# Package 'decone'

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```
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Title Cell Type Deconvolution Evaluator
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Description The evaluation platform for cell type deconvolution. This platform
      provides multiple simulation and real mixture expression datasets for evaluate
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VignetteBuilder knitr

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addNoise

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Add noise to gene expression data

## **Description**

Add noise based on negtive bionomial distribution. Please see "A benchmark for RNA-seq deconvolution analysis under dynamic testing environments" in Genome Biology.

# Usage

```
addNoise(x = NULL, pt = 0.1, type = "NB")
```

# **Arguments**

x a gene expression numeric vector.

pt parameter to control noised level. Default: 0.1

type "NB", "N" or "LN". "NB" means Negative binomial model, "N" means normal

model, "LN" means Log-normal model. Default: "NB"

# Value

the sample length vector.

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#### **Examples**

```
res <- addNoise(x = seq(100))
```

addNoiseExpr

Add noise to the simulated expression matrix

### **Description**

Add different level noise to the simulated expression matrix based on the negative binomial distribution. The related model can be found in this paper (Jin, H., Liu, Z. A benchmark for RNA-seq deconvolution analysis under dynamic testing environments. Genome Biol 22, 102 (2021). https://doi.org/10.1186/s13059-021-02290-6).

# Usage

```
addNoiseExpr(
  exprFile,
  outputPath = NULL,
  prefix = NULL,
  Pt = seq(0.1, 1, 0.1),
  type = "NB"
)
```

### **Arguments**

exprFile The input expression file. Must be in csv format. Each row is a gene, each column is a sample. rownames and colnames are required. Please check the

output of the function  ${\tt exprSim}.$ 

outputPath Output path, create if not exists. Default: a new folder based on the exprFile. For

example, if the exprFile is "/data/test.txt", the outputPath will be a new folder

"/data/test".

prefix The prefix of the output file. Note: the suffix will be added automatically based

on the noise level.

Pt Parameter to control noise level. Default: seq(0.1, 1, 0.1)

type "NB", "N" or "LN". Noise type. "NB" means Negative binomial model, "N"

means normal model, "LN" means Log-normal model. Default: "NB"

### Value

All the information will be written in the output path, and each file is the generated data in a specific noise level.

```
res <- pseudoExpr()
write.csv(x = res$mix, file = "mix.csv", row.names = T, quote = F)
addNoiseExpr(exprFile = "mix.csv", Pt = seq(0.1, 1, 0.1), type = "NB")</pre>
```

| boxplot_mult | i Boxplo | ot of deconvolution result | ts for multiple compa | rison. |
|--------------|----------|----------------------------|-----------------------|--------|
|              |          |                            |                       |        |

# Description

Boxplot for illustrating the deconvolution performance.

# Usage

```
boxplot_multi(actual, predicted, label = NULL, method, title = "Boxplot")
```

# **Arguments**

| actual    | The groundtruth proportion of cell types in matrix. row: cell types, column: samples. Also can be a csv matrix with row and column names.  |
|-----------|--|
| predicted | a vector contains all the files with the predicted proportions from different methods. Must be in csv file with row and column names. row: cell types, column: samples. For example, c("method1.csv", "method2.csv", "method3.csv"). |
| label     | a vector contains the label corresponding to the predicted proportion file. Default: base file name in parameter 'predicted'. For example, $c("method1", "method2", "method3")$ .  |
| method    | "rmse", "mape", "mae", "pearson" or "spearman". Note: for mape, cell types with read proportion 0 will be ignored.   |
| title     | Boxplot title in character.  |

### Value

The computed metrics and plot data.

```
boxplot_NcrossCompare Comparing boxplot for noise testing.
```

# Description

Boxplot for illustrating the deconvlition performance with the noised in silico data. this function is used for comparing multiple methods.

