2.3

**Suggests** knitr,  
rmarkdown,  
testthat (>= 3.0.0)

**Config/testthat/edition** 3

**VignetteBuilder** knitr

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MatCon	<i>Confusion matrix</i>
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## Description

Using the confusion matrix, various indices are calculated.

## Details

List of possible errors:

- Error type 1: Non-square matrix.
- Error type 2: Single element matrix.
- Error type 3: negative values.
- Error type 4: Sum of elements 0.
- Error type 5: Sum of rows 0.
- Error type 6: Sum of columns 0.
- Error type 7: It is not a matrix.

## Value

Object of class MatCon or an error if a matrix is not entered.

## Methods

### Public methods:

- `MatCon$new()`
- `MatCon$OverallAcc()`
- `MatCon$UserAcc()`
- `MatCon$UserAcc_i()`
- `MatCon$ProdAcc()`
- `MatCon$ProdAcc_i()`
- `MatCon$AvUserProdAcc_i()`
- `MatCon$Sucess()`
- `MatCon$Sucess_i()`
- `MatCon$AvHelldenAcc_i()`
- `MatCon$ShortAcc_i()`
- `MatCon$UserKappa_i()`
- `MatCon$ProdKappa_i()`
- `MatCon$ModKappa()`
- `MatCon$ModKappaUser_i()`
- `MatCon$ModKappaProd_i()`
- `MatCon$EntropUser_i()`
- `MatCon$EntropProd_i()`
- `MatCon$AvUserAcc()`
- `MatCon$AvProdAcc()`

- `MatCon$AvUserProdAcc()`
- `MatCon$AvHelldenAcc()`
- `MatCon$AvShortAcc()`
- `MatCon$CombUserAcc()`
- `MatCon$CombProdAcc()`
- `MatCon$CombUserProdAcc()`
- `MatCon$Kappa()`
- `MatCon$Entrop()`
- `MatCon$NormEntropUser()`
- `MatCon$NormEntropProd()`
- `MatCon$AvNormEntrop()`
- `MatCon$GeomAvNormEntrop()`
- `MatCon$AvMaxNormEntrop()`
- `MatCon$Tau()`
- `MatCon$UserProdAcc()`
- `MatCon$DetailedKappa()`
- `MatCon$DetailedCondKappa()`
- `MatCon$QES()`
- `MatCon$MTypify()`
- `MatCon$AllParameters()`
- `MatCon$MBootStrap()`
- `MatCon$MNormalize()`
- `MatCon$MPseudoZeroes()`
- `MatCon$DetailedWTau()`
- `MatCon$DetailedWKappa()`
- `MatCon$UserProdAcc_W()`
- `MatCon$clone()`

**Method new():** Public method to create an instance of the MatCon class. When creating it, values must be given to the matrix. The optional possibility of adding metadata to the matrix is offered. The creation includes a series of checks on the data that, if not met, give coded error messages. The values of the matrix must be organized in such a way that the columns represent the categories in the reference and the rows represent the categories in the product being evaluated.

*Usage:*

```
MatCon$new(values, ID = NULL, Date = NULL, Source = NULL)
```

*Arguments:*

values Confusion matrix

ID Identifier. By default, the date in YYYYMMDD format will be taken as the ID.

Date Date provided by the user. By default the date provided by the system will be taken.

Source Indicates where the matrix comes from (article, project, etc.). By default is NULL.

*Returns:* Object of class MatCon or an error if a matrix isn't entered.

*Examples:*

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
```

```
mc <- MatCon$new (A,ID=5,Date="27-10-2023",Source="Congalton, R.G., & Green, K. (2008). Assessin
```

**Method OverallAcc():** Public method to calculate the global index called Overall accuracy. The Overall accuracy for a particular classified image/map is then calculated by dividing the sum of the entries that form the major diagonal (i.e., the number of correct classifications) by the total number of samples taken. The method also offers the variance. The reference (Congalton and Green 2008) is followed for the computations.

The mathematical expression is:

$$OverallAcc = \frac{\sum_{i=1}^n x_{ii}}{\sum_{i,j=1}^n x_{ij}}$$

$$\sigma_{OverallAcc}^2 = \frac{OverallAcc \cdot (1 - OverallAcc)}{N}$$

Where:

1. OverallAcc: overall accuracy.
2. x<sub>ii</sub>: diagonal element of the matrix.
3. x<sub>ij</sub>: element of the matrix.
4. N: number of cases involved in the calculation of the index.

*Usage:*

MatCon\$OverallAcc()

*Returns:* Overall accuracy and variance as a list.

*Examples:*

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$OverallAcc()
```

**Method UserAcc():** Public method for deriving a class index called user's accuracy. The user's accuracy for the class i of thematic map is calculated by dividing the value in the diagonal of class i by the sum of all values in the row of the class i. The method also offers the variance. The reference (Congalton and Green 2008) is followed for the computations.

The mathematical expression is:

$$UserAcc = \frac{x_{ii}}{\sum_{j=1}^n x_{ij}}$$

$$\sigma_{UserAcc}^2 = \frac{UserAcc \cdot (1 - UserAcc)}{N}$$

where:

1. UserAcc: user accuracy.
2. x<sub>ii</sub>: diagonal element of the matrix.
3. x<sub>ij</sub>: element of the matrix.
4. N: number of cases involved in the calculation of the index.

*Usage:*

MatCon\$UserAcc()

*Returns:* A list with a vector of values for the user's accuracy index of all classes and another vector with their variances.

*Examples:*

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$UserAcc()
```

**Method** UserAcc\_i(): Public method where the user's accuracy index is defined for a specific class i. The user precision for class i of the thematic map is calculated by dividing the value on the diagonal of class i by the sum of all values in the row of class i. The method also offers variance. The reference (Congalton and Green 2008) is followed for the calculations.

$$UserAcc_i = \frac{x_{ii}}{\sum_{j=1}^n x_{ij}}$$

$$\sigma_{UserAcc_i}^2 = \frac{UserAcc_i \cdot (1 - UserAcc_i)}{N}$$

where:

1. UserAcc\_i: user accuracy index for class i.
2. x<sub>ii</sub>: diagonal element of the matrix.
3. x<sub>ij</sub>: element of the matrix.
4. N: number of cases involved in the calculation of the index.

*Usage:*

```
MatCon$UserAcc_i(i)
```

*Arguments:*

i User class to evaluate

*Returns:* A list of the user's accuracy index values for class i and its variance.

*Examples:*

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$UserAcc_i(2)
```

**Method** ProdAcc(): Public method for deriving a class index called producer's accuracy. The producer's accuracy for the class i of thematic map is calculated by dividing the value in the diagonal of class i by the sum of all values in the column of the class i. The method also offers the variance. The reference (Congalton and Green 2008) is followed for the computations.

$$ProdAcc = \frac{x_{jj}}{\sum_{j=1}^n x_{ij}}$$

$$\sigma_{ProdAcc}^2 = \frac{ProdAcc \cdot (1 - ProdAcc)}{N}$$

where:

1. ProdAcc: producer accuracy.
2. x<sub>jj</sub>: diagonal element of the matrix.
3. x<sub>ij</sub>: element of the matrix.
4. N: number of cases involved in the calculation of the index.

*Usage:*

```
MatCon$ProdAcc()
```

**Returns:** A list with a vector of values for the producer's accuracy index of all classes and another vector with their variances.

**Examples:**

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$ProdAcc()
```

**Method ProdAcc\_i():** Public method where the producer's accuracy index is defined for a specific class i. The user precision for class i of the thematic map is calculated by dividing the value on the diagonal of class i by the sum of all values in the column of class i. The method also offers variance. The reference (Congalton and Green 2008) is followed for the calculations.

$$ProdAcc_i = \frac{x_{jj}}{\sum_{j=1}^n x_{ij}}$$

$$\sigma_{ProdAcc_i}^2 = \frac{ProdAcc_i \cdot (1 - ProdAcc_i)}{N}$$

where:

1. ProdAcc\_i: producer accuracy index for class i.
2. x<sub>jj</sub>: diagonal element of the matrix.
3. x<sub>ij</sub>: element of the matrix.
4. N: number of cases involved in the calculation of the index.

**Usage:**

```
MatCon$ProdAcc_i(i)
```

**Arguments:**

- i Producer class to evaluate.

**Returns:** A list of the producer's accuracy index values for class i and its variance

**Examples:**

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$ProdAcc_i(1)
```

**Method AvUserProdAcc\_i():** Public method that provides the average of the accuracy rates of the user and producer of a specific class. The method also offers variance. The reference (Liu et al. 2007) is followed for the calculations.

