

# Package ‘MoonlightR’

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**Type** Package

**Title** Identify oncogenes and tumor suppressor genes from omics data

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**Depends** R (>= 3.5), doParallel, foreach

**Imports** parmigene, randomForest, SummarizedExperiment, gplots, circlize, RColorBrewer, HiveR, clusterProfiler, DOSE, Biobase, limma, grDevices, graphics, TCGAbiolinks, GEOquery, stats, RISmed, grid, utils

**Description** Motivation: The understanding of cancer mechanism requires the identification of genes playing a role in the development of the pathology and the characterization of their role (notably oncogenes and tumor suppressors). Results: We present an R/bioconductor package called MoonlightR which returns a list of candidate driver genes for specific cancer types on the basis of TCGA expression data. The method first infers gene regulatory networks and then carries out a functional enrichment analysis (FEA) (implementing an upstream regulator analysis, URA) to score the importance of well-known biological processes with respect to the studied cancer type. Eventually, by means of random forests, MoonlightR predicts two specific roles for the candidate driver genes: i) tumor suppressor genes (TSGs) and ii) oncogenes (OCGs). As a consequence, this methodology does not only identify genes playing a dual role (e.g. TSG in one cancer type and OCG in another) but also helps in elucidating the biological processes underlying their specific roles. In particular, MoonlightR can be used to discover OCGs and TSGs in the same cancer type. This may help in answering the question whether some genes change role between early stages (I, II) and late stages (III, IV) in breast cancer. In the future, this analysis could be useful to determine the causes of different resistances to

chemotherapeutic treatments.

**License** GPL (>= 3)

**biocViews** DNAMethylation, DifferentialMethylation, GeneRegulation,  
GeneExpression, MethylationArray, DifferentialExpression,  
Pathways, Network, Survival, GeneSetEnrichment,  
NetworkEnrichment

**Suggests** BiocStyle, knitr, rmarkdown, testthat, devtools, roxygen2,  
png

**VignetteBuilder** knitr

**LazyData** true

**URL** <https://github.com/ibsquare/MoonlightR>

**BugReports** <https://github.com/ibsquare/MoonlightR/issues>

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|                 |   |
|-----------------|---|
| <i>dataFilt</i> | <i>Gene Expression (Rnaseqv2) data from TCGA LUAD</i> |
|-----------------|---|

---

**Description**

A data set containing the following data:

**Usage**

`data(dataFilt)`

**Format**

A 13742x20 matrix

**Details**

- `dataFilt` matrix with 13742 rows (genes) and 20 columns samples with TCGA's barcodes (10TP, 10NT)

**Value**

a 13742x20 matrix

---

|                |   |
|----------------|---|
| <i>dataGRN</i> | <i>GRN gene regulatory network output</i> |
|----------------|---|

---

**Description**

output from GRN function

**Usage**

`data(dataGRN)`

**Format**

A large list of 2 elements

**Details**

- `dataGRN` list of 2 elements `miTFGenes`, `maxmi` from GRN function

**Value**

a large list of 2 elements

dataURA

*Output example from function Upstram Regulator Analysis*

---

**Description**

A data set containing the following data:

**Usage**

```
data(dataURA)
```

**Format**

A data frame with 100 rows and 2 variables

**Details**

- dataURA matrix with 100 rows (genes) and 2 columns "apoptosis" "proliferation of cells"

**Value**

a 100x2 matrix

---

DEGsmatrix

*DEG Differentially expressed genes*

---

**Description**

A data set containing the following data:

**Usage**

```
data(DEGsmatrix)
```

**Format**

A 3502x5 matrix

**Details**

- DEGsmatrix matrix with 3502 rows (genes) and five columns "logFC" "logCPM" "LR" "PValue" "FDR"

**Value**

the 3502x5 matrix

---

|             |  |
|-------------|--|
| DiseaseList | <i>Information on 101 biological processes</i> |
|-------------|--|

---

**Description**

A data set containing the following data:

**Usage**

```
data(DiseaseList)
```

**Format**

A list of 101 matrices

**Details**

- DiseaseList list for 101 biological processes, each containing a matrix with five columns: ID, Genes.in.dataset, Prediction based on expression direction, Log ratio, Findings

