# Package 'crossmeta'

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Title Cross Platform Meta-Analysis of Microarray Data

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Description Implements cross-platform and cross-species meta-analyses of Affymentrix, Illumina, and Agilent microarray data. This package automates common tasks such as downloading, normalizing, and annotating raw GEO data. The user then selects control and treatment samples in order to perform differential expression/pathway analyses for all comparisons. After analysing each contrast seperately, the user can select tissue sources for each contrast and specify any tissue sources that should be grouped for the subsequent meta-analyses. Finally, effect size and pathway meta-analyses can proceed and the results graphically explored.

**Depends** R (>= 3.3)

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## **Description**

User selects a tissue source for each contrast and indicates any sources that should be paired. This step is required if you would like to perform source-specific effect-size/pathway meta-analyses.

## Usage

```
add_sources(diff_exprs, data_dir = getwd())
```

## **Arguments**

diff\_exprs Previous result of diff\_expr, which can be reloaded using load\_diff.
data\_dir String specifying directory of GSE folders.

## **Details**

The **Sources** tab is used to add a source for each contrast. To do so: click the relevant contrast rows, search for a source in the *Sample source* dropdown box, and then click the *Add* button.

The **Pairs** tab is used to indicate sources that should be paired (treated as the same source for subsequent effect-size and pathway meta-analyses). To do so: select at least two sources from the *Paired sources* dropdown box, and then click the *Add* button.

For each GSE, analysis results with added sources/pairs are saved in the corresponding GSE folder (in data\_dir) that was created by get\_raw.

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#### Value

Same as diff\_expr with added slots for each GSE in diff\_exprs:

sources Named vector specifying selected sample source for each contrast. Vector names

identify the contrast.

pairs List of character vectors indicating tissue sources that should be treated as the

same source for subsequent effect-size and pathway meta-analyses.

## **Examples**

```
library(lydata)

# load result of previous call to diff_expr:
data_dir <- system.file("extdata", package = "lydata")
gse_names <- c("GSE9601", "GSE34817")
anals <- load_diff(gse_names, data_dir)

# run shiny GUI to add tissue sources
# anals <- add_sources(anals, data_dir)</pre>
```

contribute

Contribute results of meta-analysis to public database.

#### **Description**

Contributed results will be used to build a freely searchable database of gene expression metaanalyses.

## Usage

```
contribute(diff_exprs, subject)
```

## Arguments

diff\_exprs Result of call to diff\_expr.

subject String identifying meta-analysis subject (e.g. "rapamycin" or "prostate\_cancer").

## **Details**

Performs meta-analysis on diff\_exprs using es\_meta. Sends overall mean effect size values and minimal information needed to reproduce meta-analysis.

#### Value

NULL (used to contribute meta-analysis).

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#### **Examples**

```
library(lydata)

# location of data
data_dir <- system.file("extdata", package = "lydata")

# gather GSE names
gse_names <- c("GSE9601", "GSE15069", "GSE50841", "GSE34817", "GSE29689")

# load differential expression analyses
anals <- load_diff(gse_names, data_dir)

# contribute results of meta-analysis
# contribute(anals, subject = "LY294002")</pre>
```

diff\_expr

Differential expression analysis of esets.

## Description

After selecting control and test samples for each contrast, surrogate variable analysis (sva) and differential expression analysis is performed.

#### Usage

```
diff_expr(esets, data_dir = getwd(), annot = "SYMBOL",
    prev_anals = list(NULL))
```

## **Arguments**

esets List of annotated esets. Created by load\_raw.

data\_dir String specifying directory of GSE folders.

annot String, column name in fData common to all e

String, column name in fData common to all esets. For duplicated values in this column, the row with the highest interquartile range across selected samples will be kept. If meta-analysis will follow, appropriate values are "SYMBOL" (default - for gene level analysis) or, if all esets are from the same platform,

"PROBE" (for probe level analysis).

prev\_anals Previous result of diff\_expr, which can be reloaded using load\_diff. If

present, previous selections, names, and pairs will be reused.