# Usage

```
boxplot_NcrossCompare(actual, predicted, label = NULL, method, title = "")
```

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# **Arguments**

| a list of vectors. Each vector contains all the files with the predicted proportion in different noise level. Must be in csv file with row and column names. To cell types, column: samples. For example, list(method1 = c("m1_noise0.csv"m1_noise1.csv", "m1_noise2.csv"), method2 = c("m2_noise0.csv", "m2_noise2.csv")). Note: names for each method will be used for assign column a vector contains the label corresponding to the predicted proportion file. | nn:                      |
|--|--------------------------|
| label a vector contains the label corresponding to the predicted proportion file. I  | ow:<br>v",<br>ise1.csv", |
| fault: c("condition1", "condition2", "condition3",)  | )e-                      |
| method "rmse", "mape", "mae", "pearson" or "spearman". Note: for mape, cell ty with read proportion 0 will be ignored.   | oes                      |

#### Value

The computed metrics and plot data.

# **Examples**

title

```
res <- pseudoData(type = 3, outputPath = "test_file")
res <- boxplot_NcrossCompare(actual = res$actual, predicted = res$predicted,
label = res$noise_level, method = "pearson", title = "pearson")</pre>
```

Boxplot title in character.

boxplot\_NGrad Boxplot of deconvolution results for multiple test.

# Description

Boxplot for illustrating the deconvolution performance with the noised in silico data.

# Usage

```
boxplot_NGrad(actual, predicted, label = NULL, method, title = "Boxplot")
```

## **Arguments**

| actual    | The groundtruth proportion of cell types in matrix. row: cell types, column: samples. Also can be a csv matrix with row and column names.   |
|-----------|---|
| predicted | a vector contains all the files with the predicted proportions in different noise level. Must be in csv file with row and column names. row: cell types, column: samples. For example, c("noise0.csv", "noise1.csv", "noise2.csv"). |
| label     | a vector contains the label corresponding to the predicted proportion file. Default: base file name in parameter 'predicted'. For example, $c("noise0", "noise1", "noise2")$ .  |
| method    | "rmse", "mape", "mae", "pearson" or "spearman". Note: for mape, cell types with read proportion 0 will be ignored.  |
| title     | Boxplot title in character.   |

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### Value

The computed metrics and plot data.

## **Examples**

```
res <- pseudoData(type = 2)
res <-boxplot_NGrad(actual = res$actual, predicted = res$predicted,
label = res$noise_level, method = "rmse")</pre>
```

boxplot\_simple

Boxplot of deconvolution results for a specific method.

# Description

Boxplot for deconvolution results with multiple samples.

## Usage

```
boxplot_simple(actual, predicted, method = NULL)
```

### **Arguments**

actual The groundtruth proportion of cell types in matrix. row: cell types, column:

samples. Also can be a csv matrix with row and column names.

predicted The predicted proportion of cell types in matrix. row: cell types, column: sam-

ples. Also can be a csv matrix with row and column names.

method "rmse", "mape", "mae", "pearson" or "spearman". Note: for mape, cell types

with read proportion 0 will be ignored.

# Value

The computed metrics and plot data.

```
res <- pseudoData(type = 1)
res <- boxplot_simple(actual = res$actual, predicted = res$predicted, method = "mape")</pre>
```

cheatmap\_NcrossCompare

Plot circle heatmap for deconvolution results.

# Description

Usually, only one metric cannot reveal the deconvolution performance well. Therefore, the circle heatmap which contains two different dimension information can be used for a better illustration.

