Package: GRIN2 (via r-universe)

November 20, 2024

Title Genomic Random Interval (GRIN)

Version 1.0

Description Improved version of 'GRIN' software that streamlines its use in practice to analyze genomic lesion data, accelerate its computing, and expand its analysis capabilities to answer additional scientific questions including a rigorous evaluation of the association of genomic lesions with RNA expression. Pounds, Stan, et al. (2013) [<DOI:10.1093/bioinformatics/btt372>](https://doi.org/10.1093/bioinformatics/btt372).

License GPL $(>= 3)$

Encoding UTF-8

Roxygen list(markdown = TRUE)

RoxygenNote 7.3.2

Imports biomaRt, circlize, ComplexHeatmap, data.table, dplyr, EnsDb.Hsapiens.v75, ensembldb, forcats, GenomeInfoDb, ggplot2, graphics, grDevices, grid, gridGraphics, Gviz, magrittr, stats, stringr, survival, tibble, tidyselect, utils, writexl

Suggests knitr, rmarkdown

NeedsCompilation no

Maintainer Abdelrahman Elsayed <aelsayed@stjude.org>

Depends R $(>= 4.2.0)$

LazyData true

VignetteBuilder knitr

URL <https://github.com/abdel-elsayed87/GRIN2>

BugReports <https://github.com/abdel-elsayed87/GRIN2/issues>

Config/pak/sysreqs make libicu-dev libjpeg-dev libpng-dev libxml2-dev libssl-dev perl zlib1g-dev

Repository https://abdel-elsayed87.r-universe.dev

RemoteUrl https://github.com/abdel-elsayed87/grin2

RemoteRef HEAD

RemoteSha ce7ea96b1c13f63ff42cdd4e1aeb0ef3b15d1b88

Contents

Contents

Description

Function return box plots for expression data by lesion groups for selected number of genes based on a specified q-value of the kruskal-wallis test results.

Usage

alex.boxplots(out.dir, alex.data, alex.kw.results, q, gene.annotation)

Arguments

Value

Function return a PDF file with box plots for expression data by lesion groups for selected number of genes based on a specified q-value of the kruskal-wallis test results (one gene per page).

Author(s)

Abdelrahman Elsayed <abdelrahman.elsayed@stjude.org> and Stanley Pounds <stanley.pounds@stjude.org>

References

Cao, X., Elsayed, A. H., & Pounds, S. B. (2023) Statistical Methods Inspired by Challenges in Pediatric Cancer Multi-omics.

See Also

[alex.prep.lsn.expr\(\)](#page-5-1), [KW.hit.express\(\)](#page-36-1)

Examples

```
data(expr.data)
data(lesion.data)
data(hg19.gene.annotation)
```
prepare expression, lesion data and return the set of genes with both types of data available # ordered by gene IDs in rows and patient IDs in columns: alex.data=alex.prep.lsn.expr(expr.data, lesion.data,

hg19.gene.annotation, min.expr=5, min.pts.lsn=5)

run KW test for association between lesion groups and expression level of the same gene: alex.kw.results=KW.hit.express(alex.data, hg19.gene.annotation, min.grp.size=5)

return boxplots for a list of top significant genes to a pre-specified folder using 'out.dir': dir.create(resultsFolder <- file.path(tempdir(), "temp.out"))

```
boxplots=alex.boxplots(out.dir=resultsFolder,
                       alex.data, alex.kw.results,
                       1e-15, hg19.gene.annotation)
```

```
unlink(resultsFolder, recursive = TRUE)
```
alex.pathway *Associate Lesions with Expression Data on the Pathway Level*

Description

Function compute the distance between subjects in the dataset based on the lesions that affect different genes assigned to the pathway of interest and return two panels of lesion and expression data of ordered subjects based on the computed distances.

Usage

```
alex.pathway(alex.data, lsn.data, pathways, selected.pathway)
```
Arguments

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Details

Function compute the distance between subjects in th dataset based on lesions affecting different genes assigned to the pathway of interest and return two panels of lesion and expression data of ordered subjects based on the computed distances. Function also return a data.frame with lesion and expression data of the pathway genes ordered based on the hierarchical clustering analysis (same order of the subjects in the lesion and expression panels of the figure).

Value

Function will return two panels figure of lesion and expression data of ordered subjects based on the computed distances of lesions in all genes assigned to the pathway of interest. The function will also return:

ordered.path.data

data.frame with lesion and expression data of the pathway genes ordered based on the hiearchial clustering analysis (same order of the subjects in the lesion and expression panels of the figure).

Author(s)

Abdelrahman Elsayed <abdelrahman.elsayed@stjude.org> and Stanley Pounds <stanley.pounds@stjude.org>

See Also

[alex.prep.lsn.expr\(\)](#page-5-1), [stats::hclust\(\)](#page-0-0)

Examples

```
data(expr.data)
data(lesion.data)
data(hg19.gene.annotation)
data(pathways)
```
prepare expression, lesion data and return the set of genes with both types of data available # ordered by gene IDs in rows and patient IDs in columns: alex.data=alex.prep.lsn.expr(expr.data, lesion.data,

> hg19.gene.annotation, min.expr=5, min.pts.lsn=5)

use lesions in all genes assigned to the jak_pathway as an example pathway: alex.path=alex.pathway(alex.data, lesion.data, pathways, "Jak_Pathway") # extract expression and lesion data (same subjects order in the figure) alex.path

alex.prep.lsn.expr *Prepare Lesion and Expression Data for Kruskal-Wallis Test*

Description

The function prepares lesion and expression data matrices for the KW.hit.express function that runs the kruskal-Wallis test for the association between lesion groups and expression level of each gene with available lesion and expression data.

Usage

```
alex.prep.lsn.expr(
  expr.mtx,
  lsn.data,
  gene.annotation,
 min.expr = NULL,min.pts.lsn = NULL
)
```
Arguments

Details

The function use prep.lsn.type.matrix function to prepare the lesion matrix that has each gene represented in one row with all lesion types included. Next, the function will prepare lesion and expression data matrices for the KW.hit.express function that runs the kruskal-Wallis test. It only

alex.waterfall.plot 7

keep genes with both lesion and expression data with rows ordered by ensembl ID and columns ordered by patient's ID.

Value

A list with the following components:

Author(s)

Abdelrahman Elsayed <abdelrahman.elsayed@stjude.org> and Stanley Pounds <stanley.pounds@stjude.org>

References

Cao, X., Elsayed, A. H., & Pounds, S. B. (2023). Statistical Methods Inspired by Challenges in Pediatric Cancer Multi-omics.

See Also

[KW.hit.express\(\)](#page-36-1)

Examples

```
data(expr.data)
data(lesion.data)
data(hg19.gene.annotation)
```

```
# prepare expression, lesion data and return the set of genes with both types of data available
# ordered by gene IDs in rows and patient IDs in columns:
alex.data=alex.prep.lsn.expr(expr.data, lesion.data,
                             hg19.gene.annotation, min.expr=1,
                             min.pts.lsn=5)
```
alex.waterfall.plot *Prepare Waterfall Plot of Lesion and Expression Data*

Description

Function return a waterfall plot for expression data by lesion groups of a selected gene.

Usage

```
alex.waterfall.plot(waterfall.prep, lsn.data, lsn.clrs = NULL, delta = 0.5)
```
Arguments

Details

Function return a waterfall plot for expression data by lesion groups of a selected gene. This plot offers a side by side graphical representation of lesion and expression data for each patient where lesion groups are ordered alphabetically. For each lesion category, expression data is ordered from the lowest to the highest with patient with the median expression of the gene in the middle of the panel.

Value

Function return a waterfall plot for expression data by lesion groups of a selected gene.

Author(s)

Abdelrahman Elsayed <abdelrahman.elsayed@stjude.org> and Stanley Pounds <stanley.pounds@stjude.org>

References

Cao, X., Elsayed, A. H., & Pounds, S. B. (2023). Statistical Methods Inspired by Challenges in Pediatric Cancer Multi-omics.

