# Package 'InterCellar'

October 17, 2024

**Title** InterCellar: an R-Shiny app for interactive analysis and exploration of cell-cell communication in single-cell transcriptomics

**Version** 2.10.0

**Description** InterCellar is implemented as an R/Bioconductor Package containing a Shiny app that allows users to interactively analyze cell-cell communication from scRNA-seq data. Starting from precomputed ligand-receptor interactions, InterCellar provides filtering options, annotations and multiple visualizations to explore clusters, genes and functions. Finally, based on functional annotation from Gene Ontology and pathway databases, InterCellar implements data-driven analyses to investigate cell-cell communication in one or multiple conditions.

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**Imports** config, golem, shiny, DT, shinydashboard, shinyFiles, shinycssloaders, data.table, fs, dplyr, tidyr, circlize, colourpicker, dendextend, factoextra, ggplot2, plotly, plyr, shinyFeedback, shinyalert, tibble, umap, visNetwork, wordcloud2, readxl, htmlwidgets, colorspace, signal, scales, htmltools, ComplexHeatmap, grDevices, stats, tools, utils, biomaRt, rlang, fmsb, igraph

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**Suggests** testthat (>= 3.0.0), knitr, rmarkdown, glue, graphite, processx, attempt, BiocStyle, httr

**Config/testthat/edition** 3

**URL** https://github.com/martaint/InterCellar

**BugReports** https://github.com/martaint/InterCellar/issues

**VignetteBuilder** knitr

**biocViews** Software, SingleCell, Visualization, GO, Transcriptomics

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# Contents

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`annotateGO` *Perform GO annotation of input data*

---

## Description

Perform GO annotation of input data

## Usage

```
annotateGO(
  input_select_ensembl,
  input_go_evidence_exclude,
  input_go_sources_checkbox,
  input.data
)
```

## Arguments

| | |
|---|---|
| `input_select_ensembl` | ensembl version selected by user |
| `input_go_evidence_exclude` | evidence codes to exclude by user |
| `input_go_sources_checkbox` | GO sources to use by user |
| `input.data` | preprocessed input data |

## Value

GO_annotation

`annotatePathways` *Annotate pathways for input data*

## Description

Annotate pathways for input data

## Usage

```
annotatePathways(selected.db, input.data)
```

## Arguments

| | |
|---|---|
| `selected.db` | pathways sources to use |
| `input.data` | filtered input data |

## Value

pathways_annotation

`buildPairsbyFunctionMatrix` *Build binary matrix with int-pairs in rows, functions in cols*

## Description

Build binary matrix with int-pairs in rows, functions in cols

## Usage

```
buildPairsbyFunctionMatrix(functions_df)
```

## Arguments

| | |
|---|---|
| `functions_df` | annotated df (GO/path/combined) |

## Value

binary matrix

checkLL_RR *Manually change the annotation of L-L and R-R pairs*

## Description

Manually change the annotation of L-L and R-R pairs

## Usage

```
checkLL_RR(input.data)
```

## Arguments

| | |
|---|---|
| input.data | preprocessed table |

## Value

input.data

## Examples

```
data(input.data)
checked.input.data <- checkLL_RR(input.data)
```

circlePlot *Plot circle plot*

## Description

Plot circle plot

## Usage

```
circlePlot(data, cluster_colors, ipm_color, int_flow, link.color)
```

## Arguments

| | |
|---|---|
| data | subset of input data by flow / intpair module |
| cluster_colors | global |
| ipm_color | single color for chosen int-pair module |
| int_flow | string specifying the flow |
| link.color | string specifying variable by which to color links |

