

Package ‘AnaCoDa’

May 12, 2019

Type Package

Title Analysis of Codon Data under Stationarity using a Bayesian Framework

Version 0.1.3.0

Date 2019-05-11

Maintainer Cedric Landerer <cedric.landerer@gmail.com>

URL <https://github.com/clandere/AnaCoDa>

VignetteBuilder knitr

NeedsCompilation yes

Depends R (>= 3.3.0), Rcpp (>= 0.11.3), methods, mvtnorm

Suggests knitr, Hmisc, VGAM, coda, testthat, lmodel2

RcppModules Test_mod, Trace_mod, CovarianceMatrix_mod,
MCMCAlgorithm_mod, Model_mod, Parameter_mod, Genome_mod,
Gene_mod, SequenceSummary_mod

Description Is a collection of models to analyze genome scale codon data using a Bayesian framework. Provides visualization routines and checkpointing for model fittings. Currently published models to analyze gene data for selection on codon usage based on Ribosome Overhead Cost (ROC) are: ROC (Gilchrist et al. (2015) <doi:10.1093/gbe/evv087>), and ROC with phi (Wallace & Drummond (2013) <doi:10.1093/molbev/mst051>). In addition 'AnaCoDa' contains three currently unpublished models. The FONSE (First order approximation On NonSense Error) model analyzes gene data for selection on codon usage against of nonsense error rates. The PA (PAusing time) and PANSE (PAusing time + NonSense Error) models use ribosome footprinting data to analyze estimate ribosome pausing times with and without nonsense error rate from ribosome footprinting data.

License GPL (>= 2)

Imports

LinkingTo Rcpp

LazyLoad yes

LazyData yes

RoxygenNote 6.1.1

Author Cedric Landerer [aut, cre],
 Gabriel Hanas [ctb],
 Jeremy Rogers [ctb],
 Alex Cope [ctb],
 Denizhan Pak [ctb]

Repository CRAN

Date/Publication 2019-05-11 22:12:44 UTC

R topics documented:

AAToCodon	3
acfCSP	4
acfMCMC	4
addObservedSynthesisRateSet	5
aminoAcids	6
calculateSCUO	6
calculate_marginal_log_likelihood	7
codons	8
codonToAA	8
convergence.test	9
findOptimalCodon	10
geomMean	11
getCAI	12
getCAIweights	13
getCodonCounts	14
getCodonCountsForAA	15
getCSPEstimates	15
getExpressionEstimates	17
getMixtureAssignmentEstimate	18
getNames	20
getNc	20
getNcAA	21
getObservedSynthesisRateSet	22
getSelectionCoefficients	22
getTrace	23
initializeCovarianceMatrices	24
initializeGenomeObject	25
initializeMCMCObject	26
initializeModelObject	27
initializeParameterObject	29
length.Rcpp_Genome	31
loadMCMCObject	32
loadParameterObject	32

plot.Rcpp_FONSEModel	33
plot.Rcpp_FONSEParameter	34
plot.Rcpp_MCMCAlgortithm	35
plot.Rcpp_ROCModel	35
plot.Rcpp_ROCParameter	36
plot.Rcpp_Trace	37
plotCodonSpecificParameters	37
runMCMC	38
setRestartSettings	39
summary.Rcpp_Genome	40
writeMCMCObject	41
writeParameterObject	42

AAToCodon*Amino Acid to codon set*

Description

Converts one character amino acid code to the set of codon encoding that amino acid

Usage

```
AAToCodon(aa, focal = FALSE)
```

Arguments

- | | |
|-------|---|
| aa | Amino acid in single character notation |
| focal | logical, Include the alphabetically first (focal) codon |

Value

Returns the names of the codon encoding the give amino acid

See Also

[codonToAA](#)

acfCSP

*Plots ACF for codon specific parameter traces***Description**

The function calculates and by defaults plots the acf and estimates the autocorrelation in the trace

Usage

```
acfCSP(parameter, csp = "Mutation", numMixtures = 1, samples = NULL,
       lag.max = 40, plot = TRUE)
```

Arguments

parameter	object of class Parameter
csp	"Selection" or "Mutation", defaults to "Mutation"
numMixtures	indicates the number of CSP mixtures used
samples	number of samples at the end of the trace used to calculate the acf
lag.max	Maximum amount of lag to calculate acf. Default is $10 \log_{10}(N)$, where N i the number of observations.
plot	logical. If TRUE (default) a plot of the acf is created

See Also

[acfMCMC](#)

acfMCMC

*Autocorrelation function for the likelihood or posterior trace***Description**

The function calculates and by defaults plots the acf and estimates the autocorrelation in the trace.

Usage

```
acfMCMC(mcmc, type = "LogPosterior", samples = NULL, lag.max = 40,
        plot = TRUE)
```

Arguments

mcmc	object of class MCMC
type	"LogPosterior" or "LogLikelihood", defaults to "LogPosterior"
samples	number of samples at the end of the trace used to calculate the acf
lag.max	Maximum amount of lag to calculate acf. Default is $10 \log_{10}(N)$, where N i the number of observations.
plot	logical. If TRUE (default) a plot of the acf is created

See Also[acfCSP](#)

addObservedSynthesisRateSet*Add gene observed synthesis rates*

Description

`addObservedSynthesisRateSet` returns the observed synthesis rates of the genes within the genome specified.

Usage

```
addObservedSynthesisRateSet(genome, observed.expression.file,  
                           match.expression.by.id = TRUE)
```

Arguments

<code>genome</code>	A genome object initialized with initializeGenomeObject to add observed expression data.
<code>observed.expression.file</code>	A string containing the location of a file containing empirical expression rates (optional).
<code>match.expression.by.id</code>	If TRUE (default) observed expression values will be assigned by matching sequence identifier. If FALSE observed expression values will be assigned by order

Value

Returns the genome after adding the new gene expression values

Examples

```
genome_file <- system.file("extdata", "genome.fasta", package = "AnaCoDa")  
expression_file <- system.file("extdata", "expression.csv", package = "AnaCoDa")  
## reading genome  
genome <- initializeGenomeObject(file = genome_file)  
  
## add expression values after the genome was initialized,  
## or adding an additional set of expression values  
genome <- addObservedSynthesisRateSet(genome = genome,  
                                       observed.expression.file = expression_file)
```

aminoAcids	<i>Amino acids</i>
------------	--------------------

Description

Returns a vector of all amino acids

Usage

```
aminoAcids()
```

Value

Returns a vector of all amino acids

See Also

[codons](#)

calculateSCUO	<i>calculates the synonymous codon usage order (SCUO)</i>
---------------	---

Description

`calculateSCUO` calculates the SCUO value for each gene in genome. Note that if a codon is absent, this will be treated as NA and will be skipped in final calculation

Usage

```
calculateSCUO(genome)
```

Arguments

genome A genome object initialized with [initializeGenomeObject](#).

Value

returns the SCUO value for each gene in genome

Examples

```
genome_file <- system.file("extdata", "genome.fasta", package = "AnaCoDa")
## reading genome
genome <- initializeGenomeObject(file = genome_file)
scuo <- calculateSCUO(genome)
```

calculate_marginal_log_likelihood

Calculates the marginal log-likelihood for a set of parameters

Description

initializes the model object.

