

# Package ‘nprotreg’

May 9, 2026

**Title** Nonparametric Rotations for Sphere-Sphere Regression

**Version** 1.1.1

**Description** Fits sphere-sphere regression models by estimating locally weighted rotations. Simulation of sphere-sphere data according to non-rigid rotation models. Provides methods for bias reduction applying iterative procedures within a Newton-Raphson learning scheme. Cross-validation is exploited to select smoothing parameters. See Marco Di Marzio, Agnese Panzera & Charles C. Taylor (2018) <[doi:10.1080/01621459.2017.1421542](https://doi.org/10.1080/01621459.2017.1421542)>.

**Depends** R (>= 3.3.0)

**License** MIT + file LICENSE

**Encoding** UTF-8

**RoxygenNote** 7.2.3

**Suggests** testthat

**Imports** foreach, methods, stats

**NeedsCompilation** no

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**Repository** CRAN

**Date/Publication** 2023-09-28 08:40:02 UTC

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convert\_cartesian\_to\_spherical  
*Converts Cartesian to Spherical Coordinates.*

---

### Description

The Cartesian coordinates of points on a 3-dimensional sphere with unit radius and center at the origin are converted to the equivalent longitude and latitude coordinates, measured in radians.

### Usage

```
convert_cartesian_to_spherical(cartesian_coords)
```

### Arguments

cartesian\_coords  
 A matrix whose rows contain the Cartesian coordinates of the specified points.

### Value

A matrix of rows containing the longitude and latitude of specific points on a 3-dimensional sphere.

### See Also

<http://mathworld.wolfram.com/SphericalCoordinates.html>.  
 Other Conversion functions: [convert\\_spherical\\_to\\_cartesian\(\)](#)

### Examples

```
library(nprotreg)

# Define the Cartesian coordinates of the North and South Poles.

north_pole <- cbind(0, 0, 1)
south_pole <- cbind(0, 0, -1)
cartesian_coords <- rbind(north_pole, south_pole)

# Get the corresponding Spherical coordinates.

spherical_coords <- convert_cartesian_to_spherical(cartesian_coords)
```

---

`convert_spherical_to_cartesian`*Converts Spherical to Cartesian Coordinates.*

---

### Description

The longitude and latitude coordinates of points on a 3-dimensional sphere with unit radius and center at the origin are converted to the equivalent Cartesian coordinates.

### Usage

```
convert_spherical_to_cartesian(spherical_coords)
```

### Arguments

`spherical_coords`

A matrix of rows containing the longitude and latitude, measured in radians, of specific points on a 3-dimensional sphere.

### Value

A matrix whose rows contain the Cartesian coordinates of the specified points.

### See Also

<http://mathworld.wolfram.com/SphericalCoordinates.html>.

Other Conversion functions: [convert\\_cartesian\\_to\\_spherical\(\)](#)

### Examples

```
library(nprotreg)

# Define the Spherical coordinates of the North and South Poles.

north_pole <- cbind(0, pi / 2)
south_pole <- cbind(0, - pi / 2)
spherical_coords <- rbind(north_pole, south_pole)

# Get the corresponding Cartesian coordinates.

cartesian_coords <- convert_spherical_to_cartesian(spherical_coords)
```

---

cross\_validate\_concentration

*Cross-validates The Concentration Parameter In A 3D Spherical Regression.*

---

### Description

Returns a cross-validated value for the concentration parameter in a 3D regression, relating specific explanatory points to response ones, given a weighting scheme for the observed data set. This function supports the method for sphere-sphere regression proposed by Di Marzio et al. (2018).

### Usage

```
cross_validate_concentration(
    concentration_upper_bound = 10,
    explanatory_points,
    response_points,
    weights_generator = weight_explanatory_points,
    number_of_expansion_terms = 1,
    number_of_iterations = 1,
    allow_reflections = FALSE
)
```

### Arguments

**concentration\_upper\_bound**  
A scalar numeric value representing the upper end-point of the interval to be searched for the required minimizer. Defaults to 10.

**explanatory\_points**  
An  $m$ -by-3 matrix whose rows contain the Cartesian coordinates of the explanatory points used to calculate the regression estimators.

**response\_points**  
An  $m$ -by-3 matrix whose rows contain the Cartesian coordinates of the response points corresponding to the explanatory points.