**Value**

list of 101 matrices

---

|     |            |
|-----|------------|
| DPA | <i>DPA</i> |
|-----|------------|

---

**Description**

This function carries out the differential phenotypes analysis

**Usage**

```
DPA(dataType, dataFilt, dataConsortium = "TCGA", fdr.cut = 0.01,
    logFC.cut = 1, diffmean.cut = 0.25, samplesType, colDescription,
    gset, gsetFile = "gsetFile.RData")
```

**Arguments**

|                |   |
|----------------|---|
| dataType       | selected  |
| dataFilt       | obtained from getDataTCGA                                       |
| dataConsortium | is TCGA or GEO, default TCGA                                    |
| fdr.cut        | is a threshold to filter DEGs according their p-value corrected |
| logFC.cut      | is a threshold to filter DEGs according their logFC             |
| diffmean.cut   | diffmean.cut for DMR  |
| samplesType    | samplesType   |
| colDescription | colDescription  |
| gset           | gset  |
| gsetFile       | gsetFile  |

**Value**

result matrix from differential phenotype analysis

**Examples**

```
dataDEGs <- DPA(dataFile = dataFile, dataType = "Gene expression")
```

---

|         |                                |
|---------|--------------------------------|
| EAGenes | <i>Information about genes</i> |
|---------|--------------------------------|

---

**Description**

A data set containing the following data:

**Usage**

```
data(EAGenes)
```

**Format**

A 20038x5 matrix

**Details**

- EAGenes matrix with 20038 rows (genes) and five columns "ID" "Gene" "Description" "Location" "Family"

**Value**

a 20038x5 matrix

---

|     |            |
|-----|------------|
| FEA | <i>FEA</i> |
|-----|------------|

---

**Description**

This function carries out the functional enrichment analysis (FEA)

**Usage**

```
FEA(BPname = NULL, DEGsmatrix)
```

**Arguments**

|            |   |
|------------|---|
| BPname     | BPname biological process such as "proliferation of cells", "ALL" (default) if FEA should be carried out for all 101 biological processes |
| DEGsmatrix | DEGsmatrix output from DEA such as dataDEGs"  |

**Value**

matrix from FEA

**Examples**

```
dataDEGs <- DPA(dataFilt = dataFilt,  
dataType = "Gene expression")  
dataFEA <- FEA(DEGsmatrix = dataDEGs)
```

---

**GDCprojects***Information on GDC projects*

---

**Description**

A character vector of GDC projects:

**Usage**

```
data(GDCprojects)
```

**Format**

A character vector of 39 elements

**Details**

- character vector for GDC projects.

**Value**

character vector of 39 elements

---

**geneInfo***Information about genes for normalization*

---

**Description**

A data set containing the following data:

**Usage**

```
data(geneInfo)
```

**Format**

A data frame with 20531 rows and 3 variables

**Details**

- geneInfo matrix with 20531 rows (genes) and 3 columns "geneLength" "gcContent" "chr"

**Value**

a 20531x3 matrix

---

|             |  |
|-------------|--|
| GEO_TCGAtab | <i>Information on GEO data (and overlap with TCGA)#' A data set containing the following data:</i> |
|-------------|--|

---

**Description**

- GEO\_TCGAtab a 18x12 matrix that provides the GEO data set we matched to one of the 18 given TCGA cancer types

**Usage**

```
data(GEO_TCGAtab)
```

**Format**

A 101x3 matrix

**Value**

a 101x3 matrix

---

|            |                   |
|------------|-------------------|
| getDataGEO | <i>getDataGEO</i> |
|------------|-------------------|

---

**Description**

This function retrieves and prepares GEO data

**Usage**

```
getDataGEO(GEOobject = "GSE39004", platform = "GPL6244",
  TCGAtumor = NULL)
```

**Arguments**

|           |            |
|-----------|------------|
| GEOobject | GEOobject  |
| platform  | platform   |
| TCGAtumor | tumor name |