#### **Details**

The **Samples** tab is used to select control and test samples for each contrast. To do so: select rows for control samples, type a group name in the *Control group name* text input box and click the *Add Group* button. Repeat for test samples. While adding additional contrasts, a previous control group can be quickly reselected from the *Previous selections* dropdown box. After control and test samples have been added for all contrasts that you wish to include, click the *Done* button. Repeat for all GSEs.

Paired samples (e.g. the same subject before and after treatment) can be specified by selecting sample rows to pair and then clicking *Pair Samples*. The author does not usually specify paired

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samples and instead allows surrogate variable analysis to discover these inter-sample relationships from the data itself.

The **Contrasts** tab is used to view and delete contrasts that have already been added.

For each GSE, analysis results are saved in the corresponding GSE folder in data\_dir that was created by get\_raw. If analyses needs to be repeated, previous results can be reloaded with load\_diff and supplied to the prev\_anals parameter. In this case, previous selections, names, and pairs will be reused.

#### Value

List of named lists, one for each GSE. Each named list contains:

pdata data.frame with phenotype data for selected samples. Columns treatment ('ctrl' or 'test'), group, and pairs are added based on user selections.

top\_tables List with results of topTable call (one per contrast). These results account for the effects of nuissance variables discovered by surrogate variable analysis.

Results of call to eBayes with surrogate variables included in the model matrix.

Value of annot variable.

#### **Examples**

```
library(lydata)

# location of raw data
data_dir <- system.file("extdata", package = "lydata")

# gather GSE names
gse_names <- c("GSE9601", "GSE15069", "GSE50841", "GSE34817", "GSE29689")

# load first eset
esets <- load_raw(gse_names[1], data_dir)

# run analysis
# anals <- diff_expr(esets, data_dir)

# re-run analysis on first eset
prev <- load_diff(gse_names[1], data_dir)
# anals <- diff_expr(esets[1], data_dir, prev_anals = prev)</pre>
```

diff\_path

Differential expression of KEGG pathways.

#### **Description**

Performs PADOG pathway analysis using KEGG database (downloaded Feb 2017).

```
diff_path(esets, prev_anals, data_dir = getwd())
```

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#### **Arguments**

esets List of annotated esets. Created by load\_raw.

prev\_anals Previous result of diff\_expr, which can be reloaded using load\_diff.

data\_dir String specifying directory for GSE folders.

#### **Details**

If you wish to perform source-specific pathway meta-analyses, add\_sources must be used before diff\_paths.

For each GSE, analysis results are saved in the corresponding GSE folder in data\_dir that was created by get\_raw. PADOG outperforms other pathway analysis algorithms at prioritizing expected pathways (see references).

#### Value

List of named lists, one for each GSE. Each named list contains:

padog\_tables data.frames containing padog pathway analysis results for each contrast.

If add\_sources is used first:

sources Named vector specifying selected sample source for each contrast. Vector names

identify the contrast.

pairs List of character vectors indicating tissue sources that should be treated as the

same source for subsequent pathway meta-analysis.

#### References

Tarca AL, Bhatti G, Romero R. A Comparison of Gene Set Analysis Methods in Terms of Sensitivity, Prioritization and Specificity. Chen L, ed. PLoS ONE. 2013;8(11):e79217. doi:10.1371/journal.pone.0079217.

Dong X, Hao Y, Wang X, Tian W. LEGO: a novel method for gene set over-representation analysis by incorporating network-based gene weights. Scientific Reports. 2016;6:18871. doi:10.1038/srep18871.

```
library(lydata)

# location of data
data_dir <- system.file("extdata", package = "lydata")

# gather GSE names
gse_names <- c("GSE9601", "GSE15069", "GSE50841", "GSE34817", "GSE29689")

# load esets
esets <- load_raw(gse_names, data_dir)

# load previous differential expression analysis
anals <- load_diff(gse_names, data_dir)

# add tissue sources to perform seperate meta-analyses for each source (recommended)
# anals <- add_sources(anals)

# perform pathway analysis for each contrast
# path_anals <- diff_path(esets, anals, data_dir)</pre>
```

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es_meta Effect size combination meta analysis.	es_meta	Effect size combination meta analysis.	
--	---------	--	--

#### **Description**

Performs effect-size meta-analyses across all studies and seperately for each tissue source.

