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Contents

align_fPCA 4
align_new 6
beta 7
bootTB 7
boxplot.fdawarp 8
calc_shape_dist 10
classify 12
curve_boxplot 13
curve_depth 14
curve_dist 15
curve_geodesic 16
curve_pair_align 17
curve_to_q 18
curve2srvf 19
discrete2curve 19
discrete2warping 20
elastic.depth 20
elastic.distance 21
elastic.logistic 22
elastic.lpcr.regression 24
elastic.mlogistic 25
elastic.mlpcr.regression 26
elastic.pcr.regression 27
elastic.prediction 29
elastic.regression 30
elastic_amp_change_ff 31
elastic_change_fpca 32
elastic_ph_change_ff 34
f_plot 35
f_to_srvf 35
fastpns 36
fastPNSe2s 37
fastPNSs2e 38
fdasrvf 38
function_group_warp_bayes 40
function_mean_bayes 41
gam_to_h 42
gam_to_psi 43
gam_to_v 43
gauss_model 44

get_curve_centroid . . . 45
get_distance_matrix . . . 45
get_hilbert_sphere_distance . . . 46
get_identity_warping . . . 47
get_l2_distance . . . 48
get_l2_inner_product . . . 48
get_l2_norm . . . 49
get_shape_distance . . . 49
get_warping_distance . . . 51
gradient . . . 51
growth_vel . . . 52
h_to_gam . . . 53
horizFPCA . . . 53
horizFPNS . . . 54
im . . . 55
interparc . . . 55
inv_exp_map . . . 56
invertGamma . . . 57
joint_gauss_model . . . 58
jointFPCA . . . 59
jointFPCAh . . . 60
kmeans_align . . . 61
LongRunCovMatrix . . . 64
multiple_align_functions . . . 64
multiple_align_multivariate . . . 66
multivariate_karcher_cov . . . 68
multivariate_karcher_mean . . . 69
multivariate_pca . . . 71
optimum.reparam . . . 72
outlier.detection . . . 73
pair_align_functions . . . 74
pair_align_functions_bayes . . . 75
pair_align_functions_expomap . . . 77
pair_align_image . . . 79
pcaTB . . . 80
plot_curve . . . 81
ppd . . . 82
predict.curve_pca . . . 83
predict.hfpca . . . 84
predict.jfpca . . . 85
predict.jfpcah . . . 85
predict.lpcr . . . 86
predict.mlpcr . . . 87
predict.pcr . . . 88
predict.vfpca . . . 88
psi_to_gam . . . 89
q_to_curve . . . 90
reparam_curve . . . 90

reparam_image . . . 92
resamplecurve . . . 93
rgam . . . 93
rotation_pca . . . 94
sample_shapes . . . 95
shape_CI . . . 96
simu_data . . . 97
simu_warp . . . 97
simu_warp_median . . . 98
smooth.data . . . 98
SqrtMean . . . 99
SqrtMeanInverse . . . 100
SqrtMedian . . . 100
srvf_to_f . . . 101
srvf2curve . . . 102
time_warping . . . 103
to_hilbert_sphere . . . 105
toy_data . . . 105
toy_warp . . . 106
v_to_curve . . . 106
v_to_gam . . . 107
vertFPCA . . . 107
warp_curve . . . 108
warp_f_gamma . . . 109
warp_q_gamma . . . 110
warp_srvf . . . 110

Index **112**

`align_fPCA` *Group-wise function alignment and PCA Extractions*

Description

This function aligns a collection of functions while extracting principal components.

Usage

```
align_fPCA(
  f,
  time,
  num_comp = 3L,
  showplot = TRUE,
  smooth_data = FALSE,
  sparam = 25L,
  parallel = FALSE,
  cores = NULL,
  max_iter = 51L,
```

```
  lambda = 0
)
```

Arguments

f	A numeric matrix of shape $M \times N$ specifying a sample of N 1-dimensional curves observed on a grid of size M.
time	A numeric vector of length M specifying the grid on which functions f have been evaluated.
num_comp	An integer value specifying the number of principal components to extract. Defaults to 3L.
showplot	A boolean specifying whether to display plots along the way. Defaults to TRUE.
smooth_data	A boolean specifying whether to smooth data using box filter. Defaults to FALSE.
sparam	An integer value specifying the number of times to apply box filter. Defaults to 25L. This argument is only used if smooth_data == TRUE.
parallel	A boolean specifying whether computations should run in parallel. Defaults to FALSE.
cores	An integer value specifying the number of cores to use for parallel computations. Defaults to NULL in which case it uses all available cores but one. This argument is only used when parallel == TRUE.
max_iter	An integer value specifying the maximum number of iterations. Defaults to 51L.
lambda	A numeric value specifying the elasticity. Defaults to 0.0.

Value

A list with the following components:

- f0: A numeric matrix of shape $M \times N$ storing the original functions;
- fn: A numeric matrix of the same shape as f0 storing the aligned functions;
- qn: A numeric matrix of the same shape as f0 storing the aligned SRSFs;
- q0: A numeric matrix of the same shape as f0 storing the SRSFs of the original functions;
- mqn: A numeric vector of length M storing the mean SRSF;
- gam: A numeric matrix of the same shape as f0 storing the estimated warping functions;
- vfpca: A list storing information about the vertical PCA with the following components:
  - q_pca: A numeric matrix of shape $(M+1) \times 5 \times \text{num_comp}$ storing the first 3 principal directions in SRSF space; the first dimension is $M + 1$ because, in SRSF space, the original functions are represented by the SRSF and the initial value of the functions.
  - f_pca: A numeric matrix of shape $M \times 5 \times \text{num_comp}$ storing the first 3 principal directions in original space;
  - latent: A numeric vector of length $M + 1$ storing the singular values of the SVD decomposition in SRSF space;
  - coef: A numeric matrix of shape $N \times \text{num_comp}$ storing the scores of the N original functions on the first num_comp principal components;

  - U: A numeric matrix of shape $(M+1) \times (M+1)$ storing the eigenvectors associated with the SVD decomposition in SRSF space.
- Dx: A numeric vector of length max_iter storing the value of the cost function at each iteration.

References

Tucker, J. D., Wu, W., Srivastava, A., Generative models for functional data using phase and amplitude separation, Computational Statistics and Data Analysis (2012), 10.1016/j.csda.2012.12.001.

Examples

```
## Not run:
  out <- align_fPCA(simu_data$f, simu_data$time)

## End(Not run)
```

---

align_new *Align New Functions to a Fitted Template*

Description

This function aligns new functions to the template SRVF (mqn) of an fdawarp object with the same lambda, penalty and optimization method used by time_warping() to fit it. When the fit smoothed the data (smooth_data = TRUE), the new functions are smoothed the same way first.

Usage

```
align_new(x, f_new)
```

Arguments

x	An object of class fdawarp typically produced by time_warping().
f_new	A numeric matrix of shape $M \times K$ of K new functions sampled at x$time, or a numeric vector of length M for a single function.

Value

A list with components (one column per function):

- fn: $M \times K$ aligned functions;
- qn: $M \times K$ aligned SRVFs;
- gam: $M \times K$ warping functions.

Examples

```
## Not run:
out <- time_warping(simu_data$f, simu_data$time)
aligned <- align_new(out, simu_data$f[, 1:3])

## End(Not run)
```

`beta` *MPEG7 Curve Dataset*

Description

Contains the MPEG7 curve data set.

Usage

```
beta
```

Format

`beta`**:**
An array of shape $2 \times 100 \times 65 \times 20$ storing a sample of 20 curves from R to R^2 distributed in 65 different classes, evaluated on a grid of size 100.

`bootTB` *Tolerance Bound Calculation using Bootstrap Sampling*

Description

This function computes tolerance bounds for functional data containing phase and amplitude variation using bootstrap sampling

Usage

```
bootTB(
  f,
  time,
  a = 0.05,
  p = 0.99,
  B = 500,
  no = 5,
  Nsamp = 100,
  parallel = TRUE
)
```

Arguments

f	matrix of functions
time	vector describing time sampling
a	confidence level of tolerance bound (default = 0.05)
p	coverage level of tolerance bound (default = 0.99)
B	number of bootstrap samples (default = 500)
no	number of principal components (default = 5)
Nsamp	number of functions per bootstrap (default = 100)
parallel	enable parallel processing (default = TRUE)

Value

Returns a list containing

amp	amplitude tolerance bounds
ph	phase tolerance bounds

References

J. D. Tucker, J. R. Lewis, C. King, and S. Kurtek, "A Geometric Approach for Computing Tolerance Bounds for Elastic Functional Data," Journal of Applied Statistics, 10.1080/02664763.2019.1645818, 2019.

Tucker, J. D., Wu, W., Srivastava, A., Generative Models for Function Data using Phase and Amplitude Separation, Computational Statistics and Data Analysis (2012), 10.1016/j.csda.2012.12.001.

Jung, S. L. a. S. (2016). "Combined Analysis of Amplitude and Phase Variations in Functional Data." arXiv:1603.01775.

Examples

```
## Not run:
  out1 <- bootTB(simu_data$f, simu_data$time)

## End(Not run)
```

---

boxplot.fdawarp *Functional Boxplot*

Description

This function computes the required statistics for building up a boxplot of the aligned functional data. Since the process of alignment provides separation of phase and amplitude variability, the computed boxplot can focus either on amplitude variability or phase variability.

Usage

```
## S3 method for class 'fdawarp'
boxplot(
  x,
  variability_type = c("amplitude", "phase"),
  alpha = 0.05,
  range = 1,
  what = c("stats", "plot", "plot+stats"),
  ...
)

## S3 method for class 'ampbox'
boxplot(x, ...)

## S3 method for class 'phbox'
boxplot(x, ...)

## S3 method for class 'curvebox'
boxplot(x, ...)
```

Arguments

x	An object of class fdawarp typically produced by time_warping() or of class ampbox or phbox typically produced by boxplot.fdawarp().
variability_type	A string specifying which kind of variability should be displayed in the boxplot. Choices are "amplitude" or "phase". Defaults to "amplitude".
alpha	A numeric value specifying the quantile value. Defaults to 0.05 which uses the 95% quantile.
range	A positive numeric value specifying how far the plot whiskers extend out from the box. The whiskers extend to the most extreme data point which is no more than range times the interquartile range from the box. Defaults to 1.0.
what	A string specifying what the function should return. Choices are "plot", "stats" or "plot+stats". Defaults to "stats".
...	Unused here.

Details

The function boxplot.fdawarp() returns optionally an object of class either ampbox if variability_type = "amplitude" or phbox if variability_type = "phase". S3 methods specialized for objects of these classes are provided as well to avoid re-computation of the boxplot statistics.

Value

If what contains stats, a list containing the computed statistics necessary for drawing the boxplot. Otherwise, the function simply draws the boxplot and no object is returned.

Examples

```
## Not run:
out <- time_warping(simu_data$f, simu_data$time)
boxplot(out, what = "stats")

## End(Not run)
```

`calc_shape_dist` *Elastic Shape Distance*

Description

Calculates the elastic shape distance between two curves `beta1` and `beta2`.

Usage

```
calc_shape_dist(
  beta1,
  beta2,
  mode = "O",
  alignment = TRUE,
  rotation = TRUE,
  scale = TRUE,
  optim_method = "DP",
  include.length = FALSE,
  lambda = 0
)
```

Arguments

`beta1`	A numeric matrix of shape $L \times M$ specifying an L-dimensional curve evaluated on M sample points.
`beta2`	A numeric matrix of shape $L \times M$ specifying an L-dimensional curve evaluated on M sample points. This curve will be aligned to `beta1`.
`mode`	A character string specifying whether the input curves should be considered open (`mode == "O"`) or closed (`mode == "C"`). Defaults to `"O"`.
`alignment`	A boolean value specifying whether the curves should be aligned before computing the distance matrix. Defaults to `TRUE`.
`rotation`	A boolean specifying whether the distance should be made invariant by rotation. Defaults to `TRUE`.
`scale`	A boolean specifying whether the distance should be made invariant by scaling. This is effectively achieved by making SRVFs having unit length and switching to an appropriate metric on the sphere between normalized SRVFs. Defaults to `TRUE`.
`optim_method`	A string specifying the algorithm used for optimization. Choices are `"DP"` and `"DPo"` Defaults to `"DP"`.

include.length A boolean specifying whether to include information about the actual length of the SRVFs in the metric when using normalized SRVFs to achieve scale invariance. This only applies if scale == TRUE. Defaults to FALSE.

lambda A numeric value specifying the weight of a penalty term that constraints the warping function to be not too far from the identity. Defaults to 0.0.

Details

Distances are computed between the SRVFs of the original curves. Hence, they are intrinsically invariant to position. The user can then choose to make distances invariant to rotation and scaling by setting rotation and scale accordingly. Distances can also be made invariant to reparameterization by setting alignment = TRUE, in which case curves are aligned using an appropriate action of the diffeomorphism group on the space of SRVFs.

Value

A list with the following components:

- d: the amplitude (geodesic) distance;
- dx: the phase distance;
- q1: the SRVF of beta1;
- q2n: the SRVF of beta2 after alignment and possible optimal rotation and scaling;
- beta1: the input curve beta1;
- beta2n: the input curve beta2 after alignment and possible optimal rotation and scaling.
- R: the optimal rotation matrix that has been applied to the second curve;
- gam: the optimal warping function that has been applied to the second curve;
- betascale: the optimal scaling factor that has been applied to the second curve.

References

Srivastava, A., Klassen, E., Joshi, S., Jermyn, I., (2011). Shape analysis of elastic curves in euclidean spaces. Pattern Analysis and Machine Intelligence, IEEE Transactions on 33 (7), 1415-1428.

Kurtek, S., Srivastava, A., Klassen, E., and Ding, Z. (2012), "Statistical Modeling of Curves Using Shapes and Related Features," Journal of the American Statistical Association, 107, 1152–1165.

Srivastava, A., Klassen, E. P. (2016). Functional and shape data analysis, 1. New York: Springer.

Examples

```
out <- calc_shape_dist(beta[, , 1, 1], beta[, , 1, 4])
```

`classify` *Place New Functions in Boxplot Regions*

Description

Aligns new functions to the template of the fit behind an amplitude or phase boxplot (see `align_new()`) and compares their distance to the median with the boxplot regions: the 50% central region, the $(1-\alpha)$ central region, within the whiskers, or outlier.

Usage

```
classify(x, ...)

## S3 method for class 'ampbox'
classify(x, f_new, ...)

## S3 method for class 'phbox'
classify(x, f_new, ...)
```

Arguments

`x`	An object of class `ampbox` or `phbox` typically produced by `boxplot.fdawarp()` with `what = "stats"`.
`...`	Unused here.
`f_new`	A numeric matrix of shape $M \times K$ of K new functions sampled at the time grid of the fit, or a numeric vector of length M for a single function.

Details

For an `ampbox`, the distance is the amplitude (SRVF) distance of the aligned function to the median SRVF; for a `phbox`, it is the phase distance of its warping function to the median warping function.

Value

A list with components (one entry or column per function):

- `dist`: distance to the median;
- `level`: `1` = 50% central region, `2` = $(1-\alpha)$ central region, `3` = within whiskers, `4` = outlier;
- `region`: `level` as text;
- `side`: `"Q1"` or `"Q3"`, the quartile direction the function lies in;
- `fn`, `qn`, `gam`: the aligned functions, SRVFs and warping functions.

Examples

```
## Not run:
out <- time_warping(simu_data$f, simu_data$time, centroid_type = "median")
amp <- boxplot(out, variability_type = "amplitude", what = "stats")
classify(amp, simu_data$f[, 1:3])

## End(Not run)
```

`curve_boxplot` *Curve Boxplot*

Description

This function computes the required statistics for building up a boxplot of the aligned curve data. The computed boxplot focuses on the aligned curves.

Usage

```
curve_boxplot(
  x,
  alpha = 0.05,
  range = 1,
  what = c("stats", "plot", "plot+stats"),
  ...
)
```

Arguments

`x`	An object of class `fdacurve` typically produced by `multivariate_karcher_mean()`
`alpha`	A numeric value specifying the quantile value. Defaults to 0.05 which uses the 95% quantile.
`range`	A positive numeric value specifying how far the plot whiskers extend out from the box. The whiskers extend to the most extreme data point which is no more than `range` times the interquartile range from the box. Defaults to `1.0`.
`what`	A string specifying what the function should return. Choices are `"plot"`, `"stats"` or `"plot+stats"`. Defaults to `"plot"`.
`...`	Unused here.

Details

The function returns optionally an object of class either `curvebox`

Value

If `what` contains `stats`, a list containing the computed statistics necessary for drawing the boxplot. Otherwise, the function simply draws the boxplot and no object is returned.

Examples

```
## Not run:
out <- multivariate_karcher_mean(beta[, , 1, ], ms="median")
curve_boxplot(out, what = "stats")

## End(Not run)
```

---

curve_depth *Calculates elastic depth for a set of curves*

Description

This functions calculates the elastic depth between set of curves. If the curves are describing multidimensional functional data, then `rotated == FALSE` and `mode == 'O'`

Usage

```
curve_depth(beta, mode = "O", rotated = TRUE, scale = TRUE, parallel = FALSE)
```

Arguments

`beta`	Array of sizes $n \times T \times N$ for N curves of dimension n evaluated on a grid of T points
`mode`	Open (`"O"`) or Closed (`"C"`) curves
`rotated`	Include rotation (default = `TRUE`)
`scale`	scale curves to unit length (default = `TRUE`)
`parallel`	run computation in parallel (default = `TRUE`)

Value

Returns a list containing

`amp`	amplitude depth
`phase`	phase depth

References

T. Harris, J. D. Tucker, B. Li, and L. Shand, "Elastic depths for detecting shape anomalies in functional data," Technometrics, 10.1080/00401706.2020.1811156, 2020.

Examples

```
data("mpeg7")
# note: use more shapes and iterations, small for speed
out = curve_depth(beta[,,1,1:2])
```

curve_dist *Distance Matrix Computation*

Description

Computes the pairwise distance matrix between a set of curves using the elastic shape distance as computed by `calc_shape_dist()`.

