

# Package ‘NLCD’

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**Type** Package

**Title** Network Local Consistency Detection

**Version** 1.0.0

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**Description** Using Local Moran's I for Network local consistency Detection

**License** GPL >= 2

**Imports** igraph, spdep, fdrtool, GOstats, locfdr, mvtnorm, caTools

**LazyData** true

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---

cal_lmi_data	<i>calculate lmi for a network</i>
--------------	------------------------------------

---

### Description

cal\_lmi\_data() will calculate the lmi data for a input network.

### Usage

```
cal_lmi_data(gene_expr, gene_graph)
```

### Arguments

gene_expr	expression for gene
gene_graph	The graph of gene network.

### Details

this function will create a table of lmi data, and each column in this table stand for one lmi of one gene expression.

### Value

a table for lmi data. Each line of this table stand for one lmi of one gene expression.

### Examples

```
data("gene_graph")
data("gene_expr")
data("patient_raw_data")
lmi_data <- cal_lmi_data(gene_expr, gene_graph)
patient_data<-solve_patient_data(patient_raw_data)
t_data <- top3_bottom3_t_data(gene_graph, patient_data, lmi_data)
fdr_result <- gene_fdrtest(t_data)
sig_genes <- significant_genes(fdr_obj = fdr_result, thres = 0.2)
## The function is currently defined as
function (gene_expr, gene_graph)
{
  dis_mat <- n_hop_matrix(gene_graph, hop_n = 2)
  listw_obj <- spdep::mat2listw(dis_mat, style = "W")
  gene_expr <- t(apply(gene_expr, 1, function(gene_expr) (gene_expr -
    min(gene_expr))/(max(gene_expr) - min(gene_expr))))
  lmi_data <- double()
  for (i in seq(1, ncol(gene_expr), 1)) {
    lmi <- spdep::localmoran(gene_expr[, i], listw = listw_obj)
    lmi_data <- cbind(lmi_data, lmi[, 1])
  }
}
```

```
    return(lmi_data)
}
```

---

gene_expr	<i>gene expression</i>
-----------	------------------------

---

**Description**

a data set for gene expression.

**Usage**

```
data("gene_expr")
```

**Format**

A data frame with 7965 observations on the following 161 variables.

GSM258912 a numeric vector  
GSM258913 a numeric vector  
GSM258914 a numeric vector  
GSM258915 a numeric vector  
GSM258916 a numeric vector  
GSM258917 a numeric vector  
GSM258918 a numeric vector  
GSM258919 a numeric vector  
GSM258920 a numeric vector  
GSM258921 a numeric vector  
GSM258922 a numeric vector  
GSM258923 a numeric vector  
GSM258924 a numeric vector  
GSM258925 a numeric vector  
GSM258926 a numeric vector  
GSM258927 a numeric vector  
GSM258928 a numeric vector  
GSM258929 a numeric vector  
GSM258930 a numeric vector  
GSM258931 a numeric vector  
GSM258932 a numeric vector  
GSM258933 a numeric vector  
GSM258934 a numeric vector

GSM258935 a numeric vector  
GSM258936 a numeric vector  
GSM258937 a numeric vector  
GSM258938 a numeric vector  
GSM258939 a numeric vector  
GSM258940 a numeric vector  
GSM258941 a numeric vector  
GSM258942 a numeric vector  
GSM258943 a numeric vector  
GSM258944 a numeric vector  
GSM258945 a numeric vector  
GSM258946 a numeric vector  
GSM258947 a numeric vector  
GSM258948 a numeric vector  
GSM258949 a numeric vector  
GSM258950 a numeric vector  
GSM258951 a numeric vector  
GSM258952 a numeric vector  
GSM258953 a numeric vector  
GSM258954 a numeric vector  
GSM258955 a numeric vector  
GSM258956 a numeric vector  
GSM258957 a numeric vector  
GSM258958 a numeric vector  
GSM258959 a numeric vector  
GSM258960 a numeric vector  
GSM258961 a numeric vector  
GSM258962 a numeric vector  
GSM258963 a numeric vector  
GSM258964 a numeric vector  
GSM258965 a numeric vector  
GSM258966 a numeric vector  
GSM258967 a numeric vector  
GSM258968 a numeric vector  
GSM258969 a numeric vector  
GSM258970 a numeric vector  
GSM258971 a numeric vector

