Package 'octad'

June 13, 2025

Title Open Cancer TherApeutic Discovery (OCTAD)

Version 1.11.1

Description OCTAD provides a platform for virtually screening compounds targeting precise cancer patient groups. The essential idea is to identify drugs that reverse the gene expression signature of disease by tamping down over-expressed genes and stimulating weakly expressed ones. The package offers deep-learning based reference tissue selection, disease gene expression signature creation, pathway enrichment analysis, drug reversal potency scoring, cancer cell line selection, drug enrichment analysis and in silico hit validation. It currently covers ~20,000 patient tissue samples covering 50 cancer types, and expression profiles for ~12,000 distinct compounds.

```
License Artistic-2.0
Encoding UTF-8
LazyData FALSE
Roxygen list(markdown = TRUE)
RoxygenNote 7.3.2
Depends R (>= 4.2.0), magrittr, dplyr, ggplot2, edgeR, RUVSeq, DESeq2,
      limma, rhdf5, foreach, Rfast, octad.db, stats, httr, qpdf,
      ExperimentHub, AnnotationHub, Biobase, S4Vectors
Imports EDASeq, GSVA, data.table, htmlwidgets, plotly, reshape2,
     grDevices, utils
Suggests knitr, rmarkdown
VignetteBuilder knitr
biocViews Classification, GeneExpression, Pharmacogenetics,
      Pharmacogenomics, Software, GeneSetEnrichment
git_url https://git.bioconductor.org/packages/octad
git_branch devel
git_last_commit fdf7e72
git_last_commit_date 2025-04-20
Repository Bioconductor 3.22
Date/Publication 2025-06-12
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 ${\tt compute Cell Line}$

Compute Correlation between cell lines and vector of case ids.

Description

Select top CCLE cell lines sharing similar expression profiles with input case samples. Input case sample ids and output correlation scores for every cell line and/or output file. The results could be used for in-silico validation of predictions or used to weight cell lines in RGES computation. CellLineCorrelations.csv, correlation between CCLE cell lines and input disease samples.

Usage

Arguments

case_id	vector of ids from octad database. Ids can be obtained from phenoDF.
output	by default FALSE, if TRUE, file CellLineCorrelations.csv with results are produced in working directory.
outputFolder	Folder to store results.
LINCS_overlaps	vector of cell line ids from octad database. If TRUE, overlap with LINCS cells database wll be performed

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the file for the octad expression matrix. By default, set to octad.small to use only 978 landmark genes profiled in LINCS database. Use octad.whole option to compute DE on the whole transcriptome octad.counts.and.tpm.h5 file. The file should be present in the working directory or the whole path should be

included. If source is set to 'side', the expSet matrix is estimated.

expSet input expression matrix. By default set to NULL since the expSet is created based

on cases, controls and source file.

file if expSet='octad.whole', source path to expSet='octad.counts.and.tpm.h5'

file is required if it is not in working directory. By default function seeks for the

.h5 file in the working directory.

Value

topline data.frame with row.names as cell line names and column medcor containing

values for correlation between set of samples from case_id and cell lines.

See Also

runsRGES

Examples

```
#load data.frame with samples included in the OCTAD database
phenoDF=get_ExperimentHub_data('EH7274')
HCC_primary=subset(phenoDF,cancer=='liver hepatocellular carcinoma'&
sample.type == 'primary')  #select data
case_id=HCC_primary$sample.id  #select cases
cell_line_computed=computeCellLine(case_id=case_id,source='octad.small')
```

computeRefTissue

Compute correlating reference control samples.

Description

Compute reference control samples from OCTAD database using precomputed EncoderDF models.

Usage

```
computeRefTissue(case_id = NULL, adjacent = FALSE, source = "octad",
    n_varGenes = 500, method = c("varGenes", 'random'), expSet = NULL,
    control_size = length(case_id),
    outputFolder = NULL, cor_cutoff = "0", output = TRUE)
```

Arguments

case_id vector of cases used to compute references.

source by default set octad to use autoencoder results for computation. Any other input

like 'side' is expSet defined by users.

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adjacent by default set to FALSE. If TRUE, only tissue with sample.type 'adjacent' from

phenoDF would be used instead of 'normal'.

expSet input for expression matrix. By default NULL, since autoencoder results are

used.

n_varGenes number of genes used to select control cases.

method one of two options is avaliable. random will take a random number of samples

from control subset and varGenes (default) will select control samples based on

distance between cases and selected samples.

control_size number of control samples to be selected.

outputFolder path to output folder. By default, the function produces result files in working

directory.

cor_cutoff cut-off for correlation values, by default cor_cutoff='0'.

output if TRUE, two output files are produced.

Value

Return

control_id a vector of an appropriate set of control samples.

Besides, if output=TRUE, two files are created in the working directory:

case_normal_corMatrix.csv

contains pairwise correlation between case samples vs control samples.

case_normal_median_cor.csv

contains median correlation values with case samples for returned control samples.

See Also

diffExp.

Examples