Usage

```
curve_dist(
  beta,
  mode = "O",
  alignment = TRUE,
  rotation = FALSE,
  scale = FALSE,
  include.length = FALSE,
  lambda = 0,
  ncores = 1L
)
```

Arguments

`beta`	A numeric array of shape $L \times M \times N$ specifying the set of N curves of length M in L-dimensional space.
`mode`	A character string specifying whether the input curves should be considered open (`mode == "O"`) or closed (`mode == "C"`). Defaults to `"O"`.
`alignment`	A boolean value specifying whether the curves should be aligned before computing the distance matrix. Defaults to `TRUE`.
`rotation`	A boolean specifying whether the distance should be made invariant by rotation. Defaults to `TRUE`.
`scale`	A boolean specifying whether the distance should be made invariant by scaling. This is effectively achieved by making SRVFs having unit length and switching to an appropriate metric on the sphere between normalized SRVFs. Defaults to `TRUE`.
`include.length`	A boolean specifying whether to include information about the actual length of the SRVFs in the metric when using normalized SRVFs to achieve scale invariance. This only applies if `scale == TRUE`. Defaults to `FALSE`.
`lambda`	A numeric value specifying the weight of a penalty term that constraints the warping function to be not too far from the identity. Defaults to `0.0`.
`ncores`	An integer value specifying the number of cores to use for parallel computation. If `ncores` is greater than the number of available cores, a warning is issued and the maximum number of available cores is used. Defaults to `1L`.

Details

Distances are computed between the SRVFs of the original curves. Hence, they are intrinsically invariant to position. The user can then choose to make distances invariant to rotation and scaling by setting `rotation` and `scale` accordingly. Distances can also be made invariant to reparameterization by setting `alignment = TRUE`, in which case curves are aligned using an appropriate action of the diffeomorphism group on the space of SRVFs.

Value

A list of two objects, `Da` and `Dp`, each of class `dist` containing the amplitude and phase distances, respectively.

References

Srivastava, A., Klassen, E., Joshi, S., Jermyn, I., (2011). Shape analysis of elastic curves in euclidean spaces. Pattern Analysis and Machine Intelligence, IEEE Transactions on 33 (7), 1415-1428.

Kurtek, S., Srivastava, A., Klassen, E., and Ding, Z. (2012), "Statistical Modeling of Curves Using Shapes and Related Features," Journal of the American Statistical Association, 107, 1152–1165.

Srivastava, A., Klassen, E. P. (2016). Functional and shape data analysis, 1. New York: Springer.

Examples

```
out <- curve_dist(beta[, , 1, 1:4])
```

---

curve_geodesic *Form geodesic between two curves*

Description

Form geodesic between two curves using Elastic Method

Usage

```
curve_geodesic(beta1, beta2, k = 5)
```

Arguments

`beta1`	curve 1, provided as a matrix of dimensions $n \times T$ for n-dimensional curve evaluated on T sample points
`beta2`	curve 2, provided as a matrix of dimensions $n \times T$ for n-dimensional curve evaluated on T sample points
`k`	number of curves along geodesic (default 5)

Value

a list containing

geod	curves along geodesic (n,T,k)
geod_q	srvf's along geodesic

References

Srivastava, A., Klassen, E., Joshi, S., Jermyn, I., (2011). Shape analysis of elastic curves in euclidean spaces. Pattern Analysis and Machine Intelligence, IEEE Transactions on 33 (7), 1415-1428.

Examples

```
out <- curve_geodesic(beta[, , 1, 1], beta[, , 1, 5])
```

---

curve_pair_align *Pairwise align two curves*

Description

This function aligns to curves using Elastic Framework

Usage

```
curve_pair_align(beta1, beta2, mode = "O", rotation = TRUE, scale = TRUE)
```

Arguments

beta1	curve 1, provided as a matrix of dimensions $n \times T$ for n-dimensional curve evaluated on T sample points
beta2	curve 2, provided as a matrix of dimensions $n \times T$ for n-dimensional curve evaluated on T sample points
mode	Open ("O") or Closed ("C") curves
rotation	Include rotation (default = TRUE)
scale	scale curves to unit length (default = TRUE)

Value

a list containing

beta2n	aligned curve 2 to 1
q2n	aligned srvf 2 to 1
gam	warping function
q1	srvf of curve 1

beta1	centered curve 1
beta2	centered curve 2
R	rotation matrix
tau	seed

References

Srivastava, A., Klassen, E., Joshi, S., Jermyn, I., (2011). Shape analysis of elastic curves in euclidean spaces. Pattern Analysis and Machine Intelligence, IEEE Transactions on 33 (7), 1415-1428.

Examples

```
out <- curve_pair_align(beta[, , 1, 1], beta[, , 1, 5])
```

---

curve_to_q *Curve to SRVF Space*

Description

This function computes the SRVF of a given curve. The function also outputs the length of the original curve and the L^2 norm of the SRVF.

Usage

```
curve_to_q(beta, scale = TRUE)
```

Arguments

beta	A numeric matrix of shape $L \times M$ specifying a curve on an L-dimensional space observed on M points.
scale	A boolean value specifying whether the output SRVF function should be scaled to the hypersphere of L^2. When scale is TRUE, the length of the original curve becomes irrelevant. Defaults to TRUE.

Value

A list with the following components:

- q: A numeric matrix of the same shape as the input matrix beta storing the (possibly scaled) SRVF of the original curve beta;
- len: A numeric value specifying the length of the original curve;
- lenq: A numeric value specifying the L^2 norm of the SRVF of the original curve.

References

Srivastava, A., Klassen, E., Joshi, S., Jermyn, I., (2011). Shape analysis of elastic curves in Euclidean spaces. IEEE Transactions on Pattern Analysis and Machine Intelligence, **33**(7), 1415-1428.

Examples

```
q <- curve_to_q(beta[, , 1, 1])$q
```

`curve2srvf` *Converts a curve to its SRVF representation*

Description

Converts a curve to its SRVF representation

Usage

```
curve2srvf(beta, is_derivative = FALSE)
```

Arguments

`beta`	A numeric matrix of size $L \times M$ specifying a curve on an L-dimensional space observed on an evenly spaced grid of $[0, 1]$ of length M.
`is_derivative`	A boolean value specifying whether the input $beta$ is the derivative of the original curve. Defaults to `FALSE`.

Value

A function that takes a numeric vector s of values in $[0, 1]$ as input and returns the values of the SRVF of the original curve at s.

Examples

```
curve2srvf(beta[, , 1, 1])
```

`discrete2curve` *Converts a curve from matrix to functional data object*

Description

Converts a curve from matrix to functional data object

Usage

```
discrete2curve(beta)
```

Arguments

`beta`	A numeric matrix of size $L \times M$ specifying a curve on an L-dimensional space observed on an evenly spaced grid of $[0, 1]$ of length M.

Value

A function that takes a numeric vector s of values in $[0, 1]$ as input and returns the values of the original curve at s.

Examples

```
discrete2curve(beta[, , 1, 1])
```

`discrete2warping` *Converts a warping function from vector to functional data object*

Description

Converts a warping function from vector to functional data object

Usage

```
discrete2warping(gam)
```

Arguments

`gam`	A numeric vector of length M specifying a warping function on an evenly spaced grid of $[0, 1]$ of length M.

Value

A function that takes a numeric vector s of values in $[0, 1]$ as input and returns the values of the original warping function at s.

Examples

```
discrete2warping(toy_warp$gam[, 1])
```

`elastic.depth` *Calculates elastic depth*

Description

This functions calculates the elastic depth between set of functions in R^1

Usage

```
elastic.depth(f, time, lambda = 0, pen = "roughness", parallel = FALSE)
```

Arguments

f	matrix of sizes $M \times N$, specifying values of N function of M time points
time	vector of length M, specifying the sample points of functions
lambda	controls amount of warping (default = 0)
pen	alignment penalty (default = "roughness") options are second derivative ("roughness"), L^2 distance of the warping function from id ("l2gam"), L^2 distance of the SRVF of the warping function from that of id ("l2psi"), geodesic distance from id ("geodesic"), and no penalty ("none"). "norm" is kept for backward compatibility as an alias for "l2gam". The penalty is weighted by lambda, so it has no effect when lambda = 0.
parallel	run computation in parallel (default = TRUE)

Value

Returns a list containing

amp	amplitude depth
phase	phase depth
amp_dist	amplitude distance matrix
phs_dist	phase distance matrix

References

T. Harris, J. D. Tucker, B. Li, and L. Shand, "Elastic depths for detecting shape anomalies in functional data," Technometrics, 10.1080/00401706.2020.1811156, 2020.

Examples

```
depths <- elastic.depth(simu_data$f[, 1:4], simu_data$time)
```

---

elastic.distance *Calculates two elastic distance*

Description

This functions calculates the distances between functions in R^1 D_y and D_x, where function 2 is aligned to function 1

Usage

```
elastic.distance(f1, f2, time, lambda = 0, pen = "roughness")
```

Arguments

f1	sample function 1, provided as a vector of length M
f2	sample function 2, provided as a vector of length M
time	sample points of functions, provided as a vector of length M
lambda	controls amount of warping (default = 0)
pen	alignment penalty (default = "roughness") options are second derivative ("roughness"), L^2 distance of the warping function from id ("l2gam"), L^2 distance of the SRVF of the warping function from that of id ("l2psi"), geodesic distance from id ("geodesic"), and no penalty ("none"). "norm" is kept for backward compatibility as an alias for "l2gam". The penalty is weighted by lambda, so it has no effect when lambda = 0.

Value

Returns a list containing

Dy	amplitude distance
Dx	phase distance

References

Srivastava, A., Wu, W., Kurtek, S., Klassen, E., Marron, J. S., May 2011. Registration of functional data using fisher-rao metric, arXiv:1103.3817v2.

Tucker, J. D., Wu, W., Srivastava, A., Generative Models for Function Data using Phase and Amplitude Separation, Computational Statistics and Data Analysis (2012), 10.1016/j.csda.2012.12.001.

Examples

```
distances <- elastic.distance(
  f1 = simu_data$f[, 1],
  f2 = simu_data$f[, 2],
  time = simu_data$time
)
```

---

elastic.logistic *Elastic Logistic Regression*

Description

This function identifies a logistic regression model with phase-variability using elastic methods

Usage

```
elastic.logistic(
  f,
  y,
  time,
  B = NULL,
  df = 20,
  max_itr = 20,
  smooth_data = FALSE,
  sparam = 25,
  parallel = FALSE,
  cores = 2
)
```

Arguments

f	matrix (N x M) of M functions with N samples
y	vector of size M labels (1/-1)
time	vector of size N describing the sample points
B	matrix defining basis functions (default = NULL)
df	scalar controlling degrees of freedom if B=NULL (default=20)
max_itr	scalar number of iterations (default=20)
smooth_data	smooth data using box filter (default = F)
sparam	number of times to apply box filter (default = 25)
parallel	enable parallel mode using `foreach` and `doParallel` package
cores	set number of cores to use with `doParallel` (default = 2)

Value

Returns a list containing

alpha	model intercept
beta	regressor function
fn	aligned functions - matrix (N x M) of M functions with N samples
qn	aligned srvfs - similar structure to fn
gamma	warping functions - similar structure to fn
q	original srvf - similar structure to fn
B	basis matrix
b	basis coefficients
Loss	logistic loss
type	model type ('logistic')

References

Tucker, J. D., Wu, W., Srivastava, A., Elastic Functional Logistic Regression with Application to Physiological Signal Classification, Electronic Journal of Statistics (2014), submitted.

`elastic.lpcr.regression` *Elastic logistic Principal Component Regression*

Description

This function identifies a logistic regression model with phase-variability using elastic pca

Usage

```
elastic.lpcr.regression(
  f,
  y,
  time,
  pca.method = "combined",
  no = 5,
  smooth_data = FALSE,
  sparam = 25
)
```

Arguments

`f`	matrix (N x M) of M functions with N samples
`y`	vector of size M labels
`time`	vector of size N describing the sample points
`pca.method`	string specifying pca method (options = "combined", "vert", or "horiz", default = "combined")
`no`	scalar specify number of principal components (default=5)
`smooth_data`	smooth data using box filter (default = F)
`sparam`	number of times to apply box filter (default = 25)

Value

Returns a lpcr object containing

`alpha`	model intercept
`b`	regressor vector
`y`	label vector
`warp_data`	fdawarp object of aligned data
`pca`	pca object of principal components
`Loss`	logistic loss
`pca.method`	string specifying pca method used

References

J. D. Tucker, J. R. Lewis, and A. Srivastava, "Elastic Functional Principal Component Regression," Statistical Analysis and Data Mining, 10.1002/sam.11399, 2018.

`elastic.mlogistic` *Elastic Multinomial Logistic Regression*

Description

This function identifies a multinomial logistic regression model with phase-variability using elastic methods

Usage

```
elastic.mlogistic(
  f,
  y,
  time,
  B = NULL,
  df = 20,
  max_itr = 20,
  smooth_data = FALSE,
  sparam = 25,
  parallel = FALSE,
  cores = 2
)
```

Arguments

`f`	matrix (N x M) of M functions with N samples
`y`	vector of size M labels (1,2,...,m) for m classes
`time`	vector of size N describing the sample points
`B`	matrix defining basis functions (default = NULL)
`df`	scalar controlling degrees of freedom if B=NULL (default=20)
`max_itr`	scalar number of iterations (default=20)
`smooth_data`	smooth data using box filter (default = F)
`sparam`	number of times to apply box filter (default = 25)
`parallel`	enable parallel mode using `foreach` and `doParallel` package
`cores`	set number of cores to use with `doParallel` (default = 2)

Value

Returns a list containing

alpha	model intercept
beta	regressor function
fn	aligned functions - matrix (N x M) of M functions with N samples
qn	aligned srvfs - similar structure to fn
gamma	warping functions - similar structure to fn
q	original srvf - similar structure to fn
B	basis matrix
b	basis coefficients
Loss	logistic loss
type	model type ('mlogistic')

References

Tucker, J. D., Wu, W., Srivastava, A., Elastic Functional Logistic Regression with Application to Physiological Signal Classification, Electronic Journal of Statistics (2014), submitted.

---

elastic.mlpcr.regression

Elastic Multinomial logistic Principal Component Regression

Description

This function identifies a multinomial logistic regression model with phase-variability using elastic pca

Usage

```
elastic.mlpcr.regression(
  f,
  y,
  time,
  pca.method = "combined",
  no = 5,
  smooth_data = FALSE,
  sparam = 25
)
```

Arguments

f	matrix (N x M) of M functions with N samples
y	vector of size M labels
time	vector of size N describing the sample points
pca.method	string specifying pca method (options = "combined", "vert", or "horiz", default = "combined")
no	scalar specify number of principal components (default=5)
smooth_data	smooth data using box filter (default = F)
sparam	number of times to apply box filter (default = 25)

Value

Returns a mlpcr object containing

alpha	model intercept
b	regressor vector
y	label vector
Y	Coded labels
warp_data	fdawarp object of aligned data
pca	pca object of principal components
Loss	logistic loss
pca.method	string specifying pca method used

References

J. D. Tucker, J. R. Lewis, and A. Srivastava, "Elastic Functional Principal Component Regression," Statistical Analysis and Data Mining, 10.1002/sam.11399, 2018.

---

elastic.pcr.regression

Elastic Linear Principal Component Regression

Description

This function identifies a regression model with phase-variability using elastic pca

Usage

```
elastic.pcr.regression(
  f,
  y,
  time,
  pca.method = "combined",
  no = 5,
  smooth_data = FALSE,
  sparam = 25,
  parallel = F,
  C = NULL
)
```

Arguments

f	matrix (N x M) of M functions with N samples
y	vector of size M responses
time	vector of size N describing the sample points
pca.method	string specifying pca method (options = "combined", "vert", or "horiz", default = "combined")
no	scalar specify number of principal components (default = 5)
smooth_data	smooth data using box filter (default = F)
sparam	number of times to apply box filter (default = 25)
parallel	run in parallel (default = F)
C	scale balance parameter for combined method (default = NULL)

Value

Returns a pcr object containing

alpha	model intercept
b	regressor vector
y	response vector
warp_data	fdawarp object of aligned data
pca	pca object of principal components
SSE	sum of squared errors
pca.method	string specifying pca method used

References

J. D. Tucker, J. R. Lewis, and A. Srivastava, “Elastic Functional Principal Component Regression,” Statistical Analysis and Data Mining, 10.1002/sam.11399, 2018.

`elastic.prediction` *Elastic Prediction from Regression Models*

Description

This function performs prediction from an elastic regression model with phase-variability

Usage

```
elastic.prediction(f, time, model, y = NULL, smooth_data = FALSE, sparam = 25)
```

Arguments

`f`	matrix (N x M) of M functions with N samples
`time`	vector of size N describing the sample points
`model`	list describing model from elastic regression methods
`y`	responses of test matrix f (default=NULL)
`smooth_data`	smooth data using box filter (default = F)
`sparam`	number of times to apply box filter (default = 25)

Value

Returns a list containing

`y_pred`	predicted values of f or probabilities depending on model
`SSE`	sum of squared errors if linear
`y_labels`	labels if logistic model
`PC`	probability of classification if logistic

References

Tucker, J. D., Wu, W., Srivastava, A., Elastic Functional Logistic Regression with Application to Physiological Signal Classification, Electronic Journal of Statistics (2014), submitted.

`elastic.regression` *Elastic Linear Regression*

Description

This function identifies a regression model with phase-variability using elastic methods

Usage

```
elastic.regression(
  f,
  y,
  time,
  B = NULL,
  lam = 0,
  df = 20,
  max_itr = 20,
  smooth_data = FALSE,
  sparam = 25,
  parallel = FALSE,
  cores = 2
)
```

Arguments

`f`	matrix (N x M) of M functions with N samples
`y`	vector of size M responses
`time`	vector of size N describing the sample points
`B`	matrix defining basis functions (default = NULL)
`lam`	scalar regularization parameter (default=0)
`df`	scalar controlling degrees of freedom if B=NULL (default=20)
`max_itr`	scalar number of iterations (default=20)
`smooth_data`	smooth data using box filter (default = F)
`sparam`	number of times to apply box filter (default = 25)
`parallel`	enable parallel mode using `foreach` and `doParallel` package
`cores`	set number of cores to use with `doParallel` (default = 2)

Value

Returns a list containing

`alpha`	model intercept
`beta`	regressor function
`fn`	aligned functions - matrix (N x M) of M functions with N samples

qn	aligned srvfs - similar structure to fn
gamma	warping functions - similar structure to fn
q	original srvf - similar structure to fn
B	basis matrix
b	basis coefficients
SSE	sum of squared errors
type	model type ('linear')

References

Tucker, J. D., Wu, W., Srivastava, A., Elastic Functional Logistic Regression with Application to Physiological Signal Classification, Electronic Journal of Statistics (2014), submitted.

