## goseq

## October 5, 2010

getgo	Fetch GO categories
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## **Description**

Obtains all gene ontology (GO) categories associated with a set of genes using the relevant organism package.

## Usage

```
getgo(genes, genome, id,fetch.cats=c("GO:CC","GO:BP","GO:MF"))
```

## Arguments

genes	A vector or list of genes to get the associated GO categories.
genome	A string identifying the genome that genes refer to. For a list of supported organisms run supportedGenomes.
id	A string identifying the gene identifier used by genes. For a list of supported gene IDs run supportedGeneIDs.
fetch.cats	A vector specifying which categories to fetch the mapping between category names and genes for. See details for vaild options.

## **Details**

This function attempts to make use of the organism packages (org.<Genome>.<GeneID>.db) to obtain the mapping between gene ID and GO categories. As with getlength it is preferable that the same gene identifier system is used for both summarization and retrieving GO categories.

Valid options for the fetch.cats arguement are any combination of "GO:CC", "GO:BP", "GO:MF" & "KEGG". The three GO terms refer to the Cellular Component, Biological Process and Molecular Function respectively. "KEGG" refers to KEGG pathways.

Note that <code>getgo</code> is a convenience function, designed to make extracting mappings between GO categories and Gene ID. For less common organisms and/or gene ID <code>getgo</code> may fail to return a mapping even when a legitimate mapping exists in the revelant organism package. If <code>getgo</code> fails, you should always try to build the mapping yourself from the organism package (if one exists) before deciding that the information is unavailable. Further information and examples of this can be found in the package Vignette.

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#### Value

A list where each entry is named by a gene and contains a vector of all the associated GO categories. This can be used directly with the gene2cat option in goseq.

#### Author(s)

```
Matthew D. Young <myoung@wehi.edu.au>
```

#### See Also

```
supportedGenomes, supportedGeneIDs, goseq
```

## **Examples**

```
genes <- c("ENSG00000124208", "ENSG00000182463", "ENSG00000124201", "ENSG00000124205", "E
getgo(genes,'hg19','ensGene')</pre>
```

getlength

Retrieves Gene length data

## **Description**

Gets the length of each gene in a vector.

## Usage

```
getlength(genes, genome, id)
```

## **Arguments**

genes	A vector or list of the genes for which length information is required.
genome	A string identifying the genome that genes refer to. For a list of supported organisms run supportedGenomes.
id	A string identifying the gene identifier used by genes. For a list of supported gene IDs run supportedGeneIDs.

#### **Details**

Length data is obtained from data obtained from the UCSC genome browser for each combination of genome and id. As fetching this data at runtime is time consuming, a local copy of the length information for common genomes and gene ID are included in the **geneLenDataBase** package. This function uses this package to fetch the required data.

The length of a gene is taken to be the median length of all its mature, mRNA, transcripts. It is always preferable to obtain length information directly for the gene ID used to summarize your count data, rather than converting IDs and then using the supplied databases. Even when two genes have a one-to-one mapping between different identifier conventions (which is often not the case), they frequently refer to slightly different regions of the genome with different lengths. It is therefore recommended that the user perform the full analysis in terms of only one gene ID, or manually obtain their own length data for the identifier used to bin reads by gene.

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## Value

Returns a vector of the gene lengths, in the same order as genes. If length data is unavailable for a particular gene NA is returned in that position. The returned vector is intended for use with the bias.data option of the nullp function.

## Author(s)

```
Matthew D. Young <myoung@wehi.edu.au>
```

#### See Also

```
supportedGenomes, supportedGeneIDs, nullp, geneLenDataBase
```

## **Examples**

```
genes <- c("ENSG00000124208", "ENSG00000182463", "ENSG00000124201", "ENSG00000124205", "Equipment (genes,'hg19','ensGene')
```

goseq

goseq Gene Ontology analyser

## **Description**

Does selection-unbiased testing for category enrichment amongst differentially expressed (DE) genes for RNA-seq data. By default, tests gene ontology (GO) categories, but any categories may be tested.