### Usage

```
cheatmap_NcrossCompare(actual, predicted, label = NULL, method1, method2)
```

## **Arguments**

| actual    | The groundtruth proportion of cell types in matrix. row: cell types, column: samples. Also can be a csv matrix with row and column names.  |
|-----------|--|
| predicted | a list of vectors. Each vector contains all the files with the predicted proportions in different noise level. Must be in csv file with row and column names. row: cell types, column: samples. For example, list(method1 = $c("m1\_noise0.csv", "m1\_noise1.csv", "m1\_noise1.csv", "m1\_noise2.csv")$ , method2 = $c("m2\_noise0.csv", "m2\_noise1.csv", "m2\_noise2.csv")$ ). Note: names for each method will be used for assign colors. |
| label     | a vector contains the label corresponding to the predicted proportion file. Default: c("condition1", "condition2", "condition3",)  |
| method1   | "rmse", "mape", "mae", "pearson" or "spearman". Note: for mape, cell types with read proportion 0 will be ignored.   |
| method2   | "rmse", "mape", "mae", "pearson" or "spearman". Note: for mape, cell types with read proportion 0 will be ignored.   |

# Value

The computed metrics and plot data.

# **Examples**

```
res <- pseudoData(type = 3, outputPath = "test_file")
res <- cheatmap_NcrossCompare(actual = res$actual, predicted = res$predicted,
label = res$noise_level, method1 = "pearson", method2 = "rmse")</pre>
```

cheatmap\_RcrossCompare

Plot circle heatmap for deconvolution results rare cell types.

# Description

Generate circle heatmap for different rare cell type proportions. The dataset should be generated from the function rareExprSim.

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# Usage

```
cheatmap_RcrossCompare(
   actual,
   predicted,
   p_rare = c(0.001, 0.003, 0.005, 0.008, 0.01, 0.03, 0.05),
   label = NULL,
   method1,
   method2
)
```

# **Arguments**

| actual    | The groundtruth proportion of cell types in matrix or csv file. row: cell types, column: samples. Also can be a csv matrix with row and column names.   |
|-----------|---|
| predicted | a vector contains all the files with the predicted proportions in different method. Must be in csv file with row and column names. row: cell types, column: samples. For example, c("method1.csv", "method2.csv", "method3.csv"). |
| p_rare    | A vector of proportions. should be the same with function <code>rareExprSim</code> . Default: $c(0.001,0.003,0.005,0.008,0.01,0.03,0.05)$   |
| label     | a vector contains the label corresponding to the predicted proportion file. Default: $c("condition1", "condition2", "condition3",)$   |
| method1   | "rmse", "mape", "mae". Note: for mape, cell types with read proportion 0 will be ignored.   |
| method2   | "rmse", "mape", "mae". Note: for mape, cell types with read proportion 0 will be ignored.   |

# Value

The plot data.

# **Examples**

```
rareExprSim()
# after the simulation, just pass the predicted file
# to the parameter "actual" and "predicted"
```

CPM

Convert a gene expression count matrix into CPM matrix.

# Description

Convert a gene expression count matrix into CPM matrix

# Usage

```
CPM(data = NULL)
```

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### **Arguments**

data

A dataframe or matrix object. Note: Must contain a column named "Length", "Length" means gene length.

#### Value

A dataframe or matrix object.

## **Examples**

```
data <- data.frame(s1 = seq(100), Length = seq(100))
res <- CPM(data)</pre>
```

exprSim

Generate in silico mixture expression matrix

# **Description**

Generate in silico mixture expression matrix based on the internal RNA-seq database. The internal RNA-seq database is collected from multiple studies. All the samples are passed the quality filter. This function is different from pseudoExpr. Counts in pseudoExpr is randomly generated from uniform distribution. This function use the real cell type specific expression data to generate mixture data. We provide two types of simulation called "coarse" and "fine". This idea is from DREAM Challenge Tumor Deconvolution problem. https://www.synapse.org/#! Synapse:syn15589870/wiki/.

### Usage

```
exprSim(
  n_sample = 50,
  p = 2/3,
  type = "coarse",
  transform = "TPM",
  outputPath = NULL,
  mix_name = "coarse_gene_expr.csv",
  ref_name = "coarse_ref.csv",
  prop_name = "coarse_prop.csv",
  refVar_name = NULL,
  train_name = NULL
)
```

### **Arguments**

n\_sample

Sample number to be generated, default: 50.

