

# Package ‘nnNorm’

March 26, 2024

**Version** 2.66.0

**Date** 2010-04-13

**Title** Spatial and intensity based normalization of cDNA microarray data based on robust neural nets

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**Depends** R(>= 2.2.0), marray

**Imports** graphics, grDevices, marray, methods, nnet, stats

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**Description** This package allows to detect and correct for spatial and intensity biases with two-channel microarray data.  
The normalization method implemented in this package is based on robust neural networks fitting.

**biocViews** Microarray, TwoChannel, Preprocessing

**License** LGPL

**URL** <http://bioinformaticsprb.med.wayne.edu/tarca/>

**git\_url** <https://git.bioconductor.org/packages/nnNorm>

**git\_branch** RELEASE\_3\_18

**git\_last\_commit** cdf8497

**git\_last\_commit\_date** 2023-10-24

**Repository** Bioconductor 3.18

**Date/Publication** 2024-03-25

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|----------|---|
| compNorm | <i>Compares the distribution of several vectors at a time using either boxplots or density curves</i> |
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### Description

This function was conceived to easily compare several normalization methods in terms of variability of log-ratios,  $M$ . Basically it produces two plots: The first is a the density plot of the several matrices passed as arguments, while the second is a box plot. Median of absolute deviations for each method is printed on screen.

### Usage

```
compNorm(x,...,bw="AUTO",xlim=c(-3,3),titles="AUTO",type="d")
```

### Arguments

|                     |   |
|---------------------|---|
| <code>x</code>      | A vector of numerical values, e.q. the $M$ values of a data set: <code>as.vector(maM(swirl))</code> . |
| <code>...</code>    | An undefined number of objects similar with <code>x</code> .  |
| <code>bw</code>     | Band width required to compute the density distribution. "AUTO" will adjust bw to a suitable value.   |
| <code>xlim</code>   | The range for abscissa of the density plots.  |
| <code>titles</code> | Names to be displayed the charts legend. "AUTO" will use the matrices names passed as arguments. .    |
| <code>type</code>   | If set to "d", density plot will be shown; if set to "b" box plot will be shown.                      |

### Details

This function is used to compare the normalized log ratios  $M$  obtained with several normalization methods.

### Value

NULL, this function only displays charts and prints on the screen some statistics.

### Author(s)

Tarca, A.L.

### References

A. L. Tarca, J. E. K. Cooke, and J. Mackay. Robust neural networks approach for spatial and intensity dependent normalization of cDNA data. Bioinformatics. 2004,submitted.

**See Also**[maNormNN](#)**Examples**

```
# Normalize swirl data with two methods
data(swirl)
swirlLNN<-maNormNN(swirl[,1])
swirlLoess<-maNormMain(swirl[,1])
nms<-c("None", "Loess", "NNets")
#compare distributions: density plot
compNorm(as.vector(maM(swirl[,1])),as.vector(maM(swirlLoess)),as.vector(maM(swirlLNN)),xlim=c(- 2,2),bw="AUTO",
#compare distributions: box plot
compNorm(as.vector(maM(swirl[,1])),as.vector(maM(swirlLoess)),as.vector(maM(swirlLNN)),xlim=c(- 2,2),bw="AUTO",
```

detectSpatialBias

*Detecting spatial bias within the print-tips of a two channel array***Description**

This function allows to identify in two channel batch of arrays, which are the print-tips where spatial bias is present.

**Usage**

```
detectSpatialBias(mbatch, corThreshold=0.6)
```

**Arguments**

`mbatch`            A marrayRaw or marrayNorm batch of two channel arrays.  
`corThreshold`      The correlation treshold to be used.

**Details**

This function computes two matrices: `biasRow` and `biasCol`. The elements of these matrices represent the fraction of rows (columns) for which the correlation coefficient between log-ratios, `M`, and column index (row index) is higher than a user specified treshold (default `corThreshold=0.6`). The idea here is to see in which print-tip a important fraction of the rows (columns) are highly correlated with the column (row) index. Since some rows (columns) will show positive correlation while the other negative correlation, we are only interested in a sigle direction of the correlation, i.e. either positive or negative.

