

# Package ‘biosigner’

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**Type** Package

**Title** Signature discovery from omics data

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**biocViews** Classification, FeatureExtraction, Transcriptomics,  
Proteomics, Metabolomics, Lipidomics

**Description** Feature selection is critical in omics data analysis to extract restricted and meaningful molecular signatures from complex and high-dimension data, and to build robust classifiers. This package implements a new method to assess the relevance of the variables for the prediction performances of the classifier. The approach can be run in parallel with the PLS-DA, Random Forest, and SVM binary classifiers. The signatures and the corresponding 'restricted' models are returned, enabling future predictions on new datasets. A Galaxy implementation of the package is available within the Workflow4metabolomics.org online infrastructure for computational metabolomics.

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**VignetteBuilder** knitr

**License** CeCILL

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biosigner-package	<i>Molecular signature discovery from omics data</i>
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## Description

Feature selection is critical in omics data analysis to extract restricted and meaningful molecular signatures from complex and high-dimension data, and to build robust classifiers. This package implements a new method to assess the relevance of the variables for the prediction performances of the classifier. The approach can be run in parallel with the PLS-DA, Random Forest, and SVM binary classifiers. The signatures and the corresponding 'restricted' models are returned, enabling future predictions on new datasets. A Galaxy implementation of the package is available within the Workflow4metabolomics.org online infrastructure for computational metabolomics.

## Author(s)

Philippe Rinaudo <phd.rinaudo@gmail.com> and Etienne Thevenot <etienne.thevenot@cea.fr>.

Maintainer: Philippe Rinaudo <phd.rinaudo@gmail.com>

## Examples

```
## loading the diaplasm dataset

data(diaplasm)
attach(diaplasm)

## restricting to a smaller dataset for this example

featureSelV1 <- variableMetadata[, "mzmed"] >= 490 & variableMetadata[, "mzmed"] < 500
dataMatrix <- dataMatrix[, featureSelV1]
variableMetadata <- variableMetadata[featureSelV1, ]

## signature selection for all 3 classifiers
## a bootI = 5 number of bootstraps is used for this example
## we recommend to keep the default bootI = 50 value for your analyzes

set.seed(123)
diaSign <- biosign(dataMatrix, sampleMetadata[, "type"], bootI = 5)

detach(diaplasm)
```

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biosign *Builds the molecular signature.*

---

## Description

Main function of the 'biosigner' package. For each of the available classifiers (PLS-DA, Random Forest, and SVM), the significant features are selected and the corresponding models are built.

## Usage

```
biosign(x, ...)

## S4 method for signature 'ExpressionSet'
biosign(x, y, ...)

## S4 method for signature 'data.frame'
biosign(x, ...)

## S4 method for signature 'matrix'
biosign(x, y, methodVc = c("all", "plsda", "randomforest",
  "svm")[1], bootI = 50, pvalN = 0.05, permI = 1, fixRankL = FALSE,
  printL = TRUE, plotL = TRUE, .sinkC = NULL, ...)
```

## Arguments

x	Numerical data frame or matrix (observations x variables), or ExpressionSet object with non empty assayData and phenoData; NAs are allowed for PLS-DA but for SVM, samples with NA will be removed
...	Currently not used.
y	Two-level factor corresponding to the class labels, or a character indicating the name of the column of the pData to be used, when x is an ExpressionSet object
methodVc	Character vector: Either one or all of the following classifiers: Partial Least Squares Discriminant Analysis ('plsda'), or Random Forest ('randomforest'), or Support Vector Machine ('svm')
bootI	Integer: Number of bootstaps for resampling
pvalN	Numeric: To speed up the selection, only variables which significantly improve the model up to two times this threshold (to take into account potential fluctuations) are computed
permI	Integer: Random permutation are used to assess the significance of each new variable included into the model (forward selection)
fixRankL	Logical: Should the initial ranking be computed with the full model only, or as the median of the ranks from the models built on the sampled dataset?
printL	Logical: Should informations regarding the data set and the model be printed? [default = TRUE]
plotL	Logical: Should the 'summary' plot be displayed? [default = TRUE]
.sinkC	Character: Name of the file for R output diversion [default = NULL: no diversion]; Diversion of messages is required for the integration into Galaxy