р

Proportion of sample in train set, default: 0.6.

type

"coarse" or "fine". "coarse" means the simulation will be performed in a coarse level. Only 8 cell types will be used, including ("B.cells", "CD4.T.cells", "CD8.T.cells", "endothelial.cells", "macrophages", "monocytes", "neutrophils", "NK.cells"). "fine" means the simulation will be performed in a fine level. 14 cell types will be used, including ("memory.B.cells", "naive.B.cells", "memory.CD4.T.cells",

"naive.CD4.T.cells", "regulatory.T.cells", "memory.CD8.T.cells", "naive.CD8.T.cells", "NK.cells", "neutrophils", "monocytes", "myeloid.dendritic.cells", "macrophages",

"fibroblasts", "endothelial.cells")

transform "TPM", "CPM" or "NO". Transform the data into TPM, CPM or in counts.

outputPath output file save path.
mix\_name mixture output file name.

ref\_name reference output file name in csv.

prop\_name simulated proportion file name in csv.

refVar\_name reference variance file name in csv.

train\_name file name for all data in train set in csv. This data can be used for differential

gene analysis.

### Value

All the information will be written in the output path.

# **Examples**

exprSim()

 $heatmap\_NcrossCompare$  Plot heatmap for deconvolution results.

### **Description**

Generate heatmap comparison for different methods.

# Usage

heatmap\_NcrossCompare(actual, predicted, label = NULL, method)

### **Arguments**

actual The groundtruth proportion of cell types in matrix. row: cell types, column:

samples. Also can be a csv matrix with row and column names.

predicted a list of vectors. Each vector contains all the files with the predicted proportions

in different noise level. Must be in csv file with row and column names. row: cell types, column: samples. For example, list(method1 = c("m1\_noise0.csv", "m1\_noise1.csv", "m1\_noise2.csv"), method2 = c("m2\_noise0.csv", "m2\_noise1.csv",

"m2\_noise2.csv")). Note: names for each method will be used for assign colors.

label a vector contains the label corresponding to the predicted proportion file. De-

fault: c("condition1", "condition2", "condition3", ...)

method "rmse", "mape", "mae", "pearson" or "spearman". Note: for mape, cell types

with read proportion 0 will be ignored.

### Value

The computed metrics and plot data.

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## **Examples**

```
res <- pseudoData(type = 3, outputPath = "test_file")
res <- heatmap_NcrossCompare(actual = res$actual, predicted = res$predicted,
label = res$noise_level, method = "mape")</pre>
```

heatmap\_NGradCT

Plot heatmap for different cell types.

# Description

Generate heatmap comparison for different cell types.

# Usage

```
heatmap_NGradCT(actual, predicted, label = NULL, method)
```

# Arguments

| actual    | The groundtruth proportion of cell types in matrix. row: cell types, column: samples. Also can be a csv matrix with row and column names.   |
|-----------|---|
| predicted | a vector contains all the files with the predicted proportions in different noise level. Must be in csv file with row and column names. row: cell types, column: samples. For example, c("noise0.csv", "noise1.csv", "noise2.csv"). |
| label     | a vector contains the label corresponding to the predicted proportion file. Default: $c("condition1", "condition2", "condition3",)$   |
| method    | "rmse", "mape", "mae", "pearson" or "spearman". Note: for mape, cell types with read proportion 0 will be ignored.  |

# Value

The computed metrics and plot data.

```
res <- pseudoData(type = 2)
res <- heatmap_NGradCT(actual = res$actual, predicted = res$predicted,
label = res$noise_level, method = "rmse")</pre>
```

heatmap\_RcrossCompare Plot heatmap for deconvolution results.

# Description

Generate heatmap comparison for different methods.