**weights\_generator**  
A function that, given a matrix of  $n$  evaluation points, returns an  $m$ -by- $n$  matrix whose  $j$ -th column contains the weights assigned to the explanatory points while analyzing the  $j$ -th evaluation point. Defaults to `weight_explanatory_points`.

**number\_of\_expansion\_terms**  
The number of terms to be included in the expansion of the matrix exponential applied while approximating a local rotation matrix. Must be 1 or 2. Defaults to 1.

**number\_of\_iterations**  
The number of rotation fitting steps to be executed. At each step, the points estimated during the previous step are exploited as the current explanatory points. Defaults to 1.

allow\_reflections

A logical scalar value. If set to TRUE signals that reflections are allowed. Defaults to FALSE. It is ignored if number\_of\_expansion\_terms is 2.

### Details

Function weights\_generator must be prototyped as having the following three arguments:

evaluation\_points a matrix whose  $n$  rows are the Cartesian coordinates of given evaluation points.

explanatory\_points a matrix whose  $m$  rows are the Cartesian coordinates of given explanatory points.

concentration A non negative scalar whose reciprocal value is proportional to the bandwidth applied while estimating a spherical regression model.

It is also expected that weights\_generator will return a non NULL numerical  $m$ -by- $n$  matrix whose  $j$ -th column contains the weights assigned to the explanatory points while analyzing the  $j$ -th evaluation point.

### Value

A list having two components, concentration, a scalar, numeric value representing the cross-validated concentration for the specified 3D regression, and objective, the value of the cross-validating objective function at argument concentration.

### References

Marco Di Marzio, Agnese Panzera & Charles C. Taylor (2018) Nonparametric rotations for sphere-sphere regression, Journal of the American Statistical Association, <doi:10.1080/01621459.2017.1421542>.

### See Also

Other Regression functions: [fit\\_regression\(\)](#), [get\\_equally\\_spaced\\_points\(\)](#), [get\\_skew\\_symmetric\\_matrix\(\)](#), [simulate\\_regression\(\)](#), [simulate\\_rigid\\_regression\(\)](#), [weight\\_explanatory\\_points\(\)](#)

### Examples

```
library(nprotreg)

# Define a matrix of explanatory points.

number_of_explanatory_points <- 50

explanatory_points <- get_equally_spaced_points(
  number_of_explanatory_points)

# Define a matrix of response points by simulation.

local_rotation_composer <- function(point) {
  independent_components <- (1 / 2) *
    c(exp(2.0 * point[3]), - exp(2.0 * point[2]), exp(2.0 * point[1]))
```

```
}  
  
local_error_sampler <- function(point) {  
  rnorm(3)  
}  
  
response_points <- simulate_regression(explanatory_points,  
                                     local_rotation_composer,  
                                     local_error_sampler)  
  
# Define an upper bound for concentration.  
  
concentration_upper_bound <- 1  
  
# Use default weights generator.  
  
weights_generator <- weight_explanatory_points  
  
# Cross-validate concentration parameter.  
  
cv_info <- cross_validate_concentration(  
  concentration_upper_bound,  
  explanatory_points,  
  response_points,  
  weights_generator,  
  number_of_expansion_terms = 1,  
  number_of_iterations = 2,  
  allow_reflections = FALSE  
)  
  
# Get the cross-validated concentration value.  
  
cat("cross-validated concentration value: \n")  
print(cv_info$concentration)
```

---

expm

*Computes the Exponential of a 3D Skew Symmetric Matrix.*

---

### **Description**

The exponential of a skew-symmetric matrix is computed by means of the Rodrigues' formula.

### **Usage**

```
expm(skew_symmetric_matrix)
```

### **Arguments**

skew\_symmetric\_matrix  
A 3-by-3 skew-symmetric matrix.

**Value**

A 3-by-3 rotation matrix representing the exponential of the specified skew-symmetric matrix.

---

fit_regression	<i>Fits a 3D Spherical Regression.</i>
----------------	--

---

**Description**

Returns 3D spherical points obtained by locally rotating the specified evaluation points, given an approximated model for local rotations and a weighting scheme for the observed data set. This function implements the method for sphere-sphere regression proposed by Di Marzio et al. (2018).

**Usage**

```
fit_regression(
  evaluation_points,
  explanatory_points,
  response_points,
  concentration,
  weights_generator = weight_explanatory_points,
  number_of_expansion_terms = 1,
  number_of_iterations = 1,
  allow_reflections = FALSE
)
```

**Arguments**

**evaluation\_points** An  $n$ -by-3 matrix whose rows contain the Cartesian coordinates of the points at which the regression will be estimated.