#### Usage

```
es_meta(diff_exprs, cutoff = 0.3, by_source = FALSE)
```

#### **Arguments**

diff\_exprs Previous result of diff\_expr, which can be reloaded using load\_diff.

cutoff Minimum fraction of contrasts that must have measured each gene. Between 0 and 1.

by\_source Should seperate meta-analyses be performed for each tissue source added with

add\_sources?

#### **Details**

Builds on zScores function from GeneMeta by allowing for genes that were not measured in all studies. This implementation also uses moderated unbiased effect sizes calculated by effectsize from metaMA and determines false discovery rates using fdrtool.

#### Value

A list of named lists, one for each tissue source. Each list contains two named data.frames. The first, filt, has all the columns below for genes present in cutoff or more fraction of contrasts. The second, raw, has only dprime and vardprime columns, but for all genes (NAs for genes not measured by a given contrast).

dprime Unbiased effect sizes (one column per contrast).

vardprime Variances of unbiased effect sizes (one column per contrast).

mu Overall mean effect sizes.

var Variances of overall mean effect sizes.

z Overall z score = mu / sqrt(var).

fdr False discovery rates calculated by fdrtool.

```
library(lydata)
# location of data
data_dir <- system.file("extdata", package = "lydata")
# gather GSE names
gse_names <- c("GSE9601", "GSE15069", "GSE50841", "GSE34817", "GSE29689")
# load previous analysis</pre>
```

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```
anals <- load_diff(gse_names, data_dir)

# add tissue sources to perform seperate meta-analyses for each source (optional)
# anals <- add_sources(anals, data_dir)

# perform meta-analysis
es <- es_meta(anals, by_source = TRUE)</pre>
```

explore\_paths

Explore pathway meta analyses.

#### **Description**

Shiny app for interactively exploring the results of effect-size and pathway meta-analyses. The app also interfaces with the ccmap package in order to explore drugs that are predicted to reverse or mimic your signature.

#### Usage

#### Arguments

es\_res Result of call to es\_meta.

path\_res Result of call to path\_meta.

drug\_info Matrix of differential expression values for drugs (rows are genes, columns are

drugs). If NULL (default), cmap\_es is used.

type Desired direction of drug action on query signature (see details).

#### **Details**

For a given tissue source (top left dropdown box) and KEGG pathway (bottom left dropdown box, ordered by increasing false discovery rate), effect-sizes (y-axis) are plotted for each gene in the pathway (x-axis, ordered by decreasing asbsolute effect size).

For each gene, open circles give the effect-sizes for each contrast. The transparency of the open circles is proportional to the standard deviation of the effect-size for each contrast. For each gene, error bars give one standard deviation above and below the the overall meta-analysis effect-size.

The top drugs for the full signature in a given tissue (top right dropdown box, red points) and just the pathway genes (bottom right dropdown box, blue points) are orderered by decreasing (if type is 'both' or 'mimic') or increasing (if type is 'reverse') similarity. Positive and negative cosine similarities correspond to drugs that, respectively, mimic and reverse the query signature.

Drug effect sizes can be made visible by either clicking the legend entries (top left of plot) or selecting a new drug in the dropdown boxes.

When a new tissue source or pathway is selected, the top drug and pathway dropdown boxes are approriately updated.

#### Value

None

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#### **Examples**

```
library(lydata)

data_dir <- system.file("extdata", package = "lydata")
gse_names <- c("GSE9601", "GSE15069", "GSE50841", "GSE34817", "GSE29689")

# load result of previous call to diff_expr:
es_anals <- load_diff(gse_names, data_dir)

# run shiny GUI to add tissue sources
# es_anals <- add_sources(es_anals, data_dir)

# perform effect-size meta-analyses for each tissue source
es_res <- es_meta(es_anals, by_source = TRUE)

# load result of previous call to diff_path:
# path_anals <- load_path(gse_names, data_dir)

# perform pathway meta-analyses for each tissue source
# path_res <- path_meta(path_anals, ncores = 1, nperm = 100, by_source = TRUE)