## Value

circle plot

## combineAnnotations *Combine GO annotation and pathways in a unique object*

### Description

Combine GO annotation and pathways in a unique object

### Usage

```
combineAnnotations(GO_annotation, pathways_annotation)
```

### Arguments

| | |
|---|---|
| GO_annotation | data |
| pathways_annotation | data |

### Value

combined annotation dataframe

## createBarPlot1_ggplot *Create ggplot barplot to be saved in tiff*

### Description

Create ggplot barplot to be saved in tiff

### Usage

```
createBarPlot1_ggplot(
  barplotDF,
  input_cluster_selected_checkbox,
  input_num_or_weight_bar1
)
```

### Arguments

| | |
|---|---|
| barplotDF | dataframe with N interactions per cluster (auto/para) |
| input_cluster_selected_checkbox | checkbox input |
| input_num_or_weight_bar1 | number of int or weighted number by score |

### Value

ggplot barplot

## createBarPlot2_CV *Create barplot of number of interaction for selected cluster*

### Description

Create barplot of number of interaction for selected cluster

### Usage

```
createBarPlot2_CV(
  barplotDF2,
  input_cluster_selected_checkbox,
  input_clust_barplot2
)
```

### Arguments

| | |
|---|---|
| barplotDF2 | dataframe with barplot data |
| input_cluster_selected_checkbox | selected clusters to keep |
| input_clust_barplot2 | selected cluster to plot |

### Value

plotly fig

## createBarPlot2_ggplot *Create ggplot barplot of Nint per cluster selected*

### Description

Create ggplot barplot of Nint per cluster selected

### Usage

```
createBarPlot2_ggplot(
  barplotDF2,
  input_cluster_selected_checkbox,
  input_clust_barplot2
)
```

### Arguments

| | |
|---|---|
| barplotDF2 | dataframe with barplot data |
| input_cluster_selected_checkbox | selected clusters to keep |
| input_clust_barplot2 | selected cluster to plot |

### Value

ggplot barplot

---

## createBarPlot_CV *Create Barplot cluster-verse*

---

### Description

Create Barplot cluster-verse

### Usage

```
createBarPlot_CV(
  barplotDF,
  input_cluster_selected_checkbox,
  input_num_or_weight_bar1
)
```

### Arguments

| | |
|---|---|
| barplotDF | dataframe with N interactions per cluster (auto/para) |
| input_cluster_selected_checkbox | checkbox input |
| input_num_or_weight_bar1 | number of int or weighted number by score |

### Value

plotly barplot

createNetwork *Create Network of clusters*

## Description

Create Network of clusters

## Usage

```
createNetwork(data.filt.cluster, input_num_or_weight_radio, input_edge_weight)
```

## Arguments

| | |
|---|---|
| data.filt.cluster | filtered input data (by clusters) |
| input_num_or_weight_radio | either number of interactions or weighted by score |
| input_edge_weight | small,medium or large from user input |

## Value

list containing nodes and edges for network

dendroIntPairModules *Get dendrogram of int pair modules*

## Description

Get dendrogram of int pair modules

## Usage

```
dendroIntPairModules(pairs_func_matrix)
```

## Arguments

| | |
|---|---|
| pairs_func_matrix | binary matrix pairs x functions |

## Value

list with dendrogram, hclust and umap

`elbowPoint` *Determine the elbow point on a curve (from package akmedoids)*

## Description

Given a list of x, y coordinates on a curve, function determines the elbow point of the curve.

## Usage

```
elbowPoint(x, y)
```

## Arguments

| | |
|---|---|
| `x` | vector of x coordinates of points on the curve |
| `y` | vector of y coordinates of points on the curve |

## Details

highlight the maximum curvature to identify the elbow point (credit: 'github.com/agentlans')

## Value

an x, y coordinates of the elbow point.

`ensemblLink` *Get html link to ensembl*

## Description

Get html link to ensembl

## Usage

```
ensemblLink(ensembl)
```

## Arguments

| | |
|---|---|
| `ensembl` | symbol |

## Value

html link to website

## getBack2BackBarplot *Get back-to-back barplot for 2 conditions comparison*

### Description

Get back-to-back barplot for 2 conditions comparison

### Usage

```
getBack2BackBarplot(tab_c1, tab_c2, lab_c1, lab_c2)
```

### Arguments

| | |
|---|---|
| tab_c1 | barplot dataframe generated by getBarplotDF() for condition 1 |
| tab_c2 | barplot dataframe generated by getBarplotDF() for condition 1 |
| lab_c1 | label for condition 1 |
| lab_c2 | label for condition 2 |