```
#select data
#load data.frame with samples included in the OCTAD database
phenoDF=get_ExperimentHub_data('EH7274')
HCC_primary=subset(phenoDF,cancer=='Liver Hepatocellular Carcinoma'&
sample.type == 'primary'&data.source == 'TCGA')
#select cases
case_id=HCC_primary$sample.id
#computing reference tissue, by default using small autoEncoder,
#but can use custom expression set,
#by default output=TRUE and outputFolder option is empty,
#which creates control corMatrix.csv to working directory
control_id=computeRefTissue(case_id,output=TRUE,
expSet = "octad",control_size = 50)
```

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diffExp Compute differential expression	diffExp	Compute differential expression	
---	---------	---------------------------------	--

Description

Compute differential expression for case vs control samples. Will produce the file computedEmpGenes.csv listing empirically differentially expressed genes used for RNA-Seq normalization.

Usage

Arguments

annotate

case_id	vector of cases used for differential expression.	
control_id	vector of controls used for differential expression.	
source	the file for the octad expression matrix. By default, set to octad.small to use only 978 landmark genes profiled in LINCS database. Use octad.whole option to compute DE on the whole transcriptome octad.counts.and.tpm.h5 file. The file should be present in the working directory or the whole path should be included. If source is set to 'side', the expSet matrix is estimated.	
expSet	input expression matrix. By default set to NULL since the expSet is created based on cases, controls and source file.	
file	if expSet='octad.whole', source path to expSet='octad.counts.and.tpm.h5' file is required if it is not in working directory. By default function seeks for the .h5 file in the working directory.	
normalize_samples		
	if TRUE, RUVSeq normalization is applied to either EdgeR or DESeq. No normalization needed for limma+voom.	
k	eiter $k=1$ (by default), $k=2$ or $k=3$, number of factors used in model matrix construction in RUVSeq normalization if normalize_samples=TRUE.	
n_topGenes	number of empiricaly differentially expressed genes estimated for RUVSeq normalization. Default is 5000.	
DE_method	edgeR, DESeq2, limma or wilcox DE analysis.	
output	if TRUE, output files is produced.	
outputFolder	path to output folder. By default, the function produces result files in working directory.	

of the custom input contain ensembl gene ids.

if TRUE, annotation by ENSEMBL gene is performed. If TRUE, make sure row.names

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Value

res data.frame with list of differentially expressed genes.

computedEmpGenes.csv

data.frame listing empiricaly differentially expressed genes used for RNA-Seq normalization.

See Also

```
computeRefTissue, runsRGES.
```

Examples

```
#load data.frame with samples included in the OCTAD database
phenoDF=get_ExperimentHub_data('EH7274')
HCC_primary=subset(phenoDF, cancer=='liver hepatocellular carcinoma'&
sample.type == 'primary')  #select data
case_id=HCC_primary$sample.id  #select cases
HCC_adjacent=subset(phenoDF, cancer=='liver hepatocellular carcinoma'&
sample.type == 'adjacent'&data.source == 'TCGA')  #select data
control_id=HCC_adjacent$sample.id  #select cases
res=diffExp(case_id,control_id,source='octad.small',output=FALSE)
```

loadOctadCounts

Load octad expression data

Description

Create TPM or count expression matrix for the selected samples from OCTAD.

Usage

```
loadOctadCounts(sample_vector='',type='tpm',file='')
```

Arguments

sample_vector vector of samples to be selected. Use phenoDF object for sample id selection.

type either tpm (default) or counts to be returned.
file full path to octad.counts.and.tpm.h5 file.

Value

exprData matrix with either log2 corrected counts or tmp matrix for selected samples.

See Also

diffExp.

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Examples