Usage

```
calculate_marginal_log_likelihood(parameter, mcmc, mixture, n.samples,  
    divisor)
```

Arguments

parameter	An object created with <code>initializeParameterObject</code> .
mcmc	An object created with <code>initializeMCMCObject</code>
mixture	determines for which mixture the marginal log-likelihood should be calculated
n.samples	How many samples should be used for the calculation
divisor	A value > 1 in order to scale down the tails of the importance distribution

Details

`calculate_marginal_log_likelihood` Calculate marginal log-likelihood for calculation of the Bayes factor using a generalized harmonix mean estimator of the marginal likelihood. See Gronau et al. (2017) for details

Value

This function returns the model object created.

Examples

```

## Not run:
# Calculate the log-marginal likelihood
parameter <- loadParameterObject("parameter.Rda")
mcmc <- loadMCMCObject("mcmc.Rda")
calculate_marginal_likelihood(parameter, mcmc, mixture = 1,
samples = 500, scaling = 1.5)

# Calculate the Bayes factor for two models
parameter1 <- loadParameterObject("parameter1.Rda")
parameter2 <- loadParameterObject("parameter2.Rda")
mcmc1 <- loadMCMCObject("mcmc1.Rda")
mcmc2 <- loadMCMCObject("mcmc2.Rda")
mll1 <- calculate_marginal_likelihood(parameter1, mcmc1, mixture = 1,
samples = 500, scaling = 1.5)

```

```
mll2 <- calculate_marginal_likelihood(parameter2, mcmc2, mixture = 1,
samples = 500, scaling = 1.5)
cat("Bayes factor: ", mll1 - mll2, "\n")

## End(Not run)
```

codons*Codons***Description**

Returns a vector of all codons

Usage

```
codons()
```

Value

Returns a vector of all codons

See Also

[aminoAcids](#)

codonToAA*translates codon to amino acid***Description**

Translates a given codon into the amino acid encoded by it.

Usage

```
codonToAA(codon)
```

Arguments

codon	character, codon to translate
-------	-------------------------------

Value

Returns the amino acid encoded by the given codon as character

See Also

[AAToCodon](#)

<code>convergence.test</code>	<i>Convergence Test</i>
-------------------------------	-------------------------

Description

Convergence Test

Usage

```
convergence.test(object, samples = 10, frac1 = 0.1, frac2 = 0.5,
                 thin = 1, plot = FALSE, what = "Mutation", mixture = 1)
```

Arguments

<code>object</code>	an object of either class Trace or MCMC
<code>samples</code>	number of samples at the end of the trace used to determine convergence (< length of trace). Will use as starting point of convergence test. If the MCMC trace is of length x, then starting point for convergence test will be x - samples.
<code>frac1</code>	fraction to use from beginning of samples
<code>frac2</code>	fraction to use from end of samples
<code>thin</code>	the thinning interval between consecutive observations, which is used in creating a coda::mcmc object (according to the Coda documentation, users should specify if a MCMC chain has already been thinned using a the thin parameter). This does not further thin the data.
<code>plot</code>	(logical) plot result instead of returning an object
<code>what</code>	(for Trace Object only) which parameter to calculate convergence.test – current options are Selection, Mutation, MixtureProbability, Sphi, Mphi, ExpectedPhi, and AcceptanceCSP
<code>mixture</code>	(for Trace Object only) mixture for which to calculate convergence.test

Details

Be aware that `convergence.test` for Trace objects works primarily for Trace objects from the ROC parameter class. Future updates will adapt this function to work for parameters from other models and expression traces

Value

Geweke score object evaluating whether means of two fractions (`frac1` and `frac2`) differ. Convergence occurs when they don't differ significantly, i.e. `pnorm(abs(convergence.test(mcmcObj)$a,,lower.tail=FALSE)^2 > 0.05`

Examples

```

## check for convergence after a run:

genome_file <- system.file("extdata", "genome.fasta", package = "AnaCoDa")

genome <- initializeGenomeObject(file = genome_file)
sphi_init <- c(1,1)
numMixtures <- 2
geneAssignment <- c(rep(1,floor(length(genome)/2)),rep(2,ceiling(length(genome)/2)))
parameter <- initializeParameterObject(genome = genome, sphi = sphi_init,
                                         num.mixtures = numMixtures,
                                         gene.assignment = geneAssignment,
                                         mixture.definition = "allUnique")
samples <- 2500
thinning <- 50
adaptiveWidth <- 25
mcmc <- initializeMCMCObject(samples = samples, thinning = thinning,
                               adaptive.width=adaptiveWidth, est.expression=TRUE,
                               est.csp=TRUE, est.hyper=TRUE, est.mix = TRUE)
divergence.iteration <- 10
## Not run:
runMCMC(mcmc, genome = genome, model = model,
         ncores = 4, divergence.iteration = divergence.iteration)
# check if posterior trace has converged
convergence.test(object = mcmc, samples = 500, plot = TRUE)

trace <- getTrace(parameter)
# check if Mutation trace has converged
convergence.test(object = trace, samples = 500, plot = TRUE, what = "Mutation")
# check if Sphi trace has converged
convergence.test(object = trace, samples = 500, plot = TRUE, what = "Sphi")
# check if ExpectedPhi trace has converged
convergence.test(object = trace, samples = 500, plot = TRUE, what = "ExpectedPhi")

## End(Not run)

```

findOptimalCodon *Find and return list of optimal codons*

Description

`findOptimalCodon` extracrs the optimal codon for each amino acid.

Usage

```
findOptimalCodon(csp)
```

Arguments

csp a data.frame as returned by getCSPEstimates.

Value

A named list with with optimal codons for each amino acid.

Examples

```
genome_file <- system.file("extdata", "genome.fasta", package = "AnaCoDa")

genome <- initializeGenomeObject(file = genome_file)
sphi_init <- 1
numMixtures <- 1
geneAssignment <- rep(1, length(genome))
parameter <- initializeParameterObject(genome = genome, sphi = sphi_init,
                                         num.mixtures = numMixtures,
                                         gene.assignment = geneAssignment,
                                         mixture.definition = "allUnique")
model <- initializeModelObject(parameter = parameter, model = "ROC")
samples <- 2500
thinning <- 50
adaptiveWidth <- 25
mcmc <- initializeMCMCObject(samples = samples, thinning = thinning,
                               adaptive.width=adaptiveWidth, est.expression=TRUE,
                               est.csp=TRUE, est.hyper=TRUE, est.mix = TRUE)
divergence.iteration <- 10
## Not run:
runMCMC(mcmc = mcmc, genome = genome, model = model,
         ncores = 4, divergence.iteration = divergence.iteration)

csp_mat <- getCSPEstimates(parameter, CSP="Selection")
opt_codons <- findOptimalCodon(csp_mat)

## End(Not run)
```

Description

geomMean will calculate the geometric mean of a list of numerical values.

Usage

```
geomMean(x, rm.invalid = TRUE, default = 1e-05)
```

Arguments

x	A vector of numerical .
rm.invalid	Boolean value for handling 0, negative, or NA values in the vector. Default is TRUE and will not include these values in the calculation. If FALSE, these values will be replaced by the value give to default and will be included in the calculation.
default	Numerical value that serves as the value to replace 0, negative, or NA values in the calculation when rm.invalid is FALSE. Default is 1e-5.

Details

This function is a special version of the geometric mean specifically for AnaCoda. Most models in Anacoda assume a log normal distribution for phi values, thus all values in x are expectd to be positive. geomMean returns the geometric mean of a vector and can handle 0, negative, or NA values.

Value

Returns the geometric mean of a vector.

