

# Package ‘coda.base’

May 14, 2020

**Type** Package

**Title** A Basic Set of Functions for Compositional Data Analysis

**Version** 0.3.1

**Date** 2020-05-14

**Description** A minimum set of functions to perform compositional data analysis using the log-ratio approach introduced by John Aitchison (1982) <<http://www.jstor.org/stable/2345821>>. Main functions have been implemented in c++ for better performance.

**URL** <https://mcomas.github.io/coda.base>,  
<https://github.com/mcomas/coda.base>

**Depends** R (>= 3.0.4)

**Imports** Rcpp (>= 0.12.12), stats

**LinkingTo** Rcpp, RcppArmadillo

**License** GPL

**Encoding** UTF-8

**LazyData** true

**NeedsCompilation** yes

**RoxygenNote** 7.1.0

**Suggests** knitr, rmarkdown, testthat (>= 2.1.0)

**VignetteBuilder** knitr

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**Repository** CRAN

**Date/Publication** 2020-05-14 17:10:29 UTC

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alr_basis	<i>Additive log-ratio basis</i>
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**Description**

Compute the transformation matrix to express a composition using the oblique additive log-ratio coordinates.

**Usage**

```
alr_basis(dim, denominator = dim, numerator = which(denominator != 1:dim))
```

**Arguments**

dim	number of parts
denominator	part used as denominator (default behaviour is to use last part)
numerator	parts to be used as numerator. By default all except the denominator parts are chosen following original order.

**Value**

matrix

**References**

Aitchison, J. (1986) *The Statistical Analysis of Compositional Data*. Monographs on Statistics and Applied Probability. Chapman & Hall Ltd., London (UK). 416p.

**Examples**

```
alr_basis(5)
# Third part is used as denominator
alr_basis(5, 3)
# Third part is used as denominator, and
# other parts are rearranged
alr_basis(5, 3, c(1,5,2,4))
```

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basis	<i>Coordinates basis</i>
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**Description**

Obtain coordinates basis

**Usage**

```
basis(H)
```

**Arguments**

H                      coordinates for which basis should be shown

**Value**

basis used to create coordinates H

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cbalance_approx	<i>Balance generated from the first canonical correlation component</i>
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---

**Description**

Balance generated from the first canonical correlation component

**Usage**

```
cbalance_approx(Y, X)
```

**Arguments**

Y                      compositional dataset  
X                      explanatory dataset

**Value**

matrix

---

`cc_basis`*Isometric log-ratio basis based on canonical correlations*

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**Description**

Isometric log-ratio basis based on canonical correlations

**Usage**`cc_basis(Y, X)`**Arguments**

<code>Y</code>	compositional dataset
<code>X</code>	explanatory dataset

**Value**

matrix

---

`cdp_basis`*Isometric log-ratio basis based on Balances.*

---

**Description**

The function return default balances used in CoDaPack software.

**Usage**`cdp_basis(dim)`**Arguments**

<code>dim</code>	dimension to build the ILR basis based on balanced balances
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**Value**

matrix

---

cdp_partition	<i>CoDaPack's default binary partition</i>
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**Description**

Compute the default binary partition used in CoDaPack's software

**Usage**

```
cdp_partition(ncomp)
```

**Arguments**

ncomp	number of parts
-------	-----------------

**Value**

matrix

**Examples**

```
cdp_partition(4)
```

---

clr_basis	<i>Centered log-ratio basis</i>
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**Description**

Compute the transformation matrix to express a composition using the linearly dependant centered log-ratio coordinates.

**Usage**

```
clr_basis(dim)
```

**Arguments**

dim	number of parts
-----	-----------------

**Value**

matrix

**References**

Aitchison, J. (1986) *The Statistical Analysis of Compositional Data*. Monographs on Statistics and Applied Probability. Chapman & Hall Ltd., London (UK). 416p.

**Examples**

```
(B <- clr_basis(5))
# CLR coordinates are linearly dependant coordinates.
(clr_coordinates <- coordinates(c(1,2,3,4,5), B))
# The sum of all coordinates equal to zero
sum(clr_coordinates) < 1e-15
```

---

coda.base

*coda.base*


---

**Description**

A minimum set of functions to perform compositional data analysis using the log-ratio approach introduced by John Aitchison (1982) <<http://www.jstor.org/stable/2345821>>. Main functions have been implemented in c++ for better performance.

**Author(s)**

Marc Comas-Cufí

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composition

*Get composition from coordinates w.r.t. an specific basis*


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**Description**

Calculate a composition from coordinates with respect a given basis

**Usage**

```
composition(H, basis = NULL, label = "x", sparse_basis = FALSE)
```

**Arguments**

H	coordinates of a composition. Either a matrix, a data.frame or a vector
basis	basis used to calculate the coordinates
label	name given to the coordinates
sparse_basis	Is the given matrix basis sparse? If TRUE calculation are carried taking into an account sparsity (default 'FALSE')

**Value**

coordinates with respect the given basis

**See Also**

See functions [ilr\\_basis](#), [alr\\_basis](#), [clr\\_basis](#), [sbp\\_basis](#) to define different compositional basis. See function [coordinates](#) to obtain details on how to calculate coordinates of a given composition.