### Value

ggplot object

## getBarplotDF *Get dataframe for plotting barplot (all clusters)*

### Description

Get dataframe for plotting barplot (all clusters)

### Usage

```
getBarplotDF(
  data.filt.bar,
  input_cluster_selected_checkbox,
  input_num_or_weight_bar1
)
```

### Arguments

| | |
|---|---|
| data.filt.bar | filtered object (checkbox auto/para) |
| input_cluster_selected_checkbox | checkbox input |
| input_num_or_weight_bar1 | number of int or weighted number by score |

### Value

dataframe with number of interactions per cluster auto/para

---

`getBarplotDF2` *Get dataframe for barplot (by cluster)*

---

### Description

Get dataframe for barplot (by cluster)

### Usage

```
getBarplotDF2(filt.data, input_cluster_selected_checkbox, input_clust_barplot2)
```

### Arguments

| | |
|---|---|
| `filt.data` | input data filtered in cluster-verse |
| `input_cluster_selected_checkbox` | selected clusters to keep |
| `input_clust_barplot2` | selected cluster to plot |

### Value

dataframe with num int per cluster

---

`getClusterA_Names` *Get cluster names only from sender cluster A*

---

### Description

Get cluster names only from sender cluster A

### Usage

```
getClusterA_Names(input.data)
```

### Arguments

| | |
|---|---|
| `input.data` | preprocessed input data |

### Value

named list of clusters

## getClusterColors *Get colors for clusters*

### Description

Get colors for clusters

### Usage

```
getClusterColors(input.data)
```

### Arguments

| | |
|---|---|
| input.data | preprocessed input data |

### Value

named vector with colors per cluster

## getClusterNames *Get clusters names from initial input data*

### Description

Get clusters names from initial input data

### Usage

```
getClusterNames(input.data)
```

### Arguments

| | |
|---|---|
| input.data | preprocessed input data |

### Value

named list of clusters

### Examples

```
data(input.data)
cluster_list <- getClusterNames(input.data)
```

## getClusterNetwork *Creating edges dataframe for network of clusters*

### Description

Creating edges dataframe for network of clusters

### Usage

```
getClusterNetwork(input.data, input_num_or_weight_radio, input_edge_weight)
```

### Arguments

| | |
|---|---|
| input.data | preprocessed input data |
| input_num_or_weight_radio | either num of interactions or weighted by score |
| input_edge_weight | small,medium or large from user input |

### Value

edges dataframe

## getClusterSize *Get Clusters size*

### Description

Get Clusters size

### Usage

```
getClusterSize(cl, edges.df, input_num_or_weight_radio)
```

### Arguments

| | |
|---|---|
| cl | cluster name |
| edges.df | dataframe with edges for network |
| input_num_or_weight_radio | either num of interactions or weighted by score |

### Value

sum of n interactions or weighted num for that cluster

## getDistinctCouplets *Get table of unique int-pairs/clust-pairs couplets*

### Description

Get table of unique int-pairs/clust-pairs couplets

### Usage

```
getDistinctCouplets(
  data_cond1,
  data_cond2,
  data_cond3 = NULL,
  lab_c1,
  lab_c2,
  lab_c3 = NULL
)
```

### Arguments

| | |
|---|---|
| data_cond1 | filt.data() corresponding to chosen condition 1 |
| data_cond2 | filt.data() corresponding to chosen condition 2 |
| data_cond3 | filt.data() corresponding to chosen condition 3 |
| lab_c1 | data label for condition 1 |
| lab_c2 | data label for condition 2 |
| lab_c3 | data label for condition 3 |

### Value

modified filt.data containing only unique couplets

## getDotPlot_selInt *Functions to plot DotPlots*

### Description

Functions to plot DotPlots

### Usage

```
getDotPlot_selInt(
  selected_tab,
  clust.order,
  low_color = "aquamarine",
  high_color = "#131780"
)
```