**Value**

An S4 object of class 'biosign' containing the following slots: 1) 'methodVc' character vector: selected classifier(s) ('plsda', 'randomforest', and/or 'svm'), 2) 'accuracyMN' numeric matrix: balanced accuracies for the full models, and the models restricted to the 'S' and 'AS' signatures (predictions are obtained by using the resampling scheme selected with the 'bootI' and 'crossvall' arguments), 3) 'tierMC' character matrix: contains the tier ('S', 'A', 'B', 'C', 'D', or 'E') of each feature for each classifier (features with tier 'S' have been found significant in all backward selections; features with tier 'A' have been found significant in all but the last selection, and so on), 4) modelLs list: selected classifier(s) trained on the subset restricted to the 'S' features, 5) signatureLs list: 'S' signatures for each classifier; and 6) 'AS' list: 'AS' signatures and corresponding trained classifiers, in addition to the dataset restricted to tiers 'S' and 'A' ('xMN') and the labels ('yFc')

**Author(s)**

Philippe Rinaudo and Etienne Thevenot (CEA)

**See Also**

[predict.biosign](#), [plot.biosign](#)

**Examples**

```
## loading the diaplasm dataset

data(diaplasm)
attach(diaplasm)

## restricting to a smaller dataset for this example

featureSelV1 <- variableMetadata[, "mzmed"] >= 490 & variableMetadata[, "mzmed"] < 500
dataMatrix <- dataMatrix[, featureSelV1]
variableMetadata <- variableMetadata[featureSelV1, ]

## signature selection for all 3 classifiers
## a bootI = 5 number of bootstraps is used for this example
## we recommend to keep the default bootI = 50 value for your analyzes

set.seed(123)
diaSign <- biosign(dataMatrix, sampleMetadata[, "type"], bootI = 5)

detach(diaplasm)
```

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biosign-class

*Class "biosign"*

---

**Description**

The biosigner object class

**Slots**

methodVc character vector: selected classifier(s) ('plsda', 'randomforest', or 'svm')

accuracyMN numeric matrix: balanced accuracies for the full models, and the models restricted to the 'S' and 'AS' signatures

tierMC character matrix: contains the tier ('S', 'A', 'B', 'C', 'D', or 'E') of each feature for each classifier

yFc factor with two levels: response factor

modelLs list: selected classifier(s) trained on the subset restricted to the 'S' features

signatureLs list: 'S' signatures for each classifier

xSubMN matrix: dataset restricted to the 'S' tier

AS list: 'AS' signatures and corresponding trained classifiers, in addition to the dataset restricted to tiers 'S' and 'A' ('xMN') and the labels ('yFc')

**Objects from the Class**

Objects can be created by calls of the form `new("biosign", ...)` or by calling the `biosign` function

**Author(s)**

Philippe Rinaudo and Etienne Thevenot (CEA)

**See Also**

[biosign](#)

**Examples**

```
## loading the diaplasm dataset

data(diaplasm)
attach(diaplasm)

## restricting to a smaller dataset for this example

featureSelV1 <- variableMetadata[, "mzmed"] >= 490 & variableMetadata[, "mzmed"] < 500
dataMatrix <- dataMatrix[, featureSelV1]
variableMetadata <- variableMetadata[featureSelV1, ]

## signature selection for all 3 classifiers
## a bootI = 5 number of bootstraps is used for this example
## we recommend to keep the default bootI = 50 value for your analyzes

set.seed(123)
diaSign <- biosign(dataMatrix, sampleMetadata[, "type"], bootI = 5)

detach(diaplasm)
```

---

diaplasma

*Analysis of plasma from diabetic patients by LC-HRMS*

---

### Description

Plasma samples from 69 diabetic patients were analyzed by reversed-phase liquid chromatography coupled to high-resolution mass spectrometry (Orbitrap Exactive) in the negative ionization mode. The raw data were pre-processed with XCMS and CAMERA (5,501 features), corrected for signal drift, log10 transformed, and annotated with an in-house spectral database. The patient's age, body mass index, and diabetic type are recorded. These three clinical covariates are strongly associated, most of the type II patients being older and with a higher bmi than the type I individuals.

### Format

A list with the following elements:

- `dataMatrix`: a 69 samples x 5,501 features matrix of numeric type corresponding to the intensity profiles (values have been log10-transformed),
- `sampleMetadata`: a 69 x 3 data frame, with the patients' diabetic type (`'type'`, factor), age (`'age'`, numeric), and body mass index (`'bmi'`, numeric),
- `variableMetadata`: a 5,501 x 8 data frame, with the median m/z (`'mzmed'`, numeric) and the median retention time in seconds (`'rtmed'`, numeric) from XCMS, the `'isotopes'` (character), `'adduct'` (character) and `'pcgroups'` (numeric) annotations from CAMERA, and the names of the m/z and RT matching compounds from an in-house database of pure spectra from commercial metabolites (`'spiDb'`, character).