GSM258972 a numeric vector  
GSM258973 a numeric vector  
GSM258974 a numeric vector  
GSM258975 a numeric vector  
GSM258976 a numeric vector  
GSM258977 a numeric vector  
GSM258978 a numeric vector  
GSM258979 a numeric vector  
GSM258980 a numeric vector  
GSM258981 a numeric vector  
GSM258982 a numeric vector  
GSM258983 a numeric vector  
GSM258984 a numeric vector  
GSM258985 a numeric vector  
GSM258986 a numeric vector  
GSM258987 a numeric vector  
GSM258988 a numeric vector  
GSM258989 a numeric vector  
GSM258990 a numeric vector  
GSM258991 a numeric vector  
GSM258992 a numeric vector  
GSM258993 a numeric vector  
GSM258994 a numeric vector  
GSM258995 a numeric vector  
GSM258996 a numeric vector  
GSM258997 a numeric vector  
GSM258998 a numeric vector  
GSM258999 a numeric vector  
GSM259000 a numeric vector  
GSM259001 a numeric vector  
GSM259002 a numeric vector  
GSM259003 a numeric vector  
GSM259004 a numeric vector  
GSM259005 a numeric vector  
GSM259006 a numeric vector  
GSM259007 a numeric vector  
GSM259008 a numeric vector

GSM259009 a numeric vector  
GSM259010 a numeric vector  
GSM259011 a numeric vector  
GSM259012 a numeric vector  
GSM259013 a numeric vector  
GSM259014 a numeric vector  
GSM259015 a numeric vector  
GSM259016 a numeric vector  
GSM259017 a numeric vector  
GSM259018 a numeric vector  
GSM259019 a numeric vector  
GSM259020 a numeric vector  
GSM259021 a numeric vector  
GSM259022 a numeric vector  
GSM259023 a numeric vector  
GSM259024 a numeric vector  
GSM259025 a numeric vector  
GSM259026 a numeric vector  
GSM259027 a numeric vector  
GSM259028 a numeric vector  
GSM259029 a numeric vector  
GSM259030 a numeric vector  
GSM259031 a numeric vector  
GSM259032 a numeric vector  
GSM259033 a numeric vector  
GSM259034 a numeric vector  
GSM259035 a numeric vector  
GSM259036 a numeric vector  
GSM259037 a numeric vector  
GSM259038 a numeric vector  
GSM259039 a numeric vector  
GSM259040 a numeric vector  
GSM259041 a numeric vector  
GSM259042 a numeric vector  
GSM259043 a numeric vector  
GSM259044 a numeric vector  
GSM259045 a numeric vector

GSM259046 a numeric vector  
GSM259047 a numeric vector  
GSM259048 a numeric vector  
GSM259049 a numeric vector  
GSM259050 a numeric vector  
GSM259051 a numeric vector  
GSM259052 a numeric vector  
GSM259053 a numeric vector  
GSM259054 a numeric vector  
GSM259055 a numeric vector  
GSM259056 a numeric vector  
GSM259057 a numeric vector  
GSM259058 a numeric vector  
GSM259059 a numeric vector  
GSM259060 a numeric vector  
GSM259061 a numeric vector  
GSM259062 a numeric vector  
GSM259063 a numeric vector  
GSM259064 a numeric vector  
GSM259065 a numeric vector  
GSM259066 a numeric vector  
GSM259067 a numeric vector  
GSM259068 a numeric vector  
GSM259069 a numeric vector  
GSM259070 a numeric vector  
GSM259071 a numeric vector  
GSM259072 a numeric vector

**Examples**

```
data(gene_expr)
```

gene\_fdrtest

*Using locfdr to calculate fdr\_result for a t.test result***Description**

This function use locfdr function to calculate fdr\_result

**Usage**

```
gene_fdrtest(t_data)
```

**Arguments**

gene.corr            gene\_id\_all: all gene id t\_data: t\_data for each gene

**Value**

return fdr\_result for t\_data

fdr\$name            all gene id

fdr\$fdr            fdr value for gene

**Examples**

```
data("gene_graph")
data("gene_expr")
data("patient_raw_data")
lmi_data <- cal_lmi_data(gene_expr, gene_graph)
patient_data<-solve_patient_data(patient_raw_data)
t_data <- top3_bottom3_t_data(gene_graph, patient_data, lmi_data)
fdr_result <- gene_fdrtest(t_data)
sig_genes <- significant_genes(fdr_obj = fdr_result, thres = 0.2)

## The function is currently defined as
function (gene.corr)
{
  name <- gene.corr$gene_id_all
  coef_val <- gene.corr$t_data
  z.cent <- scale(coef_val)
  fdr_result <- data.frame(fdr = locfdr(z.cent, nulltype = 0,
    plot = 4)$fdr, name)
  return(fdr_result)
}
```



---

gene_graph	<i>gene graph</i>
------------	-------------------

---

**Description**

a data set of gene graph.