`elastic_amp_change_ff` *Elastic Amplitude Changepoint Detection*

Description

This function identifies a amplitude changepoint using a fully functional approach

Usage

```
elastic_amp_change_ff(
  f,
  time,
  d = 1000,
  h = 0,
  smooth_data = FALSE,
  sparam = 25,
  showplot = TRUE
)
```

Arguments

f	matrix (N x M) of M functions with N samples
time	vector of size N describing the sample points
d	number of monte carlo iterations of Brownian Bridge (default = 1000)
h	window selection of long range covariance function (default = 0)
smooth_data	smooth data using box filter (default = F)
sparam	number of times to apply box filter (default = 25)
showplot	show results plots (default = T)

Value

Returns a list object containing

pvalue	p value
change	indice of changepoint
DataBefore	functions before changepoint
DataAfter	functions after changepoint
MeanBefore	mean function before changepoint
MeanAfter	mean function after changepoint
WarpingBefore	warping functions before changepoint
WarpingAfter	warping functions after changepoint
WarpingMeanBefore	mean warping function before changepoint
WarpingMeanAfter	mean warping function after changepoint
change_fun	amplitude change function
Sn	test statistic values
mu	mean srsfs
mu_f	mean functions

References

J. D. Tucker and D. Yarger, "Elastic Functional Changepoint Detection of Climate Impacts from Localized Sources", Envirometrics, 10.1002/env.2826, 2023.

---

elastic_change_fpca *Elastic Changepoint Detection*

Description

This function identifies changepoints using a functional PCA

Usage

```
elastic_change_fpca(
  f,
  time,
  pca.method = "combined",
  pc = 0.95,
  d = 1000,
  n_pcs = 5,
  smooth_data = FALSE,
  sparam = 25,
  showplot = TRUE
)
```

Arguments

f	matrix (N x M) of M functions with N samples
time	vector of size N describing the sample points
pca.method	string specifying pca method (options = "combined", "vert", or "horiz", default = "combined")
pc	percentage of cummulation explained variance (default = 0.95)
d	number of monte carlo iterations of Brownian Bridge (default = 1000)
n_pcs	scalar specify number of principal components (default = 5)
smooth_data	smooth data using box filter (default = F)
sparam	number of times to apply box filter (default = 25)
showplot	show results plots (default = T)

Value

Returns a list object containing

pvalue	p value
change	indice of changepoint
DataBefore	functions before changepoint
DataAfter	functions after changepoint
MeanBefore	mean function before changepoint
MeanAfter	mean function after changepoint
WarpingBefore	warping functions before changepoint
WarpingAfter	warping functions after changepoint
WarpingMeanBefore	mean warping function before changepoint
WarpingMeanAfter	mean warping function after changepoint
change_fun	amplitude change function
Sn	test statistic values

References

J. D. Tucker and D. Yarger, “Elastic Functional Changepoint Detection of Climate Impacts from Localized Sources”, Envirometrics, 10.1002/env.2826, 2023.

`elastic_ph_change_ff` *Elastic Phase Changepoint Detection*

Description

This function identifies a phase changepoint using a fully functional approach

Usage

```
elastic_ph_change_ff(
  f,
  time,
  d = 1000,
  h = 0,
  smooth_data = FALSE,
  sparam = 25,
  showplot = TRUE
)
```

Arguments

`f`	matrix (N x M) of M functions with N samples
`time`	vector of size N describing the sample points
`d`	number of monte carlo iterations of Brownian Bridge (default = 1000)
`h`	window selection of long range covariance function (default = 0)
`smooth_data`	smooth data using box filter (default = F)
`sparam`	number of times to apply box filter (default = 25)
`showplot`	show results plots (default = T)

Value

Returns a list object containing

`pvalue`	p value
`change`	indice of changepoint
`DataBefore`	functions before changepoint
`DataAfter`	functions after changepoint
`MeanBefore`	mean function before changepoint
`MeanAfter`	mean function after changepoint
`WarpingBefore`	warping functions before changepoint
`WarpingAfter`	warping functions after changepoint
`WarpingMeanBefore`	mean warping function before changepoint

WarpingMeanAfter	mean warping function after changepoint
change_fun	amplitude change function
Sn	test statistic values
mu	mean shooting vectors

References

J. D. Tucker and D. Yarger, "Elastic Functional Changepoint Detection of Climate Impacts from Localized Sources", Envirometrics, 10.1002/env.2826, 2023.

---

f_plot *Plot functional data*

Description

This function plots functional data on a time grid

Usage

```
f_plot(t, f)
```

Arguments

t	A numeric vector of length M specifying the common grid on which all curves f have been observed.
f	A numeric matrix of shape $M \times N$ specifying a sample of N curves observed on a grid of size M.

---

f_to_srvf *Transformation to SRVF Space*

Description

This function transforms functions in R^1 from their original functional space to the SRVF space.

Usage

```
f_to_srvf(f, time, smooth = FALSE)
```

Arguments

f	Either a numeric vector of a numeric matrix or a numeric array specifying the functions that need to be transformed. • If a vector, it must be of shape M and it is interpreted as a single 1-dimensional curve observed on a grid of size M. • If a matrix, it must be of shape $M \times N$. In this case, it is interpreted as a sample of N curves observed on a grid of size M, unless $M = 1$ in which case it is interpreted as a single 1-dimensional curve observed on a grid of size M.
time	A numeric vector of length M specifying the grid on which the functions are evaluated.
smooth	A boolean specifying whether to differentiate a smoothing spline of f instead of the interpolating cubic spline. Defaults to FALSE. With FALSE, srvf_to_f() inverts this transformation with $O(h^4)$ accuracy for smooth f. With TRUE, high-frequency content is removed on purpose and is not recovered by srvf_to_f().

Value

A numeric array of the same shape as the input array f storing the SRVFs of the original curves.

References

Srivastava, A., Wu, W., Kurtek, S., Klassen, E., Marron, J. S., May 2011. Registration of functional data using Fisher-Rao metric, arXiv:1103.3817v2.

Tucker, J. D., Wu, W., Srivastava, A., Generative models for functional data using phase and amplitude Separation, Computational Statistics and Data Analysis (2012), 10.1016/j.csda.2012.12.001.

Examples

```
q <- f_to_srvf(simu_data$f, simu_data$time)
```

---

fastpns *Computes fast PNS decomposition*

Description

Computes fast PNS decomposition

Usage

```
fastpns(
  x,
  n.pc = "Full",
  sphere.type = "seq.test",
  mean.type = "Frechet",
  alpha = 0.1,
```

```
  R = 100,
  nlast.small.sphere = 1,
  output = TRUE
)
```

Arguments

x	data on sphere
n.pc	number of pc's (default 'Full')
sphere.type	type of sphere decompistion (default 'seq.test')
mean.type	type of mean (default 'Frechet')
alpha	significant level for test (default .1)
R	the number of bootstrap samples to be evaluated for the sequential test. The default is 100.
nlast.small.sphere	(default 1)
output	show output (default TRUE)

Value

EuclidData euclidean representation

---

fastPNSe2s *Computes fast PNS euclidean to sphere*

Description

Computes fast PNS euclidean to sphere

Usage

```
fastPNSe2s(res, pns)
```

Arguments

res	PNS embedding
pns	list of PNS structure

Value

approx sphere representation

fastPNSs2e *Computes fast PNS sphere to euclidean*

Description

Computes fast PNS sphere to euclidean

Usage

```
fastPNSs2e(spheredata, PNS)
```

Arguments

spheredata	data on sphere
PNS	list of PNS structure

Value

EuclidData euclidean representation

fdasrvf *Elastic Functional Data Analysis*

Description

A library for functional data analysis using the square root velocity framework which performs pair-wise and group-wise alignment as well as modeling using functional component analysis.

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References

Srivastava, A., Wu, W., Kurtek, S., Klassen, E., Marron, J. S., May 2011. Registration of functional data using Fisher-Rao metric, arXiv:1103.3817v2.

Tucker, J. D., Wu, W., Srivastava, A., Generative models for functional data using phase and amplitude separation, Computational Statistics and Data Analysis (2012), 10.1016/j.csda.2012.12.001.

J. D. Tucker, W. Wu, and A. Srivastava, Phase-amplitude separation of proteomics data using extended Fisher-Rao metric, Electronic Journal of Statistics, Vol 8, no. 2. pp 1724-1733, 2014.

J. D. Tucker, W. Wu, and A. Srivastava, "Analysis of signals under compositional noise with applications to SONAR data," IEEE Journal of Oceanic Engineering, Vol 29, no. 2. pp 318-330, Apr 2014.

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Kurtek, S., Srivastava, A., Klassen, E., and Ding, Z. (2012), "Statistical Modeling of Curves Using Shapes and Related Features," Journal of the American Statistical Association, 107, 1152–1165.

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W. Xie, S. Kurtek, K. Bharath, and Y. Sun, A geometric approach to visualization of variability in functional data, Journal of American Statistical Association 112 (2017), pp. 979-993.

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Lee, S. and S. Jung, 2017: Combined analysis of amplitude and phase variations in functional data. arXiv:1603.01775, 1–21.

J. D. Tucker, J. R. Lewis, and A. Srivastava, "Elastic Functional Principal Component Regression," Statistical Analysis and Data Mining, vol. 12, no. 2, pp. 101-115, 2019.

J. D. Tucker, J. R. Lewis, C. King, and S. Kurtek, "A Geometric Approach for Computing Tolerance Bounds for Elastic Functional Data," Journal of Applied Statistics, 10.1080/02664763.2019.1645818, 2019.

T. Harris, J. D. Tucker, B. Li, and L. Shand, "Elastic depths for detecting shape anomalies in functional data," Technometrics, 10.1080/00401706.2020.1811156, 2020.

J. D. Tucker and D. Yarger, "Elastic Functional Changepoint Detection of Climate Impacts from Localized Sources", Envirometrics, 10.1002/env.2826, 2023.

See Also

Useful links:

- https://github.com/jdtuck/fdasrvf_R
- Report bugs at https://github.com/jdtuck/fdasrvf_R/issues

`function_group_warp_bayes` *Bayesian Group Warping*

Description

This function aligns a set of functions using Bayesian SRSF framework

Usage

```
function_group_warp_bayes(
  f,
  time,
  iter = 50000,
  powera = 1,
  times = 5,
  tau = ceiling(times * 0.04),
  gp = seq(dim(f)[2]),
  showplot = TRUE
)
```

Arguments

`f`	matrix (N x M) of M functions with N samples
`time`	sample points of functions
`iter`	number of iterations (default = 150000)
`powera`	Dirichlet prior parameter (default 1)
`times`	factor of length of subsample points to look at (default = 5)
`tau`	standard deviation of Normal prior for increment (default ceil(times*.4))
`gp`	number of colors in plots (defaults `seq(dim(f)[2])`)
`showplot`	shows plots of functions (default = T)

Value

Returns a list containing

`f0`	original functions
`f_q`	f aligned quotient space

gam_q	warping functions quotient space
f_a	f aligned ambient space
gam_a	warping ambient space
qmn	mean srsf

References

Cheng, W., Dryden, I. L., and Huang, X. (2016). Bayesian registration of functions and curves. Bayesian Analysis, 11(2), 447-475.

Examples

```
## Not run:
  out <- function_group_warp_bayes(simu_data$f, simu_data$time)

## End(Not run)
```

---

function_mean_bayes *Bayesian Karcher Mean Calculation*

Description

This function calculates karcher mean of functions using Bayesian method

Usage

```
function_mean_bayes(f, time, times = 5, group = 1:dim(f)[2], showplot = TRUE)
```

Arguments

f	matrix (N x M) of M functions with N samples
time	sample points of functions
times	factor of length of subsample points to look at (default = 5)
group	(defaults 1:dim(f)[2])
showplot	shows plots of functions (default = T)

Value

Returns a list containing

distfamily	dist matrix
match.matrix	matrix of warping functions
position	position
mu_5	function mean
rtmatrix	rtmatrix

sumdist	sumdist
qt.fitted	aligned srsf functions
estimator	estimator
estimator2	estimator2
regfuncs	registered functions

References

Cheng, W., Dryden, I. L., and Huang, X. (2016). Bayesian registration of functions and curves. Bayesian Analysis, 11(2), 447-475.

Examples

```
## Not run:
  out <- function_mean_bayes(simu_data$f, simu_data$time)

## End(Not run)
```

---

gam_to_h *map warping function to tangent space at identity*

Description

map warping function to tangent space at identity

Usage

```
gam_to_h(gam, smooth = FALSE)
```

Arguments

gam	Either a numeric vector of a numeric matrix or a numeric array specifying the warping functions
smooth	Apply smoothing before gradient

Value

A numeric array of the same shape as the input array gamma storing the shooting vectors of gamma obtained via finite differences.

gam_to_psi *map warping function to hilbert sphere*

Description

map warping function to hilbert sphere

Usage

```
gam_to_psi(gam, smooth = FALSE)
```

Arguments

gam	Either a numeric vector of a numeric matrix or a numeric array specifying the warping functions
smooth	Apply smoothing before gradient

Value

A numeric array of the same shape as the input array gamma storing the shooting vectors of gamma obtained via finite differences.

gam_to_v *map warping function to tangent space at identity*

Description

map warping function to tangent space at identity

Usage

```
gam_to_v(gam, smooth = FALSE)
```

Arguments

gam	Either a numeric vector of a numeric matrix or a numeric array specifying the warping functions
smooth	Apply smoothing before gradient

Value

A numeric array of the same shape as the input array gamma storing the shooting vectors of gamma obtained via finite differences.

`gauss_model` *Gaussian model of functional data*

Description

This function models the functional data using a Gaussian model extracted from the principal components of the srvfs

Usage

```
gauss_model(warp_data, n = 1, sort_samples = FALSE)
```

Arguments

`warp_data`	`fdawarp` object from time_warping of aligned data
`n`	number of random samples (n = 1)
`sort_samples`	sort samples (default = F)

Value

Returns a `fdawarp` object containing

`fs`	random aligned samples
`gams`	random warping function samples
`ft`	random function samples

References

Tucker, J. D., Wu, W., Srivastava, A., Generative Models for Function Data using Phase and Amplitude Separation, Computational Statistics and Data Analysis (2012), 10.1016/j.csda.2012.12.001.

Examples

```
out1 <- gauss_model(simu_warp, n = 10)
```

get_curve_centroid *Computes the centroid of a curve*

Description

Computes the centroid of a curve

Usage

```
get_curve_centroid(betafun)
```

Arguments

betafun	A function that takes a numeric vector s of values in $[0, 1]$ as input and returns its values at s.

Value

A numeric vector of length L storing the centroid of the curve.

Examples

```
betafun <- discrete2curve(beta[, , 1, 1])
get_curve_centroid(betafun)
```

get_distance_matrix *Computes the distance matrix between a set of shapes.*

Description

Computes the distance matrix between a set of shapes.

Usage

```
get_distance_matrix(
  qfuns,
  alignment = FALSE,
  rotation = FALSE,
  scale = FALSE,
  lambda = 0,
  M = 200L,
  parallel_setup = 1L
)
```

Arguments

qfuns	A list of functions representing the shapes. Each function should be a SRVF. See curve2srvf() for more details on how to obtain the SRVF of a shape.
alignment	A boolean value specifying whether the two SRVFs should be optimally aligned before computing the distance. Defaults to FALSE.
rotation	A boolean value specifying whether the two SRVFs should be optimally rotated before computing the distance. Defaults to FALSE.
scale	A boolean value specifying whether the two SRVFs should be projected onto the Hilbert sphere before computing the distance. Defaults to FALSE.
lambda	A numeric value specifying the regularization parameter for the optimal alignment. Defaults to 0.
M	An integer value specifying the number of points to use when searching for the optimal rotation and alignment. Defaults to 200L.
parallel_setup	An integer value specifying the number of cores to use for parallel computing or an object of class cluster. In the latter case, it is used as is for parallel computing. If parallel_setup is 1L, then no parallel computing will be used. The maximum number of cores to use is the number of available cores minus 1. Defaults to 1L.

Value

An object of class dist containing the distance matrix between the shapes.

Examples

```
N <- 4L
srvfs <- lapply(1:N, \(n) curve2srvf(beta[, , 1, n]))
get_distance_matrix(srvfs, parallel_setup = 1L)
```

---

get_hilbert_sphere_distance

Computes the geodesic distance between two SRVFs on the Hilbert sphere

---

Description

Computes the geodesic distance between two SRVFs on the Hilbert sphere

Usage

```
get_hilbert_sphere_distance(q1fun, q2fun)
```

Arguments

q1fun	A function that takes a numeric vector s of values in $[0, 1]$ as input and returns the values of the first SRVF at s.
q2fun	A function that takes a numeric vector s of values in $[0, 1]$ as input and returns the values of the second SRVF at s.

Value

A numeric value storing the geodesic distance between the two SRVFs.

Examples

```
q1 <- curve2srvf(beta[, , 1, 1])
q2 <- curve2srvf(beta[, , 1, 2])
q1p <- to_hilbert_sphere(q1)
q2p <- to_hilbert_sphere(q2)
get_hilbert_sphere_distance(q1p, q2p)
```

---

get_identity_warping *Computes the identity warping function*

Description

Computes the identity warping function

Usage

```
get_identity_warping()
```

Value

A function that takes a numeric vector s of values in $[0, 1]$ as input and returns the values of the identity warping function at s.