**explanatory\_points** An  $m$ -by-3 matrix whose rows contain the Cartesian coordinates of the explanatory points used to calculate the regression estimators.

**response\_points** An  $m$ -by-3 matrix whose rows contain the Cartesian coordinates of the response points corresponding to the explanatory points.

**concentration** A non negative scalar whose reciprocal value is proportional to the bandwidth applied while estimating a spherical regression model.

**weights\_generator** A function that, given a matrix of  $n$  evaluation points, returns an  $m$ -by- $n$  matrix whose  $j$ -th column contains the weights assigned to the explanatory points while analyzing the  $j$ -th evaluation point. Defaults to [weight\\_explanatory\\_points](#).

**number\_of\_expansion\_terms** The number of terms to be included in the expansion of the matrix exponential applied while approximating a local rotation matrix. Must be 1 or 2. Defaults to 1.

**number\_of\_iterations**

The number of rotation fitting steps to be executed. At each step, the points estimated during the previous step are exploited as the current explanatory points. Defaults to 1.

**allow\_reflections**

A logical scalar value. If set to TRUE signals that reflections are allowed. Defaults to FALSE. It is ignored if `number_of_expansion_terms` is 2.

**Details**

Function `weights_generator` must be prototyped as having the following three arguments:

`evaluation_points` a matrix whose  $n$  rows are the Cartesian coordinates of given evaluation points.

`explanatory_points` a matrix whose  $m$  rows are the Cartesian coordinates of given explanatory points.

`concentration` A non negative scalar whose reciprocal value is proportional to the bandwidth applied while estimating a spherical regression model.

It is also expected that `weights_generator` will return a non NULL numerical  $m$ -by- $n$  matrix whose  $j$ -th column contains the weights assigned to the explanatory points while analyzing the  $j$ -th evaluation point.

Function `fit_regression` supports parallel execution. To setup parallelization, you can exploit the `doParallel` package. Otherwise, `fit_regression` will be executed sequentially and, when called the first time, you will receive the following

```
## Warning: executing %dopar% sequentially: no parallel backend registered
```

This is completely safe and by design.

**Value**

A `number_of_iterations`-length vector of lists, with the  $s$ -th list having two components, `fitted_response_points`, an  $n$ -by-3 matrix whose rows contain the Cartesian coordinates of the fitted points at iteration  $s$ , and `explanatory_points`, an  $m$ -by-3 matrix whose rows contain the Cartesian coordinates of the points exploited as explanatory at iteration  $s$ .

**References**

Marco Di Marzio, Agnese Panzera & Charles C. Taylor (2018) Nonparametric rotations for sphere-sphere regression, *Journal of the American Statistical Association*, <doi:10.1080/01621459.2017.1421542>.

**See Also**

Other Regression functions: [cross\\_validate\\_concentration\(\)](#), [get\\_equally\\_spaced\\_points\(\)](#), [get\\_skew\\_symmetric\\_matrix\(\)](#), [simulate\\_regression\(\)](#), [simulate\\_rigid\\_regression\(\)](#), [weight\\_explanatory\\_po](#).

