# Package 'birta'

# April 4, 2014

	1.6.0
Date 20	013-03-07
Title B	ayesian Inference of Regulation of Transcriptional Activity
Author B	enedikt Zacher, Khalid Abnaof, Stephan Gade, Erfan Younesi, Achim Tresch, Holger Froehlich
	ner Benedikt Zacher <zacher@lmb.uni-muenchen.de>, Holger roehlich<frohlich@bit.uni-bonn.de></frohlich@bit.uni-bonn.de></zacher@lmb.uni-muenchen.de>
Depend	s limma, MASS, R(>= 2.10), Biobase, methods
in cr to di w	tion Expression levels of mRNA molecules are regulated by different processes, comprisg inhibition or activation by transcription factors and post-transcriptional degradation by mioRNAs. birta (Bayesian Inference of Regulation of Transcriptional Activity) uses the regularry networks of TFs and miRNAs together with mRNA and miRNA expression data to prect switches in regulatory activity between two conditions. A Bayesian network is used to model the regulatory structure and Markov-Chain-Monte-Carlo is applied to sample the activity states.
License	GPL (>= 2)
biocVie M	ws Gicroarray, Sequencing, GeneExpression, Transcription, Bioinformatics, GraphsAndNetworks
LazyLo	ad yes
R top	ics documented:

2 birta-package

	limmaAnalysis																													g
	limmaAnalysis-meth	ods																												10
	plotConvergence .																													
	potential_swaps																													
	sim																													
	TFexpr		•		•					•												•		•						12
Index																														13
birta	n-nackage	Join	t Ba	av <sub>e</sub>	esi.	an	In	fer	en	1.C.e	2.0	f n	nik	?N	Ά	an	d'	Tra	ın:	SCI	in	tio	n	Fa	cti	or	Ac	tiv	iti	es

# Description

Expression levels of mRNA molecules are regulated by different processes, comprising inhibition or activation by transcription factors and post-transcriptional degradation by microRNAs. birta uses the regulatory networks of TFs and miRNAs together with mRNA and miRNA expression data to predict switches in regulatory activity between two conditions. A Bayesian network is used to model the regulatory structure and Markov-Chain-Monte-Carlo is applied to sample the activity states.

# **Details**

Package: birta
Type: Package
Version: 0.99.1
Date: 2012-03-6

License: GNU GENERAL PUBLIC LICENSE (GPL) >= 2

LazyLoad: yes

#### Author(s)

Benedikt Zacher, Khalid Abnaof, Stephan Gade, Erfan Younesi, Achim Tresch, Holger Frohlich Maintainer: Benedikt Zacher <zacher@lmb.uni-muenchen.de>

# References

B. Zacher, K. Abnaof, S. Gade, E. Younesi, A. Tresch, and H. Frohlich. Joint Bayesian Inference of Condition Specific miRNA and Transcription Factor Activities from Combined Gene and microRNA Expression Data. *submitted*, 2012.

birta 3

birta	Main interface for Bayesian Inference of Regulation of Transcriptional Activity.

# Description

The function estimates parameterization of the model and then executes MCMC sampling to infer activity states.

# Usage

birta(dat.mRNA, dat.miRNA, TFexpr, limmamRNA=NULL, limmamiRNA=NULL, limmaTF=NULL, nrep=NULL, fdr.mRNA

# Arguments

nburnin

dat.mRNA	mRNA expression data (ExpressionSet or matrix). IMPORTANT: Replicates must be ordered according to nrep!
dat.miRNA	miRNA expression data (ExpressionSet or matrix).
TFexpr	TF expression data (ExpressionSet or matrix).
limmamRNA	Output of limma analysis for mRNA data (list: pvalue.tab, lm.fit).
limmamiRNA	Output of limma analysis for miRNA data (list: pvalue.tab, lm.fit).
limmaTF	Output of limma analysis for TF expression data (list: pvalue.tab, lm.fit).
nrep	Vector containing four integers. Entry 1 and 2 are the number of miRNA measurement replicates for condition 1 and 2. Entry 3 and 4 are the number of mRNA measurement replicates for condition 1 and 2.
fdr.mRNA	False discovery rate (FDR) cutoff for significance of the logFC for mRNA data.
fdr.miRNA	False discovery rate (FDR) cutoff for significance of the logFC for miRNA data.
lfc.mRNA	Additional logFC cutoff for significance in mRNA data.
lfc.miRNA	Additional logFC cutoff for significance in miRNA data.
genesets	Combined TF / miRNA network. IMPORTANT: Names of TF entries must start with V\ $\$$ .
lambda	Regularization parameter for edge weights.
sample.weights	Should edge weights be adapted during sampling?
one.regulator.w	weight
	Sould weights of all edges for a regulator to be the same?
theta_TF	Expected fraction of active TFs.
theta_miRNA	Expected fraction of active miRNAs.
model	Type of model. One out of c("all-plug-in", "weight-plug-in", "no-plug-in").
niter	Number of MCMC iterations (AFTER burnin).

