Package 'rmimp'

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Title Predicting the impact of mutations on kinase-substrate

Type Package

phosphorylation
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Description MIMP is a machine learning method that predicts the impact of missense single-nucleotide variants (SNVs) on kinase-substrate interactions. MIMP analyzes kinase sequence specificities and predicts whether SNVs disrupt existing phosphorylation sites or create new sites. This helps discover mutations that modify protein function by altering kinase networks and provides insight into disease biology and therapy development.
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computeBinding Score wt and mt sequences for a pwm
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Description

Score wt and mt sequences for a pwm

Usage

```
computeBinding(obj, mut_ss, mut_location, prob.thresh = 0.5,
  log2.thresh = 1)
```

Arguments

obj	MIMP object containing PWM, GMM parameters, and etc.
mut_ss	snvs data frame containing wt and mt sequences computed from SNVs function
${\sf mut_location}$	list of mutation locations
prob.thresh	Probability threshold of gains and losses. This value should be between 0.5 and $1.$
log2.thresh	Threshold for the absolute value of log ratio between wild type and mutant scores. Anything less than this value is discarded (default: 1).

dohtml	Helper function for results2html	
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Description

Helper function for results2html

Usage

```
dohtml(x, LOGO_DIR, HL_DIR, logoExt = ".svg", .webserver = F)
```

Arguments

X	Data frame resulting from mimp call.
LOGO_DIR	Directory containing sequence logo images.
HL_DIR	Directory containing overlays
logoExt	Extension of logo files
.webserver	Request coming from webserver?

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dol		

Helper function for results2html and sh3 binding

Description

Helper function for results2html and sh3 binding

Usage

```
dohtmlSh3(x, LOGO_DIR, HL_DIR, logoExt = ".svg", .webserver = F)
```

Arguments

x Data frame resulting from mimp call.

LOGO_DIR Directory containing sequence logo images.

HL_DIR Directory containing overlays

logoExt Extension of logo files

.webserver Request coming from webserver?

mimp

Predict the impact of single variants on phosphorylation.

Description

This function takes in mutation, sequence and phosphorylation data to predict the impact the mutation has on phosphorylation.

Usage

```
mimp(muts, seqs, central = T, domain = "phos", species = "human",
   psites = NULL, terminal.range = 5, prob.thresh = 0.5, log2.thresh = 1,
   display.results = T, include.cent = F, model.data = "hconf")
```

Arguments

muts

Mutation data file: a space delimited text file OR data frame containing two columns (1) gene and (1) mutation. Example:

TP53 R282W CTNNB1 S33C CTNNB1 S37F

seqs

Sequence data file containing protein sequences in FASTA format OR named list

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of sequences where each list element is the uppercase sequence and the name of each element is that of the protein. Example: list(GENEA="ARNDGH",

GENEB="YVRRHS")

central Whether the mutation site is at the central residue of the sequence

domain Which binding domain to run mimp for

psites Phosphorylation data file (optional): a space delimited text file OR data frame

containing two columns (1) gene and (1) positions of phosphorylation sites. Ex-

ample:

TP53 280 CTNNB1 29 CTNNB1 44

terminal.range The number of amino acids used for predicting terminal domain binding.

prob. thresh Probability threshold of gains and losses. This value should be between 0.5 and

1.

log2.thresh Threshold for the absolute value of log ratio between wild type and mutant

scores. Anything less than this value is discarded (default: 1).

display.results

If TRUE results are visualised in an html document after analysis is complete

include.cent If TRUE, gains and losses caused by mutation in the central STY residue are

kept. Scores of peptides with a non-STY central residue is given a score of 0

(default: FALSE).

model.data Name of specificity model data to use, can be "hconf": individual experimen-

tal kinase specificity models used to scan for rewiring events. For experimental kinase specificity models, grouped by family, set to "hconf-fam". Both are considered high confidence. For lower confidence predicted specificity models, set to "lconf". NOTE: Predicted models are purely speculative and should be used

with caution

Value

The data is returned in a data. frame with the following columns:

gene Gene with the rewiring event

mut Mutation causing the rewiring event

psite_pos (Optional) Position of the phosphosite, if domain = "phos"

mut_dist (Optional) Distance of the mutation relative to the central residue, if domain =

"phos"

wt Sequence of the wildtype phosphosite (before the mutation). Score is NA if the

central residue is not S, T or Y

mt Sequence of the mutated phosphosite (after the mutation). Score is NA if the

central residue is not S, T or Y

score_wt Matrix similarity score of the wildtype phosphosite

score_mt Matrix similarity score of the mutated phosphosite

log_ratio

Log2 ratio between mutant and wildtype scores. A high positive log ratio represents a high confidence gain-of-phosphorylation event. A high negative log ratio represents a high confidence loss-of-phosphorylation event. This ratio is NA for mutations that affect the central phosphorylation sites

pwm Name of the kinase being rewired

pwm_fam (Optional, available only if domain = "phos") Family/subfamily of kinase being

rewired. If a kinase subfamily is available the family and subfamily will be separated by an underscore e.g. "DMPK_ROCK". If no subfamily is available,

only the family is shown e.g. "GSK"

nseqs (Optional, available only if domain = "phos") Number of sequences used to

construct the PWM. PWMs constructed with a higher number of sequences are

generally considered of better quality.

prob Joint probability of wild type sequence belonging to the foreground distribution

and mutated sequence belonging to the background distribution, for loss and

vice versa for gain.

effect Type of rewiring event, can be "loss" or "gain"