**Examples**

```
library(nproreg)

# Create 100 equally spaced design points on the sphere.

number_of_explanatory_points <- 100

explanatory_points <- get_equally_spaced_points(
  number_of_explanatory_points
)

# Define the regression model, where the rotation for a given "point"
# is obtained from the exponential of a skew-symmetric matrix with the
# following components.

local_rotation_composer <- function(point) {
  independent_components <- (1 / 8) *
    c(exp(2.0 * point[3]), - exp(2.0 * point[2]), exp(2.0 * point[1]))
}

# Define an error term given by a small rotation, similarly defined
# from a skew-symmetric matrix with random entries.

local_error_sampler <- function(point) {
  rnorm(3, sd = .01)
}

# Generate the matrix of responses, using the regression model
# and the error model.

response_points <- simulate_regression(
  explanatory_points,
  local_rotation_composer,
  local_error_sampler
)

# Create some "test data" for which the response will be predicted.

evaluation_points <- rbind(
  cbind(.5, 0, .8660254),
  cbind(-.5, 0, .8660254),
  cbind(1, 0, 0),
  cbind(0, 1, 0),
  cbind(-1, 0, 0),
  cbind(0, -1, 0),
  cbind(.5, 0, -.8660254),
  cbind(-.5, 0, -.8660254)
)

# Define a weight function for nonparametric fit.

weights_generator <- weight_explanatory_points
```

```

# Set the concentration parameter.

concentration <- 5

# Or obtain this by cross-validation: see
# the `cross_validate_concentration` function.

# Fit regression.

fitted_model <- fit_regression(
  evaluation_points,
  explanatory_points,
  response_points,
  concentration,
  weights_generator,
  number_of_expansion_terms = 1,
  number_of_iterations = 2
)

# Extract the point corresponding to the
# second evaluation point fitted at
# the first iteration.

cat("Point fitted at iteration 1 corresponding to the second evaluation point: \n")
cat(fitted_model[[1]]$fitted_response_points[2, ], "\n")

## Not run:
# Create some plots to view the results.

# 3D plot.

library(rgl)

plot3d(
  explanatory_points,
  type = "n",
  xlab = "x",
  ylab = "y",
  zlab = "z",
  box = TRUE,
  axes = TRUE
)
spheres3d(0, 0, 0, radius = 1, lit = FALSE, color = "white")
spheres3d(0, 0, 0, radius = 1.01, lit = FALSE, color = "black", front = "culled")
text3d(c(0, 0, 1), text = "N", adj = 0)

l1 <- 10
vv1 <- (l1 - (0:(l1))) / l1
vv2 <- 1 - vv1
plot3d(explanatory_points, add = TRUE, col = 2)
for (i in 1:dim(explanatory_points)[1]) {
  m <- outer(vv1, explanatory_points[i,], "*") +

```

```

    outer(vv2, response_points[i,], "*")
    m <- m / sqrt(apply(m ^ 2, 1, sum))
    lines3d(m, col = 3)
  }

plot3d(evaluation_points, add = TRUE, col = 4)

for (i in 1:dim(evaluation_points)[1]) {
  m <- outer(vv1, evaluation_points[i,], "*") +
    outer(vv2, fitted_model[[1]]$fitted_response_points[i,], "*")
  m <- m / sqrt(apply(m ^ 2, 1, sum))
  lines3d(m, col = 1)
}

# 2D plot.

explanatory_spherical_coords <- convert_cartesian_to_spherical(explanatory_points)
response_spherical_coords <- convert_cartesian_to_spherical(response_points)

plot(
  x = explanatory_spherical_coords[, 1],
  y = explanatory_spherical_coords[, 2],
  pch = 20,
  cex = .7,
  col = 2,
  xlab = "longitude",
  ylab = "latitude"
)

for (i in 1:dim(explanatory_spherical_coords)[1]) {
  column <- 1
  if ((explanatory_spherical_coords[i, 1] - response_spherical_coords[i, 1]) ^ 2 +
      (explanatory_spherical_coords[i, 2] - response_spherical_coords[i, 2]) ^ 2 > 4)
    column <- "grey"
  lines(
    c(explanatory_spherical_coords[i, 1], response_spherical_coords[i, 1]),
    c(explanatory_spherical_coords[i, 2], response_spherical_coords[i, 2]),
    col = column
  )
}

evaluation_spherical_coords <- convert_cartesian_to_spherical(
  evaluation_points
)

fitted_response_spherical_coords <- convert_cartesian_to_spherical(
  fitted_model[[1]]$fitted_response_points
)

points(
  x = evaluation_spherical_coords[, 1],
  y = evaluation_spherical_coords[, 2],
  pch = 20,

```

```

    cex = .7,
    col = 4
  )

  for (i in 1:dim(evaluation_spherical_coords)[1]) {
    column <- 3
    if ((evaluation_spherical_coords[i, 1] - fitted_response_spherical_coords[i, 1]) ^ 2 +
        (evaluation_spherical_coords[i, 2] - fitted_response_spherical_coords[i, 2]) ^ 2 > 4)
      column <- "grey"
    lines(
      c(evaluation_spherical_coords[i, 1], fitted_response_spherical_coords[i, 1]),
      c(evaluation_spherical_coords[i, 2], fitted_response_spherical_coords[i, 2]),
      col = column
    )
  }

  ## End(Not run)

```

---

```
get_equally_spaced_points
```

*Generates Equally Spaced Points On A 3D Sphere.*

---

### Description

Generates points approximately equally spaced on a 3D sphere.