Examples

```
get_identity_warping()
```

get_l2_distance *Computes the L^2 distance between two SRVFs*

Description

Computes the L^2 distance between two SRVFs

Usage

```
get_l2_distance(q1fun, q2fun, method = "quadrature")
```

Arguments

q1fun	A function that takes a numeric vector s of values in $[0, 1]$ as input and returns the values of the first SRVF at s.
q2fun	A function that takes a numeric vector s of values in $[0, 1]$ as input and returns the values of the second SRVF at s.
method	A character string specifying the method to use for computing the L^2 distance. Options are "quadrature" and "trapz". Defaults to "quadrature".

Value

A numeric value storing the L^2 distance between the two SRVFs.

Examples

```
q1 <- curve2srvf(beta[, , 1, 1])
q2 <- curve2srvf(beta[, , 1, 2])
get_l2_distance(q1, q2)
```

get_l2_inner_product *Computes the L^2 inner product between two SRVFs*

Description

Computes the L^2 inner product between two SRVFs

Usage

```
get_l2_inner_product(q1fun, q2fun)
```

Arguments

q1fun	A function that takes a numeric vector s of values in $[0, 1]$ as input and returns the values of the first SRVF at s.
q2fun	A function that takes a numeric vector s of values in $[0, 1]$ as input and returns the values of the second SRVF at s.

Value

A numeric value storing the L^2 inner product between the two SRVFs.

Examples

```
q1 <- curve2srvf(beta[, , 1, 1])
q2 <- curve2srvf(beta[, , 1, 2])
get_l2_inner_product(q1, q2)
```

`get_l2_norm` *Computes the L^2 norm of an SRVF*

Description

Computes the L^2 norm of an SRVF

Usage

```
get_l2_norm(qfun)
```

Arguments

`qfun`	A function that takes a numeric vector s of values in $[0, 1]$ as input and returns the values of the SRVF at s.

Value

A numeric value storing the L^2 norm of the SRVF.

Examples

```
q <- curve2srvf(beta[, , 1, 1])
get_l2_norm(q)
```

`get_shape_distance` *Computes the distance between two shapes*

Description

Computes the distance between two shapes

Usage

```
get_shape_distance(
  q1fun,
  q2fun,
  alignment = FALSE,
  rotation = FALSE,
  scale = FALSE,
  lambda = 0,
  M = 200L
)
```

Arguments

q1fun	A function that takes a numeric vector s of values in $[0, 1]$ as input and returns the values of the first SRVF at s.
q2fun	A function that takes a numeric vector s of values in $[0, 1]$ as input and returns the values of the second SRVF at s.
alignment	A boolean value specifying whether the two SRVFs should be optimally aligned before computing the distance. Defaults to FALSE.
rotation	A boolean value specifying whether the two SRVFs should be optimally rotated before computing the distance. Defaults to FALSE.
scale	A boolean value specifying whether the two SRVFs should be projected onto the Hilbert sphere before computing the distance. Defaults to FALSE.
lambda	A numeric value specifying the regularization parameter for the optimal alignment. Defaults to 0.
M	An integer value specifying the number of points to use when searching for the optimal rotation and alignment. Defaults to 200L.

Value

A list with the following components:

- amplitude_distance: A numeric value storing the amplitude distance between the two SRVFs.
- phase_distance: A numeric value storing the phase distance between the two SRVFs.
- optimal_warping: A function that takes a numeric vector s of values in $[0, 1]$ as input and returns the optimal warping function evaluated at s.
- q1fun_modified: A function that takes a numeric vector s of values in $[0, 1]$ as input and returns the first SRVF modified according to the optimal alignment, rotation, and scaling.
- q2fun_modified: A function that takes a numeric vector s of values in $[0, 1]$ as input and returns the second SRVF modified according to the optimal alignment, rotation, and scaling.

Examples

```
q1 <- curve2srvf(beta[, , 1, 1])
q2 <- curve2srvf(beta[, , 1, 2])
get_shape_distance(q1, q2)
```

get_warping_distance *Computes the distance between two warping functions*

Description

Computes the distance between two warping functions

Usage

```
get_warping_distance(gam1fun, gam2fun)
```

Arguments

gam1fun	A function that takes a numeric vector s of values in $[0, 1]$ as input and returns the values of the first warping function at s.
gam2fun	A function that takes a numeric vector s of values in $[0, 1]$ as input and returns the values of the second warping function at s.

Value

A numeric value storing the distance between the two warping functions.

Examples

```
gam1 <- discrete2warping(toy_warp$gam[, 1])
gam2 <- discrete2warping(toy_warp$gam[, 2])
get_warping_distance(gam1, gam2)
get_warping_distance(gam1, get_identity_warping())
```

gradient *Gradient using finite differences*

Description

This function computes the gradient of f using finite differences.

Usage

```
gradient(f, binsize, multidimensional = FALSE)
```

Arguments

f	Either a numeric vector of a numeric matrix or a numeric array specifying the curve(s) that need to be differentiated. • If a vector, it must be of shape M and it is interpreted as a single 1-dimensional curve observed on a grid of size M. • If a matrix and multidimensional == FALSE, it must be of shape $M \times N$. In this case, it is interpreted as a sample of N curves observed on a grid of size M, unless $M = 1$ in which case it is interpreted as a single 1-dimensional curve observed on a grid of size M. • If a matrix and multidimensional == TRUE,it must be of shape $L \times M$ and it is interpreted as a single L-dimensional curve observed on a grid of size M. • If a 3D array, it must be of shape $L \times M \times N$ and it is interpreted as a sample of N L-dimensional curves observed on a grid of size M.
binsize	A numeric value specifying the size of the bins for computing finite differences.
multidimensional	A boolean specifying if the curves are multi-dimensional. This is useful when f is provided as a matrix to determine whether it is a single multi-dimensional curve or a collection of uni-dimensional curves. Defaults to FALSE.

Value

A numeric array of the same shape as the input array f storing the gradient of f obtained via finite differences.

Examples

```
out <- gradient(simu_data$f[, 1], mean(diff(simu_data$time)))
```

---

growth_vel *Berkeley Growth Velocity Dataset*

Description

Combination of both boys and girls growth velocity from the Berkley dataset.

Usage

```
growth_vel
```

Format

`growth_vel`:
A list with two components:

- `f`: A numeric matrix of shape 69×93 storing a sample of size $N = 93$ of curves evaluated on a grid of size $M = 69$.
- `time`: A numeric vector of size $M = 69$ storing the grid on which the curves `f` have been evaluated.

---

h_to_gam *map shooting vector to warping function at identity*

Description

map shooting vector to warping function at identity

Usage

```
h_to_gam(h)
```

Arguments

`h`	Either a numeric vector of a numeric matrix or a numeric array specifying the shooting vectors

Value

A numeric array of the same shape as the input array `h` storing the warping functions of `h`.

---

horizFPCA *Horizontal Functional Principal Component Analysis*

Description

This function calculates vertical functional principal component analysis on aligned data

Usage

```
horizFPCA(warp_data, no = 3, var_exp = NULL, ci = c(-1, 0, 1), showplot = TRUE)
```

Arguments

`warp_data`	fdawarp object from time_warping of aligned data
`no`	number of principal components to extract
`var_exp`	compute no based on value percent variance explained (example: 0.95) will override `no`
`ci`	geodesic standard deviations (default = c(-1,0,1))
`showplot`	show plots of principal directions (default = T)

Value

Returns a hfpca object containing

gam_pca	warping functions principal directions
psi_pca	srvf principal directions
latent	latent values
U	eigenvectors
vec	shooting vectors
mu	Karcher Mean

References

Tucker, J. D., Wu, W., Srivastava, A., Generative Models for Function Data using Phase and Amplitude Separation, Computational Statistics and Data Analysis (2012), 10.1016/j.csda.2012.12.001.

Examples

```
hfpca <- horizFPCA(simu_warp, no = 3)
```

---

horizFPNS *Horizontal Functional Principal Component Analysis*

Description

This function calculates vertical functional principal nested sphere analysis on aligned data

Usage

```
horizFPNS(warp_data, var_exp = 0.99, ci = c(-1, 0, 1), showplot = TRUE)
```

Arguments

warp_data	fdawarp object from time_warping of aligned data
var_exp	compute no based on value percent variance explained (example: 0.95)
ci	geodesic standard deviations (default = c(-1,0,1))
showplot	show plots of principal directions (default = T)

Value

Returns a hfpns object containing

gam_pns	warping functions principal directions
psi_pns	srvf principal directions
PNS	PNS object

References

Yu, Q., Lu, X., and Marron, J. S. (2017), "Principal Nested Spheres for Time-Warped Functional Data Analysis," Journal of Computational and Graphical Statistics, 26, 144–151.

Examples

```
hfpns <- horizFPNS(simu_warp)
```

---

im *Example Image Data set*

Description

Contains two simulated images for registration.

Usage

```
im
```

Format

`im`**:**

A list with two components:

- `I1`: A numeric matrix of shape 64×64 storing the 1st image;
- `I2`: A numeric matrix of shape 64×64 storing the 2nd image.

---

interparc *Interpolate Equally Spaced Points Along a Closed Curve*

Description

Fits periodic cubic smoothing splines to the supplied coordinates and returns points that are equally spaced by arc length along the fitted curve. The curve is parameterized by cumulative chord length, smoothed using cyclic cubic splines (`mgcv`), and then reparameterized by arc length to obtain equally spaced samples.

Usage

```
interparc(t, ..., n.samples = 5000, k = 20, gamma = 1)
```

Arguments

t	Either: • A single integer specifying the number of equally spaced output points, or • A numeric vector of normalized arc-length positions between 0 and 1 at which to evaluate the curve.
...	Two or more equal-length coordinate vectors (e.g., x, y, or x, y, z) describing the input curve.
n.samples	Number of points used to approximate the arc-length parameterization of the fitted spline. Larger values improve accuracy but increase computation time.
k	Basis dimension for the cyclic cubic spline. Larger values allow more flexible fits.
gamma	Smoothing penalty multiplier passed to mgcv::gam(). Values greater than 1 produce smoother curves.

Details

The function first parameterizes the input points by cumulative chord length, fits a cyclic cubic spline to each coordinate using Restricted Maximum Likelihood (REML), estimates arc length from the fitted curve, and then evaluates the splines at parameter values corresponding to equally spaced arc-length positions.

Value

A matrix with one column per coordinate dimension (V1, V2, ...), containing interpolated points that are equally spaced by arc length.

Examples

```
x <- cos(seq(0, 2*pi, length.out = 20) * 1.5 - pi/2)
y <- sin(seq(0, 2*pi, length.out = 20))
plot(x, y, type = "b")
lines(interparc(1000, x, y), col = 2)
points(interparc(10, x, y), col = 2)
```

---

inv_exp_map *map square root of warping function to tangent space*

Description

map square root of warping function to tangent space

Usage

```
inv_exp_map(Psi, psi)
```

Arguments

Psi	vector describing psi function at center of tangent space
psi	vector describing psi function to map to tangent space

Value

A numeric array of the same length as the input array psi storing the shooting vector of psi

---

invertGamma *Invert Warping Function*

Description

This function calculates the inverse of gamma

Usage

```
invertGamma(gam)
```

Arguments

gam	vector of N samples

Value

Returns gamI inverted vector

References

Srivastava, A., Wu, W., Kurtek, S., Klassen, E., Marron, J. S., May 2011. Registration of functional data using fisher-rao metric, arXiv:1103.3817v2.

Tucker, J. D., Wu, W., Srivastava, A., Generative Models for Function Data using Phase and Amplitude Separation, Computational Statistics and Data Analysis (2012), 10.1016/j.csda.2012.12.001.

Examples

```
out <- invertGamma(simu_warp$warping_functions[, 1])
```

`joint_gauss_model` *Gaussian model of functional data using joint Model*

Description

This function models the functional data using a Gaussian model extracted from the principal components of the srvfs using the joint model

Usage

```
joint_gauss_model(warp_data, n = 1, no = 5)
```

Arguments

`warp_data`	fdawarp object from time_warping of aligned data
`n`	number of random samples (n = 1)
`no`	number of principal components (n=4)

Value

Returns a fdawarp object containing

`fs`	random aligned samples
`gams`	random warping function samples
`ft`	random function samples
`qs`	random srvf samples

References

Tucker, J. D., Wu, W., Srivastava, A., Generative Models for Function Data using Phase and Amplitude Separation, Computational Statistics and Data Analysis (2012), 10.1016/j.csda.2012.12.001.

Jung, S. L. a. S. (2016). "Combined Analysis of Amplitude and Phase Variations in Functional Data." arXiv:1603.01775.

Examples

```
out1 <- joint_gauss_model(simu_warp, n = 10)
```

`jointFPCA` *Joint Vertical and Horizontal Functional Principal Component Analysis*

Description

This function calculates amplitude and phase joint functional principal component analysis on aligned data

Usage

```
jointFPCA(
  warp_data,
  no = 3,
  var_exp = NULL,
  id = round(length(warp_data$time)/2),
  C = NULL,
  ci = c(-1, 0, 1),
  srvf = TRUE,
  showplot = TRUE
)
```

Arguments

`warp_data`	fdawarp object from time_warping of aligned data
`no`	number of principal components to extract (default = 3)
`var_exp`	compute no based on value percent variance explained (example: 0.95) will override `no`
`id`	integration point for f0 (default = midpoint)
`C`	balance value (default = NULL)
`ci`	geodesic standard deviations (default = c(-1,0,1))
`srvf`	use srvf (default = TRUE)
`showplot`	show plots of principal directions (default = T)

Value

Returns a list containing

`q_pca`	srvf principal directions
`f_pca`	f principal directions
`latent`	latent values
`coef`	coefficients
`U`	eigenvectors
`mu_psi`	mean psi function

mu_g	mean g function
id	point use for f(0)
C	optimized phase amplitude ratio

References

Srivastava, A., Wu, W., Kurtek, S., Klassen, E., Marron, J. S., May 2011. Registration of functional data using fisher-rao metric, arXiv:1103.3817v2.

Jung, S. L. a. S. (2016). "Combined Analysis of Amplitude and Phase Variations in Functional Data." arXiv:1603.01775.

Tucker, J. D., Wu, W., Srivastava, A., Generative Models for Function Data using Phase and Amplitude Separation, Computational Statistics and Data Analysis (2012), 10.1016/j.csda.2012.12.001.

Examples

```
jfpca <- jointFPCA(simu_warp, no = 3)
```

---

jointFPCAh *Joint Vertical and Horizontal Functional Principal Component Analysis*

Description

This function calculates amplitude and phase joint functional principal component analysis on aligned data using the SRVF framework using MFPCA and h representation

Usage

```
jointFPCAh(
  warp_data,
  var_exp = 0.99,
  id = round(length(warp_data$time)/2),
  C = NULL,
  ci = c(-1, 0, 1),
  srvf = TRUE,
  showplot = TRUE
)
```

Arguments

warp_data	fdawarp object from time_warping of aligned data
var_exp	compute no based on value percent variance explained (default: 0.99) will override no
id	integration point for f0 (default = midpoint)
C	balance value (default = NULL)

ci	geodesic standard deviations (default = c(-1,0,1))
srvf	use srvf (default = TRUE)
showplot	show plots of principal directions (default = T)

Value

Returns a list containing

q_pca	srvf principal directions
f_pca	f principal directions
latent	latent values
coef	coefficients
U	eigenvectors
mu_psi	mean psi function
mu_g	mean g function
id	point use for f(0)
C	optimized phase amplitude ratio

References

Srivastava, A., Wu, W., Kurtek, S., Klassen, E., Marron, J. S., May 2011. Registration of functional data using fisher-rao metric, arXiv:1103.3817v2.

Happ, C., Scheipl, F., Gabriel, A. & Greven, S. A general framework for multivariate functional principal component analysis of amplitude and phase variation. Stat 8, (2019).

Ma, Y., Zhou, X. & Wu, W. A stochastic process representation for time warping functions. Comput. Stat. Data Anal. 194, 107941 (2024).

Tucker, J. D., Wu, W., Srivastava, A., Generative Models for Function Data using Phase and Amplitude Separation, Computational Statistics and Data Analysis (2012), 10.1016/j.csda.2012.12.001.

Examples

```
jfpcah <- jointFPCAh(simu_warp)
```

---

kmeans_align *K-Means Clustering and Alignment*

Description

This function clusters functions and aligns using the elastic square-root velocity function (SRVF) framework.

Usage

```
kmeans_align(
  f,
  time,
  K = 1L,
  seeds = NULL,
  centroid_type = c("mean", "medoid"),
  nonempty = 0L,
  lambda = 0,
  showplot = FALSE,
  smooth_data = FALSE,
  sparam = 25L,
  parallel = FALSE,
  alignment = TRUE,
  rotation = FALSE,
  scale = TRUE,
  omethod = c("DP", "RBFGS"),
  max_iter = 50L,
  thresh = 0.01,
  use_verbose = FALSE
)
```

Arguments

f	Either a numeric matrix or a numeric 3D array specifying the functions that need to be jointly clustered and aligned. • If a matrix, it must be of shape $M \times N$. In this case, it is interpreted as a sample of N curves observed on a grid of size M. • If a 3D array, it must be of shape $L \times M \times N$ and it is interpreted as a sample of N L-dimensional curves observed on a grid of size M. If this is multidimensional functional data, it is advised that `rotation==FALSE`
time	A numeric vector of length M specifying the grid on which the curves are evaluated.
K	An integer value specifying the number of clusters. Defaults to `1L`.
seeds	An integer vector of length `K` specifying the indices of the curves in `f` which will be chosen as initial centroids. Defaults to `NULL` in which case such indices are randomly chosen.
centroid_type	A string specifying the type of centroid to compute. Choices are `"mean"` or `"medoid"`. Defaults to `"mean"`.
nonempty	An integer value specifying the minimum number of curves per cluster during the assignment step. Set it to a positive value to avoid the problem of empty clusters. Defaults to `0L`.
lambda	A numeric value specifying the elasticity. Defaults to `0.0`.
showplot	A Boolean specifying whether to show plots. Defaults to `FALSE`.
smooth_data	A Boolean specifying whether to smooth data using a box filter. Defaults to `FALSE`.

sparam	An integer value specifying the number of box filters applied. Defaults to 25L.
parallel	A Boolean specifying whether parallel mode (using foreach::foreach() and the **doParallel** package) should be activated. Defaults to FALSE.
alignment	A Boolean specifying whether to perform alignment. Defaults to TRUE.
rotation	A Boolean specifying whether to perform rotation. Defaults to to FALSE.
scale	A Boolean specifying whether to scale curves to unit length. Defaults to TRUE.
omethod	A string specifying which method should be used to solve the optimization problem that provides estimated warping functions. Choices are "DP" or "RBFGS". Defaults to "DP".
max_iter	An integer value specifying the maximum number of iterations. Defaults to 50L.
thresh	A numeric value specifying a threshold on the cost function below which convergence is assumed. Defaults to 0.01.
use_verbose	A Boolean specifying whether to display information about the calculations in the console. Defaults to FALSE.