The mathematical expression is:

$$AvUserProdAcc_i = \frac{UserAcc_i + ProdAcc_i}{2}$$

$$\sigma_{AvUserProdAcc_i}^2 = \frac{AvUserProdAcc_i \cdot (1 - AvUserProdAcc_i)}{N}$$

where:

1. AvUserProdAcc\_i: average of user's and producer's accuracy.
2. UserAcc\_i: user accuracy index for class i.
3. ProdAcc\_i: producer accuracy index for class i.
4. N: number of cases involved in the calculation of the index.

*Usage:*

MatCon\$AvUserProdAcc\_i(i)

*Arguments:*

i Class to evaluate.

*Returns:* A list with average of user's and producer's accuracy and its variance for class i.

*Examples:*

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$AvUserProdAcc_i(2)
```

**Method** `Sucess()`: Public method that provides the Classification Success Index (CSI) applies to all class and gives an overall estimation of classification effectiveness. The reference (Koukoulas and Blackburn 2001; Turk 2002) is followed for the calculations.

The mathematical expression is:

$$Sucess = 1 - (1 - AvUserAcc + 1 - AvProdAcc) = AvUserAcc + AvProdAcc - 1$$

$$VarSucess = \frac{Sucess \cdot (1 - Sucess)}{N}$$

where:

1. Sucess: classification succes index.
2. AvUserAcc: average accuracy from user's perspective.
3. AvProdAcc: average accuracy from producer's perspective.
4. N: number of cases involved in the calculation of the index.

*Usage:*

MatCon\$Sucess()

*Returns:* A list with the classification success index and its variance.

*Examples:*

```
A<-matrix(c(0.3,0.02,0.01,0.12,0.19,0.03,0.02,0.01,0.3),nrow=3,ncol=3)
p<-MatCon$new(A,Source="Labatut, V., & Cherifi, H. (2011). Evaluation of performance measures fo
p$Sucess()
```

**Method** `Sucess_i()`: Public method that provides the Individual Classification Success Index (ICSI) applies to the classification effectiveness for one particular class of interest. The reference (Koukoulas and Blackburn 2001; Turk 2002) is followed for the calculations.

The mathematical expression is:

$$Sucess_i = 1 - (1 - UserAcc_i + 1 - ProdAcc_i) = UserAcc_i + ProdAcc_i - 1$$

$$\sigma_{Sucess_i}^2 = \frac{Sucess_i \cdot (1 - Sucess_i)}{N}$$

where:

1. Sucess\_i: individual classification success index.
2. UserAcc\_i: user accuracy index for class i.
3. ProdAcc\_i: producer accuracy index for class i.

4. N: number of cases involved in the calculation of the index.

*Usage:*

MatCon\$Sucess\_i(i)

*Arguments:*

i Class to evaluate.

*Returns:* A list with the individual classification success index and its variance.

*Examples:*

```
A<-matrix(c(0.3,0.02,0.01,0.12,0.19,0.03,0.02,0.01,0.3),nrow=3,ncol=3)
```

```
p<-MatCon$new(A,Source="Labatut, V., & Cherifi, H. (2011). Evaluation of performance measures for  
p$Sucess_i(2)
```

**Method** `AvHelldenAcc_i()`: Public method that provides the Hellden' average accuracy, denotes for the probability that a randomly chosen point of a specific class on the map has a correspondence of the same class in the same position in the field and that a randomly chosen point in the field of the same class has a correspondence of the same class in the same position on the map. The method also offers variance. The reference [6] (Helldén 1980) is followed for the calculations.

$$AvHelldenAcc_i = \frac{2}{\frac{1}{UserAcc_i} + \frac{1}{ProdAcc_i}}$$

$$\sigma_{AvHelldenAcc_i}^2 = \frac{AvHelldenAcc_i \cdot (1 - AvHelldenAcc_i)}{N}$$

where:

1. AvHelldenAcc\_i: Hellden's mean accuracy.
2. UserAcc\_i: user accuracy index for class i.
3. ProdAcc\_i: producer accuracy index for class i.
4. N: number of cases involved in the calculation of the index.

*Usage:*

MatCon\$AvHelldenAcc\_i(i)

*Arguments:*

i Class to evaluate.

*Returns:* A list with Hellden's mean accuracy and its variance.

*Examples:*

```
A <- matrix(c(148,1,8,2,0,0,50,15,3,0,1,6,39,7,1,1,0,6,25,1,1,0,0,1,6),nrow=5,ncol=5)
```

```
p<-MatCon$new(A,Source="Rosenfield, G. H., & Fitzpatrick-Lins, K. (1986). A coefficient of agreement  
p$AvHelldenAcc_i(2)
```

**Method** `ShortAcc_i()`: Public method that provides Short's mapping accuracy for each class is stated as the number of correctly classified pixels (equal to the total in the correctly classified area) in terms of all pixels affected by its classification (equal to this total in the displayed area as well as the pixels involved in errors of commission and omission). The method also offers variance. The reference [6,7] is followed for the calculations.

$$ShortAcc_i = \frac{x_{ii}}{\sum_{j=1}^n x_{+j} + \sum_{i=1}^n x_{i+} - x_{ii}}$$

$$\sigma_{ShortAcc_i}^2 = \frac{ShortAcc_i \cdot (1 - ShortAcc_i)}{N}$$

where:



1. ShortAcc\_i: Short's mapping accuracy
2. x\_ii: diagonal element of the matrix.
3. x\_j+: sum of all elements in rows j.
4. x\_+j: sum of all elements in column j.
5. N: number of cases involved in the calculation of the index.

*Usage:*

MatCon\$ShortAcc\_i(i)

*Arguments:*

i Class to evaluate.

*Returns:* A list with Short's mapping accuracy and its variance.

*Examples:*

```
A <- matrix(c(148,1,8,2,0,0,50,15,3,0,1,6,39,7,1,1,0,6,25,1,1,0,0,1,6),nrow=5,ncol=5)
p <- MatCon$new(A,Source="Rosenfield, G. H., & Fitzpatrick-Lins, K. (1986). A coefficient of agreement for categorical data. Journal of the American Statistical Association, 81, 1074-1078.")
p$ShortAcc_i(2)
```

**Method** UserKappa\_i(): Public method that evaluates the kappa coefficient from the user's perspective, for a specific class i. The method also offers variance. The reference [6] is followed for the calculations.

$$UserKappa_i = \frac{UserAcc_i - \frac{\sum_{i=1}^n x_{i+}}{\sum_{i=1}^n \sum_{j=1}^n x_{ij}}}{1 - \frac{\sum_{i=1}^n x_{i+}}{\sum_{i=1}^n \sum_{j=1}^n x_{ij}}}$$

#'

$$\sigma_{UserKappa_i}^2 = \frac{UserKappa_i \cdot (1 - UserKappa_i)}{N}$$

where:

1. UserKappa\_i: coefficient kappa (user's).
2. UserAcc\_i: user accuracy index for class i.
3. x\_ii: diagonal element of the matrix.
4. x\_j+: sum of all elements in rows j.
5. x\_+j: sum of all elements in column j.
6. N: number of cases involved in the calculation of the index.

*Usage:*

MatCon\$UserKappa\_i(i)

*Arguments:*

i Class to evaluate.

*Returns:* A list with coefficient kappa (user's) and its variance.

*Examples:*

```
A <- matrix(c(73,13,5,1,0,21,32,13,3,0,16,39,35,29,13,3,5,7,28,48,1,0,2,3,17),nrow=5,ncol=5)
p <- MatCon$new(A,Source="Næsset, E. (1996). Conditional tau coefficient for assessment of producer's agreement. International Journal of Remote Sensing, 17, 1071-1081.")
p$UserKappa_i(2)
```

**Method** ProdKappa\_i(): Public method that evaluates the kappa coefficient from the producer's perspective, for a specific class i. The method also offers variance. The reference [6] is followed for the calculations.

$$ProdKappa_i = \frac{ProdAcc_i - \frac{\sum_{j=1}^n x_{+j} \sum_{j=1}^n x_{ij}}{\sum_{i=1}^n \sum_{j=1}^n x_{ij}}}{1 - \frac{\sum_{j=1}^n x_{+j} \sum_{j=1}^n x_{ij}}{\sum_{i=1}^n \sum_{j=1}^n x_{ij}}}$$

$$\sigma_{ProdKappa_i}^2 = \frac{ProdKappa_i \cdot (1 - ProdKappa_i)}{N}$$

where:

1. ProdKappa\_i: coefficient kappa (producer's).
2. ProdAcc\_i: producer accuracy index for class i.
3. x\_ii: diagonal element of the matrix.
4. x\_+j: sum of all elements in rows j.
5. x\_+j: sum of all elements in column j.
6. N: number of cases involved in the calculation of the index.

*Usage:*

MatCon\$ProdKappa\_i(i)

*Arguments:*

i Class to evaluate.

*Returns:* A list with coefficient kappa (producer's) and its variance.

*Examples:*