Examples

```
x <- c(1, 2, 3, 4)
geomMean(x)

y<- c(1, NA, 3, 4, 0, -1)
# Only take the mean of non-Na values greater than 0
geomMean(y)

# Replace values <= 0 or NAs with a default value 0.001 and then take the mean
geomMean(y, rm.invalid = FALSE, default = 0.001)
```

Description

getCAI returns the Codon Adaptation Index for a genome based on a provided reference.

Usage

```
getCAI(referenceGenome, testGenome, default.weight = 0.5)
```

Arguments

referenceGenome	A genome object initialized with <code>initializeGenomeObject</code> . Serves as reference set to calculate the necessary codon weights.
testGenome	A genome object initialized with <code>initializeGenomeObject</code> . The genome for which the CAI is supposed to be calculated
default.weight	Default weight to use if codon is missing from referenceGenome

Value

Returns a named vector with the CAI for each gene

Examples

```
genome_file1 <- system.file("extdata", "more_genes.fasta", package = "AnaCoDa")
genome_file2 <- system.file("extdata", "genome.fasta", package = "AnaCoDa")

## reading genome
referenceGenome <- initializeGenomeObject(file = genome_file1)
testGenome <- initializeGenomeObject(file = genome_file2)

cai <- getCAI(referenceGenome, testGenome)
```

getCAIweights

Calculate the CAI codon weights for a reference genome

Description

`getCAIweights` returns the weights for the Codon Adaptation Index based on a reference genome.

Usage

```
getCAIweights(referenceGenome, default.weight = 0.5)
```

Arguments

referenceGenome	A genome object initialized with <code>initializeGenomeObject</code> .
default.weight	Set default weight for any codon not observed in the reference genome

Value

Returns a named vector with the CAI weights for each codon

Examples

```
genome_file <- system.file("extdata", "genome.fasta", package = "AnaCoDa")
## reading genome
referenceGenome <- initializeGenomeObject(file = genome_file)

wi <- getCAIweights(referenceGenome)
```

getCodonCounts

Get Codon Counts For all Amino Acids

Description

provides the codon counts for a given amino acid across all genes

Usage

```
getCodonCounts(genome)
```

Arguments

genome	A genome object from which the counts of each codon can be obtained.
--------	--

Details

The returned matrix contains a row for each gene and a column for each synonymous codon of aa.

Value

Returns a data.frame storing the codon counts for each amino acid.

Examples

```
genome_file <- system.file("extdata", "genome.fasta", package = "AnaCoDa")
## reading genome
genome <- initializeGenomeObject(file = genome_file)
counts <- getCodonCounts(genome)
```

getCodonCountsForAA *Get Codon Counts For a specific Amino Acid*

Description

provides the codon counts for a given amino acid across all genes

Usage

```
getCodonCountsForAA(aa, genome)
```

Arguments

aa	One letter code of the amino acid for which the codon counts should be returned
genome	A genome object from which the counts of each codon can be obtained.

Details

The returned matrix contains a row for each gene and a column for each synonymous codon of aa.

Value

Returns a data.frame storing the codon counts for the specified amino acid.

Examples

```
genome_file <- system.file("extdata", "genome.fasta", package = "AnaCoDa")

## reading genome
genome <- initializeGenomeObject(file = genome_file)
counts <- getCodonCountsForAA("A", genome)
```

getCSPEstimates *Return Codon Specific Parameters (or write to csv) estimates as data.frame*

Description

getCSPEstimates returns the codon specific parameter estimates for a given parameter and mixture or write it to a csv file.

Usage

```
getCSPEstimates(parameter, filename = NULL, mixture = 1,
  samples = 10, relative.to.optimal.codon = T,
  report.original.ref = T)
```

Arguments

<code>parameter</code>	parameter an object created by <code>initializeParameterObject</code> .
<code>filename</code>	Posterior estimates will be written to file (format: csv). Filename will be in the format <parameter_name>_<filename>.csv.
<code>mixture</code>	estimates for which mixture should be returned
<code>samples</code>	The number of samples used for the posterior estimates.
<code>relative.to.optimal.codon</code>	Boolean determining if parameters should be relative to the preferred codon or the alphabetically last codon (Default=TRUE). Only applies to ROC and FONSE models
<code>report.original.ref</code>	Include the original reference codon (Default = TRUE). Note this is only included for the purposes of simulations, which expect the input parameter file to be in a specific format. Later version of AnaCoDa will remove this.

Value

returns a list data.frame with the posterior estimates of the models codon specific parameters or writes it directly to a csv file if `filename` is specified

Examples

```
genome_file <- system.file("extdata", "genome.fasta", package = "AnaCoDa")

genome <- initializeGenomeObject(file = genome_file)
sphi_init <- c(1,1)
numMixtures <- 2
geneAssignment <- c(rep(1,floor(length(genome)/2)),rep(2,ceiling(length(genome)/2)))
parameter <- initializeParameterObject(genome = genome, sphi = sphi_init,
                                         num.mixtures = numMixtures,
                                         gene.assignment = geneAssignment,
                                         mixture.definition = "allUnique")
model <- initializeModelObject(parameter = parameter, model = "ROC")
samples <- 2500
thinning <- 50
adaptiveWidth <- 25
mcmc <- initializeMCMCObject(samples = samples, thinning = thinning,
                               adaptive.width=adaptiveWidth, est.expression=TRUE,
                               est.csp=TRUE, est.hyper=TRUE, est.mix = TRUE)
divergence.iteration <- 10
## Not run:
runMCMC(mcmc = mcmc, genome = genome, model = model,
```

```
    ncores = 4, divergence.iteration = divergence.iteration)

    ## return estimates for codon specific parameters
    csp_mat <- getCSPEstimates(parameter)

    # write the result directly to the filesystem as a csv file. No values are returned
    getCSPEstimates(parameter, filename=file.path(tempdir(), "test.csv"))

## End(Not run)
```

getExpressionEstimates

Returns the estimated phi posterior for a gene

Description

Posterior estimates for the phi value of specified genes

Usage

```
getExpressionEstimates(parameter, gene.index, samples,
                      quantiles = c(0.025, 0.975))
```

Arguments

parameter	an object created by <code>initializeParameterObject</code> .
gene.index	a integer or vector of integers representing the gene(s) of interesst.
samples	number of samples for the posterior estimate
quantiles	vector of quantiles, (default: <code>c(0.025, 0.975)</code>)

Details

The returned vector is unnamed as gene ids are only stored in the genome object, but the `gene.index` vector can be used to match the assignment to the genome.

Value

returns a vector with the mixture assignment of each gene corresbonding to `gene.index` in the same order as the genome.