**Usage**

```
data("gene_graph")
```

**Format**

The format is: List of 10 \$ : num 7965 \$ : logi FALSE \$ : num [1:29996] 1 3 5 7 8 9 10 11 13 14 ... \$ : num [1:29996] 0 2 4 6 6 6 6 6 12 12 ... \$ : num [1:29996] 0 1 2 3 4 ... \$ : num [1:29996] 0 3323 1565 18196 18218 ... \$ : num [1:7966] 0 0 1 1 2 2 3 3 4 5 ... \$ : num [1:7966] 0 1 111 112 145 146 379 384 509 547 ... \$ :List of 4 ..\$ : num [1:3] 1 0 1 ..\$ : Named list() ..\$ :List of 1 .. ..\$ name: chr [1:7965] "27004" "4223" "151050" "10133" ... ..\$ : list() \$ :<environment: 0x362b2b0> - attr(\*, "class")= chr "igraph"

**Examples**

```
data(gene_graph)
```

---

init_simulation_gene_net	<i>Creating a simulation network for further calculation.</i>
--------------------------	---

---

**Description**

This function will create a network for NLCD. This function will change correlation of chosen gene's and its one hop neighbor.

**Usage**

```
init_simulation_gene_net(base_correlation = 0.4,  
  change_correlation = 0.8, sample_size = 100, num_gene = 5000)
```

**Arguments**

base_correlation	base correlation of network
change_correlation	change correlation of network
sample_size	multi size of patient data
num_gene	gene num in the network

**Value**

lmi\_matrix        matrix of lmi data  
 patient\_matrix        matrix of patients' status  
 gene\_code\_list   code of gene.0 for normal gene; 1 for changed gene; 2 for chosen gene  
 neigh\_list        id of gene which is changed.

**Examples**

```
##---- Should be DIRECTLY executable !! ----
##-- ==> Define data, use random,
##--or do help(data=index) for the standard data sets.

## The function is currently defined as
function (base_correlation = 0.4, change_correlation = 0.8, sample_size = 100,
  num_gene = 5000)
{
  n_hop_matrix <- function(gene_graph, hop_n) {
    dis_mat <- igraph::shortest.paths(gene_graph)
    dis_mat[dis_mat > hop_n] <- 0
    dis_mat[dis_mat == hop_n] <- 0.5
    return(dis_mat)
  }
  netdensity = 1
  gene_graph <- igraph::barabasi.game(num_gene, m = netdensity)
  deg_data <- igraph::degree(gene_graph)
  simu_sel_gene_list = numeric()
  neigh_list_3 <- numeric()
  simu_gene_num = 5
  times = 100
  while (length(simu_sel_gene_list) < simu_gene_num) {
    sel_gene <- sample(which(deg_data <= 10 & deg_data >=
      5), 1)
    if (sel_gene %in% neigh_list_3) {
      times = times - 1
      if (times < 0) {
        return(NULL)
      }
      (next)()
    }
    simu_sel_gene_list <- c(simu_sel_gene_list, sel_gene)
    neigh_list_3 <- c(neigh_list_3, unlist(igraph::neighborhood(gene_graph,
      order = 3, nodes = sel_gene, mode = "all")))
  }
  neigh_list <- numeric()
  for (sel_gene in simu_sel_gene_list) {
    neigh_list <- c(neigh_list, unlist(igraph::neighborhood(gene_graph,
      order = 1, nodes = sel_gene, mode = "all")))
  }
  gene_code_list <- rnorm(num_gene, mean = 0, sd = 0)
  for (gid in neigh_list) {
```

```

        gene_code_list[gid] <- 1
    }
    for (gid in simu_sel_gene_list) {
        gene_code_list[gid] <- 2
    }
    s <- igraph::shortest.paths(gene_graph)
    s1 <- base_correlation^(s)
    x <- t(mvtnorm::rmvnorm(n = sample_size, mean = rep(0, num_gene),
        sigma = s1, method = "chol"))
    x <- t(apply(x, 1, function(x) (x - min(x))/(max(x) - min(x))))
    y <- c(rnorm(sample_size, mean = 0, sd = 0), rnorm(sample_size,
        mean = 1, sd = 0))
    dis_mat <- n_hop_matrix(gene_graph = gene_graph, hop_n = 2)
    listw_obj <- spdep::mat2listw(dis_mat, style = "W")
    origin_lmi_data <- double()
    for (i in seq(1, ncol(x), 1)) {
        lmi <- spdep::localmoran(x[, i], listw = listw_obj)
        origin_lmi_data <- cbind(origin_lmi_data, lmi[, 1])
    }
    for (i in neigh_list) {
        for (j in neigh_list) {
            s1[i, j] <- change_correlation^s[i, j]
        }
    }
    x <- t(mvtnorm::rmvnorm(n = sample_size, mean = rep(0, num_gene),
        sigma = s1, method = "chol"))
    x <- t(apply(x, 1, function(x) (x - min(x))/(max(x) - min(x))))
    y <- c(rnorm(sample_size, mean = 0, sd = 0), rnorm(sample_size,
        mean = 1, sd = 0))
    lmi_data <- double()
    for (i in seq(1, ncol(x), 1)) {
        lmi <- spdep::localmoran(x[, i], listw = listw_obj)
        lmi_data <- cbind(lmi_data, lmi[, 1])
    }
    lmi_matrix <- cbind(lmi_data, origin_lmi_data)
    return(list(lmi_matrix = lmi_matrix, patient_matrix = y,
        gene_code_list = gene_code_list, neigh_list = neigh_list))
}