**Value**

return GEO gset

**Examples**

```
## Not run:
dataGEO <- getDataGEO(GEOobject = "GSE20347",platform = "GPL571")

## End(Not run)
```



---

|             |                    |
|-------------|--------------------|
| getDataTCGA | <i>getDataTCGA</i> |
|-------------|--------------------|

---

## Description

This function retrieves and prepares TCGA data

## Usage

```
getDataTCGA(cancerType, dataType, directory, cor.cut = 0.6,  
            qnt.cut = 0.25, nSample, stage = "ALL", subtype = 0,  
            samples = NULL)
```

## Arguments

|            |  |
|------------|--|
| cancerType | select cancer type for which analysis should be run. panCancer for all available cancer types in TCGA. Defaults to panCancer |
| dataType   | is dataType such as gene expression, cnv, methylation etc.   |
| directory  | Directory/Folder where the data was downloaded. Default: GDCdata   |
| cor.cut    | cor.cut  |
| qnt.cut    | qnt.cut  |
| nSample    | nSample  |
| stage      | stage  |
| subtype    | subtype  |
| samples    | samples  |

## Value

returns filtered TCGA data

## Examples

```
## Not run:  
dataFilt <- getDataTCGA(cancerType = "LUAD",  
                       dataType = "Gene expression", directory = "data", nSample = 4)  
  
## End(Not run)
```

GRN

*Generate network***Description**

This function carries out the gene regulatory network inference using parmigene

**Usage**

```
GRN(TFs, DEGsmatrix, DiffGenes = FALSE, normCounts, kNearest = 3,
    nGenesPerm = 10, nBoot = 10)
```

**Arguments**

|            |  |
|------------|--|
| TFs        | a vector of genes.   |
| DEGsmatrix | DEGsmatrix output from DEA such as dataDEGs  |
| DiffGenes  | if TRUE consider only diff.expr genes in GRN   |
| normCounts | is a matrix of gene expression with genes in rows and samples in columns.  |
| kNearest   | the number of nearest neighbors to consider to estimate the mutual information. Must be less than the number of columns of normCounts. |
| nGenesPerm | nGenesPerm   |
| nBoot      | nBoot  |

**Value**

an adjacent matrix

**Examples**

```
dataDEGs <- DEGsmatrix
dataGRN <- GRN(TFs = rownames(dataDEGs)[1:100],
  DEGsmatrix = dataDEGs,
  DiffGenes = TRUE,
  normCounts = dataFilt)
```

GSEA

*GSEA***Description**

This function carries out the GSEA enrichment analysis.

**Usage**

```
GSEA(DEGsmatrix, top, plot = FALSE)
```

**Arguments**

DEGsmatrix      DEGsmatrix output from DEA such as dataDEGs  
top              is the number of top BP to plot  
plot             if TRUE return a GSEA's plot

**Value**

return GSEA result

**Examples**

```
dataDEGs <- DEGsmatrix  
# dataFEA <- GSEA(DEGsmatrix = dataDEGs)
```

---

knownDriverGenes      *Information on known cancer driver gene from COSMIC*

---

**Description**

A data set containing the following data:

**Usage**

```
data(knownDriverGenes)
```

**Format**

A 101x3 matrix

**Details**

- TSG known tumor suppressor genes
- OCG known oncogenes

**Value**

a 101x3 matrix

---

|                            |                                   |
|----------------------------|-----------------------------------|
| <code>listMoonlight</code> | <i>Output list from Moonlight</i> |
|----------------------------|-----------------------------------|

---

**Description**

A list containing the following data:

**Usage**

```
data(listMoonlight)
```

**Format**

A Large list with 5 elements

**Details**

- `listMoonlight` output from moonlight's pipeline containing `dataDEGs`, `dataURA`, `listCandidates`

**Value**

output from moonlight pipeline

---

LPA

*LPA***Description**

This function carries out the literature phenotype analysis (LPA)

**Usage**

```
LPA(dataDEGs, BP, BPlist)
```

**Arguments**

|                       |                                  |
|-----------------------|----------------------------------|
| <code>dataDEGs</code> | is output from DEA               |
| <code>BP</code>       | is biological process            |
| <code>BPlist</code>   | is list of genes annotated in BP |

