

Package ‘Rsomoclu’

January 20, 2025

Version 1.7.6

Date 2022-09-20

Title Somoclu

Imports kohonen

Description Somoclu is a massively parallel implementation of self-organizing maps. It exploits multicore CPUs and it can be accelerated by CUDA. The topology of the map can be planar or toroid and the grid of neurons can be rectangular or hexagonal . Details refer to (Peter Wittek, et al (2017)) <[doi:10.18637/jss.v078.i09](https://doi.org/10.18637/jss.v078.i09)>.

URL <https://peterwittek.github.io/somoclu/>

BugReports <https://github.com/peterwittek/somoclu/issues>

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Author Peter Wittek [aut],
Shichao Gao [cre]

Maintainer Shichao Gao <xgdgsc@gmail.com>

NeedsCompilation yes

Repository CRAN

SystemRequirements C++11

LinkingTo Rcpp

LazyData true

Date/Publication 2022-09-26 07:10:07 UTC

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<code>rgbs</code>	<i>tiny rgbs data</i>
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Description

`tiny rgbs` data for testing

Usage

```
rgbs
```

Format

matrix in plain text form

<code>Rsomoclu.kohonen</code>	<i>convert Somoclu train result to kohonen class for plotting</i>
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Description

A function call to convert Somoclu train result to kohonen class for plotting.

Usage

```
Rsomoclu.kohonen(input_data, result, n.hood = NULL, toroidal = FALSE)
```

Arguments

<code>input_data</code>	input data, matrix format
<code>result</code>	The result returned by <code>Rsomoclu.train</code>
<code>n.hood</code>	Same as in <code>koohonen</code> , the shape of the neighbourhood, either "circular" or "square". The latter is the default for rectangular maps, the former for hexagonal maps.
<code>toroidal</code>	if TRUE, the edges of the map are joined. Note that in a hexagonal toroidal map, the number of rows must be even.

Value

An object of class `kohonen` for plotting.

See Also

<https://www.r-bloggers.com/2014/02/self-organising-maps-for-customer-segmentation-using-r/>

Examples

```

library('Rsomoclu')
library('kohonen')
data("rgbs", package = "Rsomoclu")
input_data <- rgbs
input_data <- data.matrix(input_data)
nSomX <- 20
nSomY <- 20
nEpoch <- 10
radius0 <- 0
radiusN <- 0
radiusCooling <- "linear"
scale0 <- 0
scaleN <- 0.01
scaleCooling <- "linear"
kernelType <- 0
mapType <- "planar"
gridType <- "rectangular"
compactSupport <- FALSE
codebook <- NULL
neighborhood <- "gaussian"
stdCoeff <- 0.5
res <- Rsomoclu.train(input_data, nEpoch, nSomX, nSomY,
                      radius0, radiusN,
                      radiusCooling, scale0, scaleN,
                      scaleCooling,
                      kernelType, mapType, gridType, compactSupport,
                      neighborhood, stdCoeff, codebook)
## Convert to kohonen object for plotting
sommap = Rsomoclu.kohonen(input_data, res)
## Show 'codebook'
plot(sommap, type="codes", main = "Codes")
## Show 'component planes'
plot(sommap, type = "property", property = sommap$codes[[1]][,1],
      main = colnames(sommap$codes)[1])
## Show 'U-Matrix'
plot(sommap, type="dist.neighbours")

```

Rsomoclu.train

Train function for Somoclu

Description

A function call to Somoclu to train the Self Organizing Map.

Usage

```
Rsomoclu.train(input_data, nEpoch, nSomX, nSomY,
               radius0, radiusN,
```

```
radiusCooling, scale0, scaleN,
scaleCooling,
kernelType, mapType, gridType, compactSupport,
neighborhood, stdCoeff, codebook, vectDistance)
```

Arguments

<code>input_data</code>	input data, matrix format
<code>nEpoch</code>	Maximum number of epochs
<code>nSomX</code>	Number of columns in map (size of SOM in direction x)
<code>nSomY</code>	Number of rows in map (size of SOM in direction y)
<code>radius0</code>	Start radius (default: half of the map in direction min(x,y))
<code>radiusN</code>	End radius (default: 1)
<code>radiusCooling</code>	Radius cooling strategy: linear or exponential (default: linear)
<code>scale0</code>	Starting learning rate (default: 1.0)
<code>scaleN</code>	Finishing learning rate (default: 0.01)
<code>scaleCooling</code>	Learning rate cooling strategy: linear or exponential (default: linear)
<code>kernelType</code>	Kernel type 0: Dense CPU 1: Dense GPU 2: Sparse CPU (default: 0)
<code>mapType</code>	Map type: planar or toroid (default: "planar")
<code>gridType</code>	Grid type: square or hexagonal (default: "rectangular")
<code>compactSupport</code>	Compact support for Gaussian neighborhood, (default:TRUE)
<code>neighborhood</code>	Neighborhood function: gaussian or bubble (default: "gaussian")
<code>codebook</code>	initial codebook, (default:NULL)
<code>stdCoeff</code>	The coefficient in the Gaussian neighborhood function $\exp(-\ x-y\ ^2/(2*(coeff*radius)^2))$, (default:0.5)
<code>vectDistance</code>	the vector distance function "norm-3", "norm-6", "norm-2"(same as default) "norm-inf", is supported with kerneltype = 0 only , (default:euclidean)

Value

a list including elements	
<code>codebook</code>	the codebook
<code>globalBmus</code>	global Best Matching Unit matrix
<code>uMatrix</code>	uMatrix

Author(s)

Peter Wittek, Shichao Gao

References

Peter Wittek, Shi Chao Gao, Ik Soo Lim, Li Zhao (2017). Somoclu: An Efficient Parallel Library for Self-Organizing Maps. Journal of Statistical Software, 78(9), 1-21. doi:10.18637/jss.v078.i09.

Examples

```
library('Rsomoclu')
data("rgbs", package = "Rsomoclu")
input_data <- rgbs
input_data <- data.matrix(input_data)
nSomX <- 10
nSomY <- 10
nEpoch <- 10
radius0 <- 0
radiusN <- 0
radiusCooling <- "linear"
scale0 <- 0
scaleN <- 0.01
scaleCooling <- "linear"
kernelType <- 0
mapType <- "planar"
gridType <- "rectangular"
compactSupport <- FALSE
codebook <- NULL
neighborhood <- "gaussian"
stdCoeff <- 0.5
vectDistance <- "euclidean"
res <- Rsomoclu.train(input_data, nEpoch, nSomX, nSomY,
                      radius0, radiusN,
                      radiusCooling, scale0, scaleN,
                      scaleCooling,
                      kernelType, mapType, gridType, compactSupport, neighborhood,
                      stdCoeff, codebook, vectDistance)
res$codebook
res$globalBmus
res$uMatrix
library('kohonen')
sommap = Rsomoclu.kohonen(input_data, res)
```

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* **datasets**

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