# explore pathway meta-analyses
# explore_paths(es_res, path_res)</pre>
```

get\_raw

Download and unpack microarray supplementary files from GEO.

## **Description**

Downloads and unpacks microarray supplementary files from GEO. Files are stored in the supplied data directory under the GSE name.

## Usage

```
get_raw(gse_names, data_dir = getwd())
```

#### **Arguments**

gse\_names Character vector of GSE names to download.

data\_dir String specifying directory for GSE folders.

#### Value

NULL (for download/unpack only).

## See Also

load\_raw.

```
get_raw("GSE41845")
```

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gs.names

Map between KEGG pathway numbers and names.

# Description

Used to map human KEGG pathway numbers to names. Updated Feb 2017.

# Usage

```
data(gs.names)
```

#### **Format**

An object of class character of length 312.

## Value

A named character vector of human KEGG pathway names. Names of vector are KEGG pathway numbers.

gslist

KEGG human pathway genes.

# Description

Genes for human KEGG pathways. Updated Feb 2017.

# Usage

```
data(gslist)
```

#### **Format**

An object of class list of length 312.

#### Value

A named list with entrez ids of genes for human KEGG pathways. List names are KEGG pathway numbers.

load\_diff

load\_diff

Load previous differential expression analyses.

## **Description**

Loads previous differential expression analyses.

## Usage

```
load_diff(gse_names, data_dir = getwd(), annot = "SYMBOL")
```

#### **Arguments**

gse\_names Character vector specifying GSE names to be loaded.

data\_dir String specifying directory of GSE folders.

annot Level of previous analysis (e.g. "SYMBOL" or "PROBE").

#### Value

Result of previous call to diff\_expr.

#### **Examples**

```
library(lydata)

data_dir <- system.file("extdata", package = "lydata")
gse_names <- c("GSE9601", "GSE34817")
prev <- load_diff(gse_names, data_dir)</pre>
```

load\_path

Load previous pathway analyses.

## **Description**

Load previous pathway analyses.

# Usage

```
load_path(gse_names, data_dir = getwd())
```

## **Arguments**

gse\_names Character vector of GSE names.

data\_dir String specifying directory for GSE folders.

## Value

Result of previous call to diff\_path.

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## **Examples**

```
library(lydata)

# location of data
data_dir <- system.file("extdata", package = "lydata")

# gather GSE names
gse_names <- c("GSE9601", "GSE15069", "GSE50841", "GSE34817", "GSE29689")

# load previous pathway analyses
# path_anals <- load_path(gse_names, data_dir)</pre>
```

load\_raw

Load and annotate raw data downloaded from GEO.

# **Description**

Loads and annotates raw data previously downloaded with get\_raw. Supported platforms include Affymetrix, Agilent, and Illumina.

## Usage

```
load_raw(gse_names, data_dir = getwd(), gpl_dir = "..", overwrite = FALSE)
```

# **Arguments**

gse_names	Character vector of GSE names.
data_dir	String specifying directory with GSE folders.
gpl_dir	String specifying parent directory to search for previously downloaded GPL.soft files.
overwrite	Do you want to overwrite saved esets from previous load_raw?

# Value

List of annotated esets.

```
library(lydata)
data_dir <- system.file("extdata", package = "lydata")
eset <- load_raw("GSE9601", data_dir = data_dir)</pre>
```

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open\_raw\_illum

Open raw Illumina microarray files.

## **Description**

Helper function to open raw Illumina microarray files in order to check that they are formatted correctly. For details on correct format, please see 'Checking Raw Illumina Data' in vignette.

# Usage

```
open_raw_illum(gse_names, data_dir = getwd())
```

## **Arguments**

gse\_names Character vector of Illumina GSE names to open.

data\_dir String specifying directory with GSE folders.

#### Value

Character vector of successfully formated Illumina GSE names.

## **Examples**

```
library(lydata)
# Illumina GSE names
illum_names <- c("GSE50841", "GSE34817", "GSE29689")
# location of raw data
data_dir <- system.file("extdata", package = "lydata")
# open raw data files with default text editor
# open_raw_illum(illum_names)</pre>
```

path\_meta

Pathway p-value meta analysis.

#### **Description**

Uses Fisher's method to combine p-values from PADOG pathway analyses.