```
A<-matrix(c(73,13,5,1,0,21,32,13,3,0,16,39,35,29,13,3,5,7,28,48,1,0,2,3,17),nrow=5,ncol=5)
p<-MatCon$new(A,Source="Næsset, E. (1996). Conditional tau coefficient for assessment of produce
p$ProdKappa_i(2)
```

**Method** ModKappa(): Public method that provides the overall modified kappa coefficient. The method also offers variance. The reference [8,20] is followed for the calculations.

$$ModKappa = \frac{OverallAcc - \frac{1}{\sqrt{M}}}{1 - \frac{1}{\sqrt{M}}}$$

$$\sigma_{ModKappa}^2 = \frac{ModKappa \cdot (1 - ModKappa)}{N}$$

where:

1. ModKappa: modified coefficient kappa.
2. OverallAcc: overall accuracy.
3. M: number of elements of the matrix.
4. N: number of cases involved in the calculation of the index.

*Usage:*

MatCon\$ModKappa(i)

*Arguments:*

i Class to evaluate.

*Returns:* A list with modified coefficient kappa and its variance.

*Examples:*

```
A<-matrix(c(317,61,2,35,23,120,4,29,0,0,60,0,0,0,0,8),nrow=4,ncol=4)
p<-MatCon$new(A,Source="Foody, G. M. (1992). On the compensation for chance agreement in image cl
p$ModKappa()
```

**Method** `ModKappaUser_i()`: Public method, derived from the general modified kappa coefficient, which provides the modified coefficient kappa for the user. The method also offers variance. The reference [8,20] is followed for the calculations.

$$ModKappaUser_i = \frac{UserAcc_i - \frac{1}{\sqrt{M}}}{1 - \frac{1}{\sqrt{M}}}$$

$$\sigma_{ModKappaUser_i}^2 = \frac{ModKappaUser_i \cdot (1 - ModKappaUser_i)}{N}$$

where:

1. `ModKappaUser_i`: modified coefficient kappa (user's).
2. `UserAcc_i`: user accuracy index for class i.
3. `M`: number of elements of the matrix.
4. `N`: number of cases involved in the calculation of the index.

*Usage:*

```
MatCon$ModKappaUser_i(i)
```

*Arguments:*

i Class to evaluate.

*Returns:* A list with modified coefficient kappa (user's) and its variance.

*Examples:*

```
A<-matrix(c(0,12,0,0,12,0,0,0,0,0,0,12,0,0,12,0),nrow=4,ncol=4)
p<-MatCon$new(A,Source="Liu et al., 2007")
p$ModKappaUser_i(2)
```

**Method** `ModKappaProd_i()`: Public method, derived from the general modified kappa coefficient, which provides the modified coefficient kappa for the producer. The method also offers variance. The reference [8,20] is followed for the calculations.

$$ModKappaProd_i = \frac{ProdAcc_i - \frac{1}{\sqrt{M}}}{1 - \frac{1}{\sqrt{M}}}$$

$$\sigma_{ModKappaProd_i}^2 = \frac{ModKappaProd_i \cdot (1 - ModKappaProd_i)}{N}$$

where:

1. `ModKappaUser_i`: modified coefficient kappa (producer's).
2. `ProdAcc_i`: producer accuracy index for class i.
3. `M`: number of elements of the matrix.
4. `N`: number of cases involved in the calculation of the index.

*Usage:*

```
MatCon$ModKappaProd_i(i)
```

*Arguments:*

i Class to evaluate.

*Returns:* A list with modified coefficient kappa (producer's) and its variance.

*Examples:*

```
A<-matrix(c(317,61,2,35,23,120,4,29,0,0,60,0,0,0,0,8),nrow=4,ncol=4)
p<-MatCon$new(A,Source="Liu et al., 2007")
p$ModKappaProd_i(2)
```

**Method** `EntropUser_i()`: Public method that calculates relative change of entropy given a category on map. That is, the degree of uncertainty of the category. The method also offers variance. The reference [9] is followed for the calculations.

$$Entrop_i(A) = - \sum_{j=1}^n \left( \left( \frac{\sum_{i=1}^n x_{i+}}{\sum_{i,j=1}^n x_{ij}} \right) \cdot \log \left( \frac{\sum_{i=1}^n x_{i+}}{\sum_{i,j=1}^n x_{ij}} \right) \right)$$

$$Entrop_i(A|b_i) = - \sum_{j=1}^n \left( \left( \frac{x_{ij}}{\sum_{j=1}^n x_{+j}} \right) \cdot \log \left( \frac{x_{ij}}{\sum_{j=1}^n x_{+j}} \right) \right)$$

$$EntropUser_i = \frac{Entrop_i(A) - Entrop_i(A|b_i)}{Entrop_i(A)}$$

$$\sigma_{EntropUser_i}^2 = \frac{EntropUser_i \cdot (1 - EntropUser_i)}{N}$$

where:

1. `EntropUser_i`: relative change of entropy given a category on map.
2. `Entrop_i(A)`: Entropy of the map with respect to the category of the map.
3. `x_+`: sum of all elements in rows `j`.
4. `x_+j`: sum of all elements in column `j`.
5. `Entrop_i(A|b_i)`: Entropy of map A knowing that the location corresponding to map B is in class `b_i`.
6. `N`: number of cases involved in the calculation of the index.

*Usage:*

```
MatCon$EntropUser_i(i, v = NULL)
```

*Arguments:*

- `i` Class to evaluate (row).
- `v` Base of the logarithm. By default `v=10`. This value is used for the entropy units, `v=10`(Hartleys), `v=2`(bits), `v=e`(nats).

*Returns:* A list with the relative change of entropy given a category on map, its variance, map entropy, and entropy of map A knowing that the location corresponding to map B is in class `b_i`.

*Examples:*

```
A<-matrix(c(0,12,0,0,12,0,0,0,0,0,12,0,0,12,0),nrow=4,ncol=4)
p<-MatCon$new(A,Source="Liu et al., 2007")
p$EntropUser_i(1)
```

**Method** `EntropProd_i()`: Public method that calculates relative change of entropy given a category on ground truthing. That is, the degree of uncertainty of the category. The method also offers variance. The reference [8 o 9?] is followed for the calculations.

$$\begin{aligned}
Entrop_i(B) &= - \sum_{i=1}^n \left( \left( \frac{\sum_{j=1}^n x_{+j}}{\sum_{i,j=1}^n x_{ij}} \right) \cdot \log \left( \frac{\sum_{j=1}^n x_{+j}}{\sum_{i,j=1}^n x_{ij}} \right) \right) \\
Entrop_i(B|a_j) &= - \sum_{j=1}^n \left( \left( \frac{x_{ij}}{\sum_{i=1}^n x_{i+}} \right) \cdot \log \left( \frac{x_{ij}}{\sum_{i=1}^n x_{i+}} \right) \right) \\
EntropProd_i &= \frac{EntropMap(B) - EntropMap(B|a_j)}{EntropMap(B)} \\
\sigma_{EntropProd_i}^2 &= \frac{EntropProd_i \cdot (1 - EntropProd_i)}{N}
\end{aligned}$$

where:

1. EntropProd\_i: relative change of entropy given a category on ground truthing.
2. Entrop\_i(B): Entropy of the map with respect to the category on ground truthing.
3. x\_+j: sum of all elements in rows j.
4. x\_+j: sum of all elements in column j.
5. Entrop\_i(B|a\_j): Entropy of map B knowing that the location corresponding to map A is in class a\_j.
6. N: number of cases involved in the calculation of the index.

*Usage:*

MatCon\$EntropProd\_i(i, v = NULL)

*Arguments:*

i Class to evaluate

v Base of the logarithm. By default v=10. This value is used for the entropy units, v=10(Hartleys), v=2(bits), v=e(nats).

*Returns:* A list of the relative change of entropy given a category on ground truthing, its variance, map entropy, and entropy of map B knowing that the location corresponding to map A is in class a\_j.

*Examples:*