### Usage

```
get_equally_spaced_points(number_of_points)
```

### Arguments

number\_of\_points

A scalar, positive integer representing the number of points to get.

### Value

A number\_of\_points-by-3 matrix whose rows contain the Cartesian coordinates of the equally spaced points.

### See Also

Other Regression functions: [cross\\_validate\\_concentration\(\)](#), [fit\\_regression\(\)](#), [get\\_skew\\_symmetric\\_matrix\(\)](#), [simulate\\_regression\(\)](#), [simulate\\_rigid\\_regression\(\)](#), [weight\\_explanatory\\_points\(\)](#)

**Examples**

```
library(nprotreg)

# Define the number of points to get.

number_of_points <- 5

# Get the Cartesian coordinates of the equally spaced points.

equally_spaced_points <- get_equally_spaced_points(number_of_points)
```

---

```
get_skew_symmetric_matrix
Gets a 3-by-3 Skew Symmetric Matrix.
```

---

**Description**

Returns the 3-by-3 skew symmetric matrix having the specified independent components.

**Usage**

```
get_skew_symmetric_matrix(independent_components)
```

**Arguments**

`independent_components`  
A vector containing the independent components of the matrix to get.

**Details**

Given a vector of components, say  $[x, y, z]$ , this function will return matrix

$$\begin{matrix} 0 & -z & y \\ z & 0 & -x \\ -y & x & 0 \end{matrix}$$

**Value**

The 3-by-3 skew symmetric matrix corresponding to the specified independent components.

**See Also**

[https://en.wikipedia.org/wiki/Skew-symmetric\\_matrix](https://en.wikipedia.org/wiki/Skew-symmetric_matrix).

Other Regression functions: [cross\\_validate\\_concentration\(\)](#), [fit\\_regression\(\)](#), [get\\_equally\\_spaced\\_points\(\)](#), [simulate\\_regression\(\)](#), [simulate\\_rigid\\_regression\(\)](#), [weight\\_explanatory\\_points\(\)](#)

**Examples**

```
library(nprotreg)

# Define a vector of independent components.

independent_components <- cbind(1, 2, 3)

# Get the corresponding 3-by-3 skew symmetric matrix.

m <- get_skew_symmetric_matrix(independent_components)
```

---

logm	<i>Computes the Logarithm of a 3D Rotation Matrix.</i>
------	--

---

**Description**

Computes the Logarithm of a 3D Rotation Matrix.

**Usage**

```
logm(rotation_matrix)
```

**Arguments**

```
rotation_matrix  
A 3-by-3 rotation matrix.
```

**Value**

A 3-by-3 skew-symmetric matrix representing the logarithm of the specified rotation matrix.

---

nprotreg	<i>nprotreg: Nonparametric Rotations for Sphere-Sphere Regression.</i>
----------	--

---

**Description**

The nprotreg package provides several categories of functions.

### Regression functions

Regression functions provide support for simulating and fitting 3-dimensional spherical regression models.

- [cross\\_validate\\_concentration](#)
- [fit\\_regression](#)
- [get\\_equally\\_spaced\\_points](#)
- [get\\_skew\\_symmetric\\_matrix](#)
- [simulate\\_regression](#)
- [simulate\\_rigid\\_regression](#)
- [weight\\_explanatory\\_points](#)

### Conversion functions

Conversion functions transform coordinates of points on a 3-dimensional sphere with unit radius and center at the origin.

- [convert\\_cartesian\\_to\\_spherical](#)
- [convert\\_spherical\\_to\\_cartesian](#)

---

`simulate_regression`     *Simulates a 3D Spherical Regression.*

---

### Description

Returns the response points corresponding to the specified explanatory points, given a model for local rotations and an error term sampler.

### Usage

```
simulate_regression(
    explanatory_points,
    local_rotation_composer,
    local_error_sampler
)
```

### Arguments

`explanatory_points`

An  $m$ -by-3 matrix whose rows contain the Cartesian coordinates of the points at which the regression will be simulated.

`local_rotation_composer`

A function that returns a 3-length numeric vector representing the independent components of a skew symmetric matrix local to an explanatory point, given its Cartesian coordinates.

`local_error_sampler`

A function that returns a 3-length numeric vector representing a sampled error term local to an explanatory point, given its Cartesian coordinates.