Value

An object of class fdakma which is a list containing:

- f0: the original functions;
- q0: the original SRSFs;
- fn: the aligned functions as matrices or a 3D arrays of the same shape than f0 by clusters in a list;
- qn: the aligned SRSFs as matrices or a 3D arrays of the same shape than f0 separated in clusters in a list;
- labels: the cluster memberships as an integer vector;
- templates: the centroids in the original functional space;
- templates.q: the centroids in SRSF space;
- distances_to_center: A numeric vector storing the distances of each observed curve to its center;
- gam: the warping functions as matrices or a 3D arrays of the same shape than f0 by clusters in a list;
- qun: cost function value.

References

Srivastava, A., Wu, W., Kurtek, S., Klassen, E., Marron, J. S., May 2011. Registration of functional data using Fisher-Rao metric, arXiv:1103.3817v2.

Tucker, J. D., Wu, W., Srivastava, A., Generative models for functional data using phase and amplitude separation, Computational Statistics and Data Analysis (2012), 10.1016/j.csda.2012.12.001.

Sangalli, L. M., et al. (2010). "k-mean alignment for curve clustering." Computational Statistics & Data Analysis 54(5): 1219-1233.

Examples

```
## Not run:
  out <- kmeans_align(growth_vel$f, growth_vel$time, K = 2)

## End(Not run)
```

`LongRunCovMatrix` *Long Run Covariance Matrix Estimation for Multivariate Time Series*

Description

This function estimates the long run covariance matrix of a given multivariate data sample.

Usage

```
LongRunCovMatrix(mdobj, h = 0, kern_type = "bartlett")
```

Arguments

`mdobj`	A multivariate data object
`h`	The bandwidth parameter. It is strictly non-zero. Choosing the bandwidth parameter to be zero is identical to estimating covariance matrix assuming iid data.
`kern_type`	Kernel function to be used for the estimation of the long run covariance matrix. The choices are `c("BT", "PR", "SP", "FT")` which are respectively, bartlett, parzen, simple and flat-top kernels. By default the function uses a `"barlett"` kernel.

Value

Returns long run covariance matrix

`multiple_align_functions`
Group-wise function alignment to specified mean

Description

This function aligns a collection of functions using the elastic square-root slope (srsf) framework.

Usage

```
multiple_align_functions(
  f,
  time,
  mu,
  lambda = 0,
  pen = "roughness",
  showplot = TRUE,
  smooth_data = FALSE,
  sparam = 25,
  parallel = FALSE,
  cores = -1,
  omethod = "DP",
  MaxItr = 20,
  iter = 2000,
  verbose = TRUE
)
```

Arguments

f	matrix (N x M) of M functions with N samples
time	vector of size N describing the sample points
mu	vector of size N that f is aligned to
lambda	controls the elasticity (default = 0)
pen	alignment penalty (default="roughness") options are second derivative ("roughness"), L^2 distance of the warping function from id ("l2gam"), L^2 distance of the SRVF of the warping function from that of id ("l2psi"), geodesic distance from id ("geodesic"), and no penalty ("none"). "norm" is kept for backward compatibility as an alias for "l2gam". The penalty is weighted by lambda, so it has no effect when lambda = 0.
showplot	shows plots of functions (default = T)
smooth_data	smooth data using box filter (default = F)
sparam	number of times to apply box filter (default = 25)
parallel	enable parallel mode using foreach and doParallel package (default=F)
cores	number of cores in parallel (default=-1, means all cores)
omethod	optimization method (DP,DPo,SIMUL,RBFGS,dBayes,expBayes)
MaxItr	maximum number of iterations
iter	bayesian number of mcmc samples (default 2000)
verbose	verbose printing (default TRUE)

Value

Returns a fdawarp object containing

f0	original functions

fn	aligned functions - matrix (N x M) of M functions with N samples
qn	aligned SRSFs - similar structure to fn
q0	original SRSF - similar structure to fn
fmean	function mean or median - vector of length N
mqn	SRSF mean or median - vector of length N
gam	warping functions - similar structure to fn
orig.var	Original Variance of Functions
amp.var	Amplitude Variance
phase.var	Phase Variance
qun	Cost Function Value

References

Srivastava, A., Wu, W., Kurtek, S., Klassen, E., Marron, J. S., May 2011. Registration of functional data using fisher-rao metric, arXiv:1103.3817v2.

Tucker, J. D., Wu, W., Srivastava, A., Generative Models for Function Data using Phase and Amplitude Separation, Computational Statistics and Data Analysis (2012), 10.1016/j.csda.2012.12.001.

`multiple_align_multivariate`
Group-wise multivariate function alignment to specified mean

Description

This function aligns a collection of functions using the elastic square-root velocity (srvf) framework.

Usage

```
multiple_align_multivariate(
  beta,
  mu,
  mode = "O",
  alignment = TRUE,
  rotation = FALSE,
  scale = FALSE,
  lambda = 0,
  ncores = 1L,
  verbose = TRUE
)
```

Arguments

beta	A numeric array of shape $L \times M \times N$ specifying an N-sample of L-dimensional functional data evaluated on a same grid of size M.
mu	array of size $L \times M$ that f is aligned to
mode	A character string specifying whether the input curves should be considered open (mode == "O") or closed (mode == "C"). Defaults to "O".
alignment	A boolean value specifying whether the curves should be aligned before computing the distance matrix. Defaults to TRUE.
rotation	A boolean specifying whether the distance should be made invariant by rotation. Defaults to TRUE.
scale	A boolean specifying whether the distance should be made invariant by scaling. This is effectively achieved by making SRVFs having unit length and switching to an appropriate metric on the sphere between normalized SRVFs. Defaults to TRUE.
lambda	A numeric value specifying the weight of a penalty term that constraints the warping function to be not too far from the identity. Defaults to 0.0.
ncores	An integer value specifying the number of cores to use for parallel computation. Defaults to 1L. The maximum number of available cores is determined by the **parallel** package. One core is always left out to avoid overloading the system.
verbose	verbose printing (default TRUE)

Value

Returns a fdacurve object containing:

- beta: A numeric array of shape $L \times M \times N$ storing the original input data.
- q: A numeric array of shape $L \times M \times N$ storing the SRVFs of the input data.
- betan: A numeric array of shape $L \times M \times N$ storing the aligned, possibly optimally rotated and optimally scaled, input data.
- qn: A numeric array of shape $L \times M \times N$ storing the SRVFs of the aligned, possibly optimally rotated and optimally scaled, input data.
- gamma: A numeric array of shape $L \times M \times N$ storing the warping functions of the aligned, possibly optimally rotated and optimally scaled, input data.
- R: A numeric array of shape $L \times L \times N$ storing the rotation matrices of the aligned, possibly optimally rotated and optimally scaled, input data.
- betamean: A numeric array of shape $L \times M$ storing the Karcher mean or median of the input data.
- qmean: A numeric array of shape $L \times M$ storing the Karcher mean or median of the SRVFs of the input data.
- type: A character string indicating whether the Karcher mean or median has been returned.
- E: A numeric vector storing the energy of the Karcher mean or median at each iteration.
- qun: A numeric vector storing the cost function of the Karcher mean or median at each iteration.

References

Srivastava, A., Klassen, E., Joshi, S., Jermyn, I., (2011). Shape analysis of elastic curves in euclidean spaces. IEEE Transactions on Pattern Analysis and Machine Intelligence, **33** (7), 1415-1428.

`multivariate_karcher_cov`

Curve Karcher Covariance

Description

Calculate Karcher Covariance of a set of curves

Usage

```
multivariate_karcher_cov(align_data)
```

Arguments

`align_data`	fdacurve object from multivariate_karcher_mean of aligned data

Value

K covariance matrix

References

Srivastava, A., Klassen, E., Joshi, S., Jermyn, I., (2011). Shape analysis of elastic curves in euclidean spaces. Pattern Analysis and Machine Intelligence, IEEE Transactions on 33 (7), 1415-1428.

Examples

```
out <- multivariate_karcher_mean(beta[, , 1, 1:2], maxit = 2)
# note: use more shapes, small for speed
K <- multivariate_karcher_cov(out)
```

multivariate_karcher_mean
Karcher Mean of Multivariate Functional Data

Description

Calculates the Karcher mean or median of a collection of multivariate functional data using the elastic square-root velocity (SRVF) framework. While most of the time, the setting does not require a metric that is invariant to rotation and scale, this can be achieved through the optional arguments `rotation` and `scale`.

Usage

```
multivariate_karcher_mean(
  beta,
  mode = "O",
  alignment = TRUE,
  rotation = FALSE,
  scale = FALSE,
  lambda = 0,
  maxit = 20L,
  ms = c("mean", "median"),
  exact_medoid = FALSE,
  ncores = 1L,
  verbose = FALSE
)
```

Arguments

`beta`	A numeric array of shape $L \times M \times N$ specifying an N-sample of L-dimensional functional data evaluated on a same grid of size M.
`mode`	A character string specifying whether the input curves should be considered open (`mode == "O"`) or closed (`mode == "C"`). Defaults to `"O"`.
`alignment`	A boolean value specifying whether the curves should be aligned before computing the distance matrix. Defaults to `TRUE`.
`rotation`	A boolean specifying whether the distance should be made invariant by rotation. Defaults to `TRUE`.
`scale`	A boolean specifying whether the distance should be made invariant by scaling. This is effectively achieved by making SRVFs having unit length and switching to an appropriate metric on the sphere between normalized SRVFs. Defaults to `TRUE`.
`lambda`	A numeric value specifying the weight of a penalty term that constraints the warping function to be not too far from the identity. Defaults to `0.0`.
`maxit`	An integer value specifying the maximum number of iterations. Defaults to `20L`.

ms	A character string specifying whether the Karcher mean ("mean") or Karcher median ("median") is returned. Defaults to "mean".
exact_medoid	A boolean specifying whether to compute the exact medoid from the distance matrix or as the input curve closest to the pointwise mean. Defaults to FALSE for saving computational time.
ncores	An integer value specifying the number of cores to use for parallel computation. Defaults to 1L. The maximum number of available cores is determined by the **parallel** package. One core is always left out to avoid overloading the system.
verbose	A boolean specifying whether to print the progress of the algorithm. Defaults to FALSE.

Value

A fdacurve object with the following components:

- beta: A numeric array of shape $L \times M \times N$ storing the original input data.
- q: A numeric array of shape $L \times M \times N$ storing the SRVFs of the input data.
- betan: A numeric array of shape $L \times M \times N$ storing the aligned, possibly optimally rotated and optimally scaled, input data.
- qn: A numeric array of shape $L \times M \times N$ storing the SRVFs of the aligned, possibly optimally rotated and optimally scaled, input data.
- gamma: A numeric array of shape $L \times M \times N$ storing the warping functions of the aligned, possibly optimally rotated and optimally scaled, input data.
- R: A numeric array of shape $L \times L \times N$ storing the rotation matrices of the aligned, possibly optimally rotated and optimally scaled, input data.
- betamean: A numeric array of shape $L \times M$ storing the Karcher mean or median of the input data.
- qmean: A numeric array of shape $L \times M$ storing the Karcher mean or median of the SRVFs of the input data.
- type: A character string indicating whether the Karcher mean or median has been returned.
- E: A numeric vector storing the energy of the Karcher mean or median at each iteration.
- qun: A numeric vector storing the cost function of the Karcher mean or median at each iteration.

References

Srivastava, A., Klassen, E., Joshi, S., Jermyn, I., (2011). Shape analysis of elastic curves in euclidean spaces. IEEE Transactions on Pattern Analysis and Machine Intelligence, **33** (7), 1415-1428.

Examples

```
out <- multivariate_karcher_mean(beta[, , 1, 1:2], maxit = 2)
# note: use more functions, small for speed
```

multivariate_pca *Curve PCA*

Description

Calculate principal directions of a set of curves

Usage

```
multivariate_pca(
  align_data,
  no = 3,
  var_exp = NULL,
  ci = c(-1, 0, 1),
  mode = "O",
  showplot = TRUE
)
```

Arguments

align_data	fdacurve object from multivariate_karcher_mean of aligned data
no	number of components
var_exp	compute no based on value percent variance explained (example: 0.95) will override no
ci	geodesic standard deviations (default = c(-1,0,1))
mode	Open ("O") or Closed ("C") curves
showplot	show plots of principal directions (default = TRUE)

Value

Returns a curve_pca object containing

latent	singular values
U	singular vectors
coef	principal coefficients
pd	principal directions

References

Srivastava, A., Klassen, E., Joshi, S., Jermyn, I., (2011). Shape analysis of elastic curves in euclidean spaces. Pattern Analysis and Machine Intelligence, IEEE Transactions on 33 (7), 1415-1428.

Examples

```
align_data <- multivariate_karcher_mean(beta[, , 1, 1:2], maxit = 2)
out <- multivariate_pca(align_data)
```

`optimum.reparam` *Align two functions*

Description

This function aligns the SRVFs of two functions in R^1 defined on an interval $[t_{\min}, t_{\max}]$ using dynamic programming or RBFGS

Usage

```
optimum.reparam(
  Q1,
  T1,
  Q2,
  T2,
  lambda = 0,
  pen = "roughness",
  method = c("DP", "DPo", "SIMUL", "RBFGS"),
  f1o = 0,
  f2o = 0,
  nbhd_dim = 10,
  scl = 3
)
```

Arguments

`Q1`	A numeric matrix of shape `n_points x n_dimensions` specifying the SRSF of the 1st `n_dimensions`-dimensional function evaluated on a grid of size `n_points` of its univariate domain.
`T1`	A numeric vector of size `n_points` specifying the grid on which the 1st SRSF is evaluated.
`Q2`	A numeric matrix of shape `n_points x n_dimensions` specifying the SRSF of the 2nd `n_dimensions`-dimensional function evaluated on a grid of size `n_points` of its univariate domain.
`T2`	A numeric vector of size `n_points` specifying the grid on which the 1st SRSF is evaluated.
`lambda`	A numeric value specifying the amount of warping. Defaults to `0.0`.
`pen`	alignment penalty (default="roughness") options are no penalty ("none"), second derivative ("roughness"), geodesic distance from id ("geodesic"), norm from id ("l2gam"), and srvf norm from id ("l2psi"). The penalty is weighted by `lambda`, so it has no effect when `lambda = 0`. It is used by the `"DP"`, `"DPo"`, and `"RBFGS"` methods; `"SIMUL"` ignores it.
`method`	A string specifying the optimization method. Choices are `"DP"`, `"DPo"`, `"SIMUL"`, or `"RBFGS"`. Defaults to `"DP"`.

f1o	A numeric vector of size n_dimensions specifying the value of the 1st function at $t = t_{\min}$. Defaults to rep(0, n_dimensions).
f2o	A numeric vector of size n_dimensions specifying the value of the 2nd function at $t = t_{\min}$. Defaults to rep(0, n_dimensions).
nbhd_dim	size of the grid for "DP" (default = 10, the 63 slopes "DPo" searches)
scl	for "DPo", the SRVFs are resampled scl times finer to evaluate the alignment cost; the warp stays on the grid T1 (default = 3, 1 turns the resampling off). "DP" integrates the linearly interpolated SRVFs exactly, so it does not use scl.

Value

A numeric vector of size n_points storing discrete evaluations of the estimated boundary-preserving warping diffeomorphism on the initial grid.

References

Srivastava, A., Wu, W., Kurtek, S., Klassen, E., Marron, J. S., May 2011. Registration of functional data using Fisher-Rao metric, arXiv:1103.3817v2.

Tucker, J. D., Wu, W., Srivastava, A., Generative models for functional data using phase and amplitude separation, Computational Statistics and Data Analysis (2012), 10.1016/j.csda.2012.12.001.

Examples

```
q <- f_to_srvf(simu_data$f, simu_data$time)
gam <- optimum.reparam(q[, 1], simu_data$time, q[, 2], simu_data$time)
```

---

outlier.detection *Outlier Detection*

Description

This function calculates outlier's using geodesic distances of the SRVFs from the median

Usage

```
outlier.detection(q, time, mq, k = 1.5)
```

Arguments

q	matrix (N x M) of M SRVF functions with N samples
time	vector of size N describing the sample points
mq	median calculated using time_warping()
k	cutoff threshold (default = 1.5)

Value

q_outlier	outlier functions

References

Srivastava, A., Wu, W., Kurtek, S., Klassen, E., Marron, J. S., May 2011. Registration of functional data using fisher-rao metric, arXiv:1103.3817v2.

Tucker, J. D., Wu, W., Srivastava, A., Generative Models for Function Data using Phase and Amplitude Separation, Computational Statistics and Data Analysis (2012), 10.1016/j.csda.2012.12.001.