Number of MCMC iterations UNTIL burnin is assumed to be finished.

4 birta

potential\_swaps

Pre-computed potential swaps (OPTIONAL, see get\_potential\_swaps).

run.pretest

Initialize miRNA and TF states via the result of a hypergeometric test in order to improve convergence (should be taken with care; advise: only use it in case of observed convergence problems!).

condition.specific.inference

Should inference on TF / miRNA activities be made only RELATIVE to a ref-

erence condition or independently in both conditions?

thin

Thinning of Markov chain: only use every thin's sample for posterior computa-

tion.

only\_switches

Should only switches be performed?

weightSampleMean

Mean for normal distribution used for sampling the omegas.

weightSampleVariance

Variance for normal distribution used for sampling the omegas.

#### Value

The function returns a list containing the following entries:

miRNAstates1 Probability of each miRNA to be active in condition 1 (only for condition spe-

cific sampling).

miRNAstates2 Probability of each miRNA to be active in condition 2 (only for condition spe-

cific sampling).

miRNAActivitySwitch

Probability of each miRNA switching its activities (non-condition specific sam-

pling).

TFstates1 Probability of each TF to be active in condition 1 (only for condition specific

sampling).

TFstates2 Probability of each TF to be active in condition 2 (only for condition specific

sampling).

miRNAactivitySwitch

Probability of each TF switching its activities (non-condition specific sampling).

log\_lik\_trace Log-likelihood trace of MCMC sampling.

TFomega Weights of TF-target graph. (effect of a TF on its targets)

miRNAomega Weights of miRNA-target graph. (effect of a miRNA on its targets)

genesetsTF TF-target network. This might be different from the network submitted to the

function, due to incompatibilities of network and experimental measurements.

Check your warnings and command line output!

genesetsmiRNA miRNA-target network. This might be different from the network submitted

to the function, due to incompatibilities of network and experimental measure-

ments. Check your warnings and command line output!

mRNA expression data. This might be different from the matrix submitted to the

function, due to incompatibilities of network and experimental measurements.

Check your warnings and command line output!

birta-methods 5

miRNA expression data. This might be different from the matrix submitted

to the function, due to incompatibilities of network and experimental measure-

ments. Check your warnings and command line output!

TF expression data (only if they are specifically included). This might be dif-

ferent from the matrix submitted to the function, due to incompatibilities of network and experimental measurements. Check your warnings and command

line output!

#### Author(s)

Holger Frohlich, Benedikt Zacher

# **Examples**

```
data(humanSim)
design = model.matrix(~0+factor(c(rep("control", 5), rep("treated", 5))))
colnames(design) = c("control", "treated")
contrasts = "treated - control"
limmamRNA = limmaAnalysis(sim$dat.mRNA, design, contrasts)
limmamiRNA = limmaAnalysis(sim$dat.miRNA, design, contrasts)
sim_result = birta(sim$dat.mRNA, sim$dat.miRNA, limmamRNA=limmamRNA,
limmamiRNA=limmamiRNA, nrep=c(5,5,5,5), genesets=genesets,
model="all-plug-in", niter=50000, nburnin=10000,
sample.weights=FALSE, potential_swaps=potential_swaps)
```

birta-methods

Methods for Function birta in Package birta

#### Description

Generic methods for function birta in package **birta**. The expression data can be stored in a matrix or an ExpressionSet.

signature(dat.mRNA = "matrix", dat.miRNA = "missing", TFexpr = "missing")