```

A<-matrix(c(0,12,0,0,12,0,0,0,0,0,12,0,0,12,0),nrow=4,ncol=4)
p<-MatCon$new(A,Source="Liu et al., 2007")
p$EntropProd_i(2)

```

**Method** AvUserAcc(): Public method that provides the user's average accuracy, which is an average of the accuracy of individual categories, in this case the categories will be taken from the user's perspective. The method also offers variance. The reference [10] is followed for the calculations.

$$\begin{aligned}
AvUserAcc &= \frac{1}{\sqrt{M}} \sum_{i=1}^n \frac{x_{ii}}{\sum_{j=1}^n x_{j+}} \\
\sigma_{AvUserAcc}^2 &= \frac{AvUserAcc \cdot (1 - AvUserAcc)}{N}
\end{aligned}$$

where:

1. AvUserAcc: average accuracy from user's perspective.
2. x\_+j: sum of all elements in rows j.
3. x\_ii: diagonal element of the matrix.

4. M: number of elements of the matrix.
5. N: number of cases involved in the calculation of the index.

*Usage:*

MatCon\$AvUserAcc()

*Returns:* A list with the average accuracy from user's perspective and its variance.

*Examples:*

```
A<-matrix(c(352,43,89,203),nrow=2,ncol=2)
p<-MatCon$new(A,Source="Tung, F., & LeDrew, E. (1988). The determination of optimal threshold lev
p$AvUserAcc()
```

**Method** AvProdAcc(): Public method that provides the producer's average accuracy, which is an average of the accuracy of individual categories, in this case the categories will be taken from the producer's perspective. The method also offers variance. The reference [10] is followed for the calculations.

$$AvProdAcc = \frac{1}{\sqrt{N}} \sum_{i=1}^n \frac{x_{ii}}{\sum_{j=1}^n x_{+j}}$$

$$\sigma_{AvProdAcc}^2 = \frac{AvProdAcc \cdot (1 - AvProdAcc)}{N}$$

where:

1. AvProdAcc: average accuracy from producer's perspective.
2. x<sub>+j</sub>: sum of all elements in column j.
3. x<sub>ii</sub>: diagonal element of the matrix.
4. M: number of elements of the matrix.
5. N: number of cases involved in the calculation of the index.

*Usage:*

MatCon\$AvProdAcc()

*Returns:* A list with the average accuracy from producer's perspective and its variance.

*Examples:*

```
A<-matrix(c(352,43,89,203),nrow=2,ncol=2)
p<-MatCon$new(A,Source="Tung, F., & LeDrew, E. (1988). The determination of optimal threshold lev
p$AvProdAcc()
```

**Method** AvUserProdAcc(): Public method that offers the average of the average precision from the perspective of the user and the producer. The method also offers variance. The reference (Liu et al. 2007) is followed for the calculations.

$$AvUserProdAcc = \frac{AvUserAcc + AvProdAcc}{2}$$

$$\sigma_{AvUserProdAcc}^2 = \frac{AvUserProdAcc \cdot (1 - AvUserProdAcc)}{N}$$

where:

1. AvUserProdAcc: average of average of user's and producer's perspective.
2. AvUserAcc: average accuracy from user's perspective.

3. AvProdAcc: average accuracy from producer's perspective.
4. N: number of cases involved in the calculation of the index.

*Usage:*

MatCon\$AvUserProdAcc()

*Returns:* A list with the values of the average of the average precision from the perspective of the user and the producer the user and producer perspective and their variance.

*Examples:*

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$AvUserProdAcc()
```

**Method** AvHelldenAcc(): Public method that provides the average value of the Hellden mean precision index. The method also offers variance. The reference (Liu et al. 2007) is followed for the calculations.

$$AvHelldenAcc = \frac{1}{\sqrt{M}} \sum_{i=1}^n \frac{2x_{ii}}{x_{+i} + x_{i+}}$$

$$\sigma_{AvHelldenAcc}^2 = \frac{AvHelldenAcc \cdot (1 - AvHelldenAcc)}{N}$$

where:

1. AvHelldenAcc: average of Hellden's mean accuracy index.
2.  $x_{+i}$ : sum of all elements in column i.
3.  $x_{i+}$ : sum of all elements in row i.
4.  $x_{ii}$ : diagonal element of the matrix.
5. M: number of elements of the matrix.
6. N: number of cases involved in the calculation of the index.

*Usage:*

MatCon\$AvHelldenAcc()

*Returns:* A list with average of Hellden's mean accuracy index and its variance.

*Examples:*

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$AvHelldenAcc()
```

**Method** AvShortAcc(): Public method that provides the average of Short's mapping accuracy index. The method also offers variance. The reference (Liu et al. 2007) is followed for the calculations.

$$AvShortAcc = \frac{1}{\sqrt{M}} \frac{\sum_{i=1}^n x_{ii}}{\sum_{j=1}^n x_{+j} + \sum_{i=1}^n x_{i+} - x_{ii}}$$

$$\sigma_{AvShortAcc}^2 = \frac{AvShortAcc \cdot (1 - AvShortAcc)}{N}$$

where:

1.  $x_{+i}$ : sum of all elements in column i.

2.  $x_{i+}$ : sum of all elements in row  $i$ .
3.  $x_{ii}$ : diagonal element of the matrix.
4.  $M$ : number of elements of the matrix.
5.  $N$ : number of cases involved in the calculation of the index.#'

*Usage:*

MatCon\$AvShortAcc()

*Returns:* A list with average of Short's mapping accuracy index and its variance.

*Examples:*

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$AvShortAcc()
```

**Method CombUserAcc():** Public method that provides the combined user accuracy that is the average of the overall accuracy and the average user accuracy. The method also offers variance. The reference [10] is followed for the calculations.

$$CombUserAcc = \frac{OverallAcc + AvUserAcc}{2}$$

$$\sigma_{CombUserAcc}^2 = \frac{CombUserAcc \cdot (1 - CombUserAcc)}{N}$$

where:

1. CombUserAcc: combined accuracy from user's perspective.
2. OverallAcc: overall accuracy.
3. AvUserAcc: average accuracy from user's perspective.
4.  $N$ : number of cases involved in the calculation of the index.

*Usage:*

MatCon\$CombUserAcc()

*Details:* Example matrix taken from Tung, F., & LeDrew, E. (1988). The determination of optimal threshold levels for change detection using various accuracy indexes. Photogrammetric Engineering and Remote Sensing, 54(10), 1449-1454.

*Returns:* A list of the combined accuracy from the user's perspective and its variation.

*Examples:*

```
A<-matrix(c(352,43,89,203),nrow=2,ncol=2)
p<-MatCon$new(A)
p$CombUserAcc()
```

**Method CombProdAcc():** Public method that provides the combined producer accuracy that is the average of the overall accuracy and the average producer accuracy. The method also offers variance. The reference [10] is followed for the calculations.

$$CombProdAcc = \frac{OverallAcc + AvProdAcc}{2}$$

$$\sigma_{CombProdAcc}^2 = \frac{CombProdAcc \cdot (1 - CombProdAcc)}{N}$$

where:



1. CombProdAcc: combined accuracy from producer's perspective.
2. OverallAcc: overall accuracy.
3. AvProdAcc: average accuracy from producer's perspective.
4. N: number of cases involved in the calculation of the index.

*Usage:*

MatCon\$CombProdAcc()

*Details:* Example matrix taken from Tung, F., & LeDrew, E. (1988). The determination of optimal threshold levels for change detection using various accuracy indexes. *Photogrammetric Engineering and Remote Sensing*, 54(10), 1449-1454.

*Returns:* A list of the combined accuracy from producer's perspective and its variance.

*Examples:*

```
A<-matrix(c(352,43,89,203),nrow=2,ncol=2)
p<-MatCon$new(A)
p$CombProdAcc()
```

**Method CombUserProdAcc():** Public method that provides the combined accuracy which is the average of the overall accuracy and the Hellden average accuracy, which refers to the average user and producer accuracies. The method also offers variation. The reference (Liu et al. 2007) is followed for the calculations.

$$CombUserProdAcc = \frac{OverallAcc + AvHelldenAcc}{2}$$

$$\sigma_{CombUserProdAcc}^2 = \frac{CombUserProdAcc \cdot (1 - CombUserProdAcc)}{N}$$

where:

1. CombUserProdAcc: combined accuracy from both user's and producer's perspectives.
2. OverallAcc: overall accuracy.
3. AvHelldenAcc: average of Hellden's mean accuracy index.
4. N: number of cases involved in the calculation of the index.

*Usage:*

MatCon\$CombUserProdAcc()

*Returns:* A list of the combined accuracy from both user's and producer's perspectives and its variance.