See Also

[alex.prep.lsn.expr\(\)](#page-5-1), [KW.hit.express\(\)](#page-36-1), [alex.waterfall.prep\(\)](#page-8-1)

Examples

```
data(expr.data)
data(lesion.data)
data(hg19.gene.annotation)
```
prepare expression, lesion data and return the set of genes with both types of data available # ordered by gene IDs in rows and patient IDs in columns:

```
alex.data=alex.prep.lsn.expr(expr.data, lesion.data,
                             hg19.gene.annotation, min.expr=1, min.pts.lsn=5)
```
run KW test for association between lesion groups and expression level of the same gene: alex.kw.results=KW.hit.express(alex.data, hg19.gene.annotation, min.grp.size=5)

To prepare lesion and expression data for a waterfall plot (WT1 gene): WT1.waterfall.prep=alex.waterfall.prep(alex.data, alex.kw.results, "WT1", lesion.data)

waterfall plot of WT1 gene: WT1.waterfall.plot=alex.waterfall.plot(WT1.waterfall.prep, lesion.data)

alex.waterfall.prep *Prepare Lesion and Expression Data for Waterfall Plots*

Description

Function prepares lesion and expression data of a selected gene for the alex.waterfall.plot function.

Usage

alex.waterfall.prep(alex.data, alex.kw.results, gene, lsn.data)

Arguments

Details

Function prepares lesion and expression data of a selected gene for the alex.waterfall.plot function. It return a data table with patient ID, lesion types that affect each patient if any and expression level of the gene of interest. It also extract the kruskal-wallis test result and all lesions that affect the gene of interest.

Value

A list of four components:

Author(s)

Abdelrahman Elsayed <abdelrahman.elsayed@stjude.org> and Stanley Pounds <stanley.pounds@stjude.org>

References

Cao, X., Elsayed, A. H., & Pounds, S. B. (2023). Statistical Methods Inspired by Challenges in Pediatric Cancer Multi-omics.

See Also

[alex.prep.lsn.expr\(\)](#page-5-1), [KW.hit.express\(\)](#page-36-1)

Examples

data(expr.data) data(lesion.data) data(hg19.gene.annotation)

prepare expression, lesion data and return the set of genes with both types of data available # ordered by gene IDs in rows and patient IDs in columns: alex.data=alex.prep.lsn.expr(expr.data, lesion.data,

hg19.gene.annotation, min.expr=1, min.pts.lsn=5)

run KW test for association between lesion groups and expression level of the same gene: alex.kw.results=KW.hit.express(alex.data, hg19.gene.annotation, min.grp.size=5)

To prepare lesion and expression data for a waterfall plot (WT1 gene): WT1.waterfall.prep=alex.waterfall.prep(alex.data, alex.kw.results, "WT1", lesion.data)

Description

Clinical data file showing demographic and clinical outcomes of 265 newly diagnosed T-cell Acute Lymphoblastic Leukemia (T-ALL) patients that was reported by Liu, Yu, et al. (2017).

Usage

clin.data

Format

clin.data: A data frame with 265 rows and 11 columns: ID Patient identifier Sex Patient gender Race Patient race Age Days Patient age in days WBC White Blood Cell (WBC) count MRD29 Minimal Residual Disease (MRD) percentage MRD.binary MRD as a categorical variable (0 if MRD \leq =0.1 or 1 if MRD $>$ 0.1) os.time Overall survival time in years (time between diagnosis and either the last follow-up or death) os.censor Survival status (0=alive at the last follow-up or, 1=dead) efs.time Event-free survival time in years

efs.censor Event indicator (0=censored without event or, 1=event)

Source

Data was extracted from the supplementary material tables of the published Liu, Yu, et al. (2017) manuscript <https://www.nature.com/articles/ng.3909#Sec27> and the publicly available clinical data on TARGET database. The two files were merged and selected list of variables were kept in the final clinical data file.

compute.gw.coordinates

Compute Genome-wide Coordinates

Description

The function assign plotting coordinates necessary for the genome-wide lesion plot.

Usage

compute.gw.coordinates(grin.res, scl = 1e+06)

Arguments

Details

The function divides each chromosome into multiple units based on the specified scl value. In addition, it orders and adds two columns x.start and x.end to the chromosme size file (x.start for chr2 is equal to x.end of chr1). Function also adds x.start and x.end columns to lesion and gene annotation data files (x.start is the start position of the lesion or the gene divided by scl and x.end is the end position of the lesion or the gene divided by scl taking into consideration that the start position of the chromosomes is added consecutively based on the chromosomes length).

Value

Function return a list of GRIN results with the following changes to allow adding genome-wide plotting coordinates:

count.hits 13

Author(s)

Stanley Pounds <stanley.pounds@stjude.org>

References

Cao, X., Elsayed, A. H., & Pounds, S. B. (2023). Statistical Methods Inspired by Challenges in Pediatric Cancer Multi-omics.

See Also

[grin.stats\(\)](#page-29-1)

Examples

data(lesion.data) data(hg19.gene.annotation) data(hg19.chrom.size)

Run GRIN model using grin.stats function grin.results=grin.stats(lesion.data, hg19.gene.annotation, hg19.chrom.size)

assign genomewide coordinates and prepare the results for the genomewide.lsn.plot function genome.coord=compute.gw.coordinates(grin.results)

count.hits *Count Gene Lesion Hits*

Description

The function computes the number of hits affecting each gene by lesion category. It also compute the number of subjects with a hit in each annotated gene by lesion category as well.

Usage

count.hits(ov.data)

Arguments

ov.data a list of six data.frames that represent the output results of the find.gene.lsn.overlaps function.

Details

The function use the output of the find.gene.lsn.overlaps function and return the number of unique subjects affected by each lesion category in the provided list of annotated genes and regulatory features (nsubj stats). It also count the number of hits affecting each loci per lesion category (nhits stats). For example, if NOTCH1 gene was found affected by three different mutations in the same subject, this patient will be considered as one subject in the nsubj stats but in the nhits stats for this event will be counted as 3 mutations that affect NOTCH1 gene.

Value

A list with the following components:

Author(s)

Stanley Pounds <stanley.pounds@stjude.org>

References

Pounds, Stan, et al. (2013) A genomic random interval model for statistical analysis of genomic lesion data.

Cao, X., Elsayed, A. H., & Pounds, S. B. (2023). Statistical Methods Inspired by Challenges in Pediatric Cancer Multi-omics.

default.grin.colors 15

See Also

[prep.gene.lsn.data\(\)](#page-46-1), [find.gene.lsn.overlaps\(\)](#page-16-1)

Examples

```
data(lesion.data)
data(hg19.gene.annotation)
```
prepare gene and lesion data for later computations: prep.gene.lsn=prep.gene.lsn.data(lesion.data, hg19.gene.annotation)

```
# determine lesions that overlap each gene (locus):
gene.lsn.overlap=find.gene.lsn.overlaps(prep.gene.lsn)
```
count number of subjects affected by different types of lesions and number of hits that affect # each locus: count.nsubj.nhits=count.hits(gene.lsn.overlap)

default.grin.colors *Default GRIN Colors*

Description

Function assigns default colors for each lesion group in the whole set of GRIN plots.

Usage

```
default.grin.colors(lsn.types)
```
Arguments

lsn.types Unique lesion types as specified in the lesion data file.

Details

The function specifies 10 colors for different lesion types. If the number of lesion types is more than 10, the user will be asked to specify the colors manually.

Value

Function return a vector of colors assigned to each unique lesion type.

Author(s)

Stanley Pounds <stanley.pounds@stjude.org>

References

Pounds, Stan, et al. (2013) A genomic random interval model for statistical analysis of genomic lesion data.

Cao, X., Elsayed, A. H., & Pounds, S. B. (2023). Statistical Methods Inspired by Challenges in Pediatric Cancer Multi-omics.

Examples

data(lesion.data)

```
lsn.types=unique(lesion.data$lsn.type)
# assign colors for different lesion categories using default.grin.colors function:
default.grin.colors(lsn.types)
```
expr.data *Example T-ALL Dataset Gene Expression Data*

Description

Gene expression data file showing log2 normalized expression level of 420 genes (rows) in 265 newly diagnosed T-cell Acute Lymphoblastic Leukemia (T-ALL) patients in columns that was reported by Liu, Yu, et al. (2017).

Usage

expr.data

Format

expr.data:

A data frame with 420 rows and 265 columns:

gene Ensembl IDs of the list of 420 genes included in the dataset ...