```

## Description

Using Local Moran's I for Network local consistency Detection

## Details

The DESCRIPTION file: This package was not yet installed at build time.

Index: This package was not yet installed at build time.  
sing Local Moran's I for Network local consistency Detection

### Author(s)

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### Examples

```
data("gene_graph")
data("gene_expr")
data("patient_raw_data")
lmi_data <- cal_lmi_data(gene_expr, gene_graph)
patient_data<-solve_patient_data(patient_raw_data)
t_data <- top3_bottom3_t_data(gene_graph, patient_data, lmi_data)
fdr_result <- gene_fdrtest(t_data)
sig_genes <- significant_genes(fdr_obj = fdr_result, thres = 0.2)
```

---

n_hop_matrix	<i>Turning a gene graph to a matrix.</i>
--------------	--

---

### Usage

```
n_hop_matrix(gene_graph, hop_n)
```

### Arguments

gene_graph	The graph of the gene network.
hop_n	the hop should be described in the matrix

### Value

a matrix describe the connection of the graph. 0.5 stand for the shortest path between two node equal to hop\_n

### See Also

cal\_lmi\_data() init\_simulation\_gene\_net()

**Examples**

```
##---- Should be DIRECTLY executable !! ----
##-- ==> Define data, use random,
##--or do help(data=index) for the standard data sets.

## The function is currently defined as
function (gene_graph, hop_n)
{
  dis_mat <- igraph::shortest.paths(gene_graph)
  dis_mat[dis_mat > hop_n] <- 0
  dis_mat[dis_mat == hop_n] <- 0.5
  return(dis_mat)
}
```

---

patient_raw_data	<i>patient raw data</i>
------------------	-------------------------