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`coordinates`*Get coordinates from compositions w.r.t. an specific basis*

---

## Description

Calculate the coordinates of a composition with respect a given basis

## Usage

```
coordinates(  
  X,  
  basis = "ilr",  
  label = ifelse(is.character(basis), basis, "h"),  
  basis_return = TRUE  
)
```

## Arguments

<code>X</code>	compositional dataset. Either a matrix, a data.frame or a vector
<code>basis</code>	basis used to calculate the coordinates. <code>basis</code> can be either a string or a matrix. Accepted values for strings are: 'ilr' (default), 'clr', 'alr', 'pc', 'pb' and 'cdp'. If <code>basis</code> is a matrix, it is expected to have log-ratio basis given in columns.
<code>label</code>	name given to the coordinates
<code>basis_return</code>	Should the basis be returned as attribute? (default: TRUE)

## Details

`coordinates` function calculates the coordinates of a compositiona w.r.t. a given basis. 'basis' parameter is used to set the basis, it can be either a matrix defining the log-contrasts in columns or a string defining some well-known log-contrast: 'alr' 'clr', 'ilr', 'pc', 'pb' and 'cdp', for the additive log-ratio, centered log-ratio, isometric log-ratio, clr principal components, clr principal balances or default's CoDaPack balances respectively.

## Value

Coordinates of composition `X` with respect the given basis.

## See Also

See functions [ilr\\_basis](#), [alr\\_basis](#), [clr\\_basis](#), [sbp\\_basis](#) to define different compositional basis. See function [composition](#) to obtain details on how to calculate a compositions from given coordinates.

**Examples**

```

coordinates(c(1,2,3,4,5))
# basis is shown if 'coda.base.basis' option is set to TRUE
options('coda.base.basis' = TRUE)
coordinates(c(1,2,3,4,5))
# Default transformation improves performance.
N = 100
K = 1000
X = matrix(exp(rnorm(N*K)), nrow=N, ncol=K)
system.time(coordinates(X, alr_basis(K)))
system.time(coordinates(X, 'alr'))

```

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**dist***Distance Matrix Computation (including Aitchison distance)*

---

**Description**

This function overwrites [dist](#) function to contain Aitchison distance between compositions.

**Usage**

```
dist(x, method = "euclidean", ...)
```

**Arguments**

x	compositions method
method	the distance measure to be used. This must be one of "aitchison", "euclidean", "maximum", "manhattan", "canberra", "binary" or "minkowski". Any unambiguous substring can be given.
...	arguments passed to <a href="#">dist</a> function

**Value**

dist returns an object of class "dist".

**See Also**

See functions [dist](#).

**Examples**

```

X = exp(matrix(rnorm(10*50), ncol=50, nrow=10))

(d <- dist(X, method = 'aitchison'))
plot(hclust(d))

# In contrast to Euclidean distance
dist(rbind(c(1,1,1), c(100, 100, 100)), method = 'euc') # method = 'euclidean'

```



```
# using Aitchison distance, only relative information is of importance
dist(rbind(c(1,1,1), c(100, 100, 100)), method = 'ait') # method = 'aitchison'
```

---

ilr\_basis *Default Isometric log-ratio basis*

---

## Description

Build an isometric log-ratio basis for a composition with  $k+1$  parts

$$h_i = \sqrt{\frac{i}{i+1}} \log \frac{\sqrt[i]{\prod_{j=1}^i x_j}}{x_{i+1}}$$

for  $i=1 \dots k$ .

## Usage

```
ilr_basis(dim, type = "default")
```

## Arguments

dim	number of components
type	if different than 'pivot' (pivot balances) or 'cdp' (codapack balances) default balances are returned, which computes a triangular Helmert matrix as defined by Egozcue et al., 2013.

## Details

Modifying parameter type (pivot or cdp) other ilr basis can be generated

## Value

matrix

## References

Egozcue, J.J., Pawlowsky-Glahn, V., Mateu-Figueras, G. and Barceló-Vidal C. (2003). *Isometric logratio transformations for compositional data analysis*. *Mathematical Geology*, **35**(3) 279-300

## Examples

```
ilr_basis(5)
```

---

parliament2017	<i>Results of catalan parliament elections in 2017 by regions.</i>
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---

**Description**

Results of catalan parliament elections in 2017 by regions.

**Usage**

```
parliament2017
```

**Format**

A data frame with 42 rows and 9 variables:

**com** Region  
**cs** Votes to Ciutadans party  
**jxcat** Votes to Junts per Catalunya party  
**erc** Votes to Esquerra republicana de Catalunya party  
**psc** Votes to Partit socialista de Catalunya party  
**catsp** Votes to Catalunya si que es pot party  
**cup** Votes to Candidatura d'unitat popular party  
**pp** Votes to Partit popular party  
**other** Votes to other parties

**Source**

<http://www.idescat.cat/tema/elecc>

---

pb_basis	<i>Isometric log-ratio basis based on Principal Balances.</i>
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---

**Description**

Exact method to calculate the principal balances of a compositional dataset. Different methods to approximate the principal balances of a compositional dataset are also included.