Source

Data was extracted from the supplementary material tables of the published Liu, Yu, et al. (2017) manuscript <https://www.nature.com/articles/ng.3909#Sec27>

find.gene.lsn.overlaps

Find Gene Lesion Overlaps

Description

The function use the output of the prep.gene.lsn.data function to find lesion-gene overlaps.

Usage

```
find.gene.lsn.overlaps(gl.data)
```
Arguments

gl.data a list of five data.frames that represent the output results of the prep.gene.lsn.data function.

Value

A list with the following components:

Author(s)

Stanley Pounds <stanley.pounds@stjude.org>

References

Pounds, Stan, et al. (2013) A genomic random interval model for statistical analysis of genomic lesion data.

Cao, X., Elsayed, A. H., & Pounds, S. B. (2023). Statistical Methods Inspired by Challenges in Pediatric Cancer Multi-omics.

See Also

[prep.gene.lsn.data\(\)](#page-46-1)

Examples

data(lesion.data) data(hg19.gene.annotation)

prepare gene and lesion data for later computations: prep.gene.lsn=prep.gene.lsn.data(lesion.data, hg19.gene.annotation)

determine lesions that overlap each gene (locus): gene.lsn.overlap=find.gene.lsn.overlaps(prep.gene.lsn)

genomewide.log10q.plot

Genomewide log10q Plot

Description

The function return a genome-wide plot based on -log(10) q-value of each of the evaluated annotated genes or lesion boundaries on each chromosome. The plot is lesion type specific (gain, loss, mutation, etc...).

Usage

```
genomewide.log10q.plot(
 grin.res,
  lsn.grps,
 lsn.colors = NULL,
 max.log10q = NULL)
```
Arguments

Value

The function return a genome-wide plot based on $-log(10)$ q-value of each of the evaluated annotated genes or lesion boundaries to be affected by a certain type of lesions (gain, loss, mutation, etc...).

Author(s)

Abdelrahman Elsayed <abdelrahman.elsayed@stjude.org> and Stanley Pounds <stanley.pounds@stjude.org>

See Also

```
grin.lsn.boundaries()
```
Examples

```
data(lesion.data)
data(hg19.gene.annotation)
data(hg19.chrom.size)
```

```
# This analysis is lesion type specific. So, user should first data extract data for a specific
# lesion group of interest for example gains from the lesion data file:
gain=lesion.data[lesion.data$lsn.type=="gain",]
# Return lesion boundaries for gains:
lsn.bound.gain=grin.lsn.boundaries(gain, hg19.chrom.size)
# Run GRIN analysis Using Lesion Boundaries as Markers Instead of the Gene Annotation File:
GRIN.results.gain.bound=grin.stats(gain, lsn.bound.gain, hg19.chrom.size)
# Return genomewide -log10q plot for association between lesion boundaries and gain:
genomewide.log10q.plot(GRIN.results.gain.bound, lsn.grps=c("gain"),
                       lsn.close = c("gain" = "red"), max.log10q = 10)# instead of lesion boundaries, users can also plot -log10q values for annotated genes using
# genes annotation data as a marker data file:
grin.results=grin.stats(lesion.data,
                        hg19.gene.annotation,
                        hg19.chrom.size)
```
genomewide.log10q.plot(grin.results, lsn.grps=c("gain"), lsn.colors=c("gain" = "red"), max.log10q = 10)

User can run this same analysis for other lesion types such as mutations and deletions.

genomewide.lsn.plot *Genome-wide Lesion Plot*

Description

Function return a genomewide lesion plot for all lesion types affecting different chromosomes.

Usage

```
genomewide.lsn.plot(
  grin.res,
  ordered = FALSE,
 pt.order = NULL,
  lsn.colors = NULL,
  max.log10q = NULL)
```
Arguments

Details

The function use the genome-wide plotting coordinates obtained from the compute.gw.coordinates function and plot the whole set of lesions affecting subjects included in the dataset in the middle panel of the figure. Two additional side panels show the number of affected subjects and -log10 q value of each locus to be affected by all different types of lesions.

Value

The function return a genome-wide lesion plot (all chromosomes) in the middle panel. For each locus, Panel on the left shows -log10 q value and the Panel on the right show the number of subjects affected by all different types of lesions color coded by lesion category.

Author(s)

Abdelrahman Elsayed <abdelrahman.elsayed@stjude.org> and Stanley Pounds <stanley.pounds@stjude.org>

References

Cao, X., Elsayed, A. H., & Pounds, S. B. (2023). Statistical Methods Inspired by Challenges in Pediatric Cancer Multi-omics.

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See Also

[compute.gw.coordinates\(\)](#page-11-1)

Examples

```
data(lesion.data)
data(hg19.gene.annotation)
data(hg19.chrom.size)
```

```
# Run GRIN model using grin.stats function
grin.results=grin.stats(lesion.data,
                        hg19.gene.annotation,
                        hg19.chrom.size)
```

```
# prepare the genomewide lesion plot using genomewide.lsn.plot function with patient IDs ordered
# alphabetically:
```

```
genomewide.plot=genomewide.lsn.plot(grin.results, max.log10q=50)
```
To pass certain patients order to the genomewide.lsn.plot function, the user should specify # a certain patients order using the pt.order argument.

get.chrom.length *Get Chromosome Length*

Description

Retrieve chromosome size data from chr.info txt files available on the UCSC genome browser based on the user specified genome assembly.

Usage

```
get.chrom.length(genome.assembly)
```
Arguments

```
genome.assembly
```
User can specify one of four supported genome assemblies that include "Human_GRCh38", "Human_GRCh37", "Mouse_HGCm39" and "Mouse_HGCm38".

Details

Based on the genome assembly specified by the user, the function will directly retrieve chromosome size data from chr.info txt file available on the UCSC genome browser.

Value

A data table with the following two columns:

Author(s)

Abdelrahman Elsayed <abdelrahman.elsayed@stjude.org> and Stanley Pounds <stanley.pounds@stjude.org>

References

Cao, X., Elsayed, A. H., & Pounds, S. B. (2023). Statistical Methods Inspired by Challenges in Pediatric Cancer Multi-omics.

See Also

[circlize::read.chromInfo\(\)](#page-0-0)

Examples

To retreive chromosome size data for hg19 genome assembly: hg19.chrom.size=get.chrom.length("Human_GRCh37") # "Human_GRCh38" can be used to retreive chromosome size data for hg38 genome assembly.

get.ensembl.annotation

Get Ensembl Gene and Regulatory Features Annotation Data

Description

Function directly retrieve gene and regulatory features annotation data from Ensembl BioMart database based on the specified genome assembly.

Usage

get.ensembl.annotation(genome.assembly)

Arguments

```
genome.assembly
```
User can specify one of four genome assemblies that include "Human_GRCh38", "Human_GRCh37", "Mouse_HGCm39" and "Mouse_HGCm38".

Details

Based on the genome assembly specified by the user, the function will directly retrieve gene and regulatory features annotation data from ensembl BioMart database. Annotation data include enesembl ID, the chromosome on which the gene is located, gene start and gene end position, gene name, gene description, biotype, chromosome strand and chromosome band. Gene classes (biotypes) include protein coding genes, long noncoding RNAs (lncRNAs), microRNAs (miRNAs), small nuclear RNAs (snRNA), small nucleolar RNAs (snoRNA), immunoglobulins (IGs), T-cell receptors (TCRs) and pseudogens. Regulatory features data retrieved from Ensembl regulatory build are categorized in 6 classes that include promoters, promoter flanking regions, predicted enhancers, CTCF binding sites, transcription factor (TF) binding sites and the open chromatin regions.Ensembl

first imports publicly available data from different large epigenomic consortia such as ENCODE, Roadmap Epigenomics and Blueprint. All high-throughput sequencing data sets are then uniformly processed using the Ensembl Regulation Sequence Analysis (ERSA) pipeline to generate signal tracks for enriched regions also referred to as annotated features or peaks. Segmentation data provide information about promoter, promoter flanking regions, enhancers and CTCF binding sites. If TF binding probability is >0 in areas outside previously mentioned regions, it takes a label of TF binding site. If any open chromatin region did not overlap with the above features, it takes a label of unannotated open chromatin. users will also have the chance to use a list of experimentally verified enhancers/transcription start sites (TSS) using the CAGE (Cap Analysis of Gene Expression) experiment on a multitude of different primary cells and tissues from the Functional Annotation of the Mouse/Mammalian Genome (FANTOM5) project.