**Value**

table with number of pubmed that affects, increase or decrease genes annotated in BP

**Examples**

```
data(DEGsmatrix)
BPselected <- c("apoptosis")
BPannotations <- DiseaseList[[match(BPselected, names(DiseaseList))]]$ID
```

moonlight

*moonlight pipeline***Description**

moonlight is a tool for identification of cancer driver genes. This function wraps the different steps of the complete analysis workflow. Providing different solutions:

1. MoonlighR::FEA
2. MoonlighR::URA
3. MoonlighR::PIA

**Usage**

```
moonlight(cancerType = "panCancer", dataType = "Gene expression",
  directory = "GDCdata", BPname = NULL, cor.cut = 0.6,
  qnt.cut = 0.25, Genelist = NULL, fdr.cut = 0.01, logFC.cut = 1,
  corThreshold = 0.6, kNearest = 3, nGenesPerm = 10,
  DiffGenes = FALSE, nBoot = 100, nTF = NULL, nSample = NULL,
  thres.role = 0, stage = NULL, subtype = 0, samples = NULL)
```

**Arguments**

|              |   |
|--------------|---|
| cancerType   | select cancer type for which analysis should be run. panCancer for all available cancer types in TCGA. Defaults to panCancer  |
| dataType     | dataType  |
| directory    | directory   |
| BPname       | biological processes to use, if NULL: all processes will be used in analysis, RF for candidate; if not NULL the candidates for these processes will be determined (no learning) |
| cor.cut      | cor.cut Threshold   |
| qnt.cut      | qnt.cut Threshold   |
| Genelist     | Genelist  |
| fdr.cut      | fdr.cut Threshold   |
| logFC.cut    | logFC.cut Threshold   |
| corThreshold | corThreshold  |
| kNearest     | kNearest  |
| nGenesPerm   | nGenesPerm  |
| DiffGenes    | DiffGenes   |
| nBoot        | nBoot   |
| nTF          | nTF   |
| nSample      | nSample   |
| thres.role   | thres.role  |
| stage        | stage   |
| subtype      | subtype   |
| samples      | samples   |

**Value**

table with cancer driver genes TSG and OCG.

**Examples**

```
dataDEGs <- DPA(dataFilt = dataFilt, dataType = "Gene expression")
# to change with moonlight
```

---

|            |                   |
|------------|-------------------|
| MoonlightR | <i>MoonlightR</i> |
|------------|-------------------|

---

**Description**

MoonlightR

---

|            |                   |
|------------|-------------------|
| plotCircos | <i>plotCircos</i> |
|------------|-------------------|

---

**Description**

This function visualize the plotCircos

**Usage**

```
plotCircos(listMoonlight, listMutation = NULL,
           additionalFilename = NULL, intensityColOCG = 0.5,
           intensityColTSG = 0.5, intensityColDual = 0.5, fontSize = 1)
```

**Arguments**

|                    |                           |
|--------------------|---------------------------|
| listMoonlight      | output Moonlight function |
| listMutation       | listMutation              |
| additionalFilename | additionalFilename        |
| intensityColOCG    | intensityColOCG           |
| intensityColTSG    | intensityColTSG           |
| intensityColDual   | intensityColDual          |
| fontSize           | fontSize                  |

**Value**

no return value, plot is saved

**Examples**

```
plotCircos(listMoonlight = listMoonlight, additionalFilename = "_ncancer5")
```

---

|         |                |
|---------|----------------|
| plotFEA | <i>plotFEA</i> |
|---------|----------------|

---

### Description

This function visualize the functional enrichment analysis (FEA)'s barplot

### Usage

```
plotFEA(dataFEA, topBP = 10, additionalFilename = NULL, height, width,  
        offsetValue = 5, angle = 90, xleg = 35, yleg = 5, titleMain,  
        minY = -5, maxY = 10, mycols = c("#8DD3C7", "#FFFFB3", "#BEBADA"))
```