```
#load data.frame with samples included in the OCTAD database
phenoDF=get_ExperimentHub_data('EH7274')
#load expression data for raw counts or tpm values.
HCC_primary=subset(phenoDF, cancer=='liver hepatocellular carcinoma'&
sample.type == 'primary') #select data
#case_id=HCC_primary$sample.id #select cases
#expression_tmp=loadOctadCounts(case_id,type='tpm',
#file='octad.counts.and.tpm.h5')
#expression_log2=loadOctadCounts(case_id,type='counts',
#file='octad.counts.and.tpm.h5')
```

octad

Open Cancer TherApeutic Discovery (OCTAD) database package

Description

Open Cancer TherApeutic Discovery (OCTAD) package implies sRGES approach for the drug discovery. The essential idea is to identify drugs that reverse the gene expression signature of a disease by tamping down over-expressed genes and stimulating weakly expressed ones. The following package contains all required precomputed data for whole OCTAD pipeline computation.

Details

The main functions are:

- computeRefTissue Compute reference control samples from OCTAD database using precomputed EncoderDF models.
- diffExp Compute differential expression for case vs control samples. Will produce the file computedEmpGenes.csv listing empirically differentially expressed genes used for RNA-Seq normalization.
- runsRGES Compute sRGES, a score indicating the reveral potency of each drug. It first computes RGES (Reverse Gene Expression Score) for individual instances and then summarizes RGES of invididual drugs (one drug may have multiple instances under different treatment conditions).
- computeCellLine Compute Correlation between cell lines and vector of case ids.
- topLineEval Evaluate predictions using pharmacogenomics data. Given a cell line, the
 function computes the correlation between sRGES and drug sensitivity data taken from CTRP.
 A higher correlation means a better prediction. The cell line could be computed from computeCellLine.
- octadDrugEnrichment Perform enrichment analysis of drug hits based on chemical structures, drug-targets, and pharmacological classifications. An enrichment score calculated using ssGSEA and a p-value computed through a permutation test are provided.

For detailed information on usage, see the package vignette, by typing vignette('octad'), or the workflow linked to on the first page of the vignette.

The code can be viewed at the GitHub repository, which also lists the contributor code of conduct:

https://github.com/Bin-Chen-Lab/OCTAD

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References

Zeng, B., Glicksberg, B.S., Newbury, P., Chekalin, E., Xing, J., Liu, K., Wen, A., Chow, C. and Chen, B., 2021. OCTAD: an open workspace for virtually screening therapeutics targeting precise cancer patient groups using gene expression features. Nature protocols, 16(2), pp.728-753. https://www.nature.com/articles/s41596-020-00430-z_PACKAGE package

octadDrugEnrichment

Compute Drug enrichment

Description

Perform enrichment analysis of drug hits based on chemical structures, drug-targets, and pharma-cological classifications. An enrichment score calculated using ssGSEA and a p-value computed through a permutation test are provided.

Usage

```
octadDrugEnrichment(sRGES = NULL, target_type = "chembl_targets",
enrichFolder = "enrichFolder", outputFolder = NULL, outputRank = FALSE)
```

Arguments

target_type

sRGES sRGES data frame produced by runsRGES.

one or several of 'chembl_targets', 'mesh', 'ChemCluster' databases se-

lected. By deafult only 'chembl_targets' will be used.

enrichFolder folder to store output.

outputFolder path where to store enrichFolder, in case of NULL will be stored in work directory.

outputRank output detailed rank if TRUE, write sRGES for selected target as vcf.

Value

Following files are created: enriched_*_targets.csv and top_enriched_*_*_targets.pdf. In the case of chemical structural analysis, additional files are created: *drugstructureClusters.csv and *misc.csv. The results provide useful information for following candidate selection and experimental design. For example, if two structurally similar drugs are both predicted as top hits, the chance of each drug as a true positive is high.

exprData matrix with either log2 corrected counts or tmp matrix for selected samples.

See Also

runsRGES

Examples

```
data("sRGES_example",package='octad') #load example sRGES
#run drug enrichment
octadDrugEnrichment(sRGES = sRGES_example, target_type = c('chembl_targets'))
```

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res_example	Differential expression example for HCC vs adjacent liver tissue computed in diffExp() function
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Description

Differential expression example for HCC vs adjacent liver tissue computed in diffExp() function

Usage

```
data(res_example)
```

Format

A data.frame with 963 rows and 18 variables:

identifier Ensg ID

log2FoldChange Log2 fold-change

logCPM log CPM value

LR LR value

pvalue p.value

padj FDR

tax_id taxon id

GeneID Gene id

Locus Tag Locus tag

chromosome Chromosome

map_location Chromosome location

description Full gene name

type type of gene

Symbol_autho HGNC symbol

other Gene function

Details

