

Package ‘cubfits’

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Title Codon Usage Bias Fits

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graphics, utils

Suggests seqinr, VGAM, EMCluster

Enhances pbdMPI (>= 0.3-1)

LazyLoad yes

LazyData yes

Description Estimating mutation and selection coefficients on synonymous codon bias usage based on models of ribosome overhead cost (ROC). Multinomial logistic regression and Markov Chain Monte Carlo are used to estimate and predict protein production rates with/without the presence of expressions and measurement errors. Work flows with examples for simulation, estimation and prediction processes are also provided with parallelization speedup. The whole framework is tested with yeast genome and gene expression data of Yassour, et al. (2009) <[doi:10.1073/pnas.0812841106](https://doi.org/10.1073/pnas.0812841106)>.

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BugReports <https://github.com/snoweye/cubfits/issues>

URL <https://github.com/snoweye/cubfits>

NeedsCompilation yes

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cubfits-package

Codon Bias Usage Fits

Description

Estimating mutation and selection coefficients on synonymous codon bias usage based on models of ribosome overhead cost (ROC). Multinomial logistic regression and Markov Chain Monte Carlo are used to estimate and predict protein production rates with/without the presence of expressions and measurement errors.

Details

Package: cubfits
Type: Package
License: Mozilla Public License 2.0
LazyLoad: yes

The install command is simply as

```
> R CMD INSTALL cubfits_*.tar.gz
```

from a command mode or

```
R> install.packages("cubfits")
```

inside an R session.

Author(s)

Wei-Chen Chen <wccsnow@gmail.com>, Russell Zaretzki, William Howell, Drew Schmidt, and Michael Gilchrist.

References

<https://github.com/snoweye/cubfits/>

See Also

[init.function\(\)](#), [cubfits\(\)](#), [cubpred\(\)](#), and [cubappr\(\)](#).

Examples

```
## Not run:  
suppressMessages(library(cubfits, quietly = TRUE))  
  
demo(roc.train, 'cubfits', ask = F, echo = F)  
demo(roc.pred, 'cubfits', ask = F, echo = F)  
demo(roc.appr, 'cubfits', ask = F, echo = F)  
  
## End(Not run)
```

 Asymmetric Laplace Distribution

The Asymmetric Laplace Distribution

Description

Density, probability, quantile, random number generation, and MLE functions for the asymmetric Laplace distribution with parameters either in $ASL(\theta, \mu, \sigma)$ or the alternative $ASL^*(\theta, \kappa, \sigma)$.

Usage

```

dasl(x, theta = 0, mu = 0, sigma = 1, log = FALSE)
dasla(x, theta = 0, kappa = 1, sigma = 1, log = FALSE)

pasl(q, theta = 0, mu = 0, sigma = 1, lower.tail = TRUE,
      log.p = FALSE)
pasla(q, theta = 0, kappa = 1, sigma = 1, lower.tail = TRUE,
       log.p = FALSE)

qasl(p, theta = 0, mu = 0, sigma = 1, lower.tail = TRUE,
      log.p = FALSE)
qasla(p, theta = 0, kappa = 1, sigma = 1, lower.tail = TRUE,
       log.p = FALSE)

rasl(n, theta = 0, mu = 0, sigma = 1)
rasla(n, theta = 0, kappa = 1, sigma = 1)

asl.optim(x)

```

Arguments

<code>x, q</code>	vector of quantiles.
<code>p</code>	vector of probabilities.
<code>n</code>	number of observations. If <code>length(n) > 1</code> , the length is taken to be the number required.
<code>theta</code>	center parameter.
<code>mu, kappa</code>	location parameters.
<code>sigma</code>	shape parameter.
<code>log, log.p</code>	logical; if TRUE, probabilities p are given as log(p).
<code>lower.tail</code>	logical; if TRUE (default), probabilities are $P[X \leq x]$ otherwise, $P[X > x]$.

Details

The density $f(x)$ of $ASL^*(\theta, \kappa, \sigma)$ is given as $\frac{\sqrt{2}}{\sigma} \frac{\kappa}{1+\kappa^2} \exp(-\frac{\sqrt{2}\kappa}{\sigma}|x-\theta|)$ if $x \geq \theta$, and $\frac{\sqrt{2}}{\sigma} \frac{\kappa}{1+\kappa^2} \exp(-\frac{\sqrt{2}}{\sigma\kappa}|x-\theta|)$ if $x < \theta$.

The parameter domains of ASL and ASL^* are $\theta \in R$, $\sigma > 0$, $\kappa > 0$, and $\mu \in R$. The relation of μ and κ are $\kappa = \frac{\sqrt{2\sigma^2 + \mu^2} - \mu}{\sqrt{2}\sigma}$ or $\mu = \frac{\sigma}{\sqrt{2}}(\frac{1}{\kappa} - \kappa)$.

Value

“dasl” and “dasla” give the densities, “pasl” and “pasla” give the distribution functions, “qasl” and “qasla” give the quantile functions, and “rasl” and “rasla” give the random numbers.

asl.optim returns the MLE of data x including theta, mu, kappa, and sigma.

Author(s)

Wei-Chen Chen <wccsnow@gmail.com>.

References

Kotz S, Kozubowski TJ, Podgorski K. (2001) “The Laplace distribution and generalizations: a revisit with applications to communications, economics, engineering, and finance.” Boston: Birkhauser.

Examples

```
## Not run:
suppressMessages(library(cubfits, quietly = TRUE))
set.seed(1234)

dasl(-2:2)
dasla(-2:2)
pasl(-2:2)
pasla(-2:2)
qasl(seq(0, 1, length = 5))
qasla(seq(0, 1, length = 5))

dasl(-2:2, log = TRUE)
dasla(-2:2, log = TRUE)
pasl(-2:2, log.p = TRUE)
pasla(-2:2, log.p = TRUE)
qasl(log(seq(0, 1, length = 5)), log.p = TRUE)
qasla(log(seq(0, 1, length = 5)), log