Examples

```

genome_file <- system.file("extdata", "genome.fasta", package = "AnaCoDa")

genome <- initializeGenomeObject(file = genome_file)
sphi_init <- c(1,1)
numMixtures <- 2
geneAssignment <- c(rep(1,floor(length(genome)/2)),rep(2,ceiling(length(genome)/2)))
parameter <- initializeParameterObject(genome = genome, sphi = sphi_init,
                                         num.mixtures = numMixtures,
                                         gene.assignment = geneAssignment,
                                         mixture.definition = "allUnique")
model <- initializeModelObject(parameter = parameter, model = "ROC")
samples <- 2500
thinning <- 50
adaptiveWidth <- 25
mcmc <- initializeMCMCObject(samples = samples, thinning = thinning,
                               adaptive.width=adaptiveWidth, est.expression=TRUE,
                               est.csp=TRUE, est.hyper=TRUE, est.mix = TRUE)
divergence.iteration <- 10
## Not run:
runMCMC(mcmc = mcmc, genome = genome, model = model,
         ncores = 4, divergence.iteration = divergence.iteration)

# get the estimated expression values for all genes based on the mixture
# they are assigned to at each step
estimatedExpression <- getExpressionEstimates(parameter, 1:length(genome), 1000)

## End(Not run)

```

getMixtureAssignmentEstimate

Returns mixture assignment estimates for each gene

Description

Posterior estimates for the mixture assignment of specified genes

Usage

```
getMixtureAssignmentEstimate(parameter, gene.index, samples)
```

Arguments

parameter	an object created by <code>initializeParameterObject</code>
gene.index	a integer or vector of integers representing the gene(s) of interesst.
samples	number of samples for the posterior estimate

Details

The returned vector is unnamed as gene ids are only stored in the genome object, but the `gene.index` vector can be used to match the assignment to the genome.

Value

returns a vector with the mixture assignment of each gene corresponding to `gene.index` in the same order as the genome.

Examples

```
genome_file <- system.file("extdata", "genome.fasta", package = "AnaCoDa")

genome <- initializeGenomeObject(file = genome_file)
sphi_init <- c(1,1)
numMixtures <- 2
geneAssignment <- c(rep(1,floor(length(genome)/2)),rep(2,ceiling(length(genome)/2)))
parameter <- initializeParameterObject(genome = genome, sphi = sphi_init,
                                         num.mixtures = numMixtures,
                                         gene.assignment = geneAssignment,
                                         mixture.definition = "allUnique")
model <- initializeModelObject(parameter = parameter, model = "ROC")
samples <- 2500
thinning <- 50
adaptiveWidth <- 25
mcmc <- initializeMCMCObject(samples = samples, thinning = thinning, adaptive.width=adaptiveWidth,
                               est.expression=TRUE, est.csp=TRUE, est.hyper=TRUE, est.mix = TRUE)
divergence.iteration <- 10
## Not run:
runMCMC(mcmc = mcmc, genome = genome, model = model,
         ncores = 4, divergence.iteration = divergence.iteration)

# get the mixture assignment for all genes
mixAssign <- getMixtureAssignmentEstimate(parameter = parameter,
                                            gene.index = 1:length(genome), samples = 1000)

# get the mixture assignment for a subsample
mixAssign <- getMixtureAssignmentEstimate(parameter = parameter,
                                            gene.index = 5:100, samples = 1000)
# or
mixAssign <- getMixtureAssignmentEstimate(parameter = parameter,
                                            gene.index = c(10, 30:50, 3, 90), samples = 1000)

## End(Not run)
```

getNames

*Gene Names of Genome***Description**

returns the identifiers of the genes within the genome specified.

Usage

```
getNames(genome, simulated = FALSE)
```

Arguments

- | | |
|-----------|---|
| genome | A genome object initialized with initializeGenomeObject . |
| simulated | A logical value denoting if the gene names to be listed are simulated or not. The default value is FALSE. |

Value

gene.names Returns the names of the genes as a vector of strings.

Examples

```
genome_file <- system.file("extdata", "genome.fasta", package = "AnaCoDa")

## reading genome
genome <- initializeGenomeObject(file = genome_file)

## return all gene ids for the genome
geneIDs <- getNames(genome, FALSE)
```

getNc

*Calculate the Effective Number of Codons***Description**

getNc returns the Effective Number of Codons for a genome.

Usage

```
getNc(genome)
```

Arguments

- | | |
|--------|---|
| genome | A genome object initialized with initializeGenomeObject . |
|--------|---|

Value

Returns a named vector with the Effective Number of Codons for each gene

Examples

```
genome_file <- system.file("extdata", "more_genes.fasta", package = "AnaCoDa")
## reading genome
genome <- initializeGenomeObject(file = genome_file)

nc <- getNc(genome)
```

getNcAA

Calculate the Effective Number of Codons for each Amino Acid

Description

getNcAA returns the Effective Number of Codons for each Amino Acid.

Usage

```
getNcAA(genome)
```

Arguments

genome A genome object initialized with [initializeGenomeObject](#).

Value

Returns an object of type `data.frame` with the Effective Number of Codons for each amino acid in each gene.

Examples

```
genome_file <- system.file("extdata", "more_genes.fasta", package = "AnaCoDa")
## reading genome
genome <- initializeGenomeObject(file = genome_file)

nc <- getNcAA(genome)
```

getObservedSynthesisRateSet*Get gene observed synthesis rates***Description**

`getObservedSynthesisRateSet` returns the observed synthesis rates of the genes within the genome specified.

Usage

```
getObservedSynthesisRateSet(genome, simulated = FALSE)
```

Arguments

- | | |
|------------------------|---|
| <code>genome</code> | A genome object initialized with initializeGenomeObject . |
| <code>simulated</code> | A logical value denoting if the synthesis rates to be listed are simulated or not.
The default value is FALSE. |

Value

Returns a data.frame with the observed expression values in genome

Examples

```
genome_file <- system.file("extdata", "genome.fasta", package = "AnaCoDa")
expression_file <- system.file("extdata", "expression.csv", package = "AnaCoDa")
## reading genome
genome <- initializeGenomeObject(file = genome_file)

## return expression values as a data.frame with gene ids in the first column.
expressionValues <- getObservedSynthesisRateSet(genome = genome)
```

getSelectionCoefficients*Calculate Selection coefficients***Description**

`getSelectionCoefficients` calculates the selection coefficient of each codon in each gene.

Usage

```
getSelectionCoefficients(genome, parameter, samples = 100)
```

Arguments

genome	A genome object initialized with <code>initializeGenomeObject</code> to add observed expression data.
parameter	an object created by <code>initializeParameterObject</code> .
samples	The number of samples used for the posterior estimates.

Value

A matrix with selection coefficients.

Examples

```
genome_file <- system.file("extdata", "genome.fasta", package = "AnaCoDa")

genome <- initializeGenomeObject(file = genome_file)
sphi_init <- 1
numMixtures <- 1
geneAssignment <- rep(1, length(genome))
parameter <- initializeParameterObject(genome = genome, sphi = sphi_init,
                                         num.mixtures = numMixtures,
                                         gene.assignment = geneAssignment,
                                         mixture.definition = "allUnique")
model <- initializeModelObject(parameter = parameter, model = "ROC")
samples <- 2500
thinning <- 50
adaptiveWidth <- 25
mcmc <- initializeMCMCObject(samples = samples, thinning = thinning,
                               adaptive.width=adaptiveWidth, est.expression=TRUE,
                               est.csp=TRUE, est.hyper=TRUE, est.mix = TRUE)
divergence.iteration <- 10
## Not run:
runMCMC(mcmc = mcmc, genome = genome, model = model,
        ncores = 4, divergence.iteration = divergence.iteration)

## return estimates for selection coefficients s for each codon in each gene
selection.coefficients <- getSelectionCoefficients(genome = genome,
                                                    parameter = parameter, samples = 1000)

## End(Not run)
```

getTrace

extracts an object of traces from a parameter object.

Description

extracts an object of traces from a parameter object.

Usage

```
getTrace(parameter)
```

Arguments

parameter A Parameter object that corresponds to one of the model types.