**Value**

This function returns a list with two matrices. `biasRow` and `biasCol`. The rows of these matrices correspond to the print tips counted metaRow wise, and the columns correspond to arrays. Values in these matrices superior to 33 point to print-tips that have more tha a third of the rows (columns) with important spatial bias.

**Author(s)**

Tarca, A.L.

**References**

A robust neural networks approach for spatial and intensity dependent normalization of cDNA microarray data, Adi. L. Tarca , Janice. E. K. Cooke, and John Mackay, *Bioinformatics*, 21, 2005, 2674 - 2683.

**See Also**

[maNormNN](#)

**Examples**

```
# detecting spatial bias in swirl data
data(swirl)
# print-tip, intensity and spatial normalization of the first slide in swirl data set
myres<-detectSpatialBias(swirl)
```

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maNormNN

*Intensity and spatial normalization using robust neural networks fitting*

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**Description**

This function normalizes a batch of cDNA arrays by removing the intensity and spatial dependent bias.

**Usage**

```
maNormNN(mbatch,w=NULL,binWidth=3,binHeight=3,model.nonlins=3,iterations=100,nFolds=10,maplots=FALSE)
```

**Arguments**

|          |   |
|----------|---|
| mbatch   | A marrayRaw or marrayNorm batch of arrays.  |
| w        | Weights to be assigned to each spot. If provided, it should be a vector with the same length as maNspots(mbatch).   |
| binWidth | Width of the bins in the $X$ direction (spot column) in which the print tip will be divided in order to account for spatial variation. Max value is maNsc(mbatch), Min value is 1. However if it is set to a number larger than maNsc(mbatch)/2 (so less than two bins in $X$ direction) the variable $X$ will not be used as predictor to estimate the bias. |

|               |  |
|---------------|--|
| binHeight     | Height of the bins in the $Y$ direction (spot row) in which the print tip will be divided in order to account for spatial variation. Max value is <code>maNsr(mbatch)</code> , Min value is 1. However if it is set to a number larger than <code>maNsr(mbatch)/2</code> (so less than two bins in $Y$ direction) the variable $Y$ will not be used as predictor to estimate the bias. |
| model.nonlins | Number of nodes in the hidden layer of the neural network model.   |
| iterations    | The number of iterations at which (if not converged) the training of the neural net will be stopped.   |
| nFolds        | Number of cross-validation folds. It represents the number of equal parts in which the data from a print tip is divided into: the model is trained on <code>nFolds-1</code> parts and the bias is estimated for one part at the time. Higher values improve the results but increase the computation time. Ideal values are between 5 and 10.  |
| maplots       | If set to "TRUE" will produce a $M - A$ plot for each slide before and after normalization.  |
| verbose       | If set to "TRUE" will show the output of the <code>nnet</code> function which is training the neural network models.   |

### Details

This function uses neural networks to model the bias in cDNA data sets.

### Value

A `marrayNorm` object containing the normalized log ratios. See `marrayNorm` class for details

### Author(s)

Tarca, A.L.

### References

A. L. Tarca, J. E. K. Cooke, and J. Mackay. Robust neural networks approach for spatial and intensity dependent normalization of cDNA data. *Bioinformatics*. 2004,submitted.

### See Also

[compNorm](#), `nnet`

### Examples

```
# Normalization of swirl data
data(swirl)
# print-tip, intensity and spatial normalization of the first slide in swirl data set
swirlNN<-maNormNN(swirl[,1])

#do not consider spatial variations, and display M-A plots before and after normalization
swirlNN<-maNormNN(swirl[,1],binWidth=maNsc(swirl),binHeight=maNsr(swirl),maplots=TRUE)
```

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