### Arguments

|                    |                            |
|--------------------|----------------------------|
| dataFEA            | dataFEA                    |
| topBP              | topBP                      |
| additionalFilename | additionalFilename         |
| height             | Figure height              |
| width              | Figure width               |
| offsetValue        | offsetValue                |
| angle              | angle                      |
| xleg               | xleg                       |
| yleg               | yleg                       |
| titleMain          | title of the plot          |
| minY               | minY                       |
| maxY               | maxY                       |
| mycols             | colors to use for the plot |

### Value

no return value, FEA result is plotted

### Examples

```
dataFEA <- FEA(DEGsmatrix = DEGsmatrix)  
plotFEA(dataFEA = dataFEA, additionalFilename = "_example", height = 20, width = 10)
```

---

plotNetworkHive      *plotNetworkHive: Hive network plot*

---

### Description

This function visualizes the GRN as a hive plot

### Usage

```
plotNetworkHive(dataGRN, namesGenes, thres, additionalFilename = NULL)
```

### Arguments

|                    |                                   |
|--------------------|-----------------------------------|
| dataGRN            | output GRN function               |
| namesGenes         | list TSG and OCG to define axes   |
| thres              | threshold of edges to be included |
| additionalFilename | additionalFilename                |

### Value

no results Hive plot is executed

### Examples

```
data(knownDriverGenes)
data(dataGRN)
plotNetworkHive(dataGRN = dataGRN, namesGenes = knownDriverGenes, thres = 0.55)
```

---

plotURA      *plotURA: Upstream regulatory analysis heatmap plot*

---

### Description

This function visualizes the URA in a heatmap

### Usage

```
plotURA(dataURA, additionalFilename = "URApplot")
```

### Arguments

|                    |                     |
|--------------------|---------------------|
| dataURA            | output URA function |
| additionalFilename | figure name         |

### Value

heatmap



**Examples**

```

data(dataURA)
dataDual <- PRA(dataURA = dataURA,
BPname = c("apoptosis","proliferation of cells"),
thres.role = 0)
TSGs_genes <- names(dataDual$TSG)
OCGs_genes <- names(dataDual$OCG)
plotURA(dataURA = dataURA[c(TSGs_genes, OCGs_genes),],additionalFilename = "_example")

```

PRA

*Pattern Recognition Analysis (PRA)***Description**

This function carries out the pattern recognition analysis

**Usage**

```
PRA(dataURA, BPname, thres.role = 0)
```

**Arguments**

|            |                     |
|------------|---------------------|
| dataURA    | output URA function |
| BPname     | BPname              |
| thres.role | thres.role          |

**Value**

returns list of TSGs and OCGs when biological processes are provided, otherwise a randomForest based classifier that can be used on new data

**Examples**

```

data(dataURA)
dataDual <- PRA(dataURA = dataURA,
BPname = c("apoptosis","proliferation of cells"),
thres.role = 0)

```

tabGrowBlock

*Information growing/blocking characteristics for 101 selected biological processes***Description**

A data set containing the following data:

**Usage**

```
data(tabGrowBlock)
```

**Format**

A 101x3 matrix

**Details**

- tabGrowBlock matrix that defines if a process is growing or blocking cancer development, for each 101 biological processing

**Value**

a 101x3 matrix

---

URA

*URA Upstream Regulator Analysis*

---

**Description**

This function carries out the upstream regulator analysis

**Usage**

```
URA(dataGRN, DEGsmatrix, BPname, nCores = 1)
```

**Arguments**

|            |                        |
|------------|------------------------|
| dataGRN    | output GNR function    |
| DEGsmatrix | output DPA function    |
| BPname     | biological processes   |
| nCores     | number of cores to use |

**Value**

an adjacent matrix

**Examples**

```
dataDEGs <- DEGsmatrix
dataGRN <- GRN(TFs = rownames(dataDEGs)[1:100],
  DEGsmatrix = dataDEGs,
  DiffGenes = TRUE,
  normCounts = dataFilt)
dataURA <- URA(dataGRN = dataGRN,
  DEGsmatrix = dataDEGs,
  BPname = c("apoptosis",
  "proliferation of cells"))
```

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