### Arguments

| | |
|---|---|
| selected_tab | selected rows of filt.data by selection from gene table |
| clust.order | how to order clusters |
| low_color | of dotplot |
| high_color | of dotplot |

### Value

list with modified selected data and ggplot2 dotplot

---

## getGeneTable *Get table for gene-verse*

---

### Description

Get table for gene-verse

### Usage

```
getGeneTable(input.data)
```

### Arguments

| | |
|---|---|
| input.data | preprocessed input data |

### Value

gene table with unique intpairs (no connection to clusters)

### Examples

```
data(input.data)
gene_table <- getGeneTable(input.data)
```

## getGObiomaRt *Connection to Ensembl via biomaRt to get GO terms*

### Description

Connection to Ensembl via biomaRt to get GO terms

### Usage

```
getGObiomaRt(input_select_ensembl, input.data)
```

### Arguments

| | |
|---|---|
| input_select_ensembl | chosen version of Ensembl |
| input.data | filtered input data |

### Value

dataframe with GO annotation

## getHitsf *Subfunction to calculate significant functions by permutation test*

### Description

Subfunction to calculate significant functions by permutation test

### Usage

```
getHitsf(mat, gpModules_assign)
```

### Arguments

| | |
|---|---|
| mat | binary matrix of functional terms by int-pairs |
| gpModules_assign | assignment of intpairs to modules |

### Value

matrix with hits

Example

## getIntFlow — *Get subset of interactions corresponding to a certain viewpoint and flow*

### Description

Get subset of interactions corresponding to a certain viewpoint and flow

### Usage

```
getIntFlow(vp, input.data, flow)
```

### Arguments

| | |
|---|---|
| vp | viewpoint cluster |
| input.data | preprocessed/filtered input data |
| flow | one among directed_out, directed_in or undirected |

### Value

subset of data

### Examples

```
data(input.data)
caf_out <- getIntFlow(vp = "CAF", input.data, flow = "directed_out")
```

## getNtermsBYdb — *Calculate number of terms of a database*

### Description

Calculate number of terms of a database

### Usage

```
getNtermsBYdb(annotation)
```

### Arguments

| | |
|---|---|
| annotation | data from either pathways, GO or combined |

### Value

number of terms by dataset

## getNumLR *Get number of unique ligands and receptors*

### Description

Get number of unique ligands and receptors

### Usage

```
getNumLR(gene.table, type)
```

### Arguments

| | |
|---|---|
| gene.table | gene table of unique int-pairs |
| type | either L or R |