Examples

```
q_outlier <- outlier.detection(
  q = toy_warp$q0,
  time = toy_data$time,
  mq = toy_warp$mqn,
  k = .1
)
```

`pair_align_functions` *Align two functions*

Description

This function aligns two functions using SRSF framework. It will align f2 to f1

Usage

```
pair_align_functions(
  f1,
  f2,
  time,
  lambda = 0,
  pen = "roughness",
  method = "DP",
  w = 0.01,
  iter = 2000
)
```

Arguments

`f1`	function 1
`f2`	function 2
`time`	sample points of functions
`lambda`	controls amount of warping (default = 0)
`pen`	alignment penalty (default="roughness") options are second derivative ("roughness"), L^2 distance of the warping function from id ("l2gam"), L^2 distance of the SRVF of the warping function from that of id ("l2psi"), geodesic distance from id ("geodesic"), and no penalty ("none"). "norm" is kept for backward compatibility as an alias for "l2gam". The penalty is weighted by `lambda`, so it has no effect when `lambda = 0`.

method	controls which optimization method (default="DP") options are Dynamic Programming ("DP"), the original Dynamic Programming implementation ("DPo"), Riemannian BFGS ("RBFGS"), Simultaneous Alignment ("SIMUL"), Dirichlet Bayesian ("dBayes"), and Expo-Map Bayesian ("expBayes")
w	controls LRBFGS (default = 0.01)
iter	number of mcmc iterations for mcmc method (default 2000)

Value

Returns a list containing

f2tilde	aligned f2
gam	warping function

References

Srivastava, A., Wu, W., Kurtek, S., Klassen, E., Marron, J. S., May 2011. Registration of functional data using fisher-rao metric, arXiv:1103.3817v2.

Tucker, J. D., Wu, W., Srivastava, A., Generative Models for Function Data using Phase and Amplitude Separation, Computational Statistics and Data Analysis (2012), 10.1016/j.csda.2012.12.001.

Cheng, W., Dryden, I. L., and Huang, X. (2016). Bayesian registration of functions and curves. Bayesian Analysis, 11(2), 447-475.

Lu, Y., Herbei, R., and Kurtek, S. (2017). Bayesian registration of functions with a Gaussian process prior. Journal of Computational and Graphical Statistics, DOI: 10.1080/10618600.2017.1336444.

Examples

```
out <- pair_align_functions(
  f1 = simu_data$f[, 1],
  f2 = simu_data$f[, 2],
  time = simu_data$time
)
```

---

pair_align_functions_bayes

Align two functions

Description

This function aligns two functions using Bayesian SRSF framework. It will align f2 to f1

Usage

```
pair_align_functions_bayes(
  f1,
  f2,
  timet,
  iter = 15000,
  times = 5,
  tau = ceiling(times * 0.4),
  powera = 1,
  showplot = TRUE,
  extrainfo = FALSE
)
```

Arguments

f1	function 1
f2	function 2
timet	sample points of functions
iter	number of iterations (default = 15000)
times	factor of length of subsample points to look at (default = 5)
tau	standard deviation of Normal prior for increment (default ceil(times*.4))
powera	Dirichlet prior parameter (default 1)
showplot	shows plots of functions (default = T)
extrainfo	T/F whether additional information is returned

Value

Returns a list containing

f1	function 1
f2_q	registered function using quotient space
gam_q	warping function quotient space
f2_a	registered function using ambient space
q2_a	warping function ambient space
match_collect	posterior samples from warping function (returned if extrainfo=TRUE)
dist_collect	posterior samples from the distances (returned if extrainfo=TRUE)
kappa_collect	posterior samples from kappa (returned if extrainfo=TRUE)
log_collect	log-likelihood of each sample (returned if extrainfo=TRUE)
pct_accept	vector of acceptance ratios for the warping function (returned if extrainfo=TRUE)

References

Cheng, W., Dryden, I. L., and Huang, X. (2016). Bayesian registration of functions and curves. Bayesian Analysis, 11(2), 447-475.

Examples

```
out <- pair_align_functions_bayes(
  f1 = simu_data$f[, 1],
  f2 = simu_data$f[, 2],
  timet = simu_data$time
)
```

`pair_align_functions_expomap` *Align two functions using geometric properties of warping functions*

Description

This function aligns two functions using Bayesian framework. It will align f2 to f1. It is based on mapping warping functions to a hypersphere, and a subsequent exponential mapping to a tangent space. In the tangent space, the Z-mixture pCN algorithm is used to explore both local and global structure in the posterior distribution.

Usage

```
pair_align_functions_expomap(
  f1,
  f2,
  timet,
  iter = 20000,
  burnin = min(5000, iter/2),
  alpha0 = 0.1,
  beta0 = 0.1,
  zpcn = list(betas = c(0.5, 0.05, 0.005, 1e-04), probs = c(0.1, 0.1, 0.7, 0.1)),
  propvar = 1,
  init.coef = rep(0, 2 * 10),
  npoints = 200,
  extrainfo = FALSE
)
```

Arguments

`f1`	observed data, numeric vector
`f2`	observed data, numeric vector
`timet`	sample points of functions
`iter`	length of the chain
`burnin`	number of burnin MCMC iterations
`alpha0`, `beta0`	IG parameters for the prior of sigma1
`zpcn`	list of mixture coefficients and prior probabilities for Z-mixture pCN algorithm of the form list(betas, probs), where betas and probs are numeric vectors of equal length

propvar	variance of proposal distribution
init.coef	initial coefficients of warping function in exponential map; length must be even
npoints	number of sample points to use during alignment
extrainfo	T/F whether additional information is returned

Details

The Z-mixture pCN algorithm uses a mixture distribution for the proposal distribution, controlled by input parameter zpcn. The zpcn$betas must be between 0 and 1, and are the coefficients of the mixture components, with larger coefficients corresponding to larger shifts in parameter space. The zpcn$probs give the probability of each shift size.

Value

Returns a list containing

f2_warped	f2 aligned to f1
gamma	Posterior mean gamma function
g.coef	matrix with iter columns, posterior draws of g.coef
psi	Posterior mean psi function
sigma1	numeric vector of length iter, posterior draws of sigma1
accept	Boolean acceptance for each sample (if extrainfo=TRUE)
betas.ind	Index of zpcn mixture component for each sample (if extrainfo=TRUE)
logl	numeric vector of length iter, posterior loglikelihood (if extrainfo=TRUE)
gamma_mat	Matrix of all posterior draws of gamma (if extrainfo=TRUE)
gamma_q025	Lower 0.025 quantile of gamma (if extrainfo=TRUE)
gamma_q975	Upper 0.975 quantile of gamma (if extrainfo=TRUE)
sigma_eff_size	Effective sample size of sigma (if extrainfo=TRUE)
psi_eff_size	Vector of effective sample sizes of psi (if extrainfo=TRUE)
xdist	Vector of posterior draws from xdist between registered functions (if extrainfo=TRUE)
ydist	Vector of posterior draws from ydist between registered functions (if extrainfo=TRUE)

References

Lu, Y., Herbei, R., and Kurtek, S. (2017). Bayesian registration of functions with a Gaussian process prior. Journal of Computational and Graphical Statistics, DOI: 10.1080/10618600.2017.1336444.

Examples

```
## Not run:
  # This is an MCMC algorithm and takes a long time to run
  myzpcn <- list(
    betas = c(0.1, 0.01, 0.005, 0.0001),
    probs = c(0.2, 0.2, 0.4, 0.2)
  )
```

```
  out <- pair_align_functions_expomap(
    f1 = simu_data$f[, 1],
    f2 = simu_data$f[, 2],
    timet = simu_data$time,
    zpcn = myzpcn,
    extrainfo = TRUE
  )
  # overall acceptance ratio
  mean(out$accept)
  # acceptance ratio by zpcn coefficient
  with(out, tapply(accept, myzpcn$betas[betas.ind], mean))

## End(Not run)
```

`pair_align_image` *Pairwise align two images This function aligns to images using the q-map framework*

Description

Pairwise align two images This function aligns to images using the q-map framework

Usage

```
pair_align_image(
  I1,
  I2,
  M = 5,
  ortho = TRUE,
  basis_type = "t",
  resizei = FALSE,
  N = 64,
  stepsize = 1e-05,
  itermax = 1000
)
```

Arguments

`I1`	reference image
`I2`	image to warp
`M`	number of basis elements (default=5)
`ortho`	orthonormalize basis (default=TRUE)
`basis_type`	("t","s","i","o"; default="t")
`resizei`	resize image (default=TRUE)
`N`	size of resized image (default=64)
`stepsize`	gradient stepsize (default=1e-5)
`itermax`	maximum number of iterations (default=1000)

Value

Returns a list containing

Inew	aligned I2
gam	warping function

References

Q. Xie, S. Kurtek, E. Klassen, G. E. Christensen and A. Srivastava. Metric-based pairwise and multiple image registration. IEEE European Conference on Computer Vision (ECCV), September, 2014

Examples

```
## Not run:
  # This is a gradient descent algorithm and takes a long time to run
  out <- pair_align_image(im$I1, im$I2)

## End(Not run)
```

---

pcaTB *Tolerance Bound Calculation using Elastic Functional PCA*

Description

This function computes tolerance bounds for functional data containing phase and amplitude variation using principal component analysis

Usage

```
pcaTB(f, time, m = 4, B = 1e+05, a = 0.05, p = 0.99)
```

Arguments

f	matrix of functions
time	vector describing time sampling
m	number of principal components (default = 4)
B	number of monte carlo iterations
a	confidence level of tolerance bound (default = 0.05)
p	coverage level of tolerance bound (default = 0.99)

Value

Returns a list containing

pca	pca output
tol	tolerance factor

References

J. D. Tucker, J. R. Lewis, C. King, and S. Kurtek, "A Geometric Approach for Computing Tolerance Bounds for Elastic Functional Data," Journal of Applied Statistics, 10.1080/02664763.2019.1645818, 2019.

Tucker, J. D., Wu, W., Srivastava, A., Generative Models for Function Data using Phase and Amplitude Separation, Computational Statistics and Data Analysis (2012), 10.1016/j.csda.2012.12.001.

Jung, S. L. a. S. (2016). "Combined Analysis of Amplitude and Phase Variations in Functional Data." arXiv:1603.01775.

Examples

```
## Not run:
  out1 <- pcaTB(simu_data$f, simu_data$time)

## End(Not run)
```

`plot_curve` *Plot Curve*

Description

This function plots open or closed curves

Usage

```
plot_curve(beta, add = FALSE, ...)
```

Arguments

`beta`	Array of sizes $n \times T$ describing a curve of dimension n evaluated on T points
`add`	add to current plot (default = `TRUE`)
`...`	additional plotting parameters

Value

Return shape confidence intervals

ppd *Compute Peak Persistence Diagram*

Description

This computes the peak persistence diagram over a range of lambda. This can help determine the proper elasticity (penalty). This can be slow and recommended to run in parallel

Usage

```
ppd(
  f,
  time,
  max_lam = 2,
  num_lam = 10,
  pt = 0.15,
  srvf = FALSE,
  mu = NaN,
  penalty_method = c("roughness", "l2gam", "l2psi", "geodesic", "none", "norm"),
  centroid_type = c("mean", "median"),
  center_warpings = TRUE,
  smooth_data = FALSE,
  sparam = 25L,
  parallel = TRUE,
  cores = -1,
  optim_method = c("DP", "DPo", "RBFGS"),
  max_iter = 20L
)
```

Arguments

f	A numeric matrix of shape $M \times N$ specifying a sample of N curves observed on a grid of size M.
time	A numeric vector of length M specifying the common grid on which all curves f have been observed.
max_lam	maximum value of lambda. Defaults to 2
num_lam	number of steps. Defaults to 10
pt	the percentile of negative curvature of raw data Defaults to .15
srvf	perform in srvf space Defaults to FALSE
mu	vector of size N that f is aligned to, Defaults to NaN
penalty_method	A string specifying the penalty term used in the formulation of the cost function to minimize for alignment. Choices are "roughness" which uses the norm of the second derivative, "l2gam" which uses the L^2 distance of the warping function to the identity, "l2psi" which uses the L^2 distance of the SRVF of the warping function to that of the identity, "geodesic" which uses the geodesic

	distance to the identity, and "none" which applies no penalty. "norm" is kept for backward compatibility as an alias for "l2gam". Defaults to "roughness".
centroid_type	A string specifying the type of centroid to align to. Choices are "mean" or "median". Defaults to "mean".
center_warpings	A boolean specifying whether to center the estimated warping functions. Defaults to TRUE.
smooth_data	A boolean specifying whether to smooth curves using a box filter. Defaults to FALSE.
sparam	An integer value specifying the number of times to apply the box filter. Defaults to 25L. This is used only when smooth_data = TRUE.
parallel	A boolean specifying whether to run calculations in parallel. Defaults to TRUE.
cores	number of cores in parallel (default=-1, means all cores)
optim_method	A string specifying the algorithm used for optimization. Choices are "DP", "DPo", and "RBFGS". Defaults to "DP".
max_iter	An integer value specifying the maximum number of iterations. Defaults to 20L.

Value

lam_opt optimal lam

References

Kim, Woo Min, Sutanoy Dasgupta, and Anuj Srivastava. "Peak-persistence diagrams for estimating shapes and functions from noisy data." arXiv preprint arXiv:2305.04826 (2023).

Examples

```
## Not run:
  out <- ppd(simu_data$f, simu_data$time)

## End(Not run)
```

---

predict.curve_pca *Elastic Prediction for curve PCA*

Description

This function performs projection of new curves on fPCA basis

Usage

```
## S3 method for class 'curve_pca'
predict(object, newdata = NULL, ...)
```

Arguments

object	Object of class inheriting from "curve_pca"
newdata	An optional matrix in which to look for functions with which to predict. If omitted, the original functions are used.
...	additional arguments affecting the predictions produced

Value

Returns a matrix

a	principal coefficients

References

Srivastava, A., Klassen, E., Joshi, S., Jermyn, I., (2011). Shape analysis of elastic curves in euclidean spaces. Pattern Analysis and Machine Intelligence, IEEE Transactions on 33 (7), 1415-1428.

---

predict.hfpca *Elastic Prediction for functional PCA*

Description

This function performs projection of new functions on fPCA basis

Usage

```
## S3 method for class 'hfpca'
predict(object, newdata = NULL, ...)
```

Arguments

object	Object of class inheriting from "horizFPCA"
newdata	An optional matrix in which to look for functions with which to predict. If omitted, the original functions are used.
...	additional arguments affecting the predictions produced

Value

Returns a matrix

a	principal coefficients

References

Tucker, J. D., Wu, W., Srivastava, A., Generative Models for Function Data using Phase and Amplitude Separation, Computational Statistics and Data Analysis (2012), 10.1016/j.csda.2012.12.001.

predict.jfpca *Elastic Prediction for functional PCA*

Description

This function performs projection of new functions on fPCA basis

Usage

```
## S3 method for class 'jfpca'
predict(object, newdata = NULL, ...)
```

Arguments

object	Object of class inheriting from "jointFPCA"
newdata	An optional matrix in which to look for functions with which to predict. If omitted, the original functions are used.
...	additional arguments affecting the predictions produced

Value

Returns a matrix

a	principal coefficients

References

Tucker, J. D., Wu, W., Srivastava, A., Generative Models for Function Data using Phase and Amplitude Separation, Computational Statistics and Data Analysis (2012), 10.1016/j.csda.2012.12.001.

predict.jfpcah *Elastic Prediction for functional PCA*

Description

This function performs projection of new functions on fPCA basis

Usage

```
## S3 method for class 'jfpcah'
predict(object, newdata = NULL, ...)
```

Arguments

object	Object of class inheriting from "jointFPCA"
newdata	An optional matrix in which to look for functions with which to predict. If omitted, the original functions are used.
...	additional arguments affecting the predictions produced

Value

Returns a matrix

a	principal coefficients

References

Tucker, J. D., Wu, W., Srivastava, A., Generative Models for Function Data using Phase and Amplitude Separation, Computational Statistics and Data Analysis (2012), 10.1016/j.csda.2012.12.001.

---

predict.lpcr *Elastic Prediction for functional logistic PCR Model*

Description

This function performs prediction from an elastic logistic fPCR regression model with phase-variability

Usage

```
## S3 method for class 'lpcr'
predict(object, newdata = NULL, y = NULL, ...)
```

Arguments

object	Object of class inheriting from "elastic.pcr.regression"
newdata	An optional matrix in which to look for variables with which to predict. If omitted, the fitted values are used.
y	An optional vector of labels to calculate PC. If omitted, PC is NULL
...	additional arguments affecting the predictions produced

Value

Returns a list containing

y_pred	predicted probabilities of the class of newdata
y_labels	class labels of newdata
PC	probability of classification

References

J. D. Tucker, J. R. Lewis, and A. Srivastava, "Elastic Functional Principal Component Regression," Statistical Analysis and Data Mining, 10.1002/sam.11399, 2018.

`predict.mlpcr` *Elastic Prediction for functional multinomial logistic PCR Model*

Description

This function performs prediction from an elastic multinomial logistic fPCR regression model with phase-variability

Usage

```
## S3 method for class 'mlpcr'
predict(object, newdata = NULL, y = NULL, ...)
```

Arguments

`object`	Object of class inheriting from "elastic.pcr.regression"
`newdata`	An optional matrix in which to look for variables with which to predict. If omitted, the fitted values are used.
`y`	An optional vector of labels to calculate PC. If omitted, PC is NULL
`...`	additional arguments affecting the predictions produced

Value

Returns a list containing

`y_pred`	predicted probabilities of the class of newdata
`y_labels`	class labels of newdata
`PC`	probability of classification per class
`PC.comb`	total probability of classification

References

J. D. Tucker, J. R. Lewis, and A. Srivastava, "Elastic Functional Principal Component Regression," Statistical Analysis and Data Mining, 10.1002/sam.11399, 2018.

predict.pcr *Elastic Prediction for functional PCR Model*

Description

This function performs prediction from an elastic pcr regression model with phase-variability

Usage

```
## S3 method for class 'pcr'
predict(object, newdata = NULL, y = NULL, ...)
```

Arguments

object	Object of class inheriting from "elastic.pcr.regression"
newdata	An optional matrix in which to look for variables with which to predict. If omitted, the fitted values are used.
y	An optional vector of responses to calculate SSE. If omitted, SSE is NULL
...	additional arguments affecting the predictions produced

Value

Returns a list containing

y_pred	predicted values of newdata
SSE	sum of squared errors

References

J. D. Tucker, J. R. Lewis, and A. Srivastava, “Elastic Functional Principal Component Regression,” Statistical Analysis and Data Mining, 10.1002/sam.11399, 2018.

predict.vfpca *Elastic Prediction for functional PCA*

Description

This function performs projection of new functions on fPCA basis

Usage

```
## S3 method for class 'vfpca'
predict(object, newdata = NULL, ...)
```

Arguments

object	Object of class inheriting from "vertFPCA"
newdata	An optional matrix in which to look for functions with which to predict. If omitted, the original functions are used.
...	additional arguments affecting the predictions produced

Value

Returns a matrix

a	principal coefficients

References

Tucker, J. D., Wu, W., Srivastava, A., Generative Models for Function Data using Phase and Amplitude Separation, Computational Statistics and Data Analysis (2012), 10.1016/j.csda.2012.12.001.