## Details

Let  $E$  be the  $m$ -by-3 matrix of explanatory points. This function will return an  $m$ -by-3 matrix whose  $i$ -th row is obtained by transposition of the following expression:

$$\exp(\Phi(\epsilon(x)))\exp(\Phi(s(x)))x$$

where  $x$  is the transpose of the  $i$ -th row of  $E$ . Terms  $\epsilon(x)$  and  $s(x)$  are obtained by evaluating at  $x$  functions `local_error_sampler` and `local_rotation_composer`, respectively, while matrix  $\Phi(c)$ , for a 3-length numeric vector  $c$ , is the skew symmetric matrix having its independent components represented by the entries of  $c$  (for a thorough discussion, see function [get\\_skew\\_symmetric\\_matrix](#)).

Functions `local_error_sampler` and `local_rotation_composer` must be prototyped as having one argument, `point`, representing the Cartesian coordinates of a point on a 3D sphere, and returning a non NULL numerical object having length equal to 3.

## Value

An  $m$ -by-3 matrix whose rows contain the Cartesian coordinates of the response points corresponding to the explanatory points.

## See Also

Other Regression functions: [cross\\_validate\\_concentration\(\)](#), [fit\\_regression\(\)](#), [get\\_equally\\_spaced\\_points\(\)](#), [get\\_skew\\_symmetric\\_matrix\(\)](#), [simulate\\_rigid\\_regression\(\)](#), [weight\\_explanatory\\_points\(\)](#)

## Examples

```
library(nprotreg)

# Define a matrix of explanatory points.

explanatory_points <- rbind(
  cbind(.5, 0, .8660254),
  cbind(-.5, 0, .8660254),
  cbind(1, 0, 0),
  cbind(0, 1, 0),
  cbind(-1, 0, 0),
  cbind(0, -1, 0),
  cbind(.5, 0, -.8660254),
  cbind(-.5, 0, -.8660254)
)

# Define a local rotation composer.

local_rotation_composer <- function(point) {
  independent_components <- (1 / 2) *
    c(exp(2.0 * point[3]), - exp(2.0 * point[2]), exp(2.0 * point[1]))
}

# Define a local error sampler.
```

```

local_error_sampler <- function(point) {
  rnorm(3)
}

# Get the corresponding 8-by-3 matrix of response points.
# Rows corresponds to explanatory points,
# columns to Cartesian coordinates.

response_points <- simulate_regression(explanatory_points,
                                       local_rotation_composer,
                                       local_error_sampler)

# Get the response point corresponding to the second
# explanatory point.

cat("Response point corresponding to the second explanatory point: \n")
cat(response_points[2, ], "\n")

```

---

```
simulate_rigid_regression
```

*Simulates a Rigid 3D Spherical Regression.*

---

## Description

Returns the response points corresponding to the specified explanatory points, given a rigid rotation model and an error term sampler.

## Usage

```

simulate_rigid_regression(
  explanatory_points,
  rotation_matrix,
  local_error_sampler
)

```

## Arguments

`explanatory_points`

An  $m$ -by-3 matrix whose rows contain the Cartesian coordinates of the points at which the regression will be simulated.

`rotation_matrix`

A 3-by-3 rotation matrix.

`local_error_sampler`

A function that returns a 3-length numeric vector representing a sampled error term local to an explanatory point, given its Cartesian coordinates.

**Details**

Let  $E$  be the  $m$ -by-3 matrix of explanatory points. This function will return an  $m$ -by-3 matrix whose  $i$ -th row is obtained by transposition of the following expression:

$$\text{exp}(\Phi(\epsilon(x)))Rx$$

where  $x$  is the transpose of the  $i$ -th row of  $E$  and  $R$  is `rotation_matrix`. Term  $\epsilon(x)$  is obtained by evaluating at  $x$  function `local_error_sampler`, while matrix  $\Phi(c)$ , for a 3-length numeric vector  $c$ , is the skew symmetric matrix having its independent components represented by the entries of  $c$  (for a thorough discussion, see function `get_skew_symmetric_matrix`).

Function `local_error_sampler` must be prototyped as having one argument, `point`, representing the Cartesian coordinates of a point on a 3D sphere, and returning a non NULL numerical object having length equal to 3.

**Value**

An  $m$ -by-3 matrix whose rows contain the Cartesian coordinates of the response points corresponding to the explanatory points.