### Value

number of L or R genes

## getPieChart *Get Pie Chart of unique couplets*

### Description

Get Pie Chart of unique couplets

### Usage

```
getPieChart(data_dotplot)
```

### Arguments

| | |
|---|---|
| data_dotplot | same data used to generate dotplot |

### Value

pie chart

getRadar_df *#' Get radar plot of relative numbers of interactions for a certain cell type #' #' @param tab_c1 barplot dataframe from Viewpoint generated by getBarplotDF2() containing data for condition 1 #' @param tab_c2 barplot dataframe from Viewpoint generated by getBarplotDF2() containing data for condition 2 #' @param tab_c3 barplot dataframe from Viewpoint generated by getBarplotDF2() containing data for condition 3 #' @param lab_c1 label for condition 1 #' @param lab_c2 label for condition 2 #' @param lab_c3 label for condition 3 #' @param cell_name label of cell type of interest #' #' @return plot #' @importFrom fmsb radarchart #' @importFrom data.table transpose getRadarPlot <- function(tab_c1, tab_c2, tab_c3, lab_c1, lab_c2, lab_c3, cell_name) if(is.null(tab_c3)) df <- merge(tab_c1, tab_c2, by = "Clusters", all = TRUE) colnames(df) <- c("Clusters", "nint_c1", "nint_c2") else df <- merge(tab_c1, tab_c2, by = "Clusters", all = TRUE) df <- merge(df, tab_c3, by = "Clusters", all = TRUE) colnames(df) <- c("Clusters", "nint_c1", "nint_c2", "nint_c3") df[is.na(df)] <- 0 cluster_names <- df$Clusters # add max and min max_nint <- max(df[, -1]) df <- add_column(df, max_nint, .after = "Clusters") df <- add_column(df, "min_nint" = 0, .after = "max_nint") radar_df <- data.table::transpose(df[, -1]) if(is.null(lab_c3)) rownames(radar_df) <- c("max", "min", lab_c1, lab_c2) else rownames(radar_df) <- c("max", "min", lab_c1, lab_c2, lab_c3) colnames(radar_df) <- cluster_names color <- c("#438ECC", "#E97778", "#00BA38") fmsb::radarchart( radar_df, axistype = 1, # Customize the polygon pcol = color, pfcol = scales::alpha(color, 0.5), plwd = 2, plty = 1, # Customize the grid cglcol = "grey", cglty = 1, cglwd = 0.8, # Customize the axis axislabcol = "grey30", # Variable labels vlcex = 1.2, vlabels = colnames(radar_df), caxislabels = round(seq(from = 0, to = radar_df["max",1], length.out = 5)), title = cell_name ) legend( x = "bottomleft", legend = rownames(radar_df[-c(1,2),]), horiz = FALSE, bty = "n", pch = 20 , col = color, text.col = "black", cex = 1, pt.cex = 1.5 ) Get radar df of relative numbers of interactions for a certain cell type*

## Description

#' Get radar plot of relative numbers of interactions for a certain cell type #' #' @param tab_c1 barplot dataframe from Viewpoint generated by getBarplotDF2() containing data for condition 1 #' @param tab_c2 barplot dataframe from Viewpoint generated by getBarplotDF2() containing data for condition 2 #' @param tab_c3 barplot dataframe from Viewpoint generated by getBarplotDF2() containing data for condition 3 #' @param lab_c1 label for condition 1 #' @param lab_c2 label for condition 2 #' @param lab_c3 label for condition 3 #' @param cell_name label of cell type of interest #' #' @return plot #' @importFrom fmsb radarchart #' @importFrom data.table transpose getRadarPlot <- function(tab_c1, tab_c2, tab_c3, lab_c1, lab_c2, lab_c3, cell_name) if(is.null(tab_c3)) df <- merge(tab_c1, tab_c2, by = "Clusters", all = TRUE) colnames(df) <- c("Clusters",

"nint_c1", "nint_c2") else df <- merge(tab_c1, tab_c2, by = "Clusters", all = TRUE) df <- merge(df, tab_c3, by = "Clusters", all = TRUE) colnames(df) <- c("Clusters", "nint_c1", "nint_c2", "nint_c3")

df[is.na(df)] <- 0

cluster_names <- df$Clusters # add max and min max_nint <- max(df[, -1]) df <- add_column(df, max_nint, .after = "Clusters") df <- add_column(df, "min_nint" = 0, .after = "max_nint")

radar_df <- data.table::transpose(df[, -1])

if(is.null(lab_c3)) rownames(radar_df) <- c("max", "min", lab_c1, lab_c2) else rownames(radar_df) <- c("max", "min", lab_c1, lab_c2, lab_c3)

colnames(radar_df) <- cluster_names

color <- c("#438ECC", "#E97778", "#00BA38")

fmsb::radarchart( radar_df, axistype = 1, # Customize the polygon pcol = color, pfcol = scales::alpha(color, 0.5), plwd = 2, plty = 1, # Customize the grid cglcol = "grey", cglty = 1, cglwd = 0.8, # Customize the axis axislabcol = "grey30", # Variable labels vlcex = 1.2, vlabels = colnames(radar_df), caxislabels = round(seq(from = 0, to = radar_df["max",1], length.out = 5)), title = cell_name ) legend( x = "bottomleft", legend = rownames(radar_df[-c(1,2),]), horiz = FALSE, bty = "n", pch = 20 , col = color, text.col = "black", cex = 1, pt.cex = 1.5 )