Value

A list of three components:

gene.annotation

A 9 columns data.frame with gene annotation data that include enesembl ID, chromosome, gene start and gene end position, gene name, gene description, biotype, chromosome strand, and chromosome band.

reg.annotation.predicted

A 5 columns data.frame with regulatory features annotation data directly retreived from Ensembl regulatory build that include enesembl ID, chromosome, description(promoter, enhancer, etc..), feature start and end positions.

reg.annotation.validated

A 5 columns data.frame with regulatory features annotation data for experimentally verified features retreived from FANTOM5 project that include feature ID, chromosome, description(enhancer, transcription start site (TSS)), feature start and end positions.

Author(s)

Abdelrahman Elsayed <abdelrahman.elsayed@stjude.org> and Stanley Pounds <stanley.pounds@stjude.org>

References

Cao, X., Elsayed, A. H., & Pounds, S. B. (2023). Statistical Methods Inspired by Challenges in Pediatric Cancer Multi-omics.

Zerbino, Daniel R., et al. (2015). The ensembl regulatory build.

Kinsella, Rhoda J., et al. (2011). Ensembl BioMarts: a hub for data retrieval across taxonomic space.

See Also

[biomaRt::useEnsembl\(\)](#page-0-0), [biomaRt::getBM\(\)](#page-0-0)

Examples

Retrieve annotation data for human hg19 genome assembly: hg19.ann=get.ensembl.annotation("Human_GRCh37") # gene annotation data: hg19.gene.annotation=hg19.ann\$gene.annotation # regulatory features annotation data retrieved from Ensembl regulatory build: hg19.reg.annotation=hg19.ann\$reg.annotation.predicted # regulatory features annotation data retrieved from FANTOM5 project: hg19.fantom.annotation=hg19.ann\$reg.annotation.validated

"Human_GRCh38" can be used instead of "Human_GRCh37" to retrieve gene and regulatory features # annotation data for human hg38 genome assembly.

grin.assoc.lsn.outcome

Associate Lesions with Clinical Outcomes

Description

The function run association analysis between the binary lesion matrix (output of prep.binary.lsn.mtx function) and clinical outcomes of interest such as Minimal Residual Disease (MRD), Event-free Survival (EFS) and Overall Survival (OS), etc...

Usage

```
grin.assoc.lsn.outcome(
  lsn.mtx,
  clin.data,
  annotation.data,
  clinvars,
  covariate = NULL
)
```
Arguments

```
lsn.mtx Binary lesion matrix in which each type of lesions affecting certain gene is rep-
                  resented in a separate row for example ENSG00000148400_gain. If the gene
                  is affected by this specific type of lesion, patient entry will be coded as 1 or 0
                  otherwisw. This matrix is the output of the prep.binary.lsn.mtx function.
clin.data Clinical data table in which the first column "ID" should has the patient ID.
annotation.data
                  Gene annotation data either provided by the user or retrieved from ensembl
                  BioMart database using get.ensembl.annotation function included in the GRIN2.0
                  library. Data.frame should has four columns: "gene" which is the ensembl ID
                  of annotated genes, "chrom" which is the chromosome on which the gene is
                  located, "loc.start" which is the gene start position, and "loc.end" the gene end
                  position.
```


Details

The function run association analysis between the binary lesion matrix in which each type of lesions affecting certain gene is represented in a separate row (output of prep.binary.lsn.mtx function) and clinical outcomes.Function will run logistic regression models for association between each genelesion pair with numeric variables such as MRD that should be coded as 0 if the patient is MRDnegative and 1 if the patient is MRD positive. Function will also run COX-Proportional hazard models for association between lesions and survival objects such as Event-free survival (EFS) and oveall survival (OS). EFS and OS should be first coded as survival objects using Surv function and added as new columns to the clinical data file. If specified, the models can be also adjusted for one or a group of covariates such as risk group assignment, gender, age, etc...

Value

Function returns a results table that has gene annotation data, and multiple columns showing results of the logistic regression model for association with binary variables such as MRD that include odds.ratio, lower and upper 95 confidence interval (CI), model p and FDR adjusted q values, in addition to the number of patients with/without lesion who experienced or did not experience the event. Results table will also include results of COXPH models for association between lesions with survival variables such as EFS, OS that include COXPH hazard ratio, lower and upper 95 CI, model p and FDR adjusted q values, in addition to the number of patients with/without the lesion who experienced or did not experience the event.

Author(s)

Abdelrahman Elsayed <abdelrahman.elsayed@stjude.org> and Stanley Pounds <stanley.pounds@stjude.org>

References

Andersen, P. and Gill, R. (1982). Cox's regression model for counting processes, a large sample study.

Therneau, T., Grambsch, P. (2000) Modeling Survival Data: Extending the Cox Model.

Dobson, A. J. (1990) An Introduction to Generalized Linear Models.

See Also

[stats::glm\(\)](#page-0-0), [survival::coxph\(\)](#page-0-0)

Examples

data(lesion.data) data(hg19.gene.annotation) data(clin.data)

```
# prepare lesion data and find gene lesion overlaps:
gene.lsn=prep.gene.lsn.data(lesion.data, hg19.gene.annotation)
gene.lsn.overlap= find.gene.lsn.overlaps(gene.lsn)
# Prepare a binary lesion matrix for genes affected by a certain type of lesion in at least
# 5 subjects using prep.binary.lsn.mtx function:
lsn.binary.mtx=prep.binary.lsn.mtx(gene.lsn.overlap, min.ngrp=5)
# Prepare EFS and OS survival objects and add two new columns to the clinical data file:
library(survival)
clin.data$EFS <- Surv(clin.data$efs.time, clin.data$efs.censor)
clin.data$OS <- Surv(clin.data$os.time, clin.data$os.censor)
# define clinical outcome variables to be included in the analysis:
clinvars=c("MRD.binary", "EFS", "OS")
# Run association analysis between lesions in the binary lesion matrix and clinical variables
# in the clinvars object:
assc.outcomes=grin.assoc.lsn.outcome(lsn.binary.mtx,
                                     clin.data,
                                     hg19.gene.annotation,
                                     clinvars)
# to adjust the models for one or a group of covariates, user can specify one or a group
# of covariates using the 'covariate' argument.
```

```
grin.barplt GRIN Bar Plot
```
Description

Function return a stacked bar plot with number of patients affected by all different types of lesions in a pre-specified list of genes of interest.

Usage

```
grin.barplt(grin.res, count.genes, lsn.colors = NULL)
```
Arguments

Details

Function will use the input list of gene names and extract the number of patients affected by all different types of lesions in those genes from the GRIN results table (output of the grin.stats function).

Value

Function return a stacked bar plot with number of patients affected by all different types of lesions in the pre-specified list of genes of interest.

Author(s)

Abdelrahman Elsayed <abdelrahman.elsayed@stjude.org> and Stanley Pounds <stanley.pounds@stjude.org>

See Also

[grin.stats\(\)](#page-29-1)

Examples

```
data(lesion.data)
data(hg19.gene.annotation)
data(hg19.chrom.size)
# run GRIN analysis using grin.stats function
grin.results=grin.stats(lesion.data,
                        hg19.gene.annotation,
                        hg19.chrom.size)
# specify a list of genes to be included in the bar plot (driver genes)
count.genes=as.vector(c("TAL1", "FBXW7", "PTEN", "IRF8","NRAS",
                        "BCL11B", "MYB", "LEF1","RB1", "MLLT3", "EZH2", "ETV6", "CTCF",
                        "JAK1", "KRAS", "RUNX1", "IKZF1", "KMT2A", "RPL11", "TCF7",
                        "WT1", "JAK2", "JAK3", "FLT3"))
# return the stacked barplot
grin.barplt(grin.results, count.genes)
```
grin.lsn.boundaries *GRIN Evaluate Lesion Boundaries*

Description

The function evaluates Copy number variations that include gain and deletions as boundaries based on unique lesion start and end positions. This analysis is lesion type specific and covers the entire genome.