#### Methods

The following signatures make sure, that the arguments, storing the expression data are in the correct format.

```
signature(dat.mRNA = "ExpressionSet", dat.miRNA = "ExpressionSet", TFexpr = "ExpressionSet")
signature(dat.mRNA = "ExpressionSet", dat.miRNA = "missing", TFexpr = "ExpressionSet")
signature(dat.mRNA = "ExpressionSet", dat.miRNA = "missing", TFexpr = "missing")
signature(dat.mRNA = "matrix", dat.miRNA = "matrix", TFexpr = "matrix")
signature(dat.mRNA = "matrix", dat.miRNA = "matrix", TFexpr = "missing")
signature(dat.mRNA = "matrix", dat.miRNA = "missing", TFexpr = "matrix")
```

6 EColiOxygen

EColiNetwork

Example TF-target graph from Regulon DB.

# **Description**

This list contains the TF-target graph used in the vignette.

# Usage

EColiNetwork

#### **Format**

A list containing the target gene sets of 160 TFs.

# **Source**

This TF-target graph was taken from (R. Castelo and A. Roverato, 2009). It is a pre-filtered version of RegulonDB 6.1.

#### References

R. Castelo and A. Roverato. Reverse engineering molecular regulatory networks from microarray data with qp-graphs. *J Comput Biol*, 16(2):213227, Feb 2009.

EColi0xygen

Example data set from E. Coli to sample TF activities.

# Description

This data set gives expression values for three experiments of the E. Coli K12 strain under aerobic and three experiments under aerobic growth. It is used in the vignette to illustrate application of birta to TFs only.

# Usage

EColi0xygen

# **Format**

ExpressionSet

# Source

The original data comes from (Covert et al., 2004) The normalized data set used here is taken from the appropriate by R. Castelo and A. Roverato.

genesets 7

#### References

M. W. Covert, E. M. Knight, J. L. Reed, M. J. Herrgard, and B. O. Palsson. Integrating high-throughput and computational data elucidates bacterial networks. *Nature*, 429(6987):9296,May 2004.

genesets

TF-target and miRNA-target networks for simulated example.

# **Description**

For miRNAs we looked at target predictions in human via miRanda (Betel et al., 2008), miRBase (Griffths-Jones et al., 2008) and miRDB (Wang and Naqa, 2008). Experimentally confirmed targets were retrieved from Tarbase (Papadopoulos et al., 2009). As trustworthy miRNA targets we considered those, which were either experimentally confirmed (i.e. listed in Tarbase) or predicted by at least two of the other three methods. In conclusion we arrived at a network with 583 miRNAs regulating between 1 and 1628 genes (median: 302).\ A TF-target gene network was compiled by computing TF binding affinities to promoter sequences of all human genes according to the TRAP model (Roideret al., 2007) via the author's R implementation. Upstream sequences of genes were retrieved here from the ENSEMBL database via biomaRt (Haider et al., 2009). We assumed that promoter sequences were located in the range 0 - 2Kbp upstream to the transcription start site of a gene. As trustworthy TF targets we considered those, for which a Holm corrected affinity p-value smaller than 0.001 was reported. In conclusion we arrived at a network with 272 TFs regulating between 1 and 16517 genes (median: 20).

# Usage

genesets

# **Format**

A list containing a TF- and miRNA-target graph.

#### Source

The networks were generated using the TRAP model on TRANSFAC matrices and miRNA-target annotations form different databases.

#### References

- D. Betel, M. Wilson, A. Gabow, D. S. Marks, and C. Sander. The microrna.org resource: targets and expression. *Nucleic Acids Res*, 36(Database issue):D149-D153, Jan 2008.
- S. Griffiths-Jones, H. K. Saini, S. van Dongen, and A. J. Enright. miRBase: tools for microRNA genomics. *Nucleic Acids Res*, 36(Database issue):D154-D158, Jan. 2008.
- X. Wang and I. M. E. Naqa. Prediction of both conserved and nonconserved microrna targets in animals. *Bioinformatics*, 24(3):325-332, Feb 2008.