*Examples:*

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$CombUserProdAcc()
```

**Method Kappa():** Public method that provides kappa coefficient, which measures the relationship between agreement beyond chance and expected disagreement. The method also offers variation. The reference [11] is followed for the calculations.

$$ExpAcc = \sum_{i=1}^n \left( \frac{x_{+i}}{\sum_{j=1}^n x_{ij}} \cdot \frac{x_{i+}}{\sum_{j=1}^n x_{ij}} \right)$$

$$Kappa = \frac{OverallAcc - ExpAcc}{1 - ExpAcc}$$

#'

$$\sigma_{Kappa}^2 = \frac{OverallAcc - ExpAcc}{(1 - ExpAcc) \cdot N}$$

where:

1. Kappa: Kappa coefficient.
2. OverallAcc: overall accuracy.
3. ExpAcc: expected accuracy of agreement if agreement were purely random.
4. x\_<sub>+</sub>i: sum of all elements in column i.
5. x\_<sub>i</sub>+: sum of all elements in row i.
6. N: number of cases involved in the calculation of the index.

*Usage:*

MatCon\$Kappa()

*Details:* Example matrix taken from Congalton, R.G., & Green, K. (2008). Assessing the Accuracy of Remotely Sensed Data: Principles and Practices, Second Edition (2nd ed.). CRC press

*Returns:* A list with kappa coefficient and its variance.

*Examples:*

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$Kappa()
```

**Method** Entrop(): Public method for calculating map entropy. Which refers to the degree of uncertainty that the map presents. The method also offers variation. The reference [9] is followed for the calculations.

$$Entrop = \sum_{i,j=1}^n \left( \frac{x_{ij}}{\sum_{i,j=1}^n x_{ij}} \cdot \log \left( \frac{x_{ij}}{\frac{\sum_{i=1}^n x_{i+} \cdot \sum_{j=1}^n x_{+j}}{\sum_{i,j=1}^n x_{ij}}} \right) \right)$$

$$\sigma_{Entrop}^2 = \frac{Entrop \cdot (1 - Entrop)}{N}$$

where:

1. Entrop: map entropy.
2. x\_<sub>+</sub>i: sum of all elements in column i.
3. x\_<sub>i</sub>+: sum of all elements in row i.
4. N: number of cases involved in the calculation of the index.

*Usage:*

MatCon\$Entrop(v = NULL)

*Arguments:*

v Base of the logarithm. By default v=10. This value is used for the entropy units, v=10(Hartleys), v=2(bits), v=e(nats).

*Returns:* A list with map entropy and its variance.

*Examples:*

```
A<-matrix(c(0,12,0,0,12,0,0,0,0,0,12,0,0,12,0),nrow=4,ncol=4)
p<-MatCon$new(A,Source="Liu et al.,2007")
p$Entrop()
```

**Method** `NormEntropUser()`: Public method that calculates normalized entropy using the map. The method also offers variation. The reference [9] is followed for the calculations.

$$Entrop_i(B) = - \sum_{i=1}^n \left( \left( \frac{\sum_{j=1}^n x_{+j}}{\sum_{i,j=1}^n x_{ij}} \right) \cdot \log \left( \frac{\sum_{j=1}^n x_{+j}}{\sum_{i,j=1}^n x_{ij}} \right) \right)$$

$$NormEntropUser = \frac{Entrop}{Entrop_i(B)}$$

$$\sigma_{NormEntropUser}^2 = \frac{NormEntropUser \cdot (1 - NormEntropUser)}{N}$$

where:

1. NormEntropUser: normalized entropy using map.
2. Entrop\_i(B): entropy of the map with respect to the category on ground truthing.
3. Entrop: map entropy.
4. x\_+i: sum of all elements in column i.
5. x\_i+: sum of all elements in row i.
6. N: number of cases involved in the calculation of the index.

*Usage:*

`MatCon$NormEntropUser(v = NULL)`

*Arguments:*

v Base of the logarithm. By default v=10. This value is used for the entropy units, v=10(Hartleys), v=2(bits), v=e(nats).

*Returns:* A list with normalized entropy using map and its variance.

*Examples:*

```
A<-matrix(c(0,12,0,0,12,0,0,0,0,0,12,0,0,12,0),nrow=4,ncol=4)
p<-MatCon$new(A,Source="Liu et al., 2007")
p$NormEntropUser()
```

**Method** `NormEntropProd()`: Public method that calculates normalized entropy using on ground truthing. The method also offers variation. The reference [9] is followed for the calculations.

$$Entrop_i(A) = - \sum_{j=1}^n \left( \left( \frac{\sum_{i=1}^n x_{i+}}{\sum_{i,j=1}^n x_{ij}} \right) \cdot \log \left( \frac{\sum_{i=1}^n x_{i+}}{\sum_{i,j=1}^n x_{ij}} \right) \right)$$

$$NormEntropProd = \frac{Entrop}{Entrop_i(A)}$$

$$\sigma_{NormEntropProd}^2 = \frac{NormEntropProd \cdot (1 - NormEntropProd)}{N}$$

where:

1. NormEntropProd: normalized mutual information using the entropy on ground truthing.
2. Entrop\_i(A): Entropy of the map with respect to the category of the map.
3. Entrop: map entropy.
4. x\_+i: sum of all elements in column i.
5. x\_i+: sum of all elements in row i.
6. N: number of cases involved in the calculation of the index.

*Usage:*

MatCon\$NormEntropProd(v = NULL)

*Arguments:*

v Base of the logarithm. By default v=10. This value is used for the entropy units, v=10(Hartleys), v=2(bits), v=e(nats).

*Returns:* A list with normalized entropy using on ground truthing and its variance.

*Examples:*

```
A<-matrix(c(0,12,0,0,12,0,0,0,0,0,12,0,0,12,0),nrow=4,ncol=4)
p<-MatCon$new(A,Source="Liu et al., 2007")
p$NormEntropProd()
```

**Method** AvNormEntrop(): Public method that calculates normalized entropy using the arithmetic mean of the entropies on the map and on ground truthing. The method also offers variation. The reference [12] is followed for the calculations.

$$\begin{aligned}
 Entrop_i(A) &= - \sum_{j=1}^n \left( \left( \frac{\sum_{i=1}^n x_{i+}}{\sum_{i,j=1}^n x_{ij}} \right) \cdot \log \left( \frac{\sum_{i=1}^n x_{i+}}{\sum_{i,j=1}^n x_{ij}} \right) \right) \\
 Entrop_i(B) &= - \sum_{i=1}^n \left( \left( \frac{\sum_{j=1}^n x_{+j}}{\sum_{i,j=1}^n x_{ij}} \right) \cdot \log \left( \frac{\sum_{j=1}^n x_{+j}}{\sum_{i,j=1}^n x_{ij}} \right) \right) \\
 AvNormEntrop &= \frac{2Entrop}{Entrop_i(A) + Entrop_i(B)} \\
 \sigma_{AvNormEntrop}^2 &= \frac{AvNormEntrop \cdot (1 - AvNormEntrop)}{N}
 \end{aligned}$$

where:

1. AvNormEntrop: normalized entropy using the arithmetic mean of the entropies on the map and on ground truthing.
2. Entrop\_i(B): entropy of the map with respect to the category on ground truthing.
3. Entrop\_i(A): Entropy of the map with respect to the category of the map.
4. Entrop: map entropy.
5. x\_+i: sum of all elements in column i.
6. x\_i+: sum of all elements in row i.
7. N: number of cases involved in the calculation of the index.

*Usage:*

MatCon\$AvNormEntrop(v = NULL)

*Arguments:*

v Base of the logarithm. By default v=10. This value is used for the entropy units, v=10(Hartleys), v=2(bits), v=e(nats).

*Returns:* normalized entropy using the arithmetic mean of the entropies on the map and on ground truthing and its variance.