Usage

grin.lsn.boundaries(lsn.data, chrom.size)

Arguments

Details

The function evaluates Copy number variations that include gain and deletions as boundaries and return a table of ordered boundaries based on the unique start and end positions of different lesions on each chromosome. If gains are splitted to gain and amplifications based on the log2Ratio value of the CNV segmentation file, the two categories can be included in the same analysis, same for homozygous and heterozygous deletions. Boundary will be the region between each unique start and end positions where large size lesions will be splitted into multiple boundaries based on other smaller size lesions that affect the same region in other patients if any. This analysis is meant to cover the entire genome, so regions without any annotated genes or regulatory features will be incuded will be assesed in the analysis. The first boundary for each chromosome will start from the first nucleotide base on the chromosome till the start position of the first lesion that affect the chromosome. Similarly, the last boundary will start from the end position of the last lesion that affect the chromosome till the last base on the chromosome.

Value

Function return a data.frame with five columns:

Author(s)

Abdelrahman Elsayed <abdelrahman.elsayed@stjude.org> and Stanley Pounds <stanley.pounds@stjude.org>

Examples

```
data(lesion.data)
data(hg19.chrom.size)
# This analysis is lesion type specific. So, user should first data extract data for a specific
```

```
# lesion group of interest for example gains from the lesion data file:
gain=lesion.data[lesion.data$lsn.type=="gain",]
# Return lesion boundaries for gains:
lsn.bound.gain=grin.lsn.boundaries(gain, hg19.chrom.size)
```
grin.oncoprint.mtx 29

Run GRIN analysis Using Lesion Boundaries markers Instead of the gene annotation file: GRIN.results.gain.bound=grin.stats(gain, lsn.bound.gain, hg19.chrom.size)

same analysis can be done for mutations, deletions and structural rearrangments.

grin.oncoprint.mtx *GRIN OncoPrint Matrix*

Description

Function use GRIN results table and prepare the lesion matrix that the user can pass to the oncoprint function from ComplexHeatmap package to geneate an OncoPrint for a selcted list of genes.

Usage

grin.oncoprint.mtx(grin.res, oncoprint.genes)

Arguments

grin.res GRIN results (output of the grin.stats function). oncoprint.genes

Vector of ensembl IDs for the selected list of genes to be added to the OncoPrint.

Details

Function will use the input list of ensembl IDs to prepare a data table of lesions that affect these genes (each row is a gene and each column is a patient ID). This lesion matrix is compatible and can be passed to oncoprint function in ComplexHeatmap library to prepare an OncoPrint for lesions in the selected list of genes.

Value

Function uses the output results of grin.stats function and return data table of lesions that affect a group of selected genes (each row is a gene and each column is a patient ID).

Author(s)

Abdelrahman Elsayed <abdelrahman.elsayed@stjude.org> and Stanley Pounds <stanley.pounds@stjude.org>

References

Cao, X., Elsayed, A. H., & Pounds, S. B. (2023). Statistical Methods Inspired by Challenges in Pediatric Cancer Multi-omics.

See Also

[grin.stats\(\)](#page-29-1)

Examples

```
data(lesion.data)
data(hg19.gene.annotation)
data(hg19.chrom.size)
# Run GRIN analysis using grin.stats function:
grin.results=grin.stats(lesion.data,
                        hg19.gene.annotation,
                        hg19.chrom.size)
# specify a list of genes to be included in the oncoprint (driver genes):
oncoprint.genes=as.vector(c("ENSG00000148400", "ENSG00000171862", "ENSG00000171843",
                            "ENSG00000156531", "ENSG00000162367", "ENSG00000096968",
                            "ENSG00000105639", "ENSG00000118513","ENSG00000102974",
                            "ENSG00000133703"))
# prepare the oncoprint lesion matrix:
oncoprint.mtx=grin.oncoprint.mtx(grin.results,
                                 oncoprint.genes)
# user can also specify a list of top significant genes in the GRIN constellation test:
# for example: select genes affected by two types of lesion with q2.nsubj<0.01:
genes.const = grin.results$gene.hits[grin.results$gene.hits$q2.nsubj < 0.01, ]
# get ensembl.ids for this list of genes
selected.genes=as.vector(genes.const$gene)
oncoprint.mtx.const=grin.oncoprint.mtx(grin.results,
                                       selected.genes)
```


grin.stats *GRIN Statistics Output*

Description

The function run the Genomic Random Interval (GRIN) analysis to determine whether a certain locus has an abundance of lesions or a constellation of multiple types of lesions that is statistically significant.

Usage

```
grin.stats(lsn.data, gene.data = NULL, chr.size = NULL, genome.version = NULL)
```
Arguments

lsn.data data.frame with lesion data prepared by the user in a GRIN compatible format. Object should has five columns that include "ID" with patient ID, "chrom" which is the chromosome on which the lesion is located, "loc.start" which is the lesion start position, "loc.end" the lesion end position and "lsn.type" which is the lesion category for example gain, loss, mutation, fusion, etc... For Single Nucleotide Variants (SNVs), loc.start will be the same as loc.end. For Copy

Details

The function run the Genomic Random Interval (GRIN) analysis and evaluates the probability of each gene locus to be affected by different types of lesions based on a convolution of independent but non-identical Bernoulli distributions to determine whether this locus has an abundance of lesions that is statistically significant.In addition, FDR-adjusted q value is computed for each locus based on Pounds & Cheng (2006) estimator of the proportion of tests with a true null (pi.hat). The function also evaluates if a certain locus is affected by a constellation of multiple types of lesions and return the GRIN results table.

Value

A list with the following components:

gene.hits data table of GRIN results that include gene annotation, number of subjects affected by each lesion type for example gain, loss, mutation, etc.., and number of hits affecting each locus. The GRIN results table will also include P and FDR adjusted q-values showing the probability of each locus of being affected by one or a constellation of multiple types of lesions.

lsn.data input lesion data

Author(s)

Stanley Pounds <stanley.pounds@stjude.org>

References

Pounds, Stan, et al. (2013) A genomic random interval model for statistical analysis of genomic lesion data.

Cao, X., Elsayed, A. H., & Pounds, S. B. (2023). Statistical Methods Inspired by Challenges in Pediatric Cancer Multi-omics.

See Also

[prep.gene.lsn.data\(\)](#page-46-1), [find.gene.lsn.overlaps\(\)](#page-16-1), [count.hits\(\)](#page-12-1), [prob.hits\(\)](#page-49-1)

Examples

```
data(lesion.data)
data(hg19.gene.annotation)
data(hg19.chrom.size)
```
if gene annotation and chromosome size files will be provided by the user: grin.results=grin.stats(lesion.data, hg19.gene.annotation, hg19.chrom.size)

to directly retrieve gene annotation and chromosome size files from Ensembl BioMart database, # and UCSC genome browsers using get.ensembl.annotation and get.chrom.length functions respectively, # users can select to specify certain genome assembly using the 'genome.version' argument: # "Human_GRCh37" can be used for the GRCH37 (hg19) genome assembly, and "Human_GRCh38" can be used # for the GRCH38 (hg38) genome assembly

grin.stats.lsn.plot *GRIN Statistics Lesions Plot*

Description

Function return a plot with all types of lesions that spans either a gene or regulatory feature of interest with GRIN statistics added.

Usage

grin.stats.lsn.plot(grin.res, feature = NULL, lsn.clrs = NULL, expand = 5e-04)

Arguments

Details

Function return a plot with all lesions that affect either a gene regulatory feature of interest. Top panel of the plot will has all different types of lesions affecting the loci color coded according to the figure legend. Lower panel of the plot has all the GRIN statistics of the feature that include number of subjects affected by each type of lesions, -log10 p, and –log10q values showing if the feature is significantly affected by the corresponding lesion category. This plot has no panel for transcripts table as regulatory features typically do not have this kind of information.

Value

Function return a plot with all types of lesions that spans either a gene or regulatory feature of interest in addition to the locus GRIN statistics without adding the transcripts panel.

Author(s)

Abdelrahman Elsayed <abdelrahman.elsayed@stjude.org> and Stanley Pounds <stanley.pounds@stjude.org>

References

Cao, X., Elsayed, A. H., & Pounds, S. B. (2023). Statistical Methods Inspired by Challenges in Pediatric Cancer Multi-omics.

