

# Package ‘RPS’

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**Type** Package

**Title** Resistant Procrustes Superimposition

**Version** 1.0.1

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**Description** Based on RPS tools, a rather complete resistant shape analysis of 2D and 3D datasets based on landmarks can be performed. In addition, landmark-based resistant shape analysis of individual asymmetry in 2D for matching or object symmetric structures is also possible.

**License** GPL-3

**Encoding** UTF-8

**LazyData** true

**RoxygenNote** 6.0.1

**Imports** geomorph, MASS, igraph, ape, matlab, Gmedian

**NeedsCompilation** no

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## R topics documented:

'RPS'	2
cmdistance_RPS	3
eucunivMDS_RPS	4
matchingsymm_RPS	5
objectsymm_RPS	5
procrustesCM_RPS	6
readlandtxtMorphJ_RPS	7
resdistance_RPS	8
resunivMDS_RPS	9

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'RPS' *Resistant Procrustes Superposition Package in R (RPS): a novel package for landmark-based resistant shape analysis*

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### Description

RPS provides a set of tools to perform a rather complete descriptive landmark-based resistant shape analysis 3D and 2D, following Torcida et al. 2014 ("An integrated approach for landmark-based resistant shape analysis in 3D", *Evol. Biol.* 41(2):351\_366). More specifically, these tools enable to obtain: i) a generalized resistant Procrustes superposition (`robgit_RPS.R`) for a set of configurations of landmarks either in 3D and 2D; ii) a resistant distance (`resdistance_RPS.R`) to quantify shape differences obtained following the resistant Procrustes superimposition, and iii) a resistant ordination (`resunivMDS_RPS.R`) of the superimposed configurations based on the universal Multidimensional Scaling from (Agarwal et al. 2010). Corresponding least squares (LS) counterparts of all these tools (`procrustesCM_RPS.R`, `cmdistance_RPS.R` and `eucunivMDS_RPS.R`, respectively) have also been implemented in `RPS_R` to offer a more complete and self-contained set of shape analysis descriptive tools. This enables the comparison of the LS and resistant superimposition results when applied to the same dataset. Also included is a rather new method for a resistant analysis of individual shape asymmetry for configurations of landmarks in 2D with bilateral symmetry (matching or object symmetry), following Torcida et al. 2016 ("A resistant method for landmark-based analysis of individual asymmetry in two dimensions", *Quant. Biol.* 4(4):270\_282). The main tools enable to estimate the resistant symmetric shape under matching symmetry (`matchingsymm_RPS.R`) and the resistant symmetric shape estimation under object symmetry (`objectsymm_RPS.R`). In both cases, a plot of the results and the table of landmarks contributions to asymmetry are also offered.

### Usage

```
robgit_RPS(X, consenso = FALSE)
```

### Arguments

X	A s-dimensional array of n x k matrices (k configurations of n landmarks), each representing the shape of an object
consenso	A logical value that determines if the consensus configuration is returned.

### Value

s-dimensional array of n x k matrices, representing the (resistant) superimposed objects

### Functions

`eucunivMDS_RPS`, `resunivMDS_RPS`, `cmdistance_RPS`, `resdistance_RPS`, `readlandtxtMorphJ_RPS`, `robgit_RPS`, `matchingsymm_RPS`, `objectsymm_RPS`, `procrustesCM_RPS`

### Author(s)

Guillermo Pacheco, Viviana Ferraggine, Sebastian Torcida

**Examples**

```

source = array(matrix(nrow = 8, ncol = 3), c(8, 3, 3), dimnames = NULL)
source[, , 1] <- matrix(c(3, 0, 0, 3, 0, 1, 3, 1, 1, 3, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1, 0, 1, 0),
  , nrow = 8, ncol = 3, byrow = TRUE)
source[, , 2] <- matrix(c(3, 0, 0, 3, 0, 0.5, 3, 1, 0.75, 3, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1, 0, 1, 0.25),
  , nrow = 8, ncol = 3, byrow = TRUE)
source[, , 3] <- matrix(c(5, 2, 1, 3, 0, 1.5, 3.4, 1, 1.75, 3, 1, 0, 0, 0, 0, 0, 2, 1, 0, 3, 1, 0, 1, 0.75),
  , nrow = 8, ncol = 3, byrow = TRUE)
result <- RPS::robgit_RPS(source, consenso = FALSE)
result

