

# Package ‘decone’

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**Type** Package

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**Title** Cell Type Deconvolution Evaluator

**Version** 1.0.0

**Description** The evaluation platform for cell type deconvolution. This platform provides multiple simulation and real mixture expression datasets for evaluate the performance of deconvlution tools. This toolkit also plots preety figure for the comparison of different deconvolution methods.

**License** GPL-3 | file LICENSE

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**LazyData** true

**RoxygenNote** 7.2.1

**Depends** R (>= 4.1.0)

**Imports** Metrics,  
stats,  
utils,  
Matrix,  
tibble,  
data.table,  
tools,  
magrittr,  
ggplot2,  
pheatmap,  
reshape2,  
grDevices,  
forcats,  
stringr,  
R.utils,  
ggpubr,  
progress

**Suggests** knitr,  
rmarkdown

**VignetteBuilder** knitr

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---

addNoise	<i>Add noise to gene expression data</i>
----------	--

---

### Description

Add noise based on negative binomial 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.

**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 <a href="#">exprSim</a> .
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.

**Examples**

```
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")
```

---

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 = "")
```

### 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. Default: c("condition1", "condition2", "condition3", ...)
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.

### Examples

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

---

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.

**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")
```

---

boxplot_simple	<i>Boxplot of deconvolution results for a specific method.</i>
----------------	--

---

**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: samples. 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.

**Examples**

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

---

```
heatmap_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**

```
heatmap_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_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 <- heatmap_NcrossCompare(actual = res$actual, predicted = res$predicted,
label = res$noise_level, method1 = "pearson", method2 = "rmse")
```

---

heatmap\_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](#).

## Usage

```
heatmap_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 <a href="#">rareExprSim</a> . 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)
```

**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)
```

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.
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.
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. 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.

**Examples**

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

---

heatmap_NGradCT	<i>Plot heatmap for different cell types.</i>
-----------------	---

---

**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.

**Examples**

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

---

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 <a href="#">rareExprSim</a> . 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.

**Examples**

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

---

pseudoData	<i>Generate pseudo data for plotting</i>
------------	--

---

**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)
```

---

pseudoExpr	<i>Generate pseudo mixture expression matrix</i>
------------	--

---

**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	Sample number to be generated, default: 50
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

**Examples**

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

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
)

```

**Arguments**

<code>p_rare</code>	A vector of proportions. Default: <code>c(0.001, 0.003, 0.005, 0.008, 0.01, 0.03, 0.05)</code> 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.
<code>p</code>	Proportion of sample in train set, default: 0.6.
<code>type</code>	"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")
<code>transform</code>	"TPM", "CPM" or "NO". Transform the data into TPM, CPM or in counts.
<code>outputPath</code>	output file save path.
<code>mix_name</code>	mixture output file name.
<code>ref_name</code>	reference output file name in csv.
<code>prop_name</code>	simulated proportion file name in csv.
<code>refVar_name</code>	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

```
rareExprSim()
```

---

<code>rarescExprSim</code>	<i>Generate in silico mixture expression matrix based on scRNA-seq with rare component.</i>
----------------------------	---

---

### 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", "Fetal-Brain", "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

<code>p_rare</code>	A vector of proportions. Default: c(0.001, 0.003, 0.005, 0.008, 0.01, 0.03, 0.05) 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.
<code>p</code>	Proportion of sample in train set, default: 2 / 3.
<code>transform</code>	"TPM", "CPM" or "NO". Transform the data into TPM, CPM or in counts.
<code>outputPath</code>	output file save path.
<code>mix_name</code>	mixture output file name.
<code>ref_name</code>	reference output file name in csv.
<code>prop_name</code>	simulated proportion file name in csv.
<code>train_name</code>	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**

```
rarescExprSim()
```

---

regMetrics	<i>Compute regression Metrics for matrix results.</i>
------------	---

---

**Description**

Compute several regression Metrics for cell type deconvolution.

**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: samples.
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.

**Examples**

```
res <- regMetrics(actual = matrix(data = seq(9), nrow = 3),
  predicted = matrix(data = seq(9), nrow = 3),
  method = "rmse")
```

---

scatter_R	<i>Scatter plot for rare component.</i>
-----------	---

---

### 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 <a href="#">rareExprSim</a> . 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	<i>Scatter plot for proportion prediction.</i>
----------------	--

---

### Description

Generate scatter plot comparison for different cell types.

### Usage

```
scatter_simple(actual, predicted, method, 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.
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")
```

---

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.

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()
```

---

TPM	<i>Convert a gene expression count matrix into TPM matrix.</i>
-----	--

---

**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.

**Examples**

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

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,
  refVar_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.

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()
```

---

unscExprSim	<i>Generate in silico mixture expression matrix based on scRNA-seq</i>
-------------	--

---

### 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.

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()
```

---

v_norm	<i>Normalize a vector</i>
--------	---------------------------

---

**Description**

Normalize a vector

**Usage**

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

**Arguments**

x                      a numeric vector

**Value**

Normalized numeric vector

**Examples**

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

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