See Also

[grin.stats\(\)](#page-29-1)

Examples

```
data(lesion.data)
data(hg19.gene.annotation)
data(hg19.chrom.size)
```
run GRIN analysis grin.results=grin.stats(lesion.data, hg19.gene.annotation, hg19.chrom.size)

```
# Plots showing different types of lesions and GRIN stats for a gene of interest (WT1):
grin.stats.lsn.plot(grin.results,
                     feature="ENSG00000184937")
```
same function can be used to plot lesion data and GRIN statistics of regulatory features # that typically do not have transcripts track to add to the plot.

hg19.chrom.size *Chromosome Length Data File*

Description

The file has the size of 22 autosomes in addition to X and Y chromosomes in base pairs directly retrieved from chr.info txt files available on the UCSC genome browser using get.chrom.length function and "Human-GRCh37" as a genome assembly option (hg19).

Usage

hg19.chrom.size

Format

hg19.chrom.size:

A data frame with 24 rows and 2 columns:

chrom The chromosome number.

size The chromosome length in base pairs.

Source

Chromosome size data directly retrieved from chr.info txt files available on the UCSC genome browser using get.chrom.length function and "Human-GRCh37" as a genome assembly option (hg19).

hg19.gene.annotation *Example Gene Annotation Data File*

Description

The file has an example annotation data of 420 genes (same set of genes in the gene expression data file) directly retrieved from Ensembl BioMart database using get.ensembl.annotation function and "Human-GRCh37" as a genome assembly option (hg19).

Usage

hg19.gene.annotation

Format

hg19.gene.annotation:

A data frame with 420 rows and 9 columns:

gene Column has the gene ensembl ID chrom The chromosome on which the gene is located

loc.start Gene start position in base pairs

loc.end Gene end position in base pairs

description Description of the gene name

gene.name Gene symbol

biotype Gene classes that include protein coding genes, long noncoding RNAs (lncRNAs), microRNAs (miRNAs), small nuclear RNAs (snRNA), small nucleolar RNAs (snoRNA), immunoglobulins (IGs), T-cell receptors (TCRs) and pseudogens.

chrom.strand The chromosome strand on which the gene is located forward (1) or reverse (-1). chrom.band The chromosome band on which the gene is located.

Source

Data was directly retrieved from Ensembl BioMart database using get.ensembl.annotation function and "Human-GRCh37" as a genome assembly option (hg19).

Description

The dataset has the start and end positions in base pairs of all 22 autosomes in addition to X and Y chromosome cytobands for Human-GRCh37 (hg19) genome assembly.

Usage

hg19_cytoband

Format

hg19_cytoband: A data frame with 862 rows and 5 columns: chrom The chromosome number. chromStart The cytoband start position on the chromosome in base pairs. chromEnd The cytoband end position on the chromosome in base pairs. name The cytoband name. gieStain The coloring scheme of the cytobands.

Source

The Chromosome cytobands data file was downloaded from the UCSC genome browser for GRCh37 genome assembly <https://hgdownload.soe.ucsc.edu/goldenPath/hg19/database/>.

hg38_cytoband *GRCh38 Chromosome Cytobands*

Description

The dataset has the start and end positions in base pairs of all 22 autosomes in addition to X and Y chromosome cytobands for Human-GRCh38 (hg38) genome assembly.

Usage

hg38_cytoband

Format

hg38_cytoband:

A data frame with 1,549 rows and 5 columns:

chrom The chromosome number.

chromStart The cytoband start position on the chromosome in base pairs.

chromEnd The cytoband end position on the chromosome in base pairs.

name The cytoband name.

gieStain The coloring scheme of the cytobands.

Source

The Chromosome cytobands data file was downloaded from the UCSC genome browser for GRCh38 genome assembly <https://hgdownload.soe.ucsc.edu/goldenPath/hg38/database/>.

Description

Function uses Kruskal-Wallis test to evaluate the association between lesion groups and expression level of the same corresponding gene.

Usage

KW.hit.express(alex.data, gene.annotation, min.grp.size = NULL)

Arguments

Details

The function uses the ensembl IDs in each row of the row.mtch file and run the Kruskal-Wallis test for association between lesion groups of the gene in the "hit.row" column with expression level of the gene in the "expr.row" column. IDs in the two columns should be the same if the KW test will be used to evaluate association between lesion groups and expression level of the same corresponding gene. If the same patient is affected with multiple types of lesions in the same gene for example gain AND mutations, the entry will be denoted as "multiple" and patients without any type of lesions will be coded as "none".

Value

A data table with multiple columns that include:

Author(s)

Abdelrahman Elsayed <abdelrahman.elsayed@stjude.org> and Stanley Pounds <stanley.pounds@stjude.org>

References

Myles Hollander and Douglas A. Wolfe (1973), Nonparametric Statistical Methods. New York: John Wiley & Sons. Pages 115–120.

Cao, X., Elsayed, A. H., & Pounds, S. B. (2023). Statistical Methods Inspired by Challenges in Pediatric Cancer Multi-omics.

See Also

[alex.prep.lsn.expr\(\)](#page-5-1)

Examples

```
data(expr.data)
data(lesion.data)
data(hg19.gene.annotation)
```
prepare expression, lesion data and return the set of genes with both types of data available # ordered by gene IDs in rows and patient IDs in columns: alex.data=alex.prep.lsn.expr(expr.data, lesion.data,

hg19.gene.annotation, min.expr=1, min.pts.lsn=5)

run Kruskal-Wallis test for association between lesion groups and expression level of the # same corresponding gene:

alex.kw.results=KW.hit.express(alex.data, hg19.gene.annotation, min.grp.size=5)

Description

Lesion data file showing copy number variations, single nucleotide variations and structrual rearrangments affecting 265 newly diagnosed T-cell Acute Lymphoblastic Leukemia (T-ALL) patients that was reported by Liu, Yu, et al. (2017).

Usage

lesion.data

Format

lesion.data:

A data frame with 6,887 rows and 5 columns:

ID patient identifier for the patient affected by the genomic lesion

chrom the chromosome on which the lesion is located

loc.start the lesion start position in base pairs

loc.end the lesion end position in base pairs

lsn.type the lesion type for example gain, loss, mutation, fusion, etc...

Source

extracted from the supplementary material tables of the published Liu, Yu, et al. (2017) manuscript <https://www.nature.com/articles/ng.3909#Sec27>

lsn.transcripts.plot *Lesions Gene Transcripts Plot*

Description

Function prepare a plot with all types of lesions that spans either a gene or a region of interest.

Usage

```
lsn.transcripts.plot(
  grin.res,
 genome,
  gene = NULL,
  transTrack = TRUE,
  lsn.clrs = NULL,
  chrom = NULL,
```

```
plot.start = NULL,
 plot.end = NULL,
 lesion.grp = NULL,
  spec.lsn.clr = NULL,
 extend. left = NULL,extend.right = NULL,
 expand = 5e-04,
 hg38.transcripts = NULL,
 hg19.cytoband = NULL,
 hg38.cytoband = NULL
\mathcal{L}
```
Arguments

lsn.transcripts.plot 41

Details

Function return a plot with all lesions that affect either a gene or a region of interest. Top panel of the regional gene plot has the transcripts track with all transcripts annotated to the gene of interest directly retrieved from ensembl database. The middle panel will has all different types of lesions affecting the gene color coded according to the figure legend. Lower panel of the plot has all the GRIN statistics of the gene that include number of subjects affected by each type of lesions, -log10 p, and –log10q values showing if the gene is significantly affected by the corresponding lesion category. If a certain locus is specified, only transcripts track and the lesion panel will be returned (GRIN results panel will not be added to the plot). In case of plots that span large genomic region such as a chromosome band or the whole chromosome and by specifying transTrack=FALSE, transcripts track will not be added to the plot as well.

Value

Function will return either a gene plot with the transcripts track, lesions panel and GRIN statistic for the gene of interest, a plot with all lesions and transcripts aligned to a certain locus of interest if chrom, plot.start and plot.end were specified or a plot with all lesions affecting a region of interest without the transcripts track.

Author(s)

Abdelrahman Elsayed <abdelrahman.elsayed@stjude.org> and Stanley Pounds <stanley.pounds@stjude.org>

References

Cao, X., Elsayed, A. H., & Pounds, S. B. (2023). Statistical Methods Inspired by Challenges in Pediatric Cancer Multi-omics.