---

psi_to_gam *map hilbert sphere to warping function*

Description

map hilbert sphere to warping function

Usage

```
psi_to_gam(psi)
```

Arguments

psi	Either a numeric vector of a numeric matrix or a numeric array specifying the shooting vectors

Value

A numeric array of the same shape as the input array `psi` storing the warping functions `psi`.

q_to_curve *Convert to curve space*

Description

This function converts SRVFs to curves

Usage

```
q_to_curve(q, scale = 1)
```

Arguments

q	either a matrix of shape $n \times T$ describing SRVF or SRVF of multidimensional functional data in R^n, where n is the dimension and T is the number of time points
scale	scale of from curve_to_q (default = 1)

Value

beta array describing curve

References

Srivastava, A., Klassen, E., Joshi, S., Jermyn, I., (2011). Shape analysis of elastic curves in euclidean spaces. Pattern Analysis and Machine Intelligence, IEEE Transactions on 33 (7), 1415-1428.

Examples

```
q <- curve_to_q(beta[, , 1, 1])$q
beta1 <- q_to_curve(q)
```

reparam_curve *Align two curves*

Description

This function aligns two SRVF functions using Dynamic Programming. If the curves beta1 and beta2 are describing multidimensional functional data, then `rotation == FALSE` and `mode == 'O'`

Usage

```
reparam_curve(
  beta1,
  beta2,
  lambda = 0,
  method = "DP",
  w = 0.01,
  rotated = TRUE,
  isclosed = FALSE,
  mode = "O"
)
```

Arguments

beta1	curve 1, provided as a matrix of dimensions $n \times M$ for n-dimensional curve evaluated on M sample points
beta2	curve 1, provided as a matrix of dimensions $n \times M$ for n-dimensional curve evaluated on M sample points
lambda	controls amount of warping (default = 0)
method	controls which optimization method. Options are Dynamic Programming ("DP"). (default = "DP")
w	controls LRBFGS (default = 0.01)
rotated	boolean if rotation is desired
isclosed	boolean if curve is closed
mode	Open ("O") or Closed ("C") curves

Value

return a List containing

gam	warping function
R	rotation matrix
tau	seed point

References

Srivastava, A., Klassen, E., Joshi, S., Jermyn, I., (2011). Shape analysis of elastic curves in euclidean spaces. Pattern Analysis and Machine Intelligence, IEEE Transactions on 33 (7), 1415-1428.

Examples

```
gam <- reparam_curve(beta[, , 1, 1], beta[, , 1, 5])$gam
```

`reparam_image` *Find optimum reparameterization between two images*

Description

Finds the optimal warping function between two images using the elastic framework

Usage

```
reparam_image(It, Im, gam, b, stepsize = 1e-05, itermax = 1000, lmark = FALSE)
```

Arguments

`It`	template image matrix
`Im`	test image matrix
`gam`	initial warping array
`b`	basis matrix
`stepsize`	gradient stepsize (default=1e-5)
`itermax`	maximum number of iterations (default=1000)
`lmark`	use landmarks (default=FALSE)

Value

Returns a list containing

`gamnew`	final warping
`Inew`	aligned image
`H`	energy
`stepsize`	final stepsize

References

Q. Xie, S. Kurtek, E. Klassen, G. E. Christensen and A. Srivastava. Metric-based pairwise and multiple image registration. IEEE European Conference on Computer Vision (ECCV), September, 2014

resamplecurve *Resample Curve*

Description

This function resamples a curve to a number of points

Usage

```
resamplecurve(x, N = 100, mode = "O")
```

Arguments

x	matrix defining curve (n,T)
N	Number of samples to re-sample curve, N usually is > T
mode	Open ("O") or Closed ("C") curves

Value

xn matrix defining resampled curve

References

Srivastava, A., Klassen, E., Joshi, S., Jermyn, I., (2011). Shape analysis of elastic curves in euclidean spaces. Pattern Analysis and Machine Intelligence, IEEE Transactions on 33 (7), 1415-1428.

Examples

```
xn <- resamplecurve(beta[, , 1, 1], 200)
```

rgam *Random Warping*

Description

Generates random warping functions

Usage

```
rgam(N, sigma, num)
```

Arguments

N	length of warping function
sigma	variance of warping functions
num	number of warping functions

Value

gam warping functions

References

Srivastava, A., Wu, W., Kurtek, S., Klassen, E., Marron, J. S., May 2011. Registration of functional data using fisher-rao metric, arXiv:1103.3817v2.

Tucker, J. D., Wu, W., Srivastava, A., Generative Models for Function Data using Phase and Amplitude Separation, Computational Statistics and Data Analysis (2012), 10.1016/j.csda.2012.12.001.

Examples

```
gam = rgam(N=101, sigma=.01, num=35)
```

---

rotation_pca *Rotation Principal Component Analysis*

Description

This function calculates functional principal component analysis on rotation data from

Usage

```
rotation_pca(align_data, no = 3, var_exp = NULL)
```

Arguments

align_data	fdacurve object from multivariate_karcher_mean of aligned data
no	number of principal components to extract
var_exp	compute no based on value percent variance explained (example: 0.95) will override no

Value

Returns a rotpca object containing

latent	latent values
coef	coefficients
U	eigenvectors

`sample_shapes` *Sample shapes from model*

Description

Sample shapes from model

Usage

```
sample_shapes(x, no = 3, numSamp = 10)
```

Arguments

`x`	An object of class `fdacurve` typically produced by `multivariate_karcher_mean()`
`no`	number of principal components
`numSamp`	number of samples

Value

Returns a `fdacurve` object containing

`betans`	random aligned curves
`qns`	random aligned srvfs
`betans`	random curves
`qs`	random srvfs
`gams`	random reparameterization functions
`R`	random rotation matrices

References

Srivastava, A., Klassen, E., Joshi, S., Jermyn, I., (2011). Shape analysis of elastic curves in euclidean spaces. Pattern Analysis and Machine Intelligence, IEEE Transactions on 33 (7), 1415-1428.

Examples

```
out <- multivariate_karcher_mean(beta[, , 1, 1:5], scale=TRUE, maxit = 2)
# note: use more shapes, small for speed
out.samples <- sample_shapes(out)
```

shape_CI *Shape Confidence Interval Calculation using Bootstrap Sampling*

Description

This function computes Confidence bounds for shapes using elastic metric

Usage

```
shape_CI(
  beta,
  a = 0.95,
  no = 5,
  Nsamp = 100,
  mode = "O",
  rotated = TRUE,
  scale = TRUE,
  lambda = 0,
  parallel = TRUE
)
```

Arguments

beta	Array of sizes $n \times T \times N$ describing N curves of dimension n evaluated on T points
a	confidence level (default = 0.95)
no	number of principal components (default = 5)
Nsamp	number of functions to generate (default = 100)
mode	Open ("O") or Closed ("C") curves
rotated	Optimize over rotation (default = TRUE)
scale	scale curves to unit length (default = TRUE)
lambda	A numeric value specifying the elasticity. Defaults to 0.0.
parallel	enable parallel processing (default = T)

Value

Return shape confidence intervals

References

J. D. Tucker, J. R. Lewis, C. King, and S. Kurtek, "A Geometric Approach for Computing Tolerance Bounds for Elastic Functional Data," Journal of Applied Statistics, 10.1080/02664763.2019.1645818, 2019.

Tucker, J. D., Wu, W., Srivastava, A., Generative Models for Function Data using Phase and Amplitude Separation, Computational Statistics and Data Analysis (2012), 10.1016/j.csda.2012.12.001.

simu_data *Simulated two Gaussian Dataset*

Description

A functional dataset where the individual functions are given by: $y_i(t) = z_{i,1}e^{-(t-1.5)^2/2} + z_{i,2}e^{-(t+1.5)^2/2}$, $t \in [-3,3]$, $i = 1,2,\ldots,21$, where $z_{i,1}$ and $z_{i,2}$ are *i.i.d.* normal with mean one and standard deviation 0.25. Each of these functions is then warped according to: $\gamma_i(t) = 6(\frac{e^{a_i(t+3)/6}-1}{e^{a_i}-1}) - 3$ if $a_i \neq 0$, otherwise $\gamma_i = \gamma_{id}$ ($gamma_{id}(t) = t$) is the identity warping). The variables are as follows: f containing the 21 functions of 101 samples and time which describes the sampling.

Usage

```
simu_data
```

Format

`simu_data`**:**

A list with 2 components:

- `f`: A numeric matrix of shape 101×21 storing a sample of size $N = 21$ of curves evaluated on a grid of size $M = 101$.
- `time`: A numeric vector of size $M = 101$ storing the grid on which the curves `f` have been evaluated.

simu_warp *Aligned Simulated two Gaussian Dataset*

Description

A functional dataset where the individual functions are given by: $y_i(t) = z_{i,1}e^{-(t-1.5)^2/2} + z_{i,2}e^{-(t+1.5)^2/2}$, $t \in [-3,3]$, $i = 1,2,\ldots,21$, where $z_{i,1}$ and $z_{i,2}$ are *i.i.d.* normal with mean one and standard deviation 0.25. Each of these functions is then warped according to: $\gamma_i(t) = 6(\frac{e^{a_i(t+3)/6}-1}{e^{a_i}-1}) - 3$ if $a_i \neq 0$, otherwise $\gamma_i = \gamma_{id}$ ($gamma_{id}(t) = t$) is the identity warping). The variables are as follows: f containing the 21 functions of 101 samples and time which describes the sampling which has been aligned.

Usage

```
simu_warp
```

Format

`simu_warp`**:**

A list which contains the output of the `time_warping()` function applied on the data set `simu_data`.

simu_warp_median *Aligned Simulated two Gaussian Dataset using Median*

Description

A functional dataset where the individual functions are given by: $y_i(t) = z_{i,1}e^{-(t-1.5)^2/2} + z_{i,2}e^{-(t+1.5)^2/2}$, $t \in [-3,3]$, $i = 1,2,\ldots,21$, where $z_{i,1}$ and $z_{i,2}$ are *i.i.d.* normal with mean one and standard deviation 0.25. Each of these functions is then warped according to: $\gamma_i(t) = 6(\frac{e^{a_i(t+3)/6}-1}{e^{a_i}-1}) - 3$ if $a_i \neq 0$, otherwise $\gamma_i = \gamma_{id}$ ($gamma_{id}(t) = t$) is the identity warping). The variables are as follows: f containing the 21 functions of 101 samples and time which describes the sampling which has been aligned.

Usage

```
simu_warp_median
```

Format

`simu_warp_median`**:**

A list which contains the output of the `time_warping()` function finding the median applied on the data set `simu_data`.

smooth.data *Smooth Functions*

Description

This function smooths functions using standard box filter

Usage

```
smooth.data(f, sparam)
```

Arguments

`f`	matrix (N x M) of M functions with N samples
`sparam`	number of times to run box filter

Value

fo smoothed functions

References

Srivastava, A., Wu, W., Kurtek, S., Klassen, E., Marron, J. S., May 2011. Registration of functional data using fisher-rao metric, arXiv:1103.3817v2.

Tucker, J. D., Wu, W., Srivastava, A., Generative Models for Function Data using Phase and Amplitude Separation, Computational Statistics and Data Analysis (2012), 10.1016/j.csda.2012.12.001.

Examples

```
fo <- smooth.data(simu_data$f, 25)
```

`SqrtMean` *SRVF transform of warping functions*

Description

This function calculates the srvf of warping functions with corresponding shooting vectors and finds the mean

Usage

```
SqrtMean(gam)
```

Arguments

`gam`	matrix (N x M) of M warping functions with N samples

Value

Returns a list containing

`mu`	Karcher mean psi function
`gam_mu`	Karcher mean warping function
`psi`	srvf of warping functions
`vec`	shooting vectors

References

Srivastava, A., Wu, W., Kurtek, S., Klassen, E., Marron, J. S., May 2011. Registration of functional data using fisher-rao metric, arXiv:1103.3817v2.

Tucker, J. D., Wu, W., Srivastava, A., Generative Models for Function Data using Phase and Amplitude Separation, Computational Statistics and Data Analysis (2012), 10.1016/j.csda.2012.12.001.

Examples

```
out <- SqrtMean(simu_warp$warping_functions)
```

SqrtMeanInverse *SRVF transform of warping functions*

Description

This function calculates the srvf of warping functions with corresponding shooting vectors and finds the inverse of mean

Usage

```
SqrtMeanInverse(gam)
```

Arguments

gam	matrix (N x M) of M warping functions with N samples

Value

gamI inverse of Karcher mean warping function

References

Srivastava, A., Wu, W., Kurtek, S., Klassen, E., Marron, J. S., May 2011. Registration of functional data using fisher-rao metric, arXiv:1103.3817v2.

Tucker, J. D., Wu, W., Srivastava, A., Generative Models for Function Data using Phase and Amplitude Separation, Computational Statistics and Data Analysis (2012), 10.1016/j.csda.2012.12.001.

Examples

```
gamI <- SqrtMeanInverse(simu_warp$warping_functions)
```

SqrtMedian *SRVF transform of warping functions*

Description

This function calculates the srvf of warping functions with corresponding shooting vectors and finds the median

Usage

```
SqrtMedian(gam)
```

Arguments

gam	matrix (N x M) of M warping functions with N samples

Value

Returns a list containing

median	Karcher median psi function
gam_median	Karcher mean warping function
psi	srvf of warping functions
vec	shooting vectors

References

Srivastava, A., Wu, W., Kurtek, S., Klassen, E., Marron, J. S., May 2011. Registration of functional data using fisher-rao metric, arXiv:1103.3817v2.

Tucker, J. D., Wu, W., Srivastava, A., Generative Models for Function Data using Phase and Amplitude Separation, Computational Statistics and Data Analysis (2012), 10.1016/j.csda.2012.12.001.

Examples

```
out <- SqrtMedian(simu_warp_median$warping_functions)
```

---

srvf_to_f *Transformation from SRSF Space*

Description

This function transforms SRVFs back to the original functional space for functions in R^1.

Usage

```
srvf_to_f(q, time, f0 = 0)
```

Arguments

q	Either a numeric vector of a numeric matrix or a numeric array specifying the SRSFs that need to be transformed. • If a vector, it must be of shape M and it is interpreted as a single 1-dimensional curve observed on a grid of size M. • If a matrix, it must be of shape $M \times N$. In this case, it is interpreted as a sample of N curves observed on a grid of size M, unless $M = 1$ in which case it is interpreted as a single 1-dimensional curve observed on a grid of size M.
time	A numeric vector of length M specifying the grid on which SRSFs are evaluated.
f0	Either a numeric value or a numeric vector of or a numeric matrix specifying the initial value of the curves in the original functional space. It must be: • a value if q represents a single SRSF. • a vector of length N if q represents a sample of N SRVFs

Value

A numeric array of the same shape as the input `q` storing the transformation of the SRVFs `q` back to the original functional space. The integration uses a trapezoid rule corrected with cubic-spline slopes of $q|q|$, which is the inverse of `f_to_srvf(smooth = FALSE)`.

References

Srivastava, A., Wu, W., Kurtek, S., Klassen, E., Marron, J. S., May 2011. Registration of functional data using fisher-rao metric, arXiv:1103.3817v2.

Tucker, J. D., Wu, W., Srivastava, A., Generative models for functional data using amplitude and phase separation, Computational Statistics and Data Analysis (2012), 10.1016/j.csda.2012.12.001.

Examples

```
q <- f_to_srvf(simu_data$f, simu_data$time)
f <- srvf_to_f(q, simu_data$time, simu_data$f[1, ])
```

---

srvf2curve *Converts from SRVF to curve representation*

Description

Converts from SRVF to curve representation

Usage

```
srvf2curve(qfun, beta0 = NULL)
```

Arguments

`qfun`	A function that takes a numeric vector s of values in $[0, 1]$ as input and returns the values of the SRVF of an underlying curve at s.
`beta0`	A numeric vector of length L specifying the initial value of the underlying curve at $s = 0$.

Value

A function that takes a numeric vector t of values in $[0, 1]$ as input and returns the values of the underlying curve at t.

Examples

```
srvf2curve(curve2srvf(beta[, , 1, 1]))
```

`time_warping` *Alignment of univariate functional data*

Description

This function aligns a collection of 1-dimensional curves that are observed on the same grid.

Usage

```
time_warping(
  f,
  time,
  lambda = 0,
  penalty_method = c("roughness", "l2gam", "l2psi", "geodesic", "none", "norm"),
  centroid_type = c("mean", "median"),
  center_warpings = TRUE,
  smooth_data = FALSE,
  sparam = 25L,
  parallel = FALSE,
  cores = -1,
  optim_method = c("DP", "DPo", "RBFGS"),
  max_iter = 20L
)
```

Arguments

`f`	A numeric matrix of shape $M \times N$ specifying a sample of N curves observed on a grid of size M.
`time`	A numeric vector of length M specifying the common grid on which all curves `f` have been observed.
`lambda`	A numeric value specifying the elasticity. Defaults to `0.0`.
`penalty_method`	A string specifying the penalty term used in the formulation of the cost function to minimize for alignment. Choices are `"roughness"` which uses the norm of the second derivative, `"l2gam"` which uses the L^2 distance of the warping function to the identity, `"l2psi"` which uses the L^2 distance of the SRVF of the warping function to that of the identity, `"geodesic"` which uses the geodesic distance to the identity, and `"none"` which applies no penalty. `"norm"` is kept for backward compatibility as an alias for `"l2gam"`. The penalty is weighted by `lambda`, so it has no effect when `lambda = 0`. Defaults to `"roughness"`.
`centroid_type`	A string specifying the type of centroid to align to. Choices are `"mean"` or `"median"`. Defaults to `"mean"`.
`center_warpings`	A boolean specifying whether to center the estimated warping functions. Defaults to `TRUE`.
`smooth_data`	A boolean specifying whether to smooth curves using a box filter. Defaults to `FALSE`.

sparam	An integer value specifying the number of times to apply the box filter. Defaults to 25L. This is used only when smooth_data = TRUE.
parallel	A boolean specifying whether to run calculations in parallel. Defaults to FALSE.
cores	number of cores in parallel (default=-1, means all cores)
optim_method	A string specifying the algorithm used for optimization. Choices are "DP", "DPo", and "RBFGS". Defaults to "DP".
max_iter	An integer value specifying the maximum number of iterations. Defaults to 20L.