Value

trace Returns an object of type Trace extracted from the given parameter object

Examples

```
genome_file <- system.file("extdata", "genome.fasta", package = "AnaCoDa")
genome <- initializeGenomeObject(file = genome_file)
sphi_init <- c(1,1)
numMixtures <- 2
geneAssignment <- c(rep(1,floor(length(genome)/2)),rep(2,ceiling(length(genome)/2)))
parameter <- initializeParameterObject(genome = genome, sphi = sphi_init,
                                         num.mixtures = numMixtures,
                                         gene.assignment = geneAssignment,
                                         mixture.definition = "allUnique")
trace <- getTrace(parameter) # empty trace object since no MCMC was performed
```

initializeCovarianceMatrices
Initialize Covariance Matrices
Description

Initialize Covariance Matrices

Usage

```
initializeCovarianceMatrices(parameter, genome, numMixtures,
                             geneAssignment, init.csp.variance = 0.0025)
```

Arguments

parameter A Parameter object that corresponds to one of the model types. Valid values are "ROC", "PA", and "FONSE".

genome An object of type Genome necessary for the initialization of the Parameter object.

<code>numMixtures</code>	The number of mixture elements for the underlying mixture distribution (<code>numMixtures > 0</code>).
<code>geneAssignment</code>	A vector holding the initial mixture assignment for each gene. The vector length has to equal the number of genes in the genome. Valid values for the vector range from 1 to <code>numMixtures</code> . It is possible but not advised to leave a mixture element empty.
<code>init.csp.variance</code>	initial proposal variance for codon specific parameter, default is 0.0025.

Value

`parameter` Returns the `Parameter` argument, now modified with initialized mutation, selection, and covariance matrices.

initializeGenomeObject*Genome Initialization***Description**

`initializeGenomeObject` initializes the Rcpp Genome object

Usage

```
initializeGenomeObject(file, genome = NULL,
                      observed.expression.file = NULL, fasta = TRUE, simulated = FALSE,
                      match.expression.by.id = TRUE, append = FALSE)
```

Arguments

<code>file</code>	A file of coding sequences in fasta or RFPData format
<code>genome</code>	A genome object can be passed in to concatenate the input file to it (optional).
<code>observed.expression.file</code>	A string containing the location of a file containing empirical expression rates (optional).
<code>fasta</code>	A boolean value which decides whether to initialize with a fasta file or an RFPData file. (TRUE for fasta, FALSE for RFPData)
<code>simulated</code>	boolean to determine if the data should be treated as a simulated data set (Default = FALSE).
<code>match.expression.by.id</code>	If TRUE (default), observed expression values will be assigned by matching sequence identifier. If FALSE, observed expression values will be assigned by order.
<code>append</code>	If TRUE (FALSE is default), function will read in additional genome data to append to an existing genome. If FALSE, genome data is cleared before reading in data (no preexisting data).

Value

This function returns the initialized Genome object.

Examples

```
genome_file <- system.file("extdata", "genome.fasta", package = "AnaCoDa")
genes_file <- system.file("extdata", "more_genes.fasta", package = "AnaCoDa")
expression_file <- system.file("extdata", "expression.csv", package = "AnaCoDa")

## reading genome
genome <- initializeGenomeObject(file = genome_file)

## reading genome and observed expression data
genome <- initializeGenomeObject(file = genome_file, observed.expression.file = expression_file)

## add additional genes to existing genome
genome <- initializeGenomeObject(file = genome_file)
genome <- initializeGenomeObject(file = genes_file, genome = genome, append = TRUE)
```

initializeMCMCObject *Initialize MCMC*

Description

`initializeMCMCObject` initializes a MCMC object to perform a model fitting for a parameter and model object.

Usage

```
initializeMCMCObject(samples, thinning = 1, adaptive.width = 100,
est.expression = TRUE, est.csp = TRUE, est.hyper = TRUE,
est.mix = TRUE)
```

Arguments

<code>samples</code>	Number of samples to be produced when running the MCMC algorithm. No default value.
<code>thinning</code>	The thinning interval between consecutive observations. If set to 1, every step will be saved as a sample. Default value is 1.
<code>adaptive.width</code>	Number that determines how often the acceptance/rejection window should be altered. Default value is 100 samples.
<code>est.expression</code>	Boolean that tells whether or not synthesis rate values should be estimated in the MCMC algorithm run. Default value is TRUE.
<code>est.csp</code>	Boolean that tells whether or not codon specific values should be estimated in the MCMC algorithm run. Default value is TRUE.

est.hyper	Boolean that tells whether or not hyper parameters should be estimated in the MCMC algorithm run. Default value is TRUE.
est.mix	Boolean that tells whether or not the genes' mixture element should be estimated in the MCMC algorithm run. Default value is TRUE.

Details

initializeMCMCObject sets up the MCMC object (monte carlo markov chain) and returns the object so a model fitting can be done. It is important to note that est.expression and est.hyper will affect one another negatively if their values differ.

Value

mcmc Returns an intialized MCMC object.

Examples

```
## initializing an object of type mcmc

samples <- 2500
thinning <- 50
adaptiveWidth <- 25

## estimate all parameter types
mcmc <- initializeMCMCObject(samples = samples, thinning = thinning, adaptive.width=adaptiveWidth,
                               est.expression=TRUE, est.csp=TRUE, est.hyper=TRUE, est.mix = TRUE)

## do not estimate expression values, initial conditions will remain constant
mcmc <- initializeMCMCObject(samples = samples, thinning = thinning, adaptive.width=adaptiveWidth,
                               est.expression=FALSE, est.csp=TRUE, est.hyper=TRUE, est.mix = TRUE)

## do not estimate hyper parameters, initial conditions will remain constant
mcmc <- initializeMCMCObject(samples = samples, thinning = thinning, adaptive.width=adaptiveWidth,
                               est.expression=TRUE, est.csp=TRUE, est.hyper=FALSE, est.mix = TRUE)
```

initializeModelObject *Model Initialization***Description**

initializes the model object.

Usage

```
initializeModelObject(parameter, model = "ROC", with.phi = FALSE,
                      fix.observation.noise = FALSE, rfp.count.column = 1)
```

Arguments

<code>parameter</code>	An object created with <code>initializeParameterObject</code> .
<code>model</code>	A string containing the model to run (ROC, FONSE, or PA), has to match parameter object.
<code>with.phi</code>	(ROC only) A boolean that determines whether or not to include empirical phi values (expression rates) for the calculations.
<code>fix.observation.noise</code>	(ROC only) Allows to fix the noise in the observed expression dataset to the initial condition. The initial condition for the observed expression noise can be set in the parameter object.
<code>rfp.count.column</code>	(PA and PANSE only) A number representing the RFP count column to use.

Details

`initializeModelObject` initializes a model. The type of model is determined based on the string passed to the `model` argument. The Parameter object has to match the model that is initialized. E.g. to initialize a ROC model, it is required that a ROC parameter object is passed to the function.

Value

This function returns the model object created.