### Value

List containing the `'dataMatrix'` matrix (numeric) of data (samples as rows, variables as columns), the `'sampleMetadata'` data frame of sample metadata, and the `variableMetadata` data frame of variable metadata. Row names of `'dataMatrix'` and `'sampleMetadata'` are identical. Column names of `'dataMatrix'` are identical to row names of `'variableMetadata'`. For details see the `'Format'` section above.

### Source

`'diaplasma'` dataset.

### References

Rinaudo P., Boudah S., Junot C. and Thevenot E.A. (2016). biosigner: a new method for the discovery of significant molecular signatures from omics data. *Frontiers in Molecular Biosciences* 3. doi:10.3389/fmolb.2016.00026

---

getAccuracyMN	<i>Accuracies of the full model and the models restricted to the signatures</i>
---------------	---

---

### Description

Balanced accuracies for the full models, and the models restricted to the 'S' and 'AS' signatures

### Usage

```
getAccuracyMN(object, ...)
```

```
## S4 method for signature 'biosign'  
getAccuracyMN(object)
```

### Arguments

object	An S4 object of class biosign, created by the biosign function.
...	Currently not used.

### Value

A numeric matrix containing the balanced accuracies for the full models, and the models restricted to the 'S' and 'AS' signatures (predictions are obtained by using the resampling scheme selected with the 'bootI' and 'crossvall' arguments)

### Author(s)

Philippe Rinaudo and Etienne Thevenot (CEA)

### Examples

```
## loading the diaplasm dataset  
  
data(diaplasm)  
attach(diaplasm)  
  
## restricting to a smaller dataset for this example  
  
featureSelV1 <- variableMetadata[, "mzmed"] >= 490 & variableMetadata[, "mzmed"] < 500  
dataMatrix <- dataMatrix[, featureSelV1]  
variableMetadata <- variableMetadata[featureSelV1, ]  
  
## signature selection for all 3 classifiers  
## a bootI = 5 number of bootstraps is used for this example  
## we recommend to keep the default bootI = 50 value for your analyzes  
  
set.seed(123)  
diaSign <- biosign(dataMatrix, sampleMetadata[, "type"], bootI = 5)  
  
## individual boxplot of the selected signatures  
  
getAccuracyMN(diaSign)
```

```
detach(diaplasma)
```

---

```
getSignatureLs          Signatures selected by the models
```

---

### Description

List of 'S' (or 'S' and 'A') signatures for each classifier

### Usage

```
getSignatureLs(object, tierC = c("S", "AS")[1], ...)

## S4 method for signature 'biosign'
getSignatureLs(object, tierC = c("S", "AS")[1])
```

### Arguments

object	An S4 object of class biosign, created by the biosign function.
tierC	Character: defines whether signatures from the 'S' tier only (default) or the ('S' and 'A') tiers should be returned
...	Currently not used.

### Value

List of 'S' (or 'S' and 'A') signatures for each classifier

### Author(s)

Philippe Rinaudo and Etienne Thevenot (CEA)

### Examples

```
## loading the diaplasma dataset

data(diaplasma)
attach(diaplasma)

## restricting to a smaller dataset for this example

featureSelV1 <- variableMetadata[, "mzmed"] >= 490 & variableMetadata[, "mzmed"] < 500
dataMatrix <- dataMatrix[, featureSelV1]
variableMetadata <- variableMetadata[featureSelV1, ]

## signature selection for all 3 classifiers
## a bootI = 5 number of bootstraps is used for this example
## we recommend to keep the default bootI = 50 value for your analyzes

set.seed(123)
diaSign <- biosign(dataMatrix, sampleMetadata[, "type"], bootI = 5)
```



```
## individual boxplot of the selected signatures  
getSignatureLs(diaSign)  
detach(diaplasma)
```

---

plot,biosign-method     *Plot method for 'biosign' signature objects*

---

## Description

Displays classifier tiers or individual boxplots from selected features

## Usage

```
## S4 method for signature 'biosign'  
plot(x, y, tierMaxC = "S", typeC = c("tier",  
  "boxplot")[1], file.pdfC = NULL, .sinkC = NULL, ...)
```

## Arguments

x	An S4 object of class biosign, created by the biosign function.
y	Currently not used.
tierMaxC	Character: Maximum level of tiers to display: Either 'S' and 'A', (for boxplot), or also 'B', 'C', 'D', and 'E' (for tiers) by decreasing number of selections
typeC	Character: Plot type; either 'tier' [default] displaying the comparison of signatures up to the selected 'tierMaxC' or 'boxplot' showing the individual boxplots of the features selected by all the classifiers
file.pdfC	Character: Figure filename ending with '.pdf'; default is NULL (no saving; displaying instead)
.sinkC	Character: Name of the file for R output diversion [default = NULL: no diversion]; Diversion of messages is required for the integration into Galaxy
...	Currently not used.