```
To generate this dataset use the following code from the octad package #load data.frame with samples included in the OCTAD database. phenoDF=.eh[['EH7274']]
```

```
phenobi - .engl Li
```

#select data

HCC_primary=subset(phenoDF, cancer=='liver hepatocellular carcinoma'&sample.type ==
'primary')

#select cases

case_id=HCC_primary\$sample.id

control_id=subset(phenoDF,biopsy.site=='LIVER'&sample.type=='normal')\$sample.id[1:50]
res=diffExp(case_id,control_id,source='octad.small',output=FALSE)

10 runsRGES

Description

Compute sRGES, a score indicating the reveral potency of each drug. It first computes RGES (Reverse Gene Expression Score) for individual instances and then summarizes RGES of invididual drugs (one drug may have multiple instances under different treatment conditions).

Usage

```
runsRGES(dz_signature=NULL,choose_fda_drugs = FALSE,max_gene_size=500,
cells=NULL,output=FALSE,outputFolder='',weight_cell_line=NULL,permutations=10000)
```

Arguments

dz_signature disease signature. Make sure input data frame has a gene Symbol column, otherwise an error is produced. It must be an UPPERCASE gene symbol.

choose_fda_drugs

if TRUE, only FDA approved drugs are used.

max_gene_size maximum number of disease genes used for drug prediction. By default 50 for

each side (up/down).

cells cell ids in lincs_sig_info file used for prediction. By default, all cell lines are

used

weight_cell_line

by default NULL, if !NULL, an output object from computeCellLine is estimated

(see example).

permutations number of permutations, by default 10000.

output if TRUE, output files is produced.

outputFolder folder path to store drug results, by default write results to working directory.

Value

The function returns RGES data.frame

containing scores and p.values for every instance. data.frame contains drug id in pert_iname collumn, n contains the number of instances for this drug, mean, median and sd of sRGES RGES sores.

Besides, a number of additional files in the sourced directory:

dz_sig_used.csv

contains genes in the disease signature used for computing reverse gene expres-

sion scores.

sRGES.csv contains the same data as returned data.frame.

all__lincs_score.csv

includes information of RGES.

See Also

```
diffExp, octadDrugEnrichment, computeCellLine, topLineEval
```

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Examples

```
#load differential expression example for HCC
#vs adjacent liver tissue computed in diffExp() function
data("res_example",package='octad')
res_example=subset(res_example,abs(log2FoldChange)>1&padj<0.001)[1:10,]
#run sRGES computation
#sRGES=runsRGES(dz_signature=res_example,max_gene_size=100,permutations=1000,output=FALSE)</pre>
```

sRGES_example

Data of computed example sRGEs for HCC vs liver adjacent tissues on octad.small dataset

Description

Data of computed example sRGEs for HCC vs liver adjacent tissues on octad.small dataset

Usage

```
data(sRGES_example)
```

Format

A tibble with 12,442 rows and 6 variables:

```
pert_iname dbl Year price was recorded
mean mean sRGES for obtained drug if n>1
n times this drug was obtained
median median sRGES for drug if n>1
sd standart deviation for obtained drug if n>1
sRGES sRGES score of the drug
```

Details

```
To generate this dataset use the following code from the octad package load differential expression example for HCC vs adjacent liver tissue computed in diffExp() function from res_example. data('res_example',package='octad.db') res=subset(res_example,abs(log2FoldChange)>1&padj<0.001) #load example expression dataset sRGES=runsRGES(res,max_gene_size=100,permutations=1000,output=FALSE)
```

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topLineEval	Evaluate cell lines

Description

Evaluate predictions using pharmacogenomics data. Given a cell line, the function computes the correlation between sRGES and drug sensitivity data taken from CTRP. A higher correlation means a better prediction. The cell line could be computed from computeCellLine.

Usage

```
topLineEval(topline=NULL,mysRGES=NULL,outputFolder="")
```

Arguments

topline list of cell lines to be used for prediction.

mysRGES sRGES data.frame produced by runsRGES.

outputFolder Path to store results.

Value

```
The function produces 3 feils in the output directory:
```

*_ic50_insilico_validation.html correlation between drug IC50 and sGRES in a related cell line.

See Also

runsRGES

Examples

```
#load example sRGES computed by runsRGES() function for HCC
#vs liver adjacent tissues on octad.small dataset
data("sRGES_example",package='octad') #load example sRGES
#Pick up cell lines
topLineEval(topline = 'HEPG2',mysRGES = sRGES_example,outputFolder=tempdir())
```

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