See Also

[grin.stats\(\)](#page-29-1)

Examples

```
data(lesion.data)
data(hg19.gene.annotation)
data(hg19.chrom.size)
data(hg19_cytoband)
data(hg38_cytoband)
# run GRIN analysis using grin.stats function
```
grin.results=grin.stats(lesion.data, hg19.gene.annotation, hg19.chrom.size)

Plots showing different types of lesions affecting a gene of interest with a transcripts # track that show all the gene transcripts retrieved from Ensembl (hg19 genome assembly): WT1.gene.plot=lsn.transcripts.plot(grin.results, genome="hg19", gene="WT1", hg19.cytoband=hg19_cytoband)

Plots showing different types of lesions affecting a region of interest with a transcripts # track added to the plot:

locus.plot=lsn.transcripts.plot(grin.results, genome="hg19", hg19.cytoband=hg19_cytoband, chrom=9, plot.start=21800000, plot.end=22200000, lesion.grp = "loss", spec.lsn.clr = "blue")

Plots Showing Different Types of Lesions Affecting a region of Interest without plotting the # transcripts track (this will allow plotting a larger locus of the chromosome such as a # chromosome band (should specify transTrack = FALSE): noTranscripts=lsn.transcripts.plot(grin.results, genome="hg19", transTrack = FALSE, hg19.cytoband=hg19_cytoband, chrom=9, plot.start=19900000, plot.end=25600000, lesion.grp = "loss", spec.lsn.clr = "blue")

Plots Showing Different Types of Lesions Affecting the whole chromosome: chrom.plot=lsn.transcripts.plot(grin.results, genome="hg19", transTrack = FALSE, hg19.cytoband=hg19_cytoband, chrom=9, plot.start=1, plot.end=141000000)

for GRCh38 (hg38) genome assembly, users should first call the AnnotationHub() web resource then # specify ah[["AH113665"]] to retrieve the human hg38 gene transcripts. This formal class # EnsDb object should be called afterwards in the 'hg38.transcripts' argument to return gene and # regional plots.

onco.print.props *Oncoprint proportions*

Description

The function order lesion types based on their average size and assign the proportion of the oncoprint rectangle that should be color filled based on the average size of each lesion type.

Usage

```
onco.print.props(lsn.data, clr = NULL, hgt = NULL)
```
Arguments

hgt Manually assign the proportion of the oncoprint rectangle that should be color filled for each lesion group.

Details

Some patients might be affected by two or more lesion types in the same gene for example gain AND mutations. To get all lesion types represented in the same rectangle in the oncoprint, this function order lesion types based on the average size of each type and assign the proportion of the oncoprint rectangle that should be color filled based on the average size of each lesion type. Color filled proportion of the oncoprint rectangles can be also specified by the user for each lesion type based on the hgt parameter.

Value

Function return a list of three lists specifying the color assigned to each lesion type, the proportion of the rectangle that should be color filled in the oncoprint based on the average size of each lesion type, and the legend parameters.

Author(s)

Lakshmi Patibandla <LakshmiAnuhya.Patibandla@stjude.org>, Abdelrahman Elsayed <abdelrahman.elsayed@stjude and Stanley Pounds <stanley.pounds@stjude.org>

References

Cao, X., Elsayed, A. H., & Pounds, S. B. (2023). Statistical Methods Inspired by Challenges in Pediatric Cancer Multi-omics.

Examples

```
data(lesion.data)
onco.props=onco.print.props(lesion.data, hgt = c("gain"=4, "loss"=3, "mutation"=2, "fusion"=1))
# if hgt argument is not specified, the lesion category "mutation" for single point mutations will
# be assigned size=1 because it has the smallest average lesion size and will have the smallest
# proportion of the filled oncoprint rectangles 1/4=0.25
```
order.index.gene.data *Order Index Gene Data*

Description

This function order and index gene annotation data by chromosome on which the gene is located, gene start, and end positions.

Usage

```
order.index.gene.data(gene.data)
```
Arguments

gene.data data.frame with gene annotation data either provided by the user or retrieved from ensembl BioMart database using get.ensembl.annotation function included in the GRIN2.0 library. data.frame should has four columns that include "gene" which is the ensembl ID of the annotated genes to which the lesion data will be overlapped, "chrom" which is the chromosome on which the gene is located, "loc.start" which is the gene start position, and "loc.end" the gene end position.

Value

A list with the following components:

Author(s)

Stanley Pounds <stanley.pounds@stjude.org>

References

Pounds, Stan, et al. (2013) A genomic random interval model for statistical analysis of genomic lesion data.

Cao, X., Elsayed, A. H., & Pounds, S. B. (2023). Statistical Methods Inspired by Challenges in Pediatric Cancer Multi-omics.

Examples

```
data(hg19.gene.annotation)
```
ordered.genes=order.index.gene.data(hg19.gene.annotation)

order.index.lsn.data *Order Index Lesion Data*

Description

This function order and index lesion data by lesion type, the chromosome on which the lesion is located , and subject.

Usage

```
order.index.lsn.data(lsn.data)
```
pathways and the contract of t

Arguments

lsn.data data.frame with lesion data prepared by the user in a GRIN compatible format. The data.frame should has five columns that include "ID" which is a column with id of the patient affected by the lesion, "chrom" which is the chromosome on which the lesion is located, "loc.start" which is the lesion start position, "loc.end" the lesion end position and "lsn.type" which is the lesion type for example gain, loss, mutation, fusion, etc...

Value

A list with the following components:

Author(s)

Stanley Pounds <stanley.pounds@stjude.org>

References

Pounds, Stan, et al. (2013) A genomic random interval model for statistical analysis of genomic lesion data.

Cao, X., Elsayed, A. H., & Pounds, S. B. (2023). Statistical Methods Inspired by Challenges in Pediatric Cancer Multi-omics.

Examples

data(lesion.data)

ordered.lsn=order.index.lsn.data(lesion.data)

pathways *List of Genes Annotated to a Group of Pathways*

Description

The dataset has a list of genes annotated to a group of different pathways.

Usage

pathways

Format

pathways: A data frame with 121 rows and 3 columns: gene.name Gene symbol. ensembl.id Gene ensembl ID. pathway The pathway to which the gene is annotated.

Source

Data was extracted from the supplementary material tables of the published Liu, Yu, et al. (2017) manuscript <https://www.nature.com/articles/ng.3909#Sec27>

prep.binary.lsn.mtx *Prepare Binary Lesion Matrix*

Description

Prepares a lesion matrix with each gene affected by a certain type of lesion as a row and each patient as a column.

Usage

prep.binary.lsn.mtx(ov.data, min.ngrp = 0)

Arguments

Details

The function uses the output results of the find.gene.lsn.overlaps function and create a binary lesion matrix with each gene affected by certain lesion type as a row and each patient as a column. Rownames are labelled as gene.ID_lesion.type (for example: ENSG00000118513_gain for gains affecting MYB gene). The entry for each patient in the table will be denoted as 1 if the patient is affected by this specific type of lesion in the gene, for example gain in MYB gene (ENSG00000118513) or 0 otherwise.

Value

The function returns a binary lesion matrix with each row labelled as gene.ID_lesion.type and each column is a patient. Entry for each patient in the table will be denoted as 1 if the gene is affected by this specific type of lesion or 0 otherwise.

Author(s)

Stanley Pounds <stanley.pounds@stjude.org>

References

Pounds, Stan, et al. (2013) A genomic random interval model for statistical analysis of genomic lesion data.

Cao, X., Elsayed, A. H., & Pounds, S. B. (2023). Statistical Methods Inspired by Challenges in Pediatric Cancer Multi-omics.

See Also

[prep.gene.lsn.data\(\)](#page-46-1), [find.gene.lsn.overlaps\(\)](#page-16-1)

Examples

```
data(lesion.data)
data(hg19.gene.annotation)
```

```
# prepare gene and lesion data for later computations:
prep.gene.lsn=prep.gene.lsn.data(lesion.data,
                                 hg19.gene.annotation)
```
determine lesions that overlap each gene (locus): gene.lsn.overlap=find.gene.lsn.overlaps(prep.gene.lsn)

prepare the lesion binary matrix with a minimum of 5 patients affected by the lesion to be # included in the final matrix: lsn.binary.mtx=prep.binary.lsn.mtx(gene.lsn.overlap, min.ngrp=5)

prep.gene.lsn.data *Prepare Gene and Lesion Data*

Description

This function prepare gene and lesion data for later GRIN computations.