## Value

A plot is created on the current graphics device.

## Author(s)

Philippe Rinaudo and Etienne Thevenot (CEA)

**Examples**

```
## loading the diaplasm dataset

data(diaplasm)
attach(diaplasm)

## restricting to a smaller dataset for this example

featureSelV1 <- variableMetadata[, "mzmed"] >= 490 & variableMetadata[, "mzmed"] < 500
dataMatrix <- dataMatrix[, featureSelV1]
variableMetadata <- variableMetadata[featureSelV1, ]

## signature selection for all 3 classifiers
## a bootI = 5 number of bootstraps is used for this example
## we recommend to keep the default bootI = 50 value for your analyzes

set.seed(123)
diaSign <- biosign(dataMatrix, sampleMetadata[, "type"], bootI = 5)

## individual boxplot of the selected signatures

plot(diaSign, typeC = "boxplot")

detach(diaplasm)
```

---

predict,biosign-method

*Predict method for 'biosign' signature objects*

---

**Description**

This function predicts values based upon biosign classifiers trained on the 'S' signature

**Usage**

```
## S4 method for signature 'biosign'
predict(object, newdata, tierMaxC = "S", ...)
```

**Arguments**

object	An S4 object of class biosign, created by biosign function.
newdata	Either a data frame or a matrix, containing numeric columns only, with column names identical to the 'x' used for model training with 'biosign'.
tierMaxC	Character: Maximum level of tiers to display: Either 'S' or 'A'.
...	Currently not used.

**Value**

Data frame with the predictions for each classifier as factor columns.

**Author(s)**

Philippe Rinaudo and Etienne Thevenot (CEA)

**Examples**

```
## loading the diaplasm dataset

data(diaplasm)
attach(diaplasm)

## restricting to a smaller dataset for this example

featureSelV1 <- variableMetadata[, "mzmed"] >= 490 & variableMetadata[, "mzmed"] < 500
dataMatrix <- dataMatrix[, featureSelV1]
variableMetadata <- variableMetadata[featureSelV1, ]

## training the classifiers
## a bootI = 5 number of bootstraps is used for this example
## we recommend to keep the default bootI = 50 value for your analyzes

set.seed(123)
diaSign <- biosign(dataMatrix, sampleMetadata[, "type"], bootI = 5)

## fitted values (for the subsets restricted to the 'S' signatures)
sFitDF <- predict(diaSign)

## confusion tables
print(lapply(sFitDF, function(predFc) table(actual = sampleMetadata[,
"type"], predicted = predFc)))

## balanced accuracies
sapply(sFitDF, function(predFc) { conf <- table(sampleMetadata[,
"type"], predFc)
conf <- sweep(conf, 1, rowSums(conf), "/")
mean(diag(conf))
})
## note that these values are slightly different from the accuracies
## returned by biosign because the latter are computed by using the
## resampling scheme selected by the bootI or crossvalI arguments
getAccuracyMN(diaSign)["S", ]

detach(diaplasm)
```

---

show,biosign-method    *Show method for 'biosign' signature objects*

---

**Description**

Prints the selected features and the accuracies of the classifiers.

**Usage**

```
## S4 method for signature 'biosign'  
show(object)
```

**Arguments**

object                    An S4 object of class biosign, created by the biosign function.

**Value**

Invisible.

**Author(s)**

Philippe Rinaudo and Etienne Thevenot (CEA)

**Examples**

```
## loading the diaplasm dataset  
  
data(diaplasm)  
attach(diaplasm)  
  
## restricting to a smaller dataset for this example  
  
featureSelV1 <- variableMetadata[, "mzmed"] >= 490 & variableMetadata[, "mzmed"] < 500  
dataMatrix <- dataMatrix[, featureSelV1]  
variableMetadata <- variableMetadata[featureSelV1, ]  
  
## signature selection for all 3 classifiers  
## a bootI = 5 number of bootstraps is used for this example  
## we recommend to keep the default bootI = 50 value for your analyzes  
  
set.seed(123)  
diaSign <- biosign(dataMatrix, sampleMetadata[, "type"], bootI = 5)  
  
diaSign  
  
detach(diaplasm)
```

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