Get radar df of relative numbers of interactions for a certain cell type

## Usage

```
getRadar_df(tab_c1, tab_c2, tab_c3, lab_c1, lab_c2, lab_c3)
```

## Arguments

| | |
|---|---|
| `tab_c1` | barplot dataframe from Viewpoint generated by getBarplotDF2() containing data for condition 1 |
| `tab_c2` | barplot dataframe from Viewpoint generated by getBarplotDF2() containing data for condition 2 |
| `tab_c3` | barplot dataframe from Viewpoint generated by getBarplotDF2() containing data for condition 3 |
| `lab_c1` | label for condition 1 |
| `lab_c2` | label for condition 2 |
| `lab_c3` | label for condition 3 |

## Value

df to be then used with fmsb radarchart

`getRankedTerms` *Get table with ranked functional terms*

## Description

Get table with ranked functional terms

## Usage

```
getRankedTerms(data.fun.annot)
```

## Arguments

| | |
|---|---|
| `data.fun.annot` | annotated df (GO/path/combined) |

## Value

table with ranking

`getSignificantFunctions` *Calculate significant function per intpair module*

## Description

Calculate significant function per intpair module

## Usage

```
getSignificantFunctions(
  subGenePairs_func_mat,
  gpModules_assign,
  rank.terms,
  input_maxPval
)
```

## Arguments

| | |
|---|---|
| `subGenePairs_func_mat` | subset of binary mat |
| `gpModules_assign` | assignment of intpairs to modules |
| `rank.terms` | table of ranked functions |
| `input_maxPval` | threshold of significance |

## Value

table with significant functions

---

`getSignificantFunctions_multiCond` *Get significance of functional terms related to unique int-pairs per condition*

---

## Description

Get significance of functional terms related to unique int-pairs per condition

## Usage

```
getSignificantFunctions_multiCond(sub_annot, unique_intpairs)
```

## Arguments

| | |
|---|---|
| `sub_annot` | annotation matrix subset to unique int-pairs |
| `unique_intpairs` | data.frame with unique int-pairs by condition |

## Value

data.frame with calculated pvalue of significance

---

`getSignif_table` *Wrapper for other functions to get significant table of func terms*

---

## Description

Wrapper for other functions to get significant table of func terms

## Usage

```
getSignif_table(
  data_cond1,
  data_cond2,
  data_cond3,
  lab_c1,
  lab_c2,
  lab_c3,
  annot_cond1,
  annot_cond2,
  annot_cond3
)
```

### Arguments

| | |
|---|---|
| data_cond1 | filt.data() corresponding to chosen condition 1 |
| data_cond2 | filt.data() corresponding to chosen condition 2 |
| data_cond3 | filt.data() corresponding to chosen condition 3 |
| lab_c1 | data label for condition 1 |
| lab_c2 | data label for condition 2 |
| lab_c3 | data label for condition 3 |
| annot_cond1 | binary matrix int-pair by functions for cond1 |
| annot_cond2 | binary matrix int-pair by functions for cond2 |
| annot_cond3 | binary matrix int-pair by functions for cond3 |

### Value

list containing pvalue_df and unique_intpairs df

---

## getSunburst *Get Sunburst plot of selected functional terms*

---

### Description

Get Sunburst plot of selected functional terms

### Usage

```
getSunburst(
  sel.data,
  func_selected,
  int_p_fun,
  cluster.colors,
  input_num_or_weight_radio
)
```

### Arguments

| | |
|---|---|
| sel.data | dataframe of selected functions |
| func_selected | the selected functional term |
| int_p_fun | dataframe with int pairs annotated to this function |
| cluster.colors | for plotting |
| input_num_or_weight_radio | either num of interactions or weighted by score |

### Value

plotly figure

## getUMAPipModules *Get UMAP for IP modules*

### Description

Get UMAP for IP modules

### Usage

```
getUMAPipModules(intPairs.dendro, gpModules_assign, ipm_colors)
```

### Arguments

| | |
|---|---|
| intPairs.dendro | list output of dendrogram |
| gpModules_assign | named vector of module assignment |
| ipm_colors | for intpair modules |