*Examples:*

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$AvNormEntrop()
```

**Method** `GeomAvNormEntrop()`: Public method that calculates normalized entropy using the geometric mean of the entropies on the map and on ground truthing. The method also offers variation. The reference [13] is followed for the calculations.

$$\begin{aligned}
 Entrop_i(A) &= - \sum_{j=1}^n \left( \left( \frac{\sum_{i=1}^n x_{i+}}{\sum_{i,j=1}^n x_{ij}} \right) \cdot \log \left( \frac{\sum_{i=1}^n x_{i+}}{\sum_{i,j=1}^n x_{ij}} \right) \right) \\
 Entrop_i(B) &= - \sum_{i=1}^n \left( \left( \frac{\sum_{j=1}^n x_{+j}}{\sum_{i,j=1}^n x_{ij}} \right) \cdot \log \left( \frac{\sum_{j=1}^n x_{+j}}{\sum_{i,j=1}^n x_{ij}} \right) \right) \\
 GeomAvNormEntrop &= \frac{Entrop}{\sqrt{Entrop_i(A) \cdot Entrop_i(B)}} \\
 \sigma_{GeomAvNormEntrop}^2 &= \frac{GeomAvNormEntrop \cdot (1 - GeomAvNormEntrop)}{N}
 \end{aligned}$$

where:

1. `GeomAvNormEntrop`: normalized entropy using the geometric mean of the entropies on map and on ground truthing.
2. `Entrop_i(B)`: entropy of the map with respect to the category on ground truthing.
3. `Entrop_i(A)`: Entropy of the map with respect to the category of the map.
4. `Entrop`: map entropy.
5. `x_+i`: sum of all elements in column i.
6. `x_i+`: sum of all elements in row i.
7. `N`: number of cases involved in the calculation of the index.

*Usage:*

`MatCon$GeomAvNormEntrop(v = NULL)`

*Arguments:*

`v` Base of the logarithm. By default `v=10`. This value is used for the entropy units, `v=10`(Hartleys), `v=2`(bits), `v=e`(nats).

*Returns:* A list with normalized entropy using the geometric mean of the entropies on map and on ground truthing and its variance.

*Examples:*

```

A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$GeomAvNormEntrop()

```

**Method** `AvMaxNormEntrop()`: Public method that provides normalized entropy using the arithmetic mean of the maximum entropies on map and on ground truthing. The method also offers variation. The reference [14] is followed for the calculations.

*Usage:*

`MatCon$AvMaxNormEntrop(v = NULL)`

*Arguments:*

`v` Base of the logarithm. By default `v=10`. This value is used for the entropy units, `v=10`(Hartleys), `v=2`(bits), `v=e`(nats).

*Returns:* A list with normalized entropy using the arithmetic mean of the maximum entropies on map and on ground truthing and its variance.

*Examples:*

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$AvMaxNormEntrop()
```

**Method** `Tau()`: Public method that calculates the tau index and its variance. Its value indicates how much the classification has improved compared to a random classification of the N elements into M groups. The method also offers the variance. The reference [15] is followed for the computations.

The mathematical expression is:

$$PrAgCoef = \frac{1}{M}$$

$$Tau = \frac{OverallAcc - CoefAccPr}{1 - PrAgCoef}$$

$$\sigma_{Tau}^2 = \frac{OverallAcc \cdot (1 - OverallAcc)}{N \cdot (1 - CoefAccPr)^2}$$

Where:

1. OverallAcc: overall accuracy.
2. PrAgCoef: a priori random agreement coefficient.
3. M: number of classes.
4. N: number of elements of the matrix, cardinal of the matrix.

*Usage:*

```
MatCon$Tau()
```

*Details:* Example matrix taken from Muñoz, J. M. S. (2016). Análisis de Calidad Cartográfica mediante el estudio de la Matriz de Confusión. Pensamiento matemático, 6(2), 9-26.

*Returns:* A list with Tau index and its variance.

*Examples:*

```
A<-matrix(c(238051,7,132,0,0,24,9,2,189,1,4086,188,0,4,16,45,1,0,939,5082,51817,0,34,500,1867),nrow=4,ncol=4)
p<-MatCon$new(A)
p$Tau()
```

**Method** `UserProdAcc()`: Public method that calculates the pressures of the user and the producer jointly. The method also offers the standard desviations. The reference (Congalton and Green 2008) is followed for the computations.

*Usage:*

```
MatCon$UserProdAcc()
```

*Details:* Example matrix taken from Congalton, R.G., & Green, K. (2008). Assessing the Accuracy of Remotely Sensed Data: Principles and Practices, Second Edition (2nd ed.). CRC press

*Returns:* A list containing the producer's and user's accuracies and their standard deviations, respectively.

*Examples:*

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A,Source="Congalton&Green,2008")
p$UserProdAcc()
```

**Method DetailedKappa():** Public method that calculates the general Kappa agreement index, its standard deviation and the test statistic to test its significance. The reference (Congalton and Green 2008) is followed for the computations.

*Usage:*

```
MatCon$DetailedKappa()
```

*Details:* Example matrix taken from Congalton, R.G., & Green, K. (2008). Assessing the Accuracy of Remotely Sensed Data: Principles and Practices, Second Edition (2nd ed.). CRC press

*Returns:* A list of the kappa coefficient, its standard deviation, and the value of its test statistic.

*Examples:*

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A,Source="Congalton&Green,2008")
p$DetailedKappa()
```

**Method DetailedCondKappa():** Public method that calculates the Kappa class agreement index (conditional Kappa) from the perspective of user (i) and producer (j) and its standard deviations. The reference (Congalton and Green 2008) is followed for the computations.

*Usage:*

```
MatCon$DetailedCondKappa()
```

*Details:* Example matrix taken from Congalton, R.G., & Green, K. (2008). Assessing the Accuracy of Remotely Sensed Data: Principles and Practices, Second Edition (2nd ed.). CRC press

*Returns:* A list with conditional Kappa index of the user and the producer, and its corresponding standard deviation.

*Examples:*

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A,Source="Congalton&Green,2008")
p$DetailedCondKappa ()
```

**Method QES():** Public method that calculates the values of quantity, change and shift. The reference [16] is followed for the computations.

*Usage:*

```
MatCon$QES(TI = NULL, SF = 1)
```

*Arguments:*

TI Time interval (default value = 1)

SF Scale factor for results (default value = 1)

*Returns:* A list of general values for the interval t of difference, quantity, shift, and shift. In addition to the differences for categories, number of components, change of categories and turn of the components.

*Examples:*

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$QES(TI=1, SF=6)
```

**Method** `MTypify()`: Public method that types the values of each cell. The total sum of the original matrix is used for typing. The resulting values can be presented as real (parameter `RaR=1`) or as a percentage (parameter `RaR !=1`)

$$MTypify = \frac{x_{ij}}{\sum_{i,j=1}^n x_{ij}}$$

where:

1. `MTypify`: typified matrix.
2. `x_ij`: matrix element.

*Usage:*

`MatCon$MTypify(RaR = NULL)`

*Arguments:*

`RaR "1"` indicates result as real, other values mean percentage as integer. By default `RaR=1`.

*Returns:* A list with original matrix and typified matrix

*Examples:*

```
A <- t(matrix(c(35, 14,11,1,4,11,3,0,12,9,38,4,2,5,12,2), nrow = 4, ncol=4))
p<-MatCon$new(A)
p$MTypify(RaR=5)
```

**Method** `AllParameters()`: Public method in which multiple parameters are calculated for the given confusion matrix. The reference [1,11,16,19] is followed for the computations.

*Usage:*

`MatCon$AllParameters()`

*Details:* Example matrix taken from Congalton, R.G., & Green, K. (2008). Assessing the Accuracy of Remotely Sensed Data: Principles and Practices, Second Edition (2nd ed.). CRC press

*Returns:* A list containing Confusion Matrix, Dimension, Total Sum of Cell Values, Overall Precision, Overall Variance Precision, Global Precision Kappa Index, Global Kappa Simplified Variance, Producer Precision by Class, User Precision by Class, k value for the calculation of pseudozeroes, Pseudoceros Matrix, L Matrix for the calculation of pseudozeroes.

*Examples:*

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$AllParameters()
```

**Method** `MBootStrap()`: Public method that provides N resamples of the confusion matrix from a `MatCon` object. The reference [17] is followed for the computations.

*Usage:*

`MatCon$MBootStrap(n)`

*Arguments:*

`n` Number of resamples.

*Details:* Example matrix taken from Ariza, F. J., Pinilla, C., & Garcia, J. L. (2011). Comparación de matrices de confusión celda a celda mediante bootstrapping.

*Returns:* A list formed by the original confusion matrix and simulated matrices, from the confusion matrix. The multinomial distribution is applied.



*Examples:*

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$MBootstrap(2)
```

**Method** `MNormalize()`: Public method that carries out an iterative process is carried out where each element is divided by the total of the sum of its row, thus obtaining new values. In the next iteration, all the elements are added by columns and each element is divided by the total of its column and they obtain new values, and so on. The reference [18,19] is followed for the computations.

*Usage:*

```
MatCon$MNormalize(n = NULL)
```

*Arguments:*

n Number of iteration. By default n=100.

*Details:* Example matrix taken from Muñoz, J. M. S. (2016). Análisis de Calidad Cartográfica mediante el estudio de la Matriz de Confusión. Pensamiento matemático, 6(2), 9-26.

*Returns:* A list formed by the original confusion matrix and the normalized matrix.

*Examples:*

```
A<-matrix(c(238051,7,132,0,0,24,9,2,189,1,4086,188,0,4,16,45,1,0,939,5082,51817,0,34,500,1867),nrow=22,ncol=22)
p<-MatCon$new(A)
p$MNormalize()$values
```

**Method** `MPseudoZeroes()`: Public method that small values are calculated for empty cells of the matrix. All non-empty cells of the matrix change their values. This function will not be applied if all the elements of the matrix are different from 0. The reference [19] is followed for the computations.