**See Also**

Other Regression functions: `cross_validate_concentration()`, `fit_regression()`, `get_equally_spaced_points()`, `get_skew_symmetric_matrix()`, `simulate_regression()`, `weight_explanatory_points()`

**Examples**

```
library(nprotreg)

# Define a matrix of explanatory points.

explanatory_points <- rbind(
  cbind(.5, 0, .8660254),
  cbind(-.5, 0, .8660254),
  cbind(1, 0, 0),
  cbind(0, 1, 0),
  cbind(-1, 0, 0),
  cbind(0, -1, 0),
  cbind(.5, 0, -.8660254),
  cbind(-.5, 0, -.8660254)
)

# Define a rotation matrix.

rotation_matrix <- rbind(
  cbind(-0.69492055764131177575, 0.71352099052778772403, 0.08929285886191218324),
  cbind(-0.19200697279199935297, -0.30378504433947051133, 0.93319235382364695841),
  cbind(0.69297816774177023458, 0.63134969938371787723, 0.34810747783026463331)
)

# Define a local error sampler.
```

```

local_error_sampler <- function(point) {
  rnorm(3)
}

# Get the corresponding 8-by-3 matrix of response points.
# Rows corresponds to explanatory points,
# columns to Cartesian coordinates.

response_points <- simulate_rigid_regression(explanatory_points,
                                           rotation_matrix,
                                           local_error_sampler)

# Get the response point corresponding to the second
# explanatory point.

cat("Response point corresponding to the second explanatory point: \n")
cat(response_points[2, ], "\n")

```

---

weight\_explanatory\_points

*Weights the Specified Explanatory Points in a 3D Spherical Regression.*

---

## Description

Returns the weights assigned to the specified explanatory points for each evaluation point under study, given a concentration parameter.

## Usage

```
weight_explanatory_points(evaluation_points, explanatory_points, concentration)
```

## Arguments

evaluation\_points

An  $n$ -by-3 matrix whose rows contain the Cartesian coordinates of the points on which the regression will be estimated.

explanatory\_points

An  $m$ -by-3 matrix whose rows contain the Cartesian coordinates of the explanatory points used to calculate the regression estimators.

concentration

A non negative scalar whose reciprocal value is proportional to the bandwidth applied while estimating a spherical regression model.

**Details**

Let  $X$  be the  $m$ -by-3 matrix of explanatory points, and  $E$  the  $n$ -by-3 matrix of evaluation points, and  $\kappa$  the concentration parameter. This function will return an  $m$ -by- $n$  matrix whose  $(i, j)$  entry is defined as follows:

$$\exp(\kappa(s(i, j) - 1))$$

where  $s(i, j)$  is the scalar product of the  $i$ -th row of  $X$  and the  $j$ -th row of  $E$ .

**Value**

An  $m$ -by- $n$  matrix whose  $j$ -th column contains the weights assigned to the explanatory points while analyzing the  $j$ -th evaluation point.

**See Also**

Other Regression functions: [cross\\_validate\\_concentration\(\)](#), [fit\\_regression\(\)](#), [get\\_equally\\_spaced\\_points\(\)](#), [get\\_skew\\_symmetric\\_matrix\(\)](#), [simulate\\_regression\(\)](#), [simulate\\_rigid\\_regression\(\)](#)

**Examples**

```
library(nproreg)

# Define a matrix of evaluation points.

north_pole <- cbind(0, 0, 1)
south_pole <- cbind(0, 0, -1)
evaluation_points <- rbind(north_pole, south_pole)

# Define a matrix of explanatory points

explanatory_points <- rbind(
  cbind(.5, 0, .8660254),
  cbind(-.5, 0, .8660254),
  cbind(1, 0, 0),
  cbind(0, 1, 0),
  cbind(-1, 0, 0),
  cbind(0, -1, 0),
  cbind(.5, 0, -.8660254),
  cbind(-.5, 0, -.8660254)
)

# Define a value for the concentration parameter.

concentration <- 1.0

# Get the corresponding 8-by-2 matrix of weights.
# Columns corresponds to evaluation points,
# rows to explanatory ones.

weights <- weight_explanatory_points(evaluation_points,
```

```
                                explanatory_points,  
                                concentration)  
  
# Get the weights assigned to the explanatory points  
# while analyzing the second evaluation point.  
  
cat("Weights assigned while analyzing the second evaluation point: \n")  
cat(weights[, 2], "\n")
```

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