```

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cmdistance_RPS	<i>This function computes the least-squares Procrustes distance between each pair of matrices (configurations of landmarks) from the input set</i>
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**Description**

This function computes the least-squares Procrustes distance between each pair of matrices (configurations of landmarks) from the input set

**Usage**

```
cmdistance_RPS(X)
```

**Arguments**

X                    The input set of nx3 matrices (objects)

**Value**

The LS Procrustes distance matrix between pairs of objects

**Author(s)**

Guillermo Pacheco, Viviana Ferraggine, Sebastian Torcida

**Examples**

```

source = array(matrix(nrow = 8, ncol = 3), c(8, 3, 3), dimnames = NULL)
source[, , 1] <- matrix(c(3, 0, 0, 3, 0, 1, 3, 1, 1, 3, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1, 0, 1, 0),
  , nrow = 8, ncol = 3, byrow = TRUE)
source[, , 2] <- matrix(c(3, 0, 0, 3, 0, 0.5, 3, 1, 0.75, 3, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1, 0, 1, 0.25),
  , nrow = 8, ncol = 3, byrow = TRUE)
source[, , 3] <- matrix(c(5, 2, 1, 3, 0, 1.5, 3.4, 1, 1.75, 3, 1, 0, 0, 0, 0, 0, 2, 1, 0, 3, 1, 0, 1, 0.75),
  , nrow = 8, ncol = 3, byrow = TRUE)
result <- RPS::robgit_RPS(source)
RPS::cmdistance_RPS(result)

```

---

eucunivMDS\_RPS      *Given a  $n \times n$  distance matrix  $D$  (not necessarily Euclidean) and a initial set  $X_0$  of  $n$  seeds in  $k$  dim (that is, an initial  $n \times k$  matrix), this function finds a set of  $n$  points in  $k$  dimensions  $X$  (a final  $n \times k$  matrix) through a least-squares criterion such that the  $n \times n$  matrix  $D_k$  of euclidean distances among these new points  $X$  is as close as possible to  $D$ .*

---

### Description

Given a  $n \times n$  distance matrix  $D$  (not necessarily Euclidean) and a initial set  $X_0$  of  $n$  seeds in  $k$  dim (that is, an initial  $n \times k$  matrix), this function finds a set of  $n$  points in  $k$  dimensions  $X$  (a final  $n \times k$  matrix) through a least-squares criterion such that the  $n \times n$  matrix  $D_k$  of euclidean distances among these new points  $X$  is as close as possible to  $D$ .

### Usage

```
eucunivMDS_RPS(D, k = 2)
```

### Arguments

$D$                     distance matrix  $n \times n$  to be approximated  
 $k$                      dimension of output results

### Value

$X$  A set of  $n$  points in  $k$  dimensions

### Author(s)

Guillermo Pacheco, Viviana Ferraggine, Sebastian Torcida

### Examples

```
source = array(matrix(nrow = 8, ncol = 3), c(8, 3, 3), dimnames = NULL)
source[, , 1] <- matrix(c(3, 0, 0, 3, 0, 1, 3, 1, 1, 3, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1, 0, 1, 0)
, nrow = 8, ncol = 3, byrow = TRUE)
source[, , 2] <- matrix(c(3, 0, 0, 3, 0, 0.5, 3, 1, 0.75, 3, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1, 0, 1, 0.25)
, nrow = 8, ncol = 3, byrow = TRUE)
source[, , 3] <- matrix(c(5, 2, 1, 3, 0, 1.5, 3.4, 1, 1.75, 3, 1, 0, 0, 0, 0, 0, 2, 1, 0, 3, 1, 0, 1, 0.75)
, nrow = 8, ncol = 3, byrow = TRUE)
result <- RPS::robfit_RPS(source, consenso = FALSE)
distance <- RPS::resdistance_RPS(result)
RPS::eucunivMDS_RPS(distance, 2)
```

---

matchingsymm_RPS	<i>This function obtains the individual resistant-symmetric shape for 2D matching-symmetry data. The input is an array A of size n (landmarks) x p (dimensions) x 2k (objects: the left-right sides for each). Configurations are ordered in this way: left side Object 1, right side Object 1, left side Object 2, right side Object 2, etc</i>
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---

### Description

This function obtains the individual resistant-symmetric shape for 2D matching-symmetry data. The input is an array A of size n (landmarks) x p (dimensions) x 2k (objects: the left-right sides for each). Configurations are ordered in this way: left side Object 1, right side Object 1, left side Object 2, right side Object 2, etc

