

# Package ‘lilikoï’

July 30, 2018

**Title** Metabolomics Personalized Pathway Analysis Tool

**Version** 0.1.0

**Description** Computes the pathway deregulation score for a given set of metabolites, selects the pathways with the highest mutual information and then uses them to build a classifier. F. Alakwaa, S. Huang, and L. Garmire (2018) <doi:10.1101/283408>.

**Imports** caret, corrplot, devtools, dplyr, e1071, gbm, ggplot2, glmnet, hash, Hmisc, infotheo, Matrix, pamr, R.oo, princurve, pROC, randomForest, reshape2, RWeka, stringr

**Depends** R (>= 3.4.0)

**License** GPL-2

**Encoding** UTF-8

**LazyData** true

**URL** <https://github.com/lanagarmire/lilikoï>

**BugReports** <https://github.com/lanagarmire/lilikoï/issues>

**RoxygenNote** 6.0.1.9000

**Suggests** testthat

**NeedsCompilation** no

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**Repository** CRAN

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data.hmdb	<i>data.hmdb</i>
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### Description

data.hmdb

### Usage

data.hmdb

### Format

data.frame \$ Accession : Factor w/ 53287 levels 'HMDB0000001',...: 1 1 2 2 2 2 3 3 3 3 ... \$ Name : Factor w/ 53287 levels '(±)-Glycerol 1,2-diacetate',...: 45 45 55 55 55 55 146 146 146 146 ... \$ Pathway\_Name: Factor w/ 48607 levels '11-beta-hydroxylase deficiency (CYP11B1)',...: 46317 46318 98 23433 46265 48580 19 23491 46174 46176 ...

### Source

[http://www.hmdb.ca/system/downloads/current/hmdb\\_metabolites.zip](http://www.hmdb.ca/system/downloads/current/hmdb_metabolites.zip)

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data.metaboAnalyst	<i>data.metaboAnalyst</i>
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### Description

data.metaboAnalyst

### Usage

data.metaboAnalyst

### Format

data.frame Classes 'tbl\_df', 'tbl' and 'data.frame': 183937 obs. of 8 variables: \$ hmdb\_id : chr 'HMDB0000001' 'HMDB0000001' 'HMDB0000001' 'HMDB0000001' ... \$ name : chr '1-Methylhistidine' '1-Methylhistidine' '1-Methylhistidine' '1-Methylhistidine' ... \$ synonym : chr '(2S)-2-amino-3-(1-Methyl-1H-imidazol-4-yl)propanoic acid' ... \$ pubchem\_id: int 92105 92105 92105 92105 92105 92105 92105 92105 92105 ... \$ chebi\_id : int 50599 50599 50599 50599 50599 50599 50599 50599 ... \$ kegg\_id : chr 'C01152' 'C01152' 'C01152' 'C01152' ... \$ metlin\_id : int 3741 3741 3741 3741 3741 3741 3741 3741 3741 3741 ...

**Source**

<http://www.metaboanalyst.ca/MetaboAnalyst/faces/docs/Resources.xhtml>

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data.smpdb

*data.smpdb*

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**Description**

data.smpdb

**Usage**

data.smpdb

**Format**

data.frame 'data.frame': 19245 obs. of 2 variables: \$ pathway: Factor w/ 705 levels '11-beta-hydroxylase deficiency (CYP11B1)',...: 39 39 39 39 39 39 39 39 39 39 ... \$ hmdb\_id: chr 'HMDB0000538' 'HMDB0000161' 'HMDB0000045' 'HMDB0000250' ...

**Source**

[http://smpdb.ca/downloads/smpdb\\_metabolites.csv.zip](http://smpdb.ca/downloads/smpdb_metabolites.csv.zip)

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lilikoi.adjust\_model

*Model Adjustment method using clinical factors*

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**Description**

This function can adjust the best performing model using additional clinical factors specified by the user. It plots ROC for three models: a model building using only the selected pathways, a model built using the clinical factors and a model built using both the selected pathways and the clinical factors.

**Usage**

```
lilikoi.adjust_model(mlResults, PDSmatrix, selected_Pathways_Weka,  
  metaboliteMeasurements, clinical_factors_data, factors)
```

**Arguments**

mlResults	The models and other results generated from machine_learning method
PDSmatrix	the PDS matrix generated using lilikoi.get_pd_scores function
selected_Pathways_Weka	selected pathway using WEKA algorithm generated from lilikoi.select_pathways function
metaboliteMeasurements	the metaboliteMeasurements for the samples
clinical_factors_data	Metadata
factors	which the users want to add to the model

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`lilikoi.get_pd_scores` A *lilikoi.get\_pd\_scores* Function

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**Description**

This function allows you to compute Pathway Dereglulation Score deriving make sure that you have the below database for the metabolites and pathway list: meta\_path.RData

**Usage**

```
lilikoi.get_pd_scores(metaboliteMeasurements, metabolitePathwayTable,
  pathwayMetaboliteDatabase = lilikoi::data.smpdb, maxit = 200)
```

**Arguments**

metaboliteMeasurements	Metabolite levels
metabolitePathwayTable	This is the metabolitePathwayTable from lilikoi.metab_to_pathway function. This table includes the metabolite ids and their corresponding HMDB ids
pathwayMetaboliteDatabase	An external database of metabolites and their associated pathways.
maxit	Max iterations for princurve in pathifier algorithm. Lowering increases speed but lowers performance

**Examples**

```
filename <- system.file("extdata", "plasma_breast_cancer.csv", package = "lilikoi")
data <- read.csv(file = filename, check.names = FALSE, row.names = 1)[, 1:20]
metaboliteNames <- colnames(data)[-1]
matches <- lilikoi.metab_to_pathway(metaboliteNames, "name")
PDSmatrix <- lilikoi.get_pd_scores(data, matches, lilikoi::data.smpdb[1:23,], maxit = 1)
```

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`lilikoimachine_learning`*A machine learning Function*

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**Description**

This function for classification using 7 different machine learning algorithms and it plots the ROC curves and the AUC, SEN, and specificity

**Usage**

```
lilikoimachine_learning(PDSmatrix, measurementLabels, significantPathways)
```

**Arguments**

PDSmatrix	from lilikoiget_pd_scores
measurementLabels	Cancer or not
significantPathways	Pathways with high significance from lilikoiselect_pathways function

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`lilikoimetab_to_pathway`*A lilikoimetab\_to\_pathway Function*

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**Description**

This function allows you to convert your metabolites id such as names, kegg ids, pubchem ids. into pathways. Metabolites which have no pathways will be excluded from any downstream analysis.

**Usage**

```
lilikoimetab_to_pathway(metaboliteNames, searchType)
```

**Arguments**

metaboliteNames	Your metabolite data
searchType	The type of the metabolites id such as 'name', 'kegg', 'hmdb', 'pubchem'

**Examples**

```
matches = lilikoimetab_to_pathway(c("Asparagine", "Hypotaurine", "5-Oxoproline"), "name")
```

```
matches = lilikoimetab_to_pathway(c("C00152", "C00519", "C01879"), "kegg")
```

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```
lilikoi.select_pathways
```

*This function will return the significant pathways that can be used for training machine learning models.*

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### **Description**

This function can reduce the pathway dimension using method specified.

### **Usage**

```
lilikoi.select_pathways(PDSmatrix, metaboliteMeasurements, threshold = 0.5,  
  method = "info")
```

### **Arguments**

PDSmatrix	from lilikoi.get_pd_scores function
metaboliteMeasurements	Levels
threshold	Cutoff for significance
method	Which gain method to use

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