# Usage

```
heatmap_RcrossCompare(
  actual,
  predicted,
  p_rare = c(0.001, 0.003, 0.005, 0.008, 0.01, 0.03, 0.05),
  label = NULL,
  method
)
```

# Arguments

| actual    | The groundtruth proportion of cell types in matrix. row: cell types, column: samples. Also can be a csv matrix with row and column names.   |
|-----------|---|
| predicted | a vector contains all the files with the predicted proportions in different method. Must be in csv file with row and column names. row: cell types, column: samples. For example, c("method1.csv", "method2.csv", "method3.csv"). |
| p_rare    | A vector of proportions. should be the same with function <code>rareExprSim</code> . Default: $c(0.001,0.003,0.005,0.008,0.01,0.03,0.05)$   |
| label     | a vector contains the label corresponding to the predicted proportion file. Default: $c("condition1", "condition2", "condition3",)$   |
| method    | "rmse", "mape", "mae". Note: for mape, cell types with read proportion 0 will be ignored.   |

## Value

The computed metrics and plot data.

```
rareExprSim()
# after the simulation, just pass the predicted file
# to the parameter "actual" and "predicted"
```

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pseudoData

Generate pseudo data for plotting

## **Description**

Generate pseudo data (prediction values) for plotting

## Usage

```
pseudoData(type = 1, outputPath = "test_file")
```

# **Arguments**

type 1, 2 or 3 for different usage outputPath path to save tmp file

#### Value

The pseudo data

## **Examples**

```
res <- pseudoData(type = 1)</pre>
```

pseudoExpr

Generate pseudo mixture expression matrix

# Description

Generate pseudo mixture expression matrix based on uniform distribution, without any noise.

## Usage

```
pseudoExpr(n_sample = 50, n_gene = 1000, n_ct = 10)
```

## **Arguments**

n\_sample
 n\_gene
 Gene number to be generated, default: 1000
 n\_ct
 Cell type number to be generated, default: 10

## Value

a list, prop means the pseudo proportion, ref means the pseudo external reference, mix is the output of ref

```
res <- pseudoExpr()
```

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rareExprSim

Generate in silico expression dataset with rare component.

# **Description**

Generate in silico mixture expression matrix based on the internal RNA-seq database. This function is different from the function exprSim. exprSim generates all the proportion randomly, while this function takes one cell type as rare component and all Other cell type proportion will be set randomly from the uniform distribution. The internal RNA-seq database is collected from multiple studies. All the samples are passed the quality filter. We provide two types of simulation called "coarse" and "fine". This idea is from DREAM Challenge Tumor Deconvolution problem. https://www.synapse.org/#!Synapse:syn15589870/wiki/.

### Usage

```
rareExprSim(
   p_rare = c(0.001, 0.003, 0.005, 0.008, 0.01, 0.03, 0.05),
   p = 0.6,
   type = "coarse",
   transform = "TPM",
   outputPath = NULL,
   mix_name = "coarse_gene_expr.csv",
   ref_name = "coarse_ref.csv",
   prop_name = "coarse_prop.csv",
   refVar_name = NULL,
   train_name = NULL
)
```

reference output file name in csv. simulated proportion file name in csv.

reference variance file name in csv.

# **Arguments**

ref\_name

prop\_name

refVar\_name

| p_rare     | A vector of proportions. Default: $c(0.001, 0.003, 0.005, 0.008, 0.01, 0.03, 0.05)$<br>Note: Every cell type will be treated as rare component. For example, if 8 cell types need to be tested, this function will generate $7*8 = 56$ samples. 7 mean 7 rare proportions and 8 means 8 cell types.  |
|------------|--|
| p          | Proportion of sample in train set, default: 0.6.   |
| type       | "coarse" or "fine". "coarse" means the simulation will be performed in a coarse level. Only 8 cell types will be used, including ("B.cells", "CD4.T.cells", "CD8.T.cells", "endothelial.cells", "macrophages", "monocytes", "neutrophils", "NK.cells"). "fine" means the simulation will be performed in a fine level. 14 cell types will be used, including ("memory.B.cells", "naive.B.cells", "memory.CD4.T.cells", "naive.CD4.T.cells", "regulatory.T.cells", "memory.CD8.T.cells", "naive.CD8.T.cells", "NK.cells", "neutrophils", "monocytes", "myeloid.dendritic.cells", "macrophages", "fibroblasts", "endothelial.cells") |
| transform  | "TPM", "CPM" or "NO". Transform the data into TPM, CPM or in counts.   |
| outputPath | output file save path.   |
| mix_name   | mixture output file name.  |