---

**Description**

a data set of patient situation

**Usage**

```
data("patient_raw_data")
```

**Format**

A data frame with 161 observations on the following 7 variables.

```
cel a factor with levels JD0001-ALL-v5-U133A JD0002-ALL-v5-U133A JD0003-ALL-v5-U133A
JD0004-ALL-v5-U133A JD0005-ALL-v5-U133A JD0006-ALL-v5-U133A JD0008-ALL-v5-U133A
JD0009-ALL-v5-U133A JD0011-ALL-v5-U133A JD0012-ALL-v5-U133A JD0013-ALL-v5-U133A
JD0015-ALL-v5-U133A JD0016-ALL-v5-U133A JD0017-ALL-v5-U133A JD0018-ALL-v5-U133A
JD0019-ALL-v5-U133A JD0020-ALL-v5-U133A JD0022-ALL-v5-U133A JD0023-ALL-v5-U133A
JD0025-ALL-v5-U133A JD0030-ALL-v5-U133A JD0031-ALL-v5-U133A JD0032-ALL-v5-U133A
JD0036-ALL-v5-U133A JD0037-ALL-v5-U133A JD0038-ALL-v5-U133A JD0039-ALL-v5-U133A
JD0041-ALL-v5-U133A JD0042-ALL-v5-U133A JD0043-ALL-v5-U133A JD0044-ALL-v5-U133A
JD0047-ALL-v5-U133A JD0048-ALL-v5-U133A JD0050-ALL-v5-U133A JD0055-ALL-v5-U133A
JD0056-ALL-v5-U133A JD0057-ALL-v5-U133A JD0058-ALL-v5-U133A JD0059-ALL-v5-U133A
JD0062-ALL-v5-U133A JD0064-ALL-v5-U133A JD0065-ALL-v5-U133A JD0066-ALL-v5-U133A
JD0067-ALL-v5-U133A JD0073-ALL-v5-U133A JD0074-ALL-v5-U133A JD0076-ALL-v5-U133A
JD0079-ALL-v5-U133A JD0080-ALL-v5-U133A JD0083-ALL-v5-U133A JD0084-ALL-v5-U133A
JD0085-ALL-v5-U133A JD0088-ALL-v5-U133A JD0095-ALL-v5-U133A JD0096-ALL-v5-U133A
JD0097-ALL-v5-U133A JD0098-ALL-v5-U133A JD0099-ALL-v5-U133A JD0100-ALL-v5-U133A
JD0101-ALL-v5-U133A JD0102-ALL-v5-U133A JD0104-ALL-v5-U133A JD0105-ALL-v5-U133A
JD0106-ALL-v5-U133A JD0107-ALL-v5-U133A JD0108-ALL-v5-U133A JD0109-ALL-v5-U133A
JD0110-ALL-v5-U133A JD0111-ALL-v5-U133A JD0117-ALL-v5-U133A JD0118-ALL-v5-U133A
JD0119-ALL-v5-U133A JD0120-ALL-v5-U133A JD0121-ALL-v5-U133A JD0122-ALL-v5-U133A
```

JD0123-ALL-v5-U133A JD0124-ALL-v5-U133A JD0125-ALL-v5-U133A JD0126-ALL-v5-U133A  
 JD0127-ALL-v5-U133A JD0130-ALL-v5-U133A JD0131-ALL-v5-U133A JD0133-ALL-v5-U133A  
 JD0135-ALL-v5-U133A JD0137-ALL-v5-U133A JD0138-ALL-v5-U133A JD0139-ALL-v5-U133A  
 JD0140-ALL-v5-U133A JD0144-ALL-v5-U133A JD0149-ALL-v5-U133A JD0151-ALL-v5-U133A  
 JD0163-ALL-v5-U133A JD0165-ALL-v5-U133A JD0166-ALL-v5-U133A JD0167-ALL-v5-U133A  
 JD0173-ALL-v5-U133A JD0188-ALL-v5-U133A JD0196-ALL-v5-U133A JD0202-ALL-v5-U133A  
 JD0203-ALL-v5-U133A JD0212-ALL-v5-U133A JD0219-ALL-v5-U133A JD0221-ALL-v5-U133A  
 JD0225-ALL-v5-U133A JD0226-ALL-v5-U133A JD0227-ALL-v5-U133A JD0228-ALL-v5-U133A  
 JD0233-ALL-v5-U133A JD0236-ALL-v5-U133A JD0245-ALL-v5-U133A JD0246-ALL-v5-U133A  
 JD0247-ALL-v5-U133A JD0249-ALL-v5-U133A JD0252-ALL-v5-U133A JD0255-ALL-v5-U133A  
 JD0256-ALL-v5-U133A JD0258-ALL-v5-U133A JD0264- ALL-v5-U133A JD0265- ALL-v5-U133A  
 JD0266- ALL-v5-U133A JD0267- ALL-v5-U133A JD0268- ALL-v5-U133A JD0269- ALL-v5-U133A  
 JD-ALD004-v5-U133A JD-ALD009-v5-U133A JD-ALD054-v5-U133A JD-ALD066-v5-U133A JD-ALD071-v5-U133A  
 JD-ALD078-v5-U133A JD-ALD083-v5-U133A JD-ALD108-v5-U133A JD-ALD113-v5-U133A JD-ALD115-v5-U133A  
 JD-ALD163-v5-U133A JD-ALD196-v5-U133A JD-ALD258-v5-U133A JD-ALD262-v5-U133A JD-ALD283-v5-U133A  
 JD-ALD360-v5-U133A JD-ALD374-v5-U133A JD-ALD375-v5-U133A JD-ALD428-v5-U133A JD-ALD431-v5-U133A  
 JD-ALD436-v5-U133A JD-ALD437-v5-U133A JD-ALD485-v5-U133A JD-ALD488-v5-U133A JD-ALD491-v5-U133A  
 JD-ALD493-v5-U133A JD-ALD494-v5-U133A JD-ALD510-v5-U133A JD-ALD515-v5-U133A JD-ALD537-v5-U133A  
 JD-ALD540-v5-U133A JD-ALD553-v5-U133A JD-ALD557-v5-U133A JD-ALD610-v5-U133A JD-ALD611-v5-U133A  
 JD-ALD612-v5-U133A JD-ALD613-v5-U133A JD-ALD619-v5-U133A