Usage

```
prep.gene.lsn.data(lsn.data, gene.data, mess.freq = 10)
```
Arguments

```
lsn.data data.frame with lesion data prepared by the user in a GRIN compatible format.
                  The data.frame should has five columns that include "ID" which is the patient
                   ID, "chrom" which is the chromosome on which the lesion is located, "loc.start"
                   which is the lesion start position, "loc.end" the lesion end position and "lsn.type"
                   which is the lesion type for example gain, loss, mutation, fusion, etc...
```


Details

This function order and index gene and lesion data for later computations. Output of this function is used to ovelap gene and lesion data using find.gene.lsn.overlaps function.

Value

A list with the following components:

Author(s)

Stanley Pounds <stanley.pounds@stjude.org>

References

Pounds, Stan, et al. (2013) A genomic random interval model for statistical analysis of genomic lesion data.

Cao, X., Elsayed, A. H., & Pounds, S. B. (2023). Statistical Methods Inspired by Challenges in Pediatric Cancer Multi-omics.

See Also

[order.index.gene.data\(\)](#page-42-1), [order.index.lsn.data\(\)](#page-43-1)

Examples

```
data(lesion.data)
data(hg19.gene.annotation)
```
prepare gene and lesion data for later computations: prep.gene.lsn=prep.gene.lsn.data(lesion.data, hg19.gene.annotation) prep.lsn.type.matrix *Prepare Lesion Type Matrix*

Description

The function prepare a lesion matrix with all types of lesions affecting certain gene as a row and each patient as a column.

Usage

prep.lsn.type.matrix(ov.data, min.ngrp = 0)

Arguments

Details

The function returns a lesion matrix with each row as a gene and each column is a patient. If a gene is affected by one type of lesions in a certain patient, the entry will be labelled by lesion type (for example: gain OR mutation). However, if the same gene is affected by more than one type of lesions in a certain patient (for example: gain AND mutation), the entry will be labelled as "multiple". If the gene is not affected by any lesion, the entry for this patient will be labelled as "none".

Value

The function returns a lesion matrix with all types of lesions affecting certain gene as a row and each patient as a column.

Author(s)

Stanley Pounds <stanley.pounds@stjude.org>

References

Pounds, Stan, et al. (2013) A genomic random interval model for statistical analysis of genomic lesion data.

Cao, X., Elsayed, A. H., & Pounds, S. B. (2023). Statistical Methods Inspired by Challenges in Pediatric Cancer Multi-omics.

See Also

[prep.gene.lsn.data\(\)](#page-46-1), [find.gene.lsn.overlaps\(\)](#page-16-1)

Examples

```
data(lesion.data)
data(hg19.gene.annotation)
# prepare gene and lesion data for later computations:
prep.gene.lsn=prep.gene.lsn.data(lesion.data,
                                 hg19.gene.annotation)
```

```
# determine lesions that overlap each gene (locus):
gene.lsn.overlap=find.gene.lsn.overlaps(prep.gene.lsn)
```
prepare the lesion matrix with a minimum of 5 patients affected by any type of lesion in the # gene to be included in the final matrix lsn.type.mtx=prep.lsn.type.matrix(gene.lsn.overlap, min.ngrp=5)

prob.hits *Find Probablity of Locus Hit*

Description

The function evaluates the probability of a locus to be affected by one or a constellation of multiple types of lesions.

Usage

```
prob.hits(hit.cnt, chr.size = NULL)
```
Arguments

Details

The function computes p-value for the probability of each locus (gene or regulatory feature) to be affected by different types of lesions based on a convolution of independent but non-identical Bernoulli distributions to determine whether a certain locus has an abundance of lesions that is statistically significant.In addition, FDR-adjusted q value is computed for each locus based on Pounds & Cheng (2006) estimator of the proportion of tests with a true null (pi.hat). The function also evaluates if a certain locus is affected by a constellation of multiple types of lesions and computes a p and adjusted q values for the locus to be affected by one type of lesions (p1), two types of lesions (p2), etc...

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Value

A list with the following components:

Author(s)

Stanley Pounds <stanley.pounds@stjude.org>

References

Pounds, Stan, et al. (2013) A genomic random interval model for statistical analysis of genomic lesion data.

Cao, X., Elsayed, A. H., & Pounds, S. B. (2023). Statistical Methods Inspired by Challenges in Pediatric Cancer Multi-omics.

See Also

[prep.gene.lsn.data\(\)](#page-46-1), [find.gene.lsn.overlaps\(\)](#page-16-1), [count.hits\(\)](#page-12-1)

Examples

data(lesion.data) data(hg19.gene.annotation) data(hg19.chrom.size)

prepare gene and lesion data for later computations: prep.gene.lsn=prep.gene.lsn.data(lesion.data, hg19.gene.annotation)

determine lesions that overlap each gene (locus): gene.lsn.overlap=find.gene.lsn.overlaps(prep.gene.lsn)

count number of subjects affected by different types of lesions and number of hits that affect

```
# each locus:
count.subj.hits=count.hits(gene.lsn.overlap)
# compute the probability of each locus to be affected by one or a constellation of multiple
# types of lesion
hits.prob=prob.hits(count.subj.hits, hg19.chrom.size)
```
top.alex.waterfall.plots

Waterfall Plots for Lesion and Expression Data of Top Significant Genes

Description

Function return waterfall plots for top significant genes in the KW results table based on the specified q value.

Usage

```
top.alex.waterfall.plots(out.dir, alex.data, alex.kw.results, q, lsn.data)
```
Arguments

Details

Function will return waterfall plots for top significant genes in the KW results table based on the user specified q-value threshold of the KW test.The plots will be added to the user specified outdir folder.

Value

Function will return waterfall plots for top significant genes in the KW results table.

Author(s)

Abdelrahman Elsayed <abdelrahman.elsayed@stjude.org> and Stanley Pounds <stanley.pounds@stjude.org>

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References

Cao, X., Elsayed, A. H., & Pounds, S. B. (2023). Statistical Methods Inspired by Challenges in Pediatric Cancer Multi-omics.

See Also

[alex.prep.lsn.expr\(\)](#page-5-1), [KW.hit.express\(\)](#page-36-1), [alex.waterfall.prep\(\)](#page-8-1), [alex.waterfall.plot\(\)](#page-6-1)

Examples

```
data(expr.data)
data(lesion.data)
data(hg19.gene.annotation)
```
prepare expression, lesion data and return the set of genes with both types of data available # ordered by gene IDs in rows and patient IDs in columns: alex.data=alex.prep.lsn.expr(expr.data, lesion.data, hg19.gene.annotation, min.expr=5, min.pts.lsn=5)

run KW test for association between lesion groups and expression level of the same gene: alex.kw.results=KW.hit.express(alex.data, hg19.gene.annotation, min.grp.size=5)

return waterfall plots for a list of top significant genes to a pre-specified folder: dir.create(resultsFolder <- file.path(tempdir(), "temp.out"))

waterfall.plts=top.alex.waterfall.plots(out.dir=resultsFolder, alex.data, alex.kw.results, 1e-15, lesion.data)

unlink(resultsFolder, recursive = TRUE)

write.grin.xlsx *Write GRIN Results*

Description

The function Write GRIN results to an excel file with multiple sheets that include GRIN results, lesion data, gene annotation data, chromosome size, gene-lesion overlap and methods paragraph.

Usage

write.grin.xlsx(grin.result, output.file)

Arguments

Value

This function return an excel file with seven sheets that include:

include a paragraph that explains the GRIN model and cite some references.

Author(s)

Stanley Pounds <stanley.pounds@stjude.org>

References

Pounds, Stan, et al. (2013) A genomic random interval model for statistical analysis of genomic lesion data.

Cao, X., Elsayed, A. H., & Pounds, S. B. (2023). Statistical Methods Inspired by Challenges in Pediatric Cancer Multi-omics.

See Also

[grin.stats\(\)](#page-29-1)

Examples

```
data(lesion.data)
data(hg19.gene.annotation)
data(hg19.chrom.size)
```
to directly retreive gene annotation and chromosome size files from Ensembl BioMart database, # UCSC genome browsers and run the GRIN analysis: grin.results=grin.stats(lesion.data,

hg19.gene.annotation, hg19.chrom.size)

Write GRIN results in to an excel sheet ".xlsx" using write.grin.xlsx function.

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