### Value

plotly umap

## getUniqueDotplot *Plot dotplot containing only unique int-pair/cluster pairs with many conditions*

### Description

Plot dotplot containing only unique int-pair/cluster pairs with many conditions

### Usage

```
getUniqueDotplot(data_dotplot, clust.order)
```

### Arguments

| | |
|---|---|
| data_dotplot | table with selected int_pairs for multiple conditions |
| clust.order | how to order clusters |

### Value

ggplot object

## getUniqueIntpairs_byCond *Get table of unique int-pairs by condition*

### Description

Get table of unique int-pairs by condition

### Usage

```
getUniqueIntpairs_byCond(
  data_cond1,
  data_cond2,
  data_cond3 = NULL,
  lab_c1,
  lab_c2,
  lab_c3 = NULL
)
```

### Arguments

| | |
|---|---|
| data_cond1 | filt.data() corresponding to chosen condition 1 |
| data_cond2 | filt.data() corresponding to chosen condition 2 |
| data_cond3 | filt.data() corresponding to chosen condition 3 |
| lab_c1 | data label for condition 1 |
| lab_c2 | data label for condition 2 |
| lab_c3 | data label for condition 3 |

### Value

modified merged filt.data containing only unique intpairs

## goLink *Get GO link*

### Description

Get GO link

### Usage

```
goLink(go_id)
```

### Arguments

| | |
|---|---|
| go_id | string |

### Value

html link to website

---

`input.data` *Input Data example*

---

### Description

A dataset obtained from Tirosh et al melanoma dataset, running CellPhoneDBv2. This data is generated by InterCellar running read.CPDBv2()

### Usage

```
input.data
```

### Format

A data frame with 5638 rows and 11 variables:

**int_pair** interaction pair name, geneA & geneB

**geneA** name, hgnc_symbol

**geneB** name, hgnc_symbol

**typeA** molecular type of geneA, either L (ligand) or R (receptor)

**typeB** molecular type of geneB, either L (ligand) or R (receptor)

**clustA** name of first cluster, either character or number

**clustB** name of second cluster, either character or number

**score** int-pair score as avg expression of geneA and geneB over clustA and clustB, decimal

**p_value** int-pair pvalue, decimal

**annotation_strategy** database from which the int-pair was retrieved

**int.type** either autocrine or paracrine

## read.cellchat — *Read dataframe of cell-cell communication from CellChat (lig- and/receptor)*

### Description

Read dataframe of cell-cell communication from CellChat (ligand/receptor)

### Usage

```
read.cellchat(file_tab)
```

### Arguments

| | |
|---|---|
| file_tab | dataframe from cellchat |

### Value

input.data formatted for InterCellar

## read.CPDBv2 — *Read output from CellPhoneDB v2.*

### Description

Output is a folder containing 4 .txt files - deconvoluted.txt: containing list of single genes and their mean expression in each cluster (not considered); - means.txt: containing list of interacting pairs with info regarding L/R, annotation strategy and mean value of all pairs over cluster couples. - pvalues.txt: same as means, but containing pvalue of each pair, for each cluster couple. - significant_means.txt: only means of those pairs that have pvalue < 0.05. Has one more column:rank. If the statistical analysis is not run, the folder would contain only deconvoluted and means

### Usage

```
read.CPDBv2(folder)
```

### Arguments

| | |
|---|---|
| folder | folder containing output |

### Value

input.data which is the pre-processed object with annotated L-R pairs

## read.customInput *Read custom input file and re-structure it with InterCellar format*

### Description

Read custom input file and re-structure it with InterCellar format

### Usage

```
read.customInput(tab, separator)
```

### Arguments

| | |
|---|---|
| tab | custom input table |
| separator | character that separates two elements of an interaction pair |

### Value

preprocessed table

## read.icellnet *Read ICELLNET dataframe*

### Description

Read ICELLNET dataframe

### Usage

```
read.icellnet(tab, input_icellnet_CC, input_icellnet_dir)
```

### Arguments

| | |
|---|---|
| tab | dataframe with int-pairs in "X" column, other columns as cell types |
| input_icellnet_CC | central cell name |
| input_icellnet_dir | direction of interaction either out or in |