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train\_name file name for all data in train set in csv. This data can be used for differential gene analysis.

### Value

All the information will be written in the output path.

## **Examples**

```
rareExprSim()
```

rarescExprSim

Generate in silico mixture expression matrix based on scRNA-seq with rare component.

## **Description**

Generate in silico mixture expression matrix based on the internal scRNA-seq database. The internal RNA-seq database is collected from PMID:29474909. All the samples are passed the quality filter. This function use the real cell type specific expression data to generate mixture data. The cell type include c("FetalStomach", "FetalLung", "FetalLiver", "FetalKidney", "FetalIntestine", "FetalBrain", "Female.fetal.Gonad")

## Usage

```
rarescExprSim(
  p_rare = c(0.001, 0.003, 0.005, 0.008, 0.01, 0.03, 0.05),
  p = 2/3,
  transform = "TPM",
  outputPath = NULL,
  mix_name = "scMouse_gene_expr.csv",
  ref_name = "scMouse_ref.csv",
  prop_name = "scMouse_prop.csv",
  train_name = "scMouse_ref_rawCount.csv"
)
```

# **Arguments**

| p_rare     | A vector of proportions. Default: $c(0.001, 0.003, 0.005, 0.008, 0.01, 0.03, 0.05)$<br>Note: Every cell type will be treated as rare component. For example, if 8 cell types need to be tested, this function will generate $7*8 = 56$ samples. 7 mean 7 rare proportions and 8 means 8 cell types. |
|------------|---|
| р          | Proportion of sample in train set, default: 2 / 3.  |
| transform  | "TPM", "CPM" or "NO". Transform the data into TPM, CPM or in counts.  |
| outputPath | output file save path.  |
| mix_name   | mixture output file name.   |
| ref_name   | reference output file name in csv.  |
| prop_name  | simulated proportion file name in csv.  |
| train_name | file name for all data in train set in csv. This data can be used for differential gene analysis.   |

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### Value

All the information will be written in the output path.

# **Examples**

```
rarescExprSim()
```

regMetrics

Compute regression Metrics for matrix results.

# Description

Compute several regression Metrics for cell type deconcolution.

# Usage

```
regMetrics(actual, predicted, method = NULL)
```

# Arguments

actual The groundcloth proportion of cell types in matrix. row: cell types, column:

samples.

predicted The predicted proportion of cell types in matrix. row: cell types, column: sam-

ples.

method "rmse", "mape", "mae", "pearson", "spearman", "pearson2", "spearman2". Note:

for mape, cell types with read proportion 0 will be ignored. For pearson2 and spearman2, all data in matrix will be taken into consideration, which mean only

one number will be reported.

### Value

a vector with the value for each sample.

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|     |       | _   |
|-----|-------|-----|
| SCA | itter | . к |

Scatter plot for rare component.

# Description

Generate scatter plot for rare component.

# Usage

```
scatter_R(
   actual,
   predicted,
   p_rare = c(0.001, 0.003, 0.005, 0.008, 0.01, 0.03, 0.05),
   celltype = TRUE
)
```

## **Arguments**

| actual    | The groundtruth proportion of cell types in matrix or csv file. row: cell types, column: samples. Also can be a csv matrix with row and column names. |
|-----------|---|
| predicted | The predicted proportion of cell types in matrix or csv file. row: cell types, column: samples. Also can be a csv matrix with row and column names.   |
| p_rare    | A vector of proportions. should be the same with function <code>rareExprSim</code> . Default: $c(0.001,0.003,0.005,0.008,0.01,0.03,0.05)$             |
| celltype  | TRUE or FALSE. Assign different type and color for different cell types. Default: TRUE.   |

# Value

The plot data.