*Usage:*

```
MatCon$MPseudoZeroes()
```

*Details:* Example matrix taken from Muñoz, J. M. S. (2016). Análisis de Calidad Cartográfica mediante el estudio de la Matriz de Confusión. Pensamiento matemático, 6(2), 9-26.

*Returns:* A list formed by the original confusion matrix and the Pseudozeroes matrix.

*Examples:*

```
A<-matrix(c(238051,7,132,0,0,24,9,2,189,1,4086,188,0,4,16,45,1,0,939,5082,51817,0,34,500,1867),nrow=22,ncol=22)
p<-MatCon$new(A)
p$MPseudoZeroes()$values
```

**Method** `DetailedWTau()`: Public method that calculates the general Tau concordance index and its standard deviation.

*Usage:*

```
MatCon$DetailedWTau(WV)
```

*Arguments:*

WV Weights vector (as matrix)

*Returns:* Overall accuracy index, producer accuracy index, O3,O4, Tau index?(mirar definicion en funcion) y its standard desviation.

*Examples:*

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
WV <-matrix(c(0.4, 0.1, 0.4, 0.1), ncol=4)
p<-MatCon$new(A)
p$DetailedWTau(WV)
```

**Method DetailedWKappa():** Public method that calculates the general Kappa agreement index (weighted) and its standard deviation. The reference (Congalton and Green 2008) is followed for the computations.

*Usage:*

```
MatCon$DetailedWKappa(WM)
```

*Arguments:*

WM Weight matrix

*Returns:* A list with the weight matrix, kappa index obtained from the original matrix and the weight matrix, its standard desviations and the value of its test statistic.

*Examples:*

```
A <- A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
WM<- t(matrix(c(1,0,0.67,1,0,1,0,0,1,0,1,1,0.91,0,0.61,1), nrow = 4, ncol=4))
p<-MatCon$new(A)
p$DetailedWKappa(WM)
```

**Method UserProdAcc\_W():** Public method that calculates the weighted accuracies and standard deviations of the user and the producer. The reference (Congalton and Green 2008) is followed for the computations.

*Usage:*

```
MatCon$UserProdAcc_W(WM)
```

*Arguments:*

WM Weight matrix

*Returns:* A list with weight matrix, Matrix formed with its original elements and their corresponding weights, general accuracy of the weight matrix obtained, accuracy of the producer and user and their standard deviations,

*Examples:*

```
A <- t(matrix(c(35, 14,11,1,4,11,3,0,12,9,38,4,2,5,12,2), nrow = 4, ncol=4))
p<-MatCon$new(A)
WM<- t(matrix(c(1,0,0.67,1,0,1,0,0,1,0,1,1,0.91,0,0.61,1), nrow = 4, ncol=4))
p$UserProdAcc_W(WM)
```

**Method clone():** The objects of this class are cloneable with this method.

*Usage:*

```
MatCon$clone(deep = FALSE)
```

*Arguments:*

deep Whether to make a deep clone.

## References

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- [19] Muñoz, J. M. S. (2016). Análisis de Calidad Cartográfica mediante el estudio de la Matriz de Confusión. *Pensamiento matemático*, 6(2), 9-26.

## Examples

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
mc <- MatCon$new (A,ID=5,Date="27-10-2023")
```

```
## -----
## Method `MatCon$new`
## -----
```

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
mc <- MatCon$new (A,ID=5,Date="27-10-2023",Source="Congalton, R.G., & Green, K. (2008). Assessing the Accuracy
```

```
## -----
## Method `MatCon$OverallAcc`
## -----
```

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$OverallAcc()
```

```
## -----
## Method `MatCon$userAcc`
## -----
```

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$userAcc()
```

```
## -----
## Method `MatCon$userAcc_i`
## -----
```

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$userAcc_i(2)
```

```
## -----
## Method `MatCon$ProdAcc`
## -----
```

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$ProdAcc()
```

```
## -----
## Method `MatCon$ProdAcc_i`
## -----
```

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$ProdAcc_i(1)
```

```

## -----
## Method `MatCon$AvUserProdAcc_i`
## -----

A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$AvUserProdAcc_i(2)

## -----
## Method `MatCon$Sucess`
## -----

A<-matrix(c(0.3,0.02,0.01,0.12,0.19,0.03,0.02,0.01,0.3),nrow=3,ncol=3)
p<-MatCon$new(A,Source="Labatut, V., & Cherifi, H. (2011). Evaluation of performance measures for classifiers o
p$Sucess()

## -----
## Method `MatCon$Sucess_i`
## -----

A<-matrix(c(0.3,0.02,0.01,0.12,0.19,0.03,0.02,0.01,0.3),nrow=3,ncol=3)
p<-MatCon$new(A,Source="Labatut, V., & Cherifi, H. (2011). Evaluation of performance measures for classifiers o
p$Sucess_i(2)

## -----
## Method `MatCon$AvHelldenAcc_i`
## -----

A <- matrix(c(148,1,8,2,0,0,50,15,3,0,1,6,39,7,1,1,0,6,25,1,1,0,0,1,6),nrow=5,ncol=5)
p<-MatCon$new(A,Source="Rosenfield, G. H., & Fitzpatrick-Lins, K. (1986). A coefficient of agreement as a measu
p$AvHelldenAcc_i(2)

## -----
## Method `MatCon$ShortAcc_i`
## -----

A <- matrix(c(148,1,8,2,0,0,50,15,3,0,1,6,39,7,1,1,0,6,25,1,1,0,0,1,6),nrow=5,ncol=5)
p<-MatCon$new(A,Source="Rosenfield, G. H., & Fitzpatrick-Lins, K. (1986). A coefficient of agreement as a measu
p$ShortAcc_i(2)

## -----
## Method `MatCon$UserKappa_i`
## -----

A<-matrix(c(73,13,5,1,0,21,32,13,3,0,16,39,35,29,13,3,5,7,28,48,1,0,2,3,17),nrow=5,ncol=5)
p<-MatCon$new(A,Source="Næsset, E. (1996). Conditional tau coefficient for assessment of producer's accuracy o
p$UserKappa_i(2)

## -----

```

```

## Method `MatCon$ProdKappa_i`
## -----

A<-matrix(c(73,13,5,1,0,21,32,13,3,0,16,39,35,29,13,3,5,7,28,48,1,0,2,3,17),nrow=5,ncol=5)
p<-MatCon$new(A,Source="Næsset, E. (1996). Conditional tau coefficient for assessment of producer's accuracy o
p$ProdKappa_i(2)

## -----
## Method `MatCon$ModKappa`
## -----

A<-matrix(c(317,61,2,35,23,120,4,29,0,0,60,0,0,0,0,8),nrow=4,ncol=4)
p<-MatCon$new(A,Source="Foody, G. M. (1992). On the compensation for chance agreement in image classification a
p$ModKappa()

## -----
## Method `MatCon$ModKappaUser_i`
## -----

A<-matrix(c(0,12,0,0,12,0,0,0,0,0,12,0,0,12,0),nrow=4,ncol=4)
p<-MatCon$new(A,Source="Liu et al., 2007")
p$ModKappaUser_i(2)

## -----
## Method `MatCon$ModKappaProd_i`
## -----

A<-matrix(c(317,61,2,35,23,120,4,29,0,0,60,0,0,0,0,8),nrow=4,ncol=4)
p<-MatCon$new(A,Source="Liu et al., 2007")
p$ModKappaProd_i(2)

## -----
## Method `MatCon$EntropUser_i`
## -----

A<-matrix(c(0,12,0,0,12,0,0,0,0,0,12,0,0,12,0),nrow=4,ncol=4)
p<-MatCon$new(A,Source="Liu et al., 2007")
p$EntropUser_i(1)

## -----
## Method `MatCon$EntropProd_i`
## -----

A<-matrix(c(0,12,0,0,12,0,0,0,0,0,12,0,0,12,0),nrow=4,ncol=4)
p<-MatCon$new(A,Source="Liu et al., 2007")
p$EntropProd_i(2)

## -----
## Method `MatCon$AvUserAcc`
## -----

```

```
A<-matrix(c(352,43,89,203),nrow=2,ncol=2)
p<-MatCon$new(A,Source="Tung, F., & LeDrew, E. (1988). The determination of optimal threshold levels for change
p$AvUserAcc()
```

```
## -----
## Method `MatCon$AvProdAcc`
## -----
```

```
A<-matrix(c(352,43,89,203),nrow=2,ncol=2)
p<-MatCon$new(A,Source="Tung, F., & LeDrew, E. (1988). The determination of optimal threshold levels for change
p$AvProdAcc()
```

```
## -----
## Method `MatCon$AvUserProdAcc`
## -----
```