### Usage

```
matchingsymm_RPS(A,ctr="gmedian",legend.loc="topleft")
```

### Arguments

A	an array of size n (landmarks) x 2 (in 2D) x 2k (left/right sides for k configurations)
ctr	Centering options: "gmedian" (the spatial or geometric median, default choice), "median" (the componentwise median), "mean" (the average)
legend.loc	The location of the legend for the plot.result function

### Author(s)

Federico Lotto, Sebastian Torcida

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objectsymm_RPS	<i>This function obtains the individual resistant-symmetric shape for 2D object-symmetry data. The input is an array A of size n (landmarks) x p (dimensions) x k (objects) Landmarks must be in this order: saggital (or unpaired) landmarks first, then left paired landmarks and finally right paired landmarks. Configurations are ordered in this way: L side Object 1 and R side Object 1, L side Object 2 and R side Object 2, etc</i>
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### Description

This function obtains the individual resistant-symmetric shape for 2D object-symmetry data. The input is an array A of size n (landmarks) x p (dimensions) x k (objects) Landmarks must be in this order: saggital (or unpaired) landmarks first, then left paired landmarks and finally right paired landmarks. Configurations are ordered in this way: L side Object 1 and R side Object 1, L side Object 2 and R side Object 2, etc

**Usage**

```
objectsymm_RPS(A,ctr="gmedian",prs.file,proj.met="msum",legend.loc="topleft")
```

**Arguments**

A	Input data: an array or matrix of size n (landmarks) x 2 (in 2D) x k (objects)
ctr	Centering options: "gmedian" (the spatial or gemetric median, default choice), "median" (the componentwise median), "mean" (the average)
prs.file	This is a .txt file indicating the L+R paired landmarks as rows: e.g. 7 15; 8 16; etc.
proj.met	The choice to compute the saggital axis: sum or median of projections
legend.loc	The location of the legend for the plot.result function

**Value**

w

**Author(s)**

Federico Lotto, Sebastian Torcida

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procrustesCM_RPS	<i>This function s simply a wrapper for the geomorph function gpagen that performs the classical least squares Procrustes superimposition of the input configurations of landmarks.</i>
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**Description**

This function s simply a wrapper for the geomorph function gpagen that performs the classical least squares Procrustes superimposition of the input configurations of landmarks.

**Usage**

```
procrustesCM_RPS(X)
```

**Arguments**

X	A s-dimensional array (s=2 or s=3) of n x k matrices, representing shapes of k objects through n landmarks in s dimensions
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**Value**

s-dimensional array of n x k matrices, representing shapes of k objects following superimposition.

**Author(s)**

Dean C.Adams, Michael Collyer

**Examples**

```

source = array(matrix(nrow = 8, ncol = 3), c(8, 3, 3), dimnames = NULL)
source[, , 1] <- matrix(c(3, 0, 0, 3, 0, 1, 3, 1, 1, 3, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1, 0, 1, 0),
                        , nrow = 8, ncol = 3, byrow = TRUE)
source[, , 2] <- matrix(c(3, 0, 0, 3, 0, 0.5, 3, 1, 0.75, 3, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1, 0, 1, 0.25),
                        , nrow = 8, ncol = 3, byrow = TRUE)
source[, , 3] <- matrix(c(5, 2, 1, 3, 0, 1.5, 3.4, 1, 1.75, 3, 1, 0, 0, 0, 0, 0, 2, 1, 0, 3, 1, 0, 1, 0.75),
                        , nrow = 8, ncol = 3, byrow = TRUE)
result <- RPS::procrustesCM_RPS(source)
result

```

---

`readlandtxtMorphJ_RPS` *Reads a MorphoJ .txt file and returns it as an array of  $n \times k$  matrices in  $s$  dimensions ( $s=2$  or  $s=3$ )*

---

**Description**

Reads a MorphoJ .txt file and returns it as an array of  $n \times k$  matrices in  $s$  dimensions ( $s=2$  or  $s=3$ )

**Usage**

```
readlandtxtMorphJ_RPS(path, dim)
```

**Arguments**

<code>path</code>	Path of file
<code>dim</code>	Dimension of the data (2D or 3D).