# **Examples**

```
rareExprSim()
# after the simulation, just pass the predicted file
# to the parameter "actual" and "predicted"
```

scatter\_simple

Scatter plot for proportion prediction.

# Description

Generate scatter plot comparison for different cell types.

# Usage

```
scatter_simple(actual, predicted, method, celltype = TRUE)
```

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### **Arguments**

actual The groundtruth proportion of cell types in matrix or csv file. row: cell types,

column: samples. Also can be a csv matrix with row and column names.

predicted The predicted proportion of cell types in matrix or csv file. row: cell types,

column: samples. Also can be a csv matrix with row and column names.

method "pearson" (default), "kendall", or "spearman".

celltype TRUE or FALSE. Assign different type and color for different cell types.

### Value

The plot data.

### **Examples**

```
res <- pseudoData(type = 1)
res <- scatter_simple(actual = res$actual, predicted = res$predicted, method = "pearson")</pre>
```

scExprSim

Generate in silico mixture expression matrix based on scRNA-seq

### **Description**

Generate in silico mixture expression matrix based on the internal scRNA-seq database. The internal RNA-seq database is collected from PMID:29474909. All the samples are passed the quality filter. This function use the real cell type specific expression data to generate mixture data. The cell type include c("FetalStomach", "FetalLung", "FetalLiver", "FetalKidney", "FetalIntestine", "FetalBrain", "Female.fetal.Gonad")

### Usage

```
scExprSim(
   n_sample = 50,
   p = 2/3,
   transform = "TPM",
   outputPath = NULL,
   mix_name = "scMouse_gene_expr.csv",
   ref_name = "scMouse_ref.csv",
   prop_name = "scMouse_prop.csv",
   train_name = "scMouse_ref_rawCount.csv")
```

# **Arguments**

n\_sample Sample number to be generated, default: 50.
p Proportion of sample in train set, default: 2 / 3.

transform "TPM", "CPM" or "NO". Transform the data into TPM, CPM or in counts.

outputPath output file save path.
mix\_name mixture output file name.

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ref\_name reference output file name in csv.

prop\_name simulated proportion file name in csv.

train\_name file name for all data in train set in csv. This data can be used for differential

gene analysis.

# Value

All the information will be written in the output path.

# **Examples**

```
res <- scExprSim()</pre>
```

 $\mathsf{TPM}$ 

Convert a gene expression count matrix into TPM matrix.

# Description

Convert a gene expression count matrix into TPM matrix

# Usage

```
TPM(data = NULL)
```

# **Arguments**

data

A dataframe or matrix object. Note: Must contain a column named "Length", "Length" means gene length.

# Value

A dataframe or matrix object.

```
data <- data.frame(s1 = seq(100), Length = seq(100)) res <- TPM(data)
```

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unExprSim

Generate in silico expression dataset with unknown component.

#### **Description**

Generate in silico mixture expression matrix based on the internal RNA-seq database. This function is different from the function exprSim. exprSim generates all the proportion randomly, while this function takes one cell type as unknown component. The internal RNA-seq database is collected from multiple studies. All the samples are passed the quality filter. We provide two types of simulation called "coarse" and "fine". This idea is from DREAM Challenge Tumor Deconvolution problem. https://www.synapse.org/#!Synapse:syn15589870/wiki/.