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$AvUserProdAcc()
```

```
## -----
## Method `MatCon$AvHelldenAcc`
## -----
```

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$AvHelldenAcc()
```

```
## -----
## Method `MatCon$AvShortAcc`
## -----
```

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$AvShortAcc()
```

```
## -----
## Method `MatCon$CombUserAcc`
## -----
```

```
A<-matrix(c(352,43,89,203),nrow=2,ncol=2)
p<-MatCon$new(A)
p$CombUserAcc()
```

```
## -----
## Method `MatCon$CombProdAcc`
## -----
```

```
A<-matrix(c(352,43,89,203),nrow=2,ncol=2)
p<-MatCon$new(A)
p$CombProdAcc()
```

```

## -----
## Method `MatCon$CombUserProdAcc`
## -----

A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$CombUserProdAcc()

## -----
## Method `MatCon$Kappa`
## -----

A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$Kappa()

## -----
## Method `MatCon$Entrop`
## -----

A<-matrix(c(0,12,0,0,12,0,0,0,0,0,12,0,0,12,0),nrow=4,ncol=4)
p<-MatCon$new(A,Source="Liu et al., 2007")
p$Entrop()

## -----
## Method `MatCon$NormEntropUser`
## -----

A<-matrix(c(0,12,0,0,12,0,0,0,0,0,12,0,0,12,0),nrow=4,ncol=4)
p<-MatCon$new(A,Source="Liu et al., 2007")
p$NormEntropUser()

## -----
## Method `MatCon$NormEntropProd`
## -----

A<-matrix(c(0,12,0,0,12,0,0,0,0,0,12,0,0,12,0),nrow=4,ncol=4)
p<-MatCon$new(A,Source="Liu et al., 2007")
p$NormEntropProd()

## -----
## Method `MatCon$AvNormEntrop`
## -----

A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$AvNormEntrop()

## -----

```



```
## Method `MatCon$GeomAvNormEntrop`
```

```
## -----
```

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$GeomAvNormEntrop()
```

```
## -----
```

```
## Method `MatCon$AvMaxNormEntrop`
```

```
## -----
```

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$AvMaxNormEntrop()
```

```
## -----
```

```
## Method `MatCon$Tau`
```

```
## -----
```

```
A<-matrix(c(238051,7,132,0,0,24,9,2,189,1,4086,188,0,4,16,45,1,0,939,5082,51817,0,34,500,1867,325,17,0,0,5),nrow=4,ncol=4)
p<-MatCon$new(A)
p$Tau()
```

```
## -----
```

```
## Method `MatCon$UserProdAcc`
```

```
## -----
```

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A,Source="Congalton&Green,2008")
p$UserProdAcc()
```

```
## -----
```

```
## Method `MatCon$DetailedKappa`
```

```
## -----
```

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A,Source="Congalton&Green,2008")
p$DetailedKappa()
```

```
## -----
```

```
## Method `MatCon$DetailedCondKappa`
```

```
## -----
```

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A,Source="Congalton&Green,2008")
p$DetailedCondKappa ()
```

```
## -----
```

```
## Method `MatCon$QES`
```

```
## -----
```

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$QES(TI=1, SF=6)
```

```
## -----
## Method `MatCon$MTypify`
## -----
```

```
A <- t(matrix(c(35, 14,11,1,4,11,3,0,12,9,38,4,2,5,12,2), nrow = 4, ncol=4))
p<-MatCon$new(A)
p$MTypify(RaR=5)
```

```
## -----
## Method `MatCon$AllParameters`
## -----
```

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$AllParameters()
```

```
## -----
## Method `MatCon$MBootstrap`
## -----
```

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
p<-MatCon$new(A)
p$MBootstrap(2)
```

```
## -----
## Method `MatCon$MNormalize`
## -----
```

```
A<-matrix(c(238051,7,132,0,0,24,9,2,189,1,4086,188,0,4,16,45,1,0,939,5082,51817,0,34,500,1867,325,17,0,0,5),nrow=4,ncol=4)
p<-MatCon$new(A)
p$MNormalize()$values
```

```
## -----
## Method `MatCon$MPseudoZeroes`
## -----
```

```
A<-matrix(c(238051,7,132,0,0,24,9,2,189,1,4086,188,0,4,16,45,1,0,939,5082,51817,0,34,500,1867,325,17,0,0,5),nrow=4,ncol=4)
p<-MatCon$new(A)
p$MPseudoZeroes()$values
```

```
## -----
## Method `MatCon$DetailedWTau`
## -----
```

```
A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
WV <-matrix(c(0.4, 0.1, 0.4, 0.1), ncol=4)
p<-MatCon$new(A)
p$DetailedWTau(WV)
```

```
## -----
## Method `MatCon$DetailedWKappa`
## -----

A <- A<-matrix(c(65,6,0,4,4,81,11,7,22,5,85,3,24,8,19,90),nrow=4,ncol=4)
WM<- t(matrix(c(1,0,0.67,1,0,1,0,0,1,0,1,1,0.91,0,0.61,1), nrow = 4, ncol=4))
p<-MatCon$new(A)
p$DetailedWKappa(WM)

## -----
## Method `MatCon$UserProdAcc_W`
## -----

A <- t(matrix(c(35, 14,11,1,4,11,3,0,12,9,38,4,2,5,12,2), nrow = 4, ncol=4))
p<-MatCon$new(A)
WM<- t(matrix(c(1,0,0.67,1,0,1,0,0,1,0,1,1,0.91,0,0.61,1), nrow = 4, ncol=4))
p$UserProdAcc_W(WM)
```

---

QCCS

---

*Quality Control Columns Set*


---

## Description

The p value is calculated using the multinomial distribution from vectors and their corresponding probabilities.

## Value

Object of class QCCS

## Methods

### Public methods:

- [QCCS\\$new\(\)](#)
- [QCCS\\$QCCS\(\)](#)
- [QCCS\\$clone\(\)](#)

**Method new():** Creates a new instance of this [R6][R6::R6Class] class.

*Usage:*

```
QCCS$new(vectors, prob, ID = NULL, Date = NULL, Source = NULL)
```

*Arguments:*

vectors vector list.

prob probabilities list.

ID Identifier. By default ID is a date in YYYYMMDD format

Date Date provided by the user. By default the date provided by the system will be taken.

Source Indicates where the matrix comes from (article, project, etc.). By default is NULL.

*Examples:*

```
vectors<-list(c(47,4,0),c(40,5,3),c(45,6,2),c(48,0))
prob<-list(c(0.95,0.04,0.01),c(0.88,0.1,0.02),c(0.9,0.08,0.02),c(0.99,0.01))
A <- QCCS$new(vectors,prob)
```

**Method QCCS():** Public method that, using a list of vectors and their corresponding probabilities, through a multinomial distribution, calculates the p value using each of the vectors. The reference (Ariza-López et al. 2019) is followed for the computations.

*Usage:*

```
QCCS$QCCS()
```

*Returns:* The p value is obtained for each vector, and using the Bonferroni criterion it is decided whether the elements are well classified or not.

*Examples:*

```
vectors<-list(c(47,4,0),c(40,5,3),c(45,6,2),c(48,0))
prob<-list(c(0.95,0.04,0.01),c(0.88,0.1,0.02),c(0.9,0.08,0.02),c(0.99,0.01))
A <- QCCS$new(vectors,prob,Source="Ariza et al.,2019")
A$QCCS()
```

**Method clone():** The objects of this class are cloneable with this method.

*Usage:*

```
QCCS$clone(deep = FALSE)
```

*Arguments:*

deep Whether to make a deep clone.

## References

Ariza-López FJ, Rodríguez-Avi J, Alba-Fernández MV, García-Balboa JL (2019). "Thematic accuracy quality control by means of a set of multinomials." *Applied Sciences*, **9**(20), 4240.

## Examples

```
## -----
## Method `QCCS$new`
## -----

vectors<-list(c(47,4,0),c(40,5,3),c(45,6,2),c(48,0))
prob<-list(c(0.95,0.04,0.01),c(0.88,0.1,0.02),c(0.9,0.08,0.02),c(0.99,0.01))
A <- QCCS$new(vectors,prob)

## -----
## Method `QCCS$QCCS`
## -----

vectors<-list(c(47,4,0),c(40,5,3),c(45,6,2),c(48,0))
prob<-list(c(0.95,0.04,0.01),c(0.88,0.1,0.02),c(0.9,0.08,0.02),c(0.99,0.01))
A <- QCCS$new(vectors,prob,Source="Ariza et al.,2019")
A$QCCS()
```

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