**Usage**

```
pb_basis(  
  X,  
  method,  
  constrained.complete_up = FALSE,  
  cluster.method = "ward.D2",  
  ordering = TRUE,  
  ...  
)
```

**Arguments**

X	compositional dataset
method	method to be used with Principal Balances. Methods available are: 'exact', 'constrained' or 'cluster'.
constrained.complete_up	When searching up, should the algorithm try to find possible siblings for the current balance (TRUE) or build a parent directly forcing current balance to be part of the next balance (default: FALSE). While the first is more exhaustive and given better results the second is faster and can be used with high dimensional datasets.
cluster.method	Method to be used with the hclust function (default: 'ward.D2') or any other method available in hclust function
ordering	should the principal balances found be returned ordered? (first column, first principal balance and so on)
...	parameters passed to hclust function

**Value**

matrix

**References**

Martín-Fernández, J.A., Pawłowsky-Glahn, V., Egozcue, J.J., Tolosana-Delgado R. (2018). Advances in Principal Balances for Compositional Data. *Mathematical Geosciences*, 50, 273-298.

**Examples**

```
set.seed(1)
X = matrix(exp(rnorm(5*100)), nrow=100, ncol=5)

# Optimal variance obtained with Principal components
(v1 <- apply(coordinates(X, 'pc'), 2, var))
# Optimal variance obtained with Principal balances
(v2 <- apply(coordinates(X,pb_basis(X, method='exact')), 2, var))
# Solution obtained using constrained method
(v3 <- apply(coordinates(X,pb_basis(X, method='constrained')), 2, var))
# Solution obtained using Ward method
(v4 <- apply(coordinates(X,pb_basis(X, method='cluster')), 2, var))

# Plotting the variances
barplot(rbind(v1,v2,v3,v4), beside = TRUE, ylim = c(0,2),
        legend = c('Principal Components', 'PB (Exact method)',
                  'PB (Constrained)', 'PB (Ward approximation)'),
        names = paste0('Comp.', 1:4), args.legend = list(cex = 0.8), ylab = 'Variance')
```

---

`pc_basis`*Isometric log-ratio basis based on Principal Components.*

---

**Description**

Different approximations to approximate the principal balances of a compositional dataset.

**Usage**

```
pc_basis(X)
```

**Arguments**

X                    compositional dataset

**Value**

matrix

---

`print.coda`*Printing coordinates*

---

**Description**

The function hides the basis attribute. An option is included to show such basis.

**Usage**

```
## S3 method for class 'coda'  
print(x, ..., basis = getOption("coda.base.basis"))
```

**Arguments**

x                    coordinates  
...                   parameters passed to print function  
basis                boolean to show or not the basis with the output

---

sbp_basis	<i>Isometric log-ratio basis based on Balances Build an <a href="#">ilr_basis</a> using a sequential binary partition or a generic coordinate system based on balances.</i>
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---

## Description

Isometric log-ratio basis based on Balances Build an [ilr\\_basis](#) using a sequential binary partition or a generic coordinate system based on balances.

## Usage

```
sbp_basis(..., data = NULL, silent = F)
```

## Arguments

...	balances to consider
data	composition from where name parts are extracted
silent	inform about orthogonality

## Value

matrix

## Examples

```
X = data.frame(a=1:2, b=2:3, c=4:5, d=5:6, e=10:11, f=100:101, g=1:2)
sbp_basis(b1 = a~b+c+d+e+f+g,
          b2 = b~c+d+e+f+g,
          b3 = c~d+e+f+g,
          b4 = d~e+f+g,
          b5 = e~f+g,
          b6 = f~g, data = X)
sbp_basis(b1 = a~b,
          b2 = b1~c,
          b3 = b2~d,
          b4 = b3~e,
          b5 = b4~f,
          b6 = b5~g, data = X)
# A non-orthogonal basis can also be calculated.
sbp_basis(b1 = a+b+c~e+f+g,
          b2 = d~a+b+c,
          b3 = d~e+g,
          b4 = a~e+b,
          b5 = b~f,
          b6 = c~g, data = X)
```

---

variation_array	<i>Variation array is returned.</i>
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**Description**

Variation array is returned.

**Usage**

```
variation_array(X, only_variation = FALSE)
```

**Arguments**

`X` Compositional dataset  
`only_variation` if TRUE only the variation matrix is calculated

**Value**

variation array matrix

**Examples**

```
set.seed(1)  
X = matrix(exp(rnorm(5*100)), nrow=100, ncol=5)  
variation_array(X)  
variation_array(X, only_variation = TRUE)
```

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