# Usage

```
unExprSim(
  unknown = NULL,
  n_sample = 50,
  p = 0.6,
  type = "coarse",
  transform = "TPM",
  outputPath = NULL,
  mix_name = "coarse_gene_expr.csv",
  ref_name = "coarse_ref.csv",
  prop_name = "coarse_prop.csv",
  ref_total_name = NULL,
  prop_total_name = NULL,
  train_name = NULL)
```

#### **Arguments**

|  | unknown | Character to select one cell type as unknown component. | Default: NULL, mean |
|--|---------|---|---------------------|
|--|---------|---|---------------------|

randomly select one cell type to drop.

n\_sample Sample number to be generated, default: 50.
p Proportion of sample in train set, default: 0.6.

type "coarse" or "fine". "coarse" means the simulation will be performed in a coarse

level. Only 8 cell types will be used, including ("B.cells", "CD4.T.cells", "CD8.T.cells", "endothelial.cells", "macrophages", "monocytes", "neutrophils", "NK.cells"). "fine" means the simulation will be performed in a fine level. 14 cell types will be used, including ("memory.B.cells", "naive.B.cells", "memory.CD4.T.cells", "naive.CD4.T.cells", "regulatory.T.cells", "memory.CD8.T.cells", "naive.CD8.T.cells", "NK.cells", "neutrophils", "monocytes", "myeloid.dendritic.cells", "macrophages",

"fibroblasts", "endothelial.cells")

transform "TPM", "CPM" or "NO". Transform the data into TPM, CPM or in counts.

outputPath output file save path.
mix\_name mixture output file name.

ref\_name reference output file name in csv.

prop\_name simulated proportion file name in csv, unknown component will be removed.

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```
ref_total_name simulated reference file name in csv, unknown component is contained.

prop_total_name simulated proportion file name in csv, unknown component is contained.

refVar_name reference variance file name in csv.

train_name file name for all data in train set in csv. This data can be used for differential gene analysis.
```

### Value

All the information will be written in the output path.

### **Examples**

```
res <- unExprSim()</pre>
```

unscExprSim

Generate in silico mixture expression matrix based on scRNA-seq

## **Description**

Generate in silico mixture expression matrix based on the internal scRNA-seq database. The internal RNA-seq database is collected from PMID:29474909. All the samples are passed the quality filter. This function use the real cell type specific expression data to generate mixture data. The cell type include c("FetalStomach", "FetalLung", "FetalLiver", "FetalKidney", "FetalIntestine", "FetalBrain", "Female.fetal.Gonad")

# Usage

```
unscExprSim(
  unknown = NULL,
  n_sample = 50,
  p = 2/3,
  transform = "TPM",
  outputPath = NULL,
  mix_name = "scMouse_gene_expr.csv",
  ref_name = "scMouse_ref.csv",
  prop_name = "scMouse_prop.csv",
  train_name = "scMouse_ref_rawCount.csv",
  ref_total_name = NULL,
  prop_total_name = NULL
)
```

### **Arguments**

| unknown   | Character to select one cell type as unknown component. Default: NULL, mean randomly select one cell type to drop. |
|-----------|--|
| n_sample  | Sample number to be generated, default: 50.  |
| p         | Proportion of sample in train set, default: 2 / 3.   |
| transform | "TPM", "CPM" or "NO". Transform the data into TPM, CPM or in counts.   |

v\_norm

outputPath output file save path.

mix\_name mixture output file name.

ref\_name reference output file name in csv.

prop\_name simulated proportion file name in csv.

train\_name file name for all data in train set in csv. This data can be used for differential

gene analysis.

ref\_total\_name simulated reference file name in csv, unknown component is contained.

prop\_total\_name

simulated proportion file name in csv, unknown component is contained.

### Value

All the information will be written in the output path.

# **Examples**

```
res <- unscExprSim()</pre>
```

v\_norm

Normalize a vector

# Description

Normalize a vector

## Usage

```
v_norm(x = NULL)
```

# **Arguments**

Χ

a numeric vector

### Value

Normalized numeric vector

```
res <- v_norm(seq(100))
```

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