GEO.id a numeric vector

WBCCCountDay0 a numeric vector

WBCCCountDay3 a numeric vector

WBCCCountDay0Log a numeric vector

WBCCCountDay3Log a numeric vector

residuals.WBCCCountDay3Log.WBCCCountDay0Log. a numeric vector

## Examples

```
data(patient_raw_data)
```

---

significant_genes	<i>Selecting significant genes according to fdr result</i>
-------------------	--

---

## Usage

```
significant_genes(fdr_obj = fdr_result, thres = 0.2)
```

## Arguments

fdr_obj	fdr result come from function gene_fdrtest
thres	threshold to identify significant genes

## Value

ID of significant genes

**Examples**

```

data("gene_graph")
data("gene_expr")
data("patient_raw_data")
lmi_data <- cal_lmi_data(gene_expr, gene_graph)
patient_data<-solve_patient_data(patient_raw_data)
t_data <- top3_bottom3_t_data(gene_graph, patient_data, lmi_data)
fdr_result <- gene_fdrtest(t_data)
sig_genes <- significant_genes(fdr_obj = fdr_result, thres = 0.2)

## The function is currently defined as
function (fdr_obj, thres)
{
  sel.entrez <- fdr_obj$name[which(fdr_obj$fdr < thres)]
  return(sel.entrez)
}

```

simulate\_NLCD

*simulate NLCD***Description**

simulate NLCD

**Usage**

```

simulate_NLCD(base_correlation = 0.4,
              change_correlation = 0.8, sample_size = c(50, 100), num_gene = 5000)

```

**Arguments**

base_correlation	base correlation of network
change_correlation	change correlation of network
sample_size	multi size of patient data
num_gene	gene num in the network

**Value**

picture for fdr\_result

## Examples

```
##---- Should be DIRECTLY executable !! ----
##-- ==> Define data, use random,
##--or do help(data=index) for the standard data sets.

## The function is currently defined as
function (base_correlation = 0.4, change_correlation = 0.8, sample_size = c(50,
  100), num_gene = 5000)
{
  simulation <- init_simulation_gene_net(base_correlation = base_correlation,
    change_correlation = change_correlation, sample_size = max(sample_size),
    num_gene = num_gene)
  over = 0
  while (is.null(simulation)) {
    over = over + 1
    simulation <- init_simulation_gene_net(base_correlation = base_correlation,
      change_correlation = change_correlation, sample_size = max(sample_size),
      num_gene = num_gene)
  }
  lmi_data = simulation$lmi_matrix
  patient_matrix = simulation$patient_matrix
  gene_code_list = simulation$gene_code_list
  neigh_list = simulation$neigh_list
  labels = gene_code_list
  simu_sel_gene = which(gene_code_list == 2)
  labels[which(labels == 2)] = 1
  gene_ids <- seq(1, num_gene, 1)
  sample_size = sort(sample_size, decreasing = TRUE)
  left = seq(1, max(sample_size))
  right = seq(1, max(sample_size)) + max(sample_size)
  for (sz in sample_size) {
    left = sample(left, sz)
    right = sample(right, sz)
    pl = cbind(left, right)
    fdr_result <- sim_cal_fdr_result(gene_ids, patient_matrix[pl],
      lmi_data[, pl])
  }
}
```

---

sim_cal_fdr_result	<i>calucating fdr result for simulation network</i>
--------------------	---

---

## Description

calucating fdr result for simulation network. The reason why we create this function is the simulation data structure is different from real data set.