Examples

```
# Get the path to example mutation data
mut.file = system.file("extdata", "mutation_data.txt", package = "rmimp")
# Get the path to example FASTA sequence data
seq.file = system.file("extdata", "sequence_data.txt", package = "rmimp")
# View the files in a text editor
browseURL(mut.file)
browseURL(seq.file)
# Run rewiring analysis
results = mimp(mut.file, seq.file, display.results=TRUE)
# Show head of results
head(results)
```

predictKinasePhosphosites

Compute posterior probability of wild type phosphosites for kinases

Description

Compute posterior probability of wild type phosphosites for kinases

Usage

```
predictKinasePhosphosites(psites, seqs, model.data = "hconf",
   posterior_thresh = 0.8, intermediate = F, kinases)
```

Arguments

psites phosphorylation data, see ?mimp for details seqs sequence data, see ?mimp for details model.data MIMP model used, see ?mimp for details

posterior_thresh

posterior probability threshold that the score belongs to the foreground distribution of the kinase, probabilities below this value are discarded (default 0.8)

intermediate if TRUE intermediate MSS scores and likelihoods are reported (default FALSE) vector of kinases used for the scoring (e.g. c("AURKB", "CDK2")), if this isn't

provided all kinases will be used.

Value

The data is returned in a data. frame with the following columns:

gene Gene with the rewiring event pos Position of the phosphosite

wt Sequence of the wildtype phosphosite

score_wt (intermediate value) matrix similarity score of sequence

1.wt.fg (intermediate value) likelihood of score given foreground distribution1.wt.bg (intermediate value) likelihood of score given background distribution

post.wt.fg posterior probability of score in foreground distribution post.wt.bg posterior probability of score in background distribution

pwm Name of the predicted kinase

pwm_fam Family/subfamily of the predicted kinase. If a kinase subfamily is available the

family and subfamily will be seprated by an underscore e.g. "DMPK_ROCK".

If no subfamily is available, only the family is shown e.g. "GSK"

If no predictions were made, function returns NULL

Examples

```
# Get the path to example phosphorylation data
psite.file = system.file("extdata", "sample_phosphosites.tab", package = "rmimp")

# Get the path to example FASTA sequence data
seq.file = system.file("extdata", "sample_seqs.fa", package = "rmimp")

# Run for all kinases
results_all = predictKinasePhosphosites(psite.file, seq.file)

# Run for select kinases
results_select = predictKinasePhosphosites(psite.file, seq.file, kinases=c("AURKB", "CDK2"))
```

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results2html

Display MIMP results interactively in browser

Description

Display MIMP results interactively in browser

Usage

```
results2html(x, domain = "phos", max.rows = 5000)
```

Arguments

x Data frame resulting from mimp call.

domain Which binding domain to run mimp for

max.rows If data contains more rows than this value, results won't be displayed.

scoreArrayRolling

Get weight/probability for each amino acid in a sequence

Description

Gets weight/probability for the amino acid at each position of the sequence as an array.

Usage

```
scoreArrayRolling(seqs, pwm)
```

Arguments

seqs One or more sequences to be processed

pwm Position weight matrix

Examples

No Examples

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scoreWTSequence	Score wt sequence using PWMs in the model	

Description

Score wt sequence using PWMs in the model

Usage

```
scoreWTSequence(wt_seqs, central = T, domain = "phos", species = "human",
  model.data = "hconf", cores = 2)
```

Arguments

wt_seqs	A list of sequences to be scored
central	Whether the mutation site is at the central residue of the sequence

cores Number of cores the function could use

SNVs	Find non-central variants (SNVs)	

Description

Given mutation data, find variants that exist in the flanking regions of the psite

Usage

```
SNVs(md, seqdata, flank)
```

Arguments

md	Mutation data as data frame of two columns (1) name of gene or protein (2) mutation in the format X123Y, where X is the reference amino acid and Y is the alternative amino acid.
seqdata	Phosphorylation data as a data frame of two columns (1) name of gene or protein (2) Position of the phosphorylated residue
flank	Number of amino acids flanking the site to be considered

Examples

```
# No examples
```

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trainModel	Train GMM model and return as a list to be used later. If file is passed,
	the model will also be save to a .mimp file.

Description

Train GMM model and return as a list to be used later. If file is passed, the model will also be save to a .mimp file.

Usage

```
trainModel(pos.dir, neg.dir, kinase.domain = F, cores = 2, file = NULL,
    threshold = 10, min.auc = 0.65, priors)
```

Arguments

pos.dir	the path to the directory contains positive entries
neg.dir	the path to the directory contains negative entries
kinase.domain	Whether the domain to be trained is a kinase domain.
cores	(optional) the number of CPU cores that can be used to train the model
file	(optional) the path to save the model
threshold	(optional) the minimum number of scores needed for each domain to train the model
min.auc	(optional) the minimum number of AUC needed for each domain to train the model

Value

a GMM model

Examples

No examples

tSNVs

tSNVs	Find terminal variants (tSNVs)	

Description

Given mutation data, find variants that exist in the flanking regions of the psite

Usage

```
tSNVs(md, seqdata, terminal)
```

Arguments

Mutation data as data frame of two columns (1) name of gene or protein (2)

mutation in the format X123Y, where X is the reference amino acid and Y is the

alternative amino acid.

seqdata Phosphorylation data as a data frame of two columns (1) name of gene or protein

(2) Position of the phosphorylated residue

terminal Number of amino acids flanking the site to be considered

Examples

No examples