8 get\_potential\_swaps

G. L. Papadopoulos, M. Reczko, V. A. Simossis, P. Sethupathy, and A. G. Hatzigeorgiou. The database of experimentally supported targets: a functional update of tarbase. *Nucleic Acids Res*, 37(Database issue):D155-D158, Jan 2009.

get\_potential\_swaps Calculate swap partner for TF-/miRNA-target graph.

# **Description**

Calculates for TF-/miRNA-target graph all potential swap partner.

# Usage

get\_potential\_swaps(genesetsTF=NULL, genesetsmiRNA=NULL, perc.overlap.cutoff = 0.8, integer.id=TRUE,

# **Arguments**

genesetsTF Each entry corresponds to a TF and contains its target genes.

genesetsmiRNA Each entry corresponds to a miRNA and contains its target genes.

perc.overlap.cutoff

Percentage cutoff of minimal overlap between two miRNAs or TFs to be possi-

ble swap partner.

integer . id If TRUE, the swap partner are not outputted as characters, but as integer indices.

verbose print process or not.

# Value

The function returns a list, with the following entries:

T\_potential\_swaps

Contains for each TF all potential swap partner.

S\_potential\_swaps

Contains for each miRNA all potential swap partner.

#### Author(s)

Benedikt Zacher (zacher@lmb.uni-muenchen.de)

# Examples

```
genesetsTF = list("V$1"=c("A", "B", "C", "D", "E"), "V$2"=c("A", "B", "C", "D"), "V$3"=c("A", "B", "C")) genesetsmiRNA = list("miR-1"=c("C", "E", "D", "G", "H"), "miR-2"=c("C", "E", "D", "G"), "miR-3"=c("C", "E")) get_potential_swaps(genesetsTF, genesetsmiRNA, integer.id=FALSE, perc.overlap.cutoff=0.7)
```

limmaAnalysis 9

limmaAnalysis	Perform a limma analysis on expression data.
---------------	--

# **Description**

Needed for the parameterization of the model as well as the Fisher (pre-)test.

# Usage

```
limmaAnalysis(dat, design, contrasts)
```

# **Arguments**

dat A matrix or ExpressionSet containing the expression values.

design A design matrix.

contrasts Contrast for the linear model.

#### Value

Returns a list containing the following entries:

pvalue.tab Containing the result of the topTable function from limma.

lm.fit Linear fit to the model.design The design used.contrast The contrasts used.

#### Author(s)

Benedikt Zacher (zacher@lmb.uni-muenchen.de), Holger Frohlich

# References

G. K. Smyth. Limma: Linear Models for Microarray Data. Bioinformatics, (2005):397-420.

#### See Also

```
lmFit, topTable
```

# **Examples**

```
data(humanSim)
design = model.matrix(~0+factor(c(rep("control", 5), rep("treated", 5))))
colnames(design) = c("control", "treated")
contrasts = "treated - control"
limmamRNA = limmaAnalysis(sim$dat.mRNA, design, contrasts)
```

10 plotConvergence

limmaAnalysis-methods *Methods for Function* limmaAnalysis *in Package* **birta** 

# Description

Generic methods for function limmaAnalysis in package **birta**. The expression data can be stored in a matrix or an ExpressionSet.

# Methods

```
signature(dat = "ExpressionSet", design = "matrix", contrasts = "character") Generic
function for objects of class ExpressionSet.
```

signature(dat = "matrix", design = "matrix", contrasts = "character") Generic function for objects of class matrix.

plotConvergence

Plotting the likelihood along MCMC sampling.

# **Description**

Plots the log likelihood along MCMC sampling.

# Usage

```
plotConvergence(res, nburnin=NULL, title="")
```

# Arguments

res The result from birta.run (a list).

nburnin Number of iterations used for the burn in.

title Optional title of the plot.