Value

An object of class fdawarp which is a list with the following components:

- time: a numeric vector of length M storing the original grid;
- f0: a numeric matrix of shape $M \times N$ storing the original sample of N functions observed on a grid of size M;
- q0: a numeric matrix of the same shape as f0 storing the original SRSFs;
- fn: a numeric matrix of the same shape as f0 storing the aligned functions;
- qn: a numeric matrix of the same shape as f0 storing the aligned SRSFs;
- fmean: a numeric vector of length M storing the mean or median curve;
- mqn: a numeric vector of length M storing the mean or median SRSF;
- warping_functions: a numeric matrix of the same shape as f0 storing the estimated warping functions;
- original_variance: a numeric value storing the variance of the original sample;
- amplitude_variance: a numeric value storing the variance in amplitude of the aligned sample;
- phase_variance: a numeric value storing the variance in phase of the aligned sample;
- qun: a numeric vector of maximum length max_iter storing the relative change of the template at each iteration;
- lambda: the input parameter lambda which specifies the elasticity;
- centroid_type: the input centroid type;
- optim_method: the input optimization method;
- inverse_average_warping_function: the inverse of the mean estimated warping function;
- rsamps: TO DO.

References

Srivastava, A., Wu, W., Kurtek, S., Klassen, E., Marron, J. S., May 2011. Registration of functional data using Fisher-Rao metric, arXiv:1103.3817v2.

Tucker, J. D., Wu, W., Srivastava, A., Generative models for functional data using phase and amplitude Separation, Computational Statistics and Data Analysis (2012), 10.1016/j.csda.2012.12.001.

Examples

```
## Not run:
  out <- time_warping(simu_data$f, simu_data$time)

## End(Not run)
```

`to_hilbert_sphere` *Projects an SRVF onto the Hilbert sphere*

Description

Projects an SRVF onto the Hilbert sphere

Usage

```
to_hilbert_sphere(qfun, qnorm = get_l2_norm(qfun))
```

Arguments

`qfun`	A function that takes a numeric vector s of values in $[0, 1]$ as input and returns the values of the SRVF at s.
`qnorm`	A numeric value specifying the L^2 norm of the SRVF. Defaults to $\sqrt{\langle q, q \rangle}$.

Value

A function that takes a numeric vector s of values in $[0, 1]$ as input and returns the values of the SRVF projected onto the Hilbert sphere.

Examples

```
q <- curve2srvf(beta[, , 1, 1])
to_hilbert_sphere(q)
```

`toy_data` *Distributed Gaussian Peak Dataset*

Description

A functional dataset where the individual functions are given by a Gaussian peak with locations along the x-axis. The variables are as follows: f containing the 29 functions of 101 samples and time which describes the sampling.

Usage

```
toy_data
```

Format

`toy_data`:
A list with two components:

- `f`: A numeric matrix of shape 101×29 storing a sample of size $N = 29$ of curves evaluated on a grid of size $M = 101$.
- `time`: A numeric vector of size $M = 101$ storing the grid on which the curves `f` have been evaluated.

---

toy_warp *Aligned Distributed Gaussian Peak Dataset*

Description

A functional dataset where the individual functions are given by a Gaussian peak with locations along the x-axis. The variables are as follows: f containing the 29 functions of 101 samples and time which describes the sampling which as been aligned.

Usage

```
toy_warp
```

Format

`toy_warp`:
A list which contains the output of the `time_warping()` function applied on the data set `toy_data`.

---

v_to_curve *map shooting vector to curve at mean*

Description

map shooting vector to curve at mean

Usage

```
v_to_curve(v, mu, mode = "O", scale = 1)
```

Arguments

`v`	Either a numeric vector of a numeric matrix or a numeric array specifying the shooting vectors
`mu`	vector describing the mean
`mode`	A character string specifying whether the input curves should be considered open (`mode == "O"`) or closed (`mode == "C"`). Defaults to `"O"`.
`scale`	original scale of curve

Value

A numeric array of the same shape as the input array v storing the curves of v.

---

v_to_gam *map shooting vector to warping function at identity*

Description

map shooting vector to warping function at identity

Usage

```
v_to_gam(v)
```

Arguments

v	Either a numeric vector of a numeric matrix or a numeric array specifying the shooting vectors

Value

A numeric array of the same shape as the input array v storing the warping functions v.

---

vertFPCA *Vertical Functional Principal Component Analysis*

Description

This function calculates vertical functional principal component analysis on aligned data

Usage

```
vertFPCA(
  warp_data,
  no = 3,
  var_exp = NULL,
  id = round(length(warp_data$time)/2),
  ci = c(-1, 0, 1),
  showplot = TRUE
)
```

Arguments

warp_data	fdawarp object from time_warping of aligned data
no	number of principal components to extract
var_exp	compute no based on value percent variance explained (example: 0.95) will override no
id	point to use for f(0) (default = midpoint)
ci	geodesic standard deviations (default = c(-1,0,1))
showplot	show plots of principal directions (default = T)

Value

Returns a vfpca object containing

q_pca	srvf principal directions
f_pca	f principal directions
latent	latent values
coef	coefficients
U	eigenvectors
id	point used for f(0)

References

Tucker, J. D., Wu, W., Srivastava, A., Generative Models for Function Data using Phase and Amplitude Separation, Computational Statistics and Data Analysis (2012), 10.1016/j.csda.2012.12.001.

Examples

```
vfpca <- vertFPCA(simu_warp, no = 3)
```

---

warp_curve *Applies a warping function to a given curve*

Description

Applies a warping function to a given curve

Usage

```
warp_curve(betafun, gamfun)
```

Arguments

betafun	A function that takes a numeric vector s of values in $[0, 1]$ as input and returns the values of the underlying curve at s.
gamfun	A function that takes a numeric vector s of values in $[0, 1]$ as input and returns the values of a diffeomorphic warping function at s.

Value

A function that takes a numeric vector s of values in $[0, 1]$ as input and returns the values of the warped curve at s.

Examples

```
curv <- discrete2curve(beta[, , 1, 1])
gamf <- discrete2warping(seq(0, 1, length = 100)^2)
warp_curve(curv, gamf)
```

`warp_f_gamma` *Warp Function*

Description

This function warps function f by γ

Usage

```
warp_f_gamma(f, time, gamma, spl.int = FALSE)
```

Arguments

`f`	vector function
`time`	time
`gamma`	vector warping function
`spl.int`	use spline interpolation (default F)

Value

fnew warped function

References

Srivastava, A., Wu, W., Kurtek, S., Klassen, E., Marron, J. S., May 2011. Registration of functional data using fisher-rao metric, arXiv:1103.3817v2.

Tucker, J. D., Wu, W., Srivastava, A., Generative Models for Function Data using Phase and Amplitude Separation, Computational Statistics and Data Analysis (2012), 10.1016/j.csda.2012.12.001.

Examples

```
fnew <- warp_f_gamma(
  f = simu_data$f[, 1],
  time = simu_data$time,
  gamma = seq(0, 1, length.out = 101)
)
```

warp_q_gamma *Warp SRSF*

Description

This function warps srsf q by γ

Usage

```
warp_q_gamma(q, time, gamma, spl.int = FALSE)
```

Arguments

q	vector
time	time
gamma	vector warping function
spl.int	use spline interpolation (default F)

Value

qnew warped function

References

Srivastava, A., Wu, W., Kurtek, S., Klassen, E., Marron, J. S., May 2011. Registration of functional data using fisher-rao metric, arXiv:1103.3817v2.

Tucker, J. D., Wu, W., Srivastava, A., Generative Models for Function Data using Phase and Amplitude Separation, Computational Statistics and Data Analysis (2012), 10.1016/j.csda.2012.12.001.

Examples

```
q <- f_to_srvf(simu_data$f, simu_data$time)
qnew <- warp_q_gamma(q[, 1], simu_data$time, seq(0, 1, length.out = 101))
```

warp_srvf *Applies a warping function to a given SRVF*

Description

Applies a warping function to a given SRVF

Usage

```
warp_srvf(qfun, gamfun, betafun = NULL)
```

Arguments

qfun	A function that takes a numeric vector s of values in $[0, 1]$ as input and returns the values of the SRVF of an underlying curve at s.
gamfun	A function that takes a numeric vector s of values in $[0, 1]$ as input and returns the values of a diffeomorphic warping function at s.
betafun	A function that takes a numeric vector s of values in $[0, 1]$ as input and returns the values of the underlying curve at s. Defaults to NULL.

Value

A function that takes a numeric vector s of values in $[0, 1]$ as input and returns the values of the warped SRVF.

Examples

```
q <- curve2srvf(beta[, , 1, 1])
warp_srvf(q, get_identity_warping())
```

Index

∗ **alignment**
curve_geodesic, 16
curve_pair_align, 17
curve_to_q, 18
elastic.logistic, 22
elastic.lpcr.regression, 24
elastic.mlogistic, 25
elastic.mlpcr.regression, 26
elastic.pcr.regression, 27
elastic.prediction, 29
elastic.regression, 30
elastic_amp_change_ff, 31
elastic_change_fpca, 32
elastic_ph_change_ff, 34
f_to_srvf, 35
gam_to_h, 42
gam_to_psi, 43
gam_to_v, 43
gradient, 51
h_to_gam, 53
horizFPCA, 53
horizFPNS, 54
inv_exp_map, 56
invertGamma, 57
jointFPCA, 59
jointFPCAh, 60
kmeans_align, 61
multiple_align_functions, 64
multiple_align_multivariate, 66
multivariate_karcher_cov, 68
multivariate_karcher_mean, 69
multivariate_pca, 71
optimum.reparam, 72
pair_align_functions, 74
pair_align_image, 79
ppd, 82
predict.curve_pca, 83
predict.hfpca, 84
predict.jfpca, 85
predict.jfpcah, 85
predict.lpcr, 86
predict.mlpcr, 87
predict.pcr, 88
predict.vfpca, 88
psi_to_gam, 89
q_to_curve, 90
reparam_curve, 90
reparam_image, 92
resamplecurve, 93
sample_shapes, 95
smooth.data, 98
SqrtMean, 99
SqrtMeanInverse, 100
SqrtMedian, 100
srvf_to_f, 101
time_warping, 103
v_to_curve, 106
v_to_gam, 107
vertFPCA, 107
warp_f_gamma, 109
warp_q_gamma, 110
∗ **bayesian**
function_group_warp_bayes, 40
function_mean_bayes, 41
pair_align_functions_bayes, 75
∗ **bootstrap**
bootTB, 7
plot_curve, 81
shape_CI, 96
∗ **bounds**
bootTB, 7
shape_CI, 96
∗ **changepoint**
elastic_amp_change_ff, 31
elastic_change_fpca, 32
elastic_ph_change_ff, 34
∗ **clustering**
kmeans_align, 61

∗ **datasets**
beta, 7
growth_vel, 52
im, 55
simu_data, 97
simu_warp, 97
simu_warp_median, 98
toy_data, 105
toy_warp, 106
∗ **depth**
curve_depth, 14
elastic.depth, 20
∗ **detection**
outlier.detection, 73
∗ **diffeomorphism**
rgam, 93
∗ **distances**
calc_shape_dist, 10
curve_dist, 15
elastic.distance, 21
∗ **function**
rgam, 93
∗ **image**
pair_align_image, 79
reparam_image, 92
∗ **outlier**
outlier.detection, 73
∗ **pca tolerance bounds**
pcaTB, 80
∗ **pca**
align_fPCA, 4
gauss_model, 44
joint_gauss_model, 58
∗ **regression**
elastic.logistic, 22
elastic.lpcr.regression, 24
elastic.mlogistic, 25
elastic.mlpcr.regression, 26
elastic.pcr.regression, 27
elastic.prediction, 29
elastic.regression, 30
predict.curve_pca, 83
predict.hfpca, 84
predict.jfpca, 85
predict.jfpcah, 85
predict.lpcr, 86
predict.mlpcr, 87
predict.pcr, 88
predict.vfpca, 88
∗ **srsf alignment**
function_group_warp_bayes, 40
function_mean_bayes, 41
pair_align_functions_bayes, 75
∗ **srsf**
kmeans_align, 61
multiple_align_functions, 64
pair_align_functions, 74
ppd, 82
time_warping, 103
∗ **srvf alignment**
align_fPCA, 4
curve_depth, 14
elastic.depth, 20
elastic.distance, 21
∗ **srvf**
curve_geodesic, 16
curve_pair_align, 17
curve_to_q, 18
elastic.logistic, 22
elastic.lpcr.regression, 24
elastic.mlogistic, 25
elastic.mlpcr.regression, 26
elastic.pcr.regression, 27
elastic.prediction, 29
elastic.regression, 30
elastic_amp_change_ff, 31
elastic_change_fpca, 32
elastic_ph_change_ff, 34
f_to_srvf, 35
gam_to_h, 42
gam_to_psi, 43
gam_to_v, 43
gradient, 51
h_to_gam, 53
horizFPCA, 53
horizFPNS, 54
inv_exp_map, 56
invertGamma, 57
jointFPCA, 59
jointFPCAh, 60
multiple_align_multivariate, 66
multivariate_karcher_cov, 68
multivariate_karcher_mean, 69
multivariate_pca, 71
optimum.reparam, 72
outlier.detection, 73

predict.curve_pca, 83
predict.hfpca, 84
predict.jfpca, 85
predict.jfpcah, 85
predict.lpcr, 86
predict.mlpcr, 87
predict.pcr, 88
predict.vfpca, 88
psi_to_gam, 89
q_to_curve, 90
reparam_curve, 90
resamplecurve, 93
sample_shapes, 95
smooth.data, 98
SqrtMean, 99
SqrtMeanInverse, 100
SqrtMedian, 100
srvf_to_f, 101
v_to_curve, 106
v_to_gam, 107
vertFPCA, 107
warp_f_gamma, 109
warp_q_gamma, 110

∗ **tolerance**
bootTB, 7

∗ **warping**
rgam, 93

align_fPCA, 4
align_new, 6
align_new(), *12*

beta, 7
bootTB, 7
boxplot.ampbox *(*boxplot.fdawarp*)*, 8
boxplot.curvebox *(*boxplot.fdawarp*)*, 8
boxplot.fdawarp, 8
boxplot.fdawarp(), *9*, *12*
boxplot.phbox *(*boxplot.fdawarp*)*, 8

calc_shape_dist, 10
calc_shape_dist(), *15*
classify, 12
curve2srvf, 19
curve2srvf(), *46*
curve_boxplot, 13
curve_depth, 14
curve_dist, 15
curve_geodesic, 16
curve_pair_align, 17
curve_to_q, 18

discrete2curve, 19
discrete2warping, 20

elastic.depth, 20
elastic.distance, 21
elastic.logistic, 22
elastic.lpcr.regression, 24
elastic.mlogistic, 25
elastic.mlpcr.regression, 26
elastic.pcr.regression, 27
elastic.prediction, 29
elastic.regression, 30
elastic_amp_change_ff, 31
elastic_change_fpca, 32
elastic_ph_change_ff, 34

f_plot, 35
f_to_srvf, 35
fastpns, 36
fastPNSe2s, 37
fastPNSs2e, 38
fdasrvf, 38
fdasrvf-package *(*fdasrvf*)*, 38
foreach::foreach(), *63*
function_group_warp_bayes, 40
function_mean_bayes, 41

gam_to_h, 42
gam_to_psi, 43
gam_to_v, 43
gauss_model, 44
get_curve_centroid, 45
get_distance_matrix, 45
get_hilbert_sphere_distance, 46
get_identity_warping, 47
get_l2_distance, 48
get_l2_inner_product, 48
get_l2_norm, 49
get_shape_distance, 49
get_warping_distance, 51
gradient, 51
growth_vel, 52

h_to_gam, 53
horizFPCA, 53
horizFPNS, 54

im, 55
interparc, 55
inv_exp_map, 56
invertGamma, 57

joint_gauss_model, 58
jointFPCA, 59
jointFPCAh, 60

kmeans_align, 61

LongRunCovMatrix, 64

multiple_align_functions, 64
multiple_align_multivariate, 66
multivariate_karcher_cov, 68
multivariate_karcher_mean, *68*, 69, *71*, *94*
multivariate_karcher_mean(), *13*, *95*
multivariate_pca, 71

optimum.reparam, 72
outlier.detection, 73

pair_align_functions, 74
pair_align_functions_bayes, 75
pair_align_functions_expomap, 77
pair_align_image, 79
pcaTB, 80
plot_curve, 81
ppd, 82
predict.curve_pca, 83
predict.hfpca, 84
predict.jfpca, 85
predict.jfpcah, 85
predict.lpcr, 86
predict.mlpcr, 87
predict.pcr, 88
predict.vfpca, 88
psi_to_gam, 89

q_to_curve, 90

reparam_curve, 90
reparam_image, 92
resamplecurve, 93
rgam, 93
rotation_pca, 94

sample_shapes, 95
shape_CI, 96
simu_data, 97
simu_warp, 97
simu_warp_median, 98
smooth.data, 98
SqrtMean, 99
SqrtMeanInverse, 100
SqrtMedian, 100
srvf2curve, 102
srvf_to_f, 101
srvf_to_f(), *36*

time_warping, *44*, *53*, *54*, *58–60*, 103, *108*
time_warping(), *6*, *9*, *73*, *97*, *98*, *106*
to_hilbert_sphere, 105
toy_data, 105
toy_warp, 106

v_to_curve, 106
v_to_gam, 107
vertFPCA, 107

warp_curve, 108
warp_f_gamma, 109
warp_q_gamma, 110
warp_srvf, 110