## Usage

```
sim_cal_fdr_result(gene_ids, patient_matrix, lmi_data)
```



**Arguments**

gene\_ids            id for all genes  
 patient\_matrix    patients' data  
 lmi\_data           lmi data for each gene

**Value**

fdr result for gene

**Examples**

```
##---- Should be DIRECTLY executable !! ----
##-- ==> Define data, use random,
##--or do help(data=index) for the standard data sets.

## The function is currently defined as
function (gene_ids, patient_matrix, lmi_data)
{
  gene_fdrtest <- function(gene.corr) {
    name <- gene.corr$gene_id_all
    coef_val <- gene.corr$t_data
    z.cent <- coef_val - median(coef_val)
    fdr_result <- data.frame(fdr = fdrtool::fdrtool(z.cent, statistic = "normal"),
      name)
    return(fdr_result)
  }
  t_gene_test <- function(clinic_data, gene_id, lmi_data) {
    gene.corr <- data.frame(clinic = clinic_data, lmi = lmi_data[which(gene_ids ==
      gene_id), ])
    result <- t.test(gene.corr$lmi[which(gene.corr$clinic ==
      0)], gene.corr$lmi[which(gene.corr$clinic == 1)])
    return(result$statistic)
  }
  top3_bottom3_t_data <- function(gene_id_all, patient_matrix,
    lmi_data) {
    t_data = numeric()
    for (single_gene in gene_id_all) {
      t_ts <- t_gene_test(clinic_data = patient_matrix,
        gene_id = single_gene, lmi_data = lmi_data)
      t_data <- append(t_data, t_ts)
    }
    return(data.frame(gene_id_all = gene_id_all, t_data = t_data))
  }
  p <- top3_bottom3_t_data(gene_ids, patient_matrix, lmi_data)
  fdr_result <- gene_fdrtest(p)
  return(fdr_result)
}
```

---

solve_patient_data	<i>Changing the sturcture of raw patient data</i>
--------------------	---

---

**Description**

Changing the sturcture of raw patient data

**Usage**

```
solve_patient_data(patient_raw_data)
```

**Arguments**

patient\_raw\_data  
Unprocessed patient data

**Value**

data of patient which can be used in t.test calculation process

**Examples**

```
data("gene_graph")
data("gene_expr")
data("patient_raw_data")
lmi_data <- cal_lmi_data(gene_expr, gene_graph)
patient_data<-solve_patient_data(patient_raw_data)
t_data <- top3_bottom3_t_data(gene_graph, patient_data, lmi_data)
fdr_result <- gene_fdrtest(t_data)
sig_genes <- significant_genes(fdr_obj = fdr_result, thres = 0.2)

## The function is currently defined as
function (patient_raw_data)
{
  patient_data <- patient_raw_data$WBCCCountDay3Log - patient_raw_data$WBCCCountDay0Log
  return(patient_data)
}
```

---

top3_bottom3_t_data	<i>Using t.test to calculate the t_data</i>
---------------------	---

---

**Description**

using t.test to calculate the t\_data.

**Usage**

```
top3_bottom3_t_data(gene_graph, patient_data, lmi_data)
```

**Arguments**

gene_graph	The graph of gene network.
patient_data	The matrix of patient data
lmi_data	The data of local moran's I

**Value**

gene\_id\_all: id of all genes t\_data: matrix of t.test result

**Examples**

```
data("gene_graph")
data("gene_expr")
data("patient_raw_data")
lmi_data <- cal_lmi_data(gene_expr, gene_graph)
patient_data <- solve_patient_data(patient_raw_data)
t_data <- top3_bottom3_t_data(gene_graph, patient_data, lmi_data)
fdr_result <- gene_fdrtest(t_data)
sig_genes <- significant_genes(fdr_obj = fdr_result, thres = 0.2)

## The function is currently defined as
function (gene_graph, patient_data, lmi_data)
{
  gene_ids <- igraph::V(gene_graph)$name
  t_data = numeric()
  for (gene_id in gene_ids) {
    sel = which(gene_ids == gene_id)
    lmi = lmi_data[sel, ]
    gene.corr <- data.frame(clinic = patient_data, lmi = lmi)
    gg <- gene.corr[order(gene.corr$clinic), ]
    t.result <- t.test(gg[1:53, 2], gg[106:161, 2])
    t_data <- append(t_data, t.result$statistic)
  }
  return(data.frame(gene_id_all = gene_ids, t_data = t_data))
}
```

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