Examples

```
#initializing a model object

genome_file <- system.file("extdata", "genome.fasta", package = "AnaCoDa")
expression_file <- system.file("extdata", "expression.csv", package = "AnaCoDa")

genome <- initializeGenomeObject(file = genome_file,
                                   observed.expression.file = expression_file)
sphi_init <- c(1,1)
numMixtures <- 2
geneAssignment <- c(rep(1,floor(length(genome)/2)),rep(2,ceiling(length(genome)/2)))
parameter <- initializeParameterObject(genome = genome, sphi = sphi_init,
                                         num.mixtures = numMixtures,
                                         gene.assignment = geneAssignment,
                                         mixture.definition = "allUnique")

# initializing a model object assuming we have observed expression (phi)
# values stored in the genome object.
initializeModelObject(parameter = parameter, model = "ROC", with.phi = TRUE)

# initializing a model object ignoring observed expression (phi)
# values stored in the genome object.
initializeModelObject(parameter = parameter, model = "ROC", with.phi = FALSE)
```

initializeParameterObject
Initialize Parameter

Description

`initializeParameterObject` initializes a new parameter object or reconstructs one from a restart file

Usage

```
initializeParameterObject(genome = NULL, sphi = NULL,  
  num.mixtures = 1, gene.assignment = NULL,  
  initial.expression.values = NULL, model = "ROC",  
  split.serine = TRUE, mixture.definition = "allUnique",  
  mixture.definition.matrix = NULL, init.with.restart.file = NULL,  
  mutation.prior.sd = 0.35, init.csp.variance = 0.0025,  
  init.sepsilon = 0.1, init.w.obs.phi = FALSE)
```

Arguments

<code>genome</code>	An object of type Genome necessary for the initialization of the Parameter object. The default value is NULL.
<code>sphi</code>	Initial values for sphi. Expected is a vector of length numMixtures. The default value is NULL.
<code>num.mixtures</code>	The number of mixtures elements for the underlying mixture distribution (numMixtures > 0). The default value is 1.
<code>gene.assignment</code>	A vector holding the initial mixture assignment for each gene. The vector length has to equal the number of genes in the genome. Valid values for the vector range from 1 to numMixtures. It is possible but not advised to leave a mixture element empty. The default Value is NULL.
<code>initial.expression.values</code>	(Optional) A vector with intial phi values. The length of the vector has to equal the number of genes in the Genome object. The default value is NULL.
<code>model</code>	Specifies the model used. Valid options are "ROC", "PA", "PANSE", or "FONSE". The default model is "ROC". ROC is described in Gilchrist et al. 2015. PA, PANSE and FONSE are currently unpublished.
<code>split.serine</code>	Whether serine should be considered as one or two amino acids when running the model. TRUE and FALSE are the only valid values. The default value for split.serine is TRUE.
<code>mixture.definition</code>	A string describing how each mixture should be treated with respect to mutation and selection. Valid values consist of "allUnique", "mutationShared", and "selectionShared". The default value for mixture.definition is "allUnique". See details for more information.

```

mixture.definition.matrix
    A matrix representation of how the mutation and selection categories correspond
    to the mixtures. The default value for mixture.definition.matrix is NULL. If
    provided, the model will use the matrix to initialize the mutation and selection
    categories instead of the definition listed directly above. See details for more
    information.

init.with.restart.file
    File name containing information to reinitialize a previous Parameter object. If
    given, all other arguments will be ignored. The default value for init.with.restart.file
    is NULL.

mutation.prior.sd
    Controlling the standard deviation of the normal prior on the mutation parame-
    ters

init.csp.variance
    specifies the initial proposal width for codon specific parameter (default is 0.0025).
    The proposal width adapts during the runtime to reach a target acceptance rate of
    ~0.25

init.sepsilon
    specifies the initial value for sepsilon. default is 0.1

init.w.obs.phi
    TRUE: initialize phi values with observed phi values (data from RNAseq, mass
    spectrometry, ribosome footprinting) Default is FALSE. If multiple observed
    phi values exist for a gene, the geometric mean of these values is used as initial
    phi. When using this function, one should remove any genes with missing phi
    values, as these genes will not have an initial phi value.

```

Details

initializeParameterObject checks the values of the arguments given to insure the values are valid.

The mixture definition and mixture definition matrix describes how the mutation and selection categories are set up with respect to the number of mixtures. For example, if `mixture.definition = "allUnique"` and `numMixtures = 3`, a matrix representation would be `matrix(c(1,2,3,1,2,3), ncol=2)` where each row represents a mixture, the first column represents the mutation category, and the second column represents the selection category. Another example would be `mixture.definition = "selectionShared"` and `numMixtures = 4` (`matrix(c(1,2,3,4,1,1,1,1), ncol=2)`). In this case, the selection category is the same for every mixture. If a matrix is given, and it is valid, then the mutation/selection relationship will be defined by the given matrix and the keyword will be ignored. A matrix should only be given in cases where the keywords would not create the desired matrix.

Value

`parameter` Returns an initialized Parameter object.

Examples

```

genome_file <- system.file("extdata", "genome.fasta", package = "AnaCoDa")
restart_file <- system.file("extdata", "restart_file.rst", package = "AnaCoDa")

genome <- initializeGenomeObject(file = genome_file)

```

```

## initialize a new parameter object
sphi_init <- 1
numMixtures <- 1
geneAssignment <- rep(1, length(genome))
parameter <- initializeParameterObject(genome = genome, sphi = sphi_init,
                                         num.mixtures = numMixtures,
                                         gene.assignment = geneAssignment,
                                         mixture.definition = "allUnique")

## re-initialize a parameter object from a restart file. Useful for checkpointing
parameter <- initializeParameterObject(init.with.restart.file = restart_file)

## initialize a parameter object with a custom mixture definition matrix
def.matrix <- matrix(c(1,1,1,2), ncol=2)
geneAssignment <- c(rep(1,floor(length(genome)/2)),rep(2,ceiling(length(genome)/2)))
parameter <- initializeParameterObject(genome = genome, sphi = c(0.5, 2), num.mixtures = 2,
                                         gene.assignment = geneAssignment,
                                         mixture.definition.matrix = def.matrix)

```

length.Rcpp_Genome *Length of Genome*

Description

length gives the length of a genome

Usage

```

## S3 method for class 'Rcpp_Genome'
length(x)

```

Arguments

x A genome object initialized with [initializeGenomeObject](#).

Value

returns the number of genes in a genome

Examples

```

genome_file <- system.file("extdata", "genome.fasta", package = "AnaCoDa")

## reading genome
genome <- initializeGenomeObject(file = genome_file)
length(genome) # 10

```

`loadMCMCObject` *Load MCMC Object*

Description

`loadMCMCObject` creates a new MCMC object and fills it with the information in the file given.

Usage

```
loadMCMCObject(files)
```

Arguments

<code>files</code>	The filenames where the data will be stored.
--------------------	--

Details

This MCMC object is not intended to be used to do another model fitting, only to graph the stored results.

Value

This function has no return value.

Examples

```
## loading mcmc objects from the filesystem
## Not run:
# load one mcmc object
mcmc <- loadMCMCObject(files = "mcmc.Rda")

# load and combine multiple mcmc objects. Useful when using checkpointing
mcmc <- loadMCMCObject(files = c("mcmc1.Rda", "mcmc2.Rda"))

## End(Not run)
```

`loadParameterObject` *Load Parameter Object*

Description

`loadParameterObject` will load a parameter object from the filesystem

Usage

```
loadParameterObject(files)
```

Arguments

<code>files</code>	A list of parameter filenames to be loaded. If multiple files are given, the parameter objects will be concatenated in the order provided
--------------------	---

Details

The function loads one or multiple files. In the case of multiple file, e.g. due to the use of check pointing, the files will be concatenated to one parameter object. See [writeParameterObject](#) for the writing of parameter objects

Value

Returns an initialized Parameter object.

Examples

```
## Not run:
# load a single parameter object
parameter <- loadParameterObject("parameter.Rda")

# load and concatenate multiple parameter object
parameter <- loadParameterObject(c("parameter1.Rda", "parameter2.Rda"))

## End(Not run)
```

`plot.Rcpp_FONSEModel` *Plot Model Object*

Description

Plots traces from the model object such as synthesis rates for each gene. Will work regardless of whether or not expression/synthesis rate levels are being estimated. If you wish to plot observed/empirical values, these values MUST be set using the `initial.expression.values` parameter found in `initializeParameterObject`. Otherwise, the expression values plotted will just be SCUO values estimated upon initialization of the Parameter object.

Usage

```
## S3 method for class 'Rcpp_FONSEModel'
plot(x, genome, samples = 100, mixture = 1,
      simulated = FALSE, ...)
```

Arguments

x	An Rcpp model object initialized with <code>initializeModelObject</code> .
genome	An Rcpp genome object initialized with <code>initializeGenomeObject</code> .
samples	The number of samples in the trace
mixture	The mixture for which to graph values.
simulated	A boolean value that determines whether to use the simulated genome.
...	Optional, additional arguments. For this function, a possible title for the plot in the form of a list if set with "main".

Value

This function has no return value.

plot.Rcpp_FONSEParameter
Plot Parameter

Description

`plot` graphs the mutation or selection parameter for a ROC or FONSE parameter object for each mixture element.

Usage

```
## S3 method for class 'Rcpp_FONSEParameter'
plot(x, what = "Mutation", samples = 100,
      mixture.name = NULL, with.ci = TRUE, ...)
```

Arguments

x	A parameter object
what	Which aspect of the parameter to plot. Default value is "Mutation".
samples	Number of samples to plot using the posterior mean. Default value is 100.
mixture.name	a vector with names/descriptions of the mixture distributions in the parameter object
with.ci	Plot with or without confidence intervals. Default value is TRUE
...	Arguments to be passed to methods, such as graphical parameters.

Details

Graphs are based off the last # samples for the posterior mean.

Value

This function has no return value.

plot.Rcpp_MCMCAlgortithm
Plot MCMC algorithm

Description

This function will plot the logLikelihood trace, and if the Hmisc package is installed, it will plot a subplot of the logLikelihood trace with the first few samples removed.

Usage

```
## S3 method for class 'Rcpp_MCMCAlgortithm'
plot(x, what = "LogPosterior",
      zoom.window = NULL, ...)
```

Arguments

- | | |
|-------------|---|
| x | An Rcpp_MCMC object initialized with initializeMCMCObject. |
| what | character defining if log(Posterior) (Default) or log(Likelihood) options are:
LogPosterior or logLikelihood |
| zoom.window | A vector describing the start and end of the zoom window. |
| ... | Arguments to be passed to methods, such as graphical parameters. |

Value

This function has no return value.

plot.Rcpp_ROCModel *Plot Model Object*

Description

Plots traces from the model object such as synthesis rates for each gene. Will work regardless of whether or not expression/synthesis rate levels are being estimated. If you wish to plot observed/empirical values, these values MUST be set using the initial.expression.values parameter found in initializeParameterObject. Otherwise, the expression values plotted will just be SCUO values estimated upon initialization of the Parameter object.

Usage

```
## S3 method for class 'Rcpp_ROCModel'
plot(x, genome = NULL, samples = 100,
      mixture = 1, simulated = FALSE, ...)
```

Arguments

x	An Rcpp model object initialized with <code>initializeModelObject</code> .
genome	An Rcpp genome object initialized with <code>initializeGenomeObject</code> .
samples	The number of samples in the trace
mixture	The mixture for which to graph values.
simulated	A boolean value that determines whether to use the simulated genome.
...	Optional, additional arguments. For this function, a possible title for the plot in the form of a list if set with "main".

Value

This function has no return value.

plot.Rcpp_ROCParameter
Plot Parameter

Description

`plot` graphs the mutation or selection parameter for a ROC or FONSE parameter object for each mixture element.

Usage

```
## S3 method for class 'Rcpp_ROCParameter'
plot(x, what = "Mutation", samples = 100,
      mixture.name = NULL, with.ci = TRUE, ...)
```

Arguments

x	A parameter object
what	Which aspect of the parameter to plot. Default value is "Mutation".
samples	Number of samples to plot using the posterior mean. Default value is 100.
mixture.name	a vector with names/descriptions of the mixture distributions in the parameter object
with.ci	Plot with or without confidence intervals. Default value is TRUE
...	Arguments to be passed to methods, such as graphical parameters.

Details

Graphs are based off the last # samples for the posterior mean.

Value

This function has no return value.

`plot.Rcpp_Trace` *Plot Trace Object*

Description

Plots different traces, specified with the `what` parameter.

Usage

```
## S3 method for class 'Rcpp_Trace'
plot(x, what = c("Mutation", "Selection",
  "MixtureProbability", "Sphi", "Mphi", "Aphi", "Sepsilon", "ExpectedPhi",
  "Expression"), geneIndex = 1, mixture = 1, ...)
```

Arguments

<code>x</code>	An Rcpp trace object initialized with <code>initializeTraceObject</code> .
<code>what</code>	A string containing one of the following to graph: Mutation, Selection, Alpha, LambdaPrime, Mean, MixtureProbability, Sphi, Mphi, Aphi, Spesilon, ExpectedPhi, Expression.
<code>geneIndex</code>	When plotting expression, the index of the gene to be plotted.
<code>mixture</code>	The mixture for which to plot values.
<code>...</code>	Optional, additional arguments. For this function, may be a logical value determining if the trace is ROC-based or not.

Value

This function has no return value.

`plotCodonSpecificParameters`
Plot Codon Specific Parameter

Description

Plots a codon-specific set of traces, specified with the `type` parameter.

Usage

```
plotCodonSpecificParameters(trace, mixture, type = "Mutation",
  main = "Mutation Parameter Traces", ROC = TRUE)
```

Arguments

trace	An Rcpp trace object initialized with <code>initializeTraceObject</code> .
mixture	The mixture for which to plot values.
type	A string containing one of the following to graph: <code>Mutation</code> , <code>Selection</code> , <code>Alpha</code> , <code>LambdaPrime</code> , <code>Mean</code> .
main	The title of the plot.
ROC	A logical value determining if the Parameter was ROC or not.

Value

This function has no return value.

runMCMC*Run MCMC***Description**

`runMCMC` will run a monte carlo markov chain algorithm for the given `mcmc`, `genome`, and `model` objects to perform a model fitting.

Usage

```
runMCMC(mcmc, genome, model, ncores = 1, divergence.iteration = 0)
```

Arguments

<code>mcmc</code>	MCMC object that will run the model fitting algorithm.
<code>genome</code>	Genome that the model fitting will run on. Should be the same genome associated with the parameter and model objects.
<code>model</code>	Model to run the fitting on. Should be associated with the given genome.
<code>ncores</code>	Number of cores to perform the model fitting with. Default value is 1.
<code>divergence.iteration</code>	Number of steps that the initial conditions can diverge from the original conditions given. Default value is 0.

Details

`runMCMC` will run for the number of samples times the number thinning given when the `mcmc` object is initialized. Updates are provided every 100 steps, and the state of the chain is saved every thinning steps.

Value

This function has no return value.

Examples

```
#fitting a model to a genome using the runMCMC function

genome_file <- system.file("extdata", "genome.fasta", package = "AnaCoDa")

genome <- initializeGenomeObject(file = genome_file)
sphi_init <- c(1,1)
numMixtures <- 2
geneAssignment <- c(rep(1,floor(length(genome)/2)),rep(2,ceiling(length(genome)/2)))
parameter <- initializeParameterObject(genome = genome, sphi = sphi_init,
                                         num.mixtures = numMixtures,
                                         gene.assignment = geneAssignment,
                                         mixture.definition = "allUnique")
model <- initializeModelObject(parameter = parameter, model = "ROC")
samples <- 2500
thinning <- 50
adaptiveWidth <- 25
mcmc <- initializeMCMCObject(samples = samples, thinning = thinning,
                               adaptive.width=adaptiveWidth, est.expression=TRUE,
                               est.csp=TRUE, est.hyper=TRUE, est.mix = TRUE)
divergence.iteration <- 10
## Not run:
runMCMC(mcmc = mcmc, genome = genome, model = model,
        ncores = 4, divergence.iteration = divergence.iteration)

## End(Not run)
```

setRestartSettings *Set Restart Settings*

Description

`setRestartSettings` sets the needed information (what the file is called, how often the file should be written) to write information to restart the MCMC algorithm from a given point.

Usage

```
setRestartSettings(mcmc, filename, samples, write.multiple = TRUE)
```

Arguments

<code>mcmc</code>	MCMC object that will run the model fitting algorithm.
<code>filename</code>	Filename for the restart files to be written.
<code>samples</code>	Number of samples that should occur before a file is written.
<code>write.multiple</code>	Boolean that determines if multiple restart files are written. Default value is TRUE.

Details

`setRestartSettings` writes a restart file every set amount of samples that occur. Also, if `write.multiple` is true, instead of overwriting the previous restart file, the sample number is prepended onto the file name and multiple restart files are generated for a run.

Value

This function has no return value.

Examples

```
## set restart settings for checkpointing

samples <- 2500
thinning <- 50
adaptiveWidth <- 25

## estimate all parameter types
mcmc <- initializeMCMCObject(samples = samples, thinning = thinning,
                               adaptive.width=adaptiveWidth, est.expression=TRUE,
                               est.csp=TRUE, est.hyper=TRUE, est.mix = TRUE)

# prompts the mcmc to write a restart file every 100 samples during the run.
setRestartSettings(mcmc = mcmc, filename = "test_restart", samples = 100)

# prompts the mcmc to write a restart file every 100 samples during the run,
# but will overwrite it each time.
setRestartSettings(mcmc = mcmc, filename = "test_restart", samples = 100,
                   write.multiple = FALSE)
```

Description

`summary` summarizes the description of a genome, such as number of genes and average gene length.

Usage

```
## S3 method for class 'Rcpp_Genome'
summary(object, ...)
```

Arguments

- | | |
|---------------------|---|
| <code>object</code> | A genome object initialized with initializeGenomeObject . |
| ... | Optional, additional arguments to be passed to the main <code>summary</code> function that affect the summary produced. |

Value

This function returns by default an object of class c("summaryDefault", "table").

`writeMCMCObject` *Write MCMC Object*

Description

`writeMCMCObject` stores the MCMC information from the model fitting run in a file.

Usage

```
writeMCMCObject(mcmc, file)
```

Arguments

<code>mcmc</code>	MCMC object that has run the model fitting algorithm.
<code>file</code>	A filename where the data will be stored.

Value

This function has no return value.

Examples

```
## saving the MCMC object after model fitting
genome_file <- system.file("extdata", "genome.fasta", package = "AnaCoDa")

genome <- initializeGenomeObject(file = genome_file)
sphi_init <- c(1,1)
numMixtures <- 2
geneAssignment <- c(rep(1,floor(length(genome)/2)),rep(2,ceiling(length(genome)/2)))
parameter <- initializeParameterObject(genome = genome, sphi = sphi_init,
                                         num.mixtures = numMixtures,
                                         gene.assignment = geneAssignment,
                                         mixture.definition = "allUnique")
samples <- 2500
thinning <- 50
adaptiveWidth <- 25
mcmc <- initializeMCMCObject(samples = samples, thinning = thinning,
                               adaptive.width=adaptiveWidth, est.expression=TRUE,
                               est.csp=TRUE, est.hyper=TRUE, est.mix = TRUE)
divergence.iteration <- 10
## Not run:
runMCMC(mcmc = mcmc, genome = genome, model = model,
        ncores = 4, divergence.iteration = divergence.iteration)
writeMCMCObject(mcmc = mcmc, file = file.path(tempdir(), "file.Rda"))
```

```
## End(Not run)
```

writeParameterObject Write Parameter Object to a File

Description

`writeParameterObject` will write the parameter object as binary to the filesystem

Usage

```
writeParameterObject(parameter, file)
```

Arguments

<code>parameter</code>	parameter on object created by <code>initializeParameterObject</code> .
<code>file</code>	A filename that where the data will be stored.

Details

As Rcpp object are not serializable with the default R save function, therefore this custom save function is provided (see [loadParameterObject](#)).

Value

This function has no return value.

Examples

```
## Not run:

genome_file <- system.file("extdata", "genome.fasta", package = "AnaCoDa")

genome <- initializeGenomeObject(file = genome_file)
sphi_init <- c(1,1)
numMixtures <- 2
geneAssignment <- c(rep(1,floor(length(genome)/2)),rep(2,ceiling(length(genome)/2)))
parameter <- initializeParameterObject(genome = genome, sphi = sphi_init,
                                         num.mixtures = numMixtures,
                                         gene.assignment = geneAssignment,
                                         mixture.definition = "allUnique")

## writing an empty parameter object as the runMCMC routine was not called yet
writeParameterObject(parameter = parameter, file = file.path(tempdir(), "file.Rda"))

## End(Not run)
```

Index

AAToCodon, 3, 8
acfCSP, 4, 5
acfMCMC, 4, 4
addObservedSynthesisRateSet, 5
aminoAcids, 6, 8

calculate_marginal_log_likelihood, 7
calculateSCU0, 6
codons, 6, 8
codonToAA, 3, 8
convergence.test, 9

findOptimalCodon, 10

geomMean, 11
getCAI, 12
getCAIweights, 13
getCodonCounts, 14
getCodonCountsForAA, 15
getCSPEstimates, 15
getExpressionEstimates, 17
getMixtureAssignmentEstimate, 18
getNames, 20
getNc, 20
getNCAA, 21
getObservedSynthesisRateSet, 22
getSelectionCoefficients, 22
getTrace, 23

initializeCovarianceMatrices, 24
initializeGenomeObject, 5, 6, 13, 20–23,
 25, 31, 40
initializeMCMCObject, 26
initializeModelObject, 27
initializeParameterObject, 29

length.Rcpp_Genome, 31
loadMCMCObject, 32
loadParameterObject, 32, 42

plot.Rcpp_FONSEModel, 33
plot.Rcpp_FONSEParameter, 34
plot.Rcpp_MCMCAlgorithm, 35
plot.Rcpp_ROCModel, 35
plot.Rcpp_ROCParameter, 36
plot.Rcpp_Trace, 37
plotCodonSpecificParameters, 37

runMCMC, 38

setRestartSettings, 39
summary.Rcpp_Genome, 40

writeMCMCObject, 41
writeParameterObject, 33, 42