# Author(s)

Benedikt Zacher <zacher@lmb.uni-muenchen.de>

#### See Also

birta

potential\_swaps 11

# **Examples**

```
data(humanSim)
data(humanSim)
design = model.matrix(~0+factor(c(rep("control", 5), rep("treated", 5))))
colnames(design) = c("control", "treated")
contrasts = "treated - control"
limmamRNA = limmaAnalysis(sim$dat.mRNA, design, contrasts)
limmamiRNA = limmaAnalysis(sim$dat.miRNA, design, contrasts)
sim_result = birta(sim$dat.mRNA, sim$dat.miRNA, limmamRNA=limmamRNA,
limmamiRNA=limmamiRNA, nrep=c(5,5,5,5), genesets=genesets,
model="all-plug-in", niter=50000, nburnin=10000,
sample.weights=FALSE, potential_swaps=potential_swaps)
plotConvergence(sim_result, nburnin=10000, title="simulation")
```

potential\_swaps

Potential swap moves for TF-target and miRNA-target networks (see genestes data set).

# Description

Potential swap moves for TF-target and miRNA-target networks (see genestes data set) used for the simulation in the vignette.

# Usage

```
potential_swaps
```

# **Format**

A list containing potential swap moves for the TF- and miRNA-target graph used in the simulation.

sim

Simulated expression data for mRNAs and miRNAs.

# **Description**

This data set contains simulated expression values of miRNAs and mRNAs, together with the associated TF- and miRNA-target networks.

# Usage

sim

#### **Format**

A list containing the entries dat.mRNA (mRNA expression), dat.miRNA (miRNA expression), miRNAstates (activity states of miRNAs), TFstates (activity states of TFs, TRANSFAC), genesetsmiRNA (miRNA-target graph, targetScan), genesetsTFs (TF-taget graph).

TFexpr

# **Source**

Simulation (see reference for details).

#### References

B. Zacher, K. Abnaof, S. Gade, E. Younesi, A. Tresch, and H. Frohlich. Joint Bayesian Inference of Condition Specific miRNA and Transcription Factor Activities from Combined Gene and microRNA Expression Data. *submitted*, 2012.

TFexpr

Transcription factor expression values for the aerobic-anaerobic growth experiment.

# **Description**

This data set gives expression values for the 160 TF of the TF-target graph EColiNetwork used in the vignette.

# Usage

**TFexpr** 

# **Format**

ExpressionSet. Rownames in the assayData correspond to entries in TF-target graph.

# **Source**

See EColiOxygen and EColiNetwork (see reference for details).

#### References

M. W. Covert, E. M. Knight, J. L. Reed, M. J. Herrgard, and B. O. Palsson. Integrating high-throughput and computational data elucidates bacterial networks. *Nature*, 429(6987):92-96, May 2004.

# **Index**

```
*Topic datasets
                                                 birta-package, 2
    EColiNetwork, 6
                                                 EColiNetwork, 6
    EColiOxygen, 6
                                                 EColiOxygen, 6
    genesets, 7
    potential_swaps, 11
                                                 genesets, 7
    sim, 11
                                                 get_potential_swaps, 8
    TFexpr, 12
*Topic hplot
                                                 limmaAnalysis, 9
    plotConvergence, 10
                                                 limma Analysis, {\tt ExpressionSet}, {\tt matrix}, {\tt character-method}
*Topic htest
                                                          (limmaAnalysis-methods), 10
    birta, 3
                                                 limmaAnalysis, matrix, matrix, character-method
    birta-methods, 5
                                                          (limmaAnalysis-methods), 10
*Topic manip
                                                 limmaAnalysis-methods, 10
    get_potential_swaps, 8
                                                 lmFit, 9
    limmaAnalysis, 9
    limmaAnalysis-methods, 10
                                                 plotConvergence, 10
*Topic methods
                                                 potential_swaps, 11
    birta-methods, 5
    limmaAnalysis-methods, 10
                                                 sim, 11
*Topic package
    birta-package, 2
                                                 TFexpr, 12
                                                 topTable, 9
birta, 3, 10
birta, ExpressionSet, ExpressionSet, ExpressionSet-method
        (birta-methods), 5
birta, ExpressionSet, ExpressionSet, missing-method
         (birta-methods), 5
birta, ExpressionSet, missing, ExpressionSet-method
        (birta-methods), 5
\verb|birta, ExpressionSet, missing, missing-method|\\
        (birta-methods), 5
birta, matrix, matrix-method
        (birta-methods), 5
birta, matrix, matrix, missing-method
        (birta-methods), 5
birta, matrix, missing, matrix-method
        (birta-methods), 5
birta, matrix, missing, missing-method
        (birta-methods), 5
birta-methods, 5
```