```
path_meta(path_anals, ncores = parallel::detectCores(), nperm = ncores *
10000, by_source = FALSE)
```

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#### **Arguments**

path\_anals Previous result of diff\_path, which can be reloaded using load\_path.

ncores Number of cores to use. Default is all available.

nperm Number of permutation to perform to calculate p-values.

by\_source Should separate meta-analyses be performed for each tissue source added with

add\_sources?

#### **Details**

Permutation p-values are determined by shuffling pathway names associated with PADOG p-values prior to meta-analysis. Permutation p-values are then adjusted using the Benjamini & Hochberg method to obtain false discovery rates.

#### Value

A list of matrices, one for each tissue source. Each matrix contains a column of PADOG p-values for each contrast and permutation p- and fdr-values for the meta analysis.

#### See Also

```
sumlog, padog.
```

#### **Examples**

```
library(lydata)

# location of data
data_dir <- system.file("extdata", package = "lydata")

# gather GSE names
gse_names <- c("GSE9601", "GSE15069", "GSE50841", "GSE34817", "GSE29689")

# load previous pathway analyses
# path_anals <- load_path(gse_names, data_dir)

# perform pathway meta analysis
# path_res <- path_meta(path_anals, ncores = 1, nperm = 100)</pre>
```

setup\_prev

Setup selections when many samples.

# Description

Function is useful when number of samples makes manual selection with diff\_expr error prone and time-consuming. This is often true for large clinical data sets.

```
setup_prev(eset, contrasts)
```

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#### **Arguments**

eset List containing one expression set with pData 'group' and 'pairs' (optional)

columns. Name of eset should be the GSE name.

contrasts Character vector specifying contrasts to analyse. Each contrast must take the

form "B-A" where both "B" and "A" are present in eset pData 'group' column.

"B" is the treatment group and "A" is the control group.

#### Value

List containing necessary information for prev\_anal parameter of diff\_expr.

#### **Examples**

```
library(lydata)
library(Biobase)
# location of raw data
data_dir <- system.file("extdata", package = "lydata")</pre>
# load eset
gse_name <- c("GSE34817")</pre>
eset <- load_raw(gse_name, data_dir)</pre>
# inspect pData of eset
# View(pData(eset$GSE34817)) # if using RStudio
head(pData(eset$GSE34817))
                                # otherwise
# get group info from pData (differs based on eset)
group <- pData(eset$GSE34817)$characteristics_ch1.1</pre>
# make group names concise and valid
group <- gsub("treatment: ", "", group)</pre>
group <- make.names(group)</pre>
# add group to eset pData
pData(eset$GSE34817)$group <- group
# setup selections
sel <- setup_prev(eset, contrasts = "LY-DMSO")</pre>
# run differential expression analysis
# anal <- diff_expr(eset, data_dir, prev_anal = sel)</pre>
```

symbol\_annot

Add hgnc symbol to expression set.

#### **Description**

Function first maps entrez gene ids to homologous human entrez gene ids and then to hgnc symbols.

```
symbol_annot(eset, gse_name = "")
```

symbol\_annot

## **Arguments**

eset Expression set to annotate.

gse\_name GSE name for eset.

## **Details**

Initial entrez gene ids are obtained from bioconductor annotation data packages or from feature data of supplied expression set. Homologous human entrez ids are obtained from homologene and then mapped to hgnc symbols using org.Hs.eg.db. Expression set is expanded if 1:many mappings occur.

## Value

Expression set with hgnc symbols ("SYMBOL") and row names ("PROBE") added to fData slot.

## See Also

load\_raw.

```
library(lydata)

# location of raw data
data_dir <- system.file("extdata", package = "lydata")

# load eset
eset <- load_raw("GSE9601", data_dir)[[1]]

# annotate eset (need if load_raw failed to annotate)
eset <- symbol_annot(eset)</pre>
```

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