**Value**

A  $s$ -dimensional array of  $n \times k$  matrices and a list of the corresponding object's names

**Author(s)**

Guillermo Pacheco, Viviana Ferraggine, Sebastian Torcida

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resdistance_RPS	<i>This function computes the a resistant distance between each pair of matrices from the input set</i>
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## Description

This function computes the a resistant distance between each pair of matrices from the input set

## Usage

```
resdistance_RPS(X)
```

## Arguments

X                    The input set of nx3 matrices (objects)

## Value

This function computes the sum of non-squared euclidean distances across landmarks for each pair of matrices from the input set

## Author(s)

Guillermo A. Pacheco, Viviana Ferraggine, Sebastian Torcida

## Examples

```
source = array(matrix(nrow = 8, ncol = 3), c(8, 3, 3), dimnames = NULL)
source[, , 1] <- matrix(c(3, 0, 0, 3, 0, 1, 3, 1, 1, 3, 1, 0, 0, 0, 0, 0, 1, 0, 1, 1, 0, 1, 0),
  , nrow = 8, ncol = 3, byrow = TRUE)
source[, , 2] <- matrix(c(3, 0, 0, 3, 0, 0.5, 3, 1, 0.75, 3, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1, 0, 1, 0.25),
  , nrow = 8, ncol = 3, byrow = TRUE)
source[, , 3] <- matrix(c(5, 2, 1, 3, 0, 1.5, 3.4, 1, 1.75, 3, 1, 0, 0, 0, 0, 0, 2, 1, 0, 3, 1, 0, 1, 0.75),
  , nrow = 8, ncol = 3, byrow = TRUE)
result <- RPS::robfit_RPS(source, consenso = FALSE)
RPS::resdistance_RPS(result)
```



---

resunivMDS_RPS	<i>Given a <math>n \times n</math> distance matrix <math>D</math> (not necessarily Euclidean) and a initial set <math>X_0</math> (that is, a <math>n \times k</math> matrix) of <math>n</math> seeds in <math>k</math> dim, this function finds a set of <math>n</math> points in <math>k</math> dimensions <math>X</math> (that is, a <math>k \times n</math> matrix) using a resistant criterion such that the <math>n \times n</math> matrix <math>D_k</math> of euclidean distances among these new points <math>X</math> is as close as possible to <math>D</math>.</i>
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---

### Description

Given a  $n \times n$  distance matrix  $D$  (not necessarily Euclidean) and a initial set  $X_0$  (that is, a  $n \times k$  matrix) of  $n$  seeds in  $k$  dim, this function finds a set of  $n$  points in  $k$  dimensions  $X$  (that is, a  $k \times n$  matrix) using a resistant criterion such that the  $n \times n$  matrix  $D_k$  of euclidean distances among these new points  $X$  is as close as possible to  $D$ .

### Usage

```
resunivMDS_RPS(D,k)
```

### Arguments

D	distance matrix $n \times n$ to be approximated
k	dimension of output results

### Value

X A set of  $n$  points in  $k$  dimensions

### Author(s)

Guillermo Pacheco, Viviana Ferraggine, Sebastian Torcida

### Examples

```
source = array(matrix(nrow = 8, ncol = 3), c(8, 3, 3), dimnames = NULL)
source[, , 1] <- matrix(c(3, 0, 0, 3, 0, 1, 3, 1, 1, 3, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1, 0, 1, 0)
, nrow = 8, ncol = 3, byrow = TRUE)
source[, , 2] <- matrix(c(3, 0, 0, 3, 0, 0.5, 3, 1, 0.75, 3, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1, 0, 1, 0.25)
, nrow = 8, ncol = 3, byrow = TRUE)
source[, , 3] <- matrix(c(5, 2, 1, 3, 0, 1.5, 3.4, 1, 1.75, 3, 1, 0, 0, 0, 0, 0, 2, 1, 0, 3, 1, 0, 1, 0.75)
, nrow = 8, ncol = 3, byrow = TRUE)
result <- RPS::robfit_RPS(source, consenso = FALSE)
distance <- RPS::resdistance_RPS(result)
RPS::resunivMDS_RPS(distance, 2)
```

# Index

'RPS', 2  
'RPS'-package ('RPS'), 2  
  
cmdistance\_RPS, 3  
  
eucunivMDS\_RPS, 4  
  
matchingsymm\_RPS, 5  
  
objectsymm\_RPS, 5  
  
procrustesCM\_RPS, 6  
  
readlandtxtMorphJ\_RPS, 7  
resdistance\_RPS, 8  
resunivMDS\_RPS, 9  
robgit\_RPS ('RPS'), 2