### Value

pre-processed input data

## read.SCsignalR *Read output from SingleCellSignalR*

### Description

SCSR description: the output folder is a collection of txt files, one for each clusters pair considered. The "paracrine" option looks for ligands expressed in cluster A and their associated receptors according to LRdb that are expressed in any other cluster but A. These interactions are labelled "paracrine". The interactions that involve a ligand and a receptor, both differentially expressed in their respective cell clusters according to the **edgeR** analysis performed by the **cluster_analysis()** function, are labelled "specific". The "autocrine" option searches for ligands expressed in cell cluster A and their associated receptors also expressed in A. These interactions are labelled "autocrine". Additionally, it searches for those associated receptors in the other cell clusters (not A) to cover the part of the signaling that is "autocrine" and "paracrine" simultaneously. These interactions are labelled "autocrine/paracrine". This file is a 4-column table: ligands, receptors, interaction types ("paracrine", "autocrine", "autocrine/paracrine" and "specific"), and the associated LRscore. InterCellar: rename autocrine|paracrine to paracrine

### Usage

```
read.SCsignalR(folder)
```

### Arguments

| | |
|---|---|
| folder | containing output from SingleCellSignalR, named cell-signaling |

### Value

input.data: preprocessed object with annotated L-R pairs

## run_app *Run the Shiny Application*

### Description

Run the Shiny Application

### Usage

```
run_app(reproducible = TRUE)
```

### Arguments

| | |
|---|---|
| reproducible | boolean for setting a seed, making plots reproducible |

## Value

a running instance of InterCellar

## Examples

```
## Not run:
run_app()

## End(Not run)
```

---

`subsetAnnot_multiCond` *Subset int-pair by function matrices to unique int-pairs by condition*

---

## Description

Subset int-pair by function matrices to unique int-pairs by condition

## Usage

```
subsetAnnot_multiCond(
  annot_cond1,
  annot_cond2,
  annot_cond3,
  unique_intpairs,
  lab_c1,
  lab_c2,
  lab_c3
)
```

## Arguments

| | |
|---|---|
| `annot_cond1` | binary matrix int-pair by functions for cond1 |
| `annot_cond2` | binary matrix int-pair by functions for cond2 |
| `annot_cond3` | binary matrix int-pair by functions for cond3 |
| `unique_intpairs` | table of unique int-pairs by condition |
| `lab_c1` | label cond1 |
| `lab_c2` | label cond2 |
| `lab_c3` | label cond3 |

## Value

subset merged matrix

## subsetFuncMatBYFlow *Subset pairs-function matrix by selected flow*

### Description

Subset pairs-function matrix by selected flow

### Usage

```
subsetFuncMatBYFlow(pairs_func_matrix, flow_df)
```

### Arguments

| | |
|---|---|
| pairs_func_matrix | binary |
| flow_df | subset of input data by flow |

### Value

subset of binary mat

## swap.RLint *Swaps interaction pairs that are R-L to L-R*

### Description

Swaps interaction pairs that are R-L to L-R

### Usage

```
swap.RLint(RLint)
```

### Arguments

| | |
|---|---|
| RLint | subset of R-L interactions |

### Value

input data with ordered L-R pairs and L-L/R-R

`uniprotLink` *Get html link to uniprot*

## Description

Get html link to uniprot

## Usage

```
uniprotLink(uniprot)
```

## Arguments

| | |
|---|---|
| `uniprot` | symbol |

## Value

html link to website

`updateInputLR` *Function that orders all interaction pairs as L-R. Leaves unchanged the R-R and L-L*

## Description

Function that orders all interaction pairs as L-R. Leaves unchanged the R-R and L-L

## Usage

```
updateInputLR(input.data)
```

## Arguments

| | |
|---|---|
| `input.data` | uploaded data |

## Value

ordered input data

# Index