## Usage

```
goseq(DEgenes, pwf, genome, id, gene2cat = NULL,
test.cats=c("GO:CC", "GO:BP", "GO:MF"),
method = "Wallenius", repcnt = 2000)
```

## **Arguments**

DEgenes	A named binary vector where 1 represents DE, 0 not DE and the names are gene IDs.
pwf	A vector of the same length as DEgenes containing the weighting to be applied to each gene in DEgenes. Usually produced by nullp.
genome	A string identifying the genome that genes refer to. For a list of supported organisms run supportedGenomes.
id	A string identifying the gene identifier used by genes. For a list of supported gene IDs run supportedGeneIDs.
gene2cat	A data frame with two columns containing the mapping between genes and the categories of interest. Alternatively, a list where the names are genes and each entry is a vector containing GO categories associated with that gene (this is the output produced by getgo). If set to NULL goseq attempts to fetch GO categories automatically using getgo.
test.cats	A vector specifying which categories to test for over representation amongst DE genes. See details for vaild options.

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method The method to use to calculate the unbiased category enrichment scores. Valid

options are "Wallenius", "Sampling" & "Hypergeometric". "Hypergeometric"

should almost never be used (see details).

report Number of random samples to be calculated when random sampling is used.

Ignored unless method="Sampling".

#### Details

goseq obtains length data from UCSC and GO mappings from the organim packages (see link {getgo} and getlength for details). If your data is in an unsupported format you will need to obtain the GO category mapping and supply them to the goseq function using the gene2cat argument.

To use your own gene to category mapping with <code>goseq</code>, use the <code>gene2cat</code> arguement. This arguement takes a data.frame, with one column containing gene IDs and the other containing the associated categories. As the mapping from gene <-> category is in general many to many there will be multiple rows containing the same gene identifier. Alternatively, <code>gene2cat</code> can take a list, where the names are the genes and the entries are the GO categories associated with the genes. This is the format produced by the <code>getgo</code> function and is more space efficient than the data.frame representation.

If gene2cat is left as NULL, goseq attempts to use getgo to fetch GO catgeory to gene identifier mappings.

The genome and id arguements are only required for fetching go category mappings. If you are providing your own, these arguements are not required and will be ignored if specified.

The PWF is usually calculated using the nullp function to correct for length bias. However, goseq will work with any vector of weights. Any bias can be accounted for so long as a weight for each gene is supplied using this arguement. NAs are allowed in pwf (these usually occur as a result of missing length data for some genes). Any entry which is NA is set to the weighting of the median gene.

Valid options for the test.cats arguement are any combination of "GO:CC", "GO:BP", "GO:MF" & "KEGG". The three GO terms refer to the Cellular Component, Biological Process and Molecular Function respectively. "KEGG" refers to KEGG pathways.

The three methods, "Wallenius", "Sampling" & "Hypergeometric", calculate the p-values as follows.

"Wallenius" approximates the true distribution of numbers of members of a category amongst DE genes by the Wallenius non-central hypergeometric distribution. This distribution assumes that within a category all genes have the same probability of being chosen. Therefore, this approximation works best when the range in probabilities obtained by the probability weighting function is small.

"Sampling" uses random sampling to approximate the true distribution and uses it to calculate the p-values for over (and under) representation of categories. Although this is the most accurate method given a high enough value of repent, its use quickly becomes computationally prohibitive.

CAUTION: "Hypergeometric" should NEVER be used for producing results for biological interpretation. If there is genuinly no bias in power to detect DE in your experiment, the PWF will reflect this and the other methods will produce accuracte results.

"Hypergeometric" assumes there is no bias in power to detect differential expression at all and calculates the p-values using a standard hypergeometric distribution. Useful if you wish to test the effect of selection bias on your results.

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## Value

goseq returns a data frame with 3 columns. The first column gives the name of the category, the second gives the p-value for the associated category being over represented amongst DE genes. The third and final column gives the p-value for the associated category being under represented amongst DE genes. The p-values have not been corrected for multiple hypothesis testing.

## Author(s)

```
Matthew D. Young <myoung@wehi.edu.au>
```

#### References

Young, M. D., Wakefield, M. J., Smyth, G. K., Oshlack, A. (2010) *Gene ontology analysis for RNA-seq: accounting for selection bias* Genome Biology Date: Feb 2010 Vol: 11 Issue: 2 Pages: R14

## See Also

```
nullp, getgo, getlength
```

## **Examples**

```
data(prostate)
pwf <- nullp(genes,'hg19','ensGene')
pvals <- goseq(genes,pwf,'hg19','ensGene')
head(pvals)</pre>
```

makespline

Monotonic Spline

## **Description**

Fits a monotonic cubic spline to the data provided, using the penalized constrained least squares method from the mgcv package.

## Usage

```
makespline(x, y, newX=NULL, nKnots = 6)
```

## **Arguments**

x The predictor variable.

Y The response variable. Must be the same length as x.

newX The points at which to return the value on the fitted spline. If not specified x is

used.

nknots The number of knots to use in fitting the spline.

## **Details**

This uses the pcls function from the **mgcv** package to produce the fit. The montonicity constraint is enforced using mono.con from the same package.

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## Value

Returns a vector of values containing the value of the fit at each point newX.

## Author(s)

```
Matthew D. Young <myoung@wehi.edu.au>.
```

## References

Package **mgcv**. In particular this function is a modification of an example given in the man page for pcls.

## **Examples**

```
y <-c( rbinom(50,p=0.4,size=1), rbinom(50,p=0.6,size=1) ) x <-1:100 plot(x,y) p <- makespline(x,y) lines(x,p)
```

nullp

Probability Weighting Function

## Description

Calculates a Probability Weighting Function for a set of genes based on a given set of biased data (usually gene length) and each genes status as differentially expressed or not.

## Usage

```
nullp(DEgenes, genome, id, bias.data=NULL,plot.fit=TRUE)
```

## Arguments

DEgenes	A named binary vector where 1 represents DE, 0 not DE and the names are gene IDs.
genome	A string identifying the genome that genes refer to. For a list of supported organisms run supportedGenomes.
id	A string identifying the gene identifier used by genes. For a list of supported gene IDs run supportedGeneIDs.
bias.data	A numeric vector containing the data on which the DE may depend. Usually this is the median transcript length of each gene in bp. If set to NULL nullp will attempt to fetch length using getlength.
plot.fit	Plot the PWF or not?

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#### **Details**

It is essential that the entire analysis pipeline, from summarizing raw reads through to using goseq be done in just one gene identifier format. If your data is in a different format you will need to obtain the gene lengths and supply them to the nullp function using the bias.data arguement. Converting to a supported format from another format should be avoided whenever possible as this will almost always result in data loss.

NAs are allowed in the bias.data vector if you do not have information about a certain gene. Setting a gene to NA is preferable to removing it from the analysis.

If bias.data is left as NULL, nullp attempts to use getlength to fetch GO catgeory to gene identifier mappings.

It is recommended you review the fit produced by the nullp function before proceeding by leaving plot.fit as TRUE.

#### Value

A numeric vector containing the value on the probability weighting function for each gene. This is usually passed to the function goseq via the pwf arguement.

#### Author(s)

Matthew D. Young <myoung@wehi.edu.au>

#### References

Young, M. D., Wakefield, M. J., Smyth, G. K., Oshlack, A. (2010) *Gene ontology analysis for RNA-seq: accounting for selection bias* Genome Biology Date: Feb 2010 Vol: 11 Issue: 2 Pages: R14

## See Also

supportedGenomes, supportedGeneIDs, goseq, getlength

## **Examples**

```
data(prostate)
pwf <- nullp(genes, 'hg19', 'ensGene')</pre>
```

prostate

Androgen stimulation of prostate cancer Cell lines.

## **Description**

This data set gives the RNA-seq data from an experiment measuring the effects of androgen stimulation on prostate cancer. Information is given about all (ENSEMBL) genes for which there was at least one mapping read in either the treated or untreated RNA-seq experiment. The edgeR package was used to determine which genes were differentially expressed. The details of the analysis can be found in the goseq vignette.

## Usage

```
data(prostate)
```

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## **Format**

A named vector of ENSEMBL genes, with 1 representing differential expression.

## **Source**

Determination of tag density required for digital transcriptome analysis: application to an androgensensitive prostate cancer model, 2008, Li et. al.

## References

Li, H., Lovci, M. T., Kwon, Y. S., Rosenfeld, M. G., Fu, X. D., Yeo, G. W. (2008) *Determination of tag density required for digital transcriptome analysis: application to an androgen-sensitive prostate cancer model* Proceedings of the National Academy of Sciences of the United States of America Date: Dec 23 Vol: 105 Issue: 51 Pages: 20179-84

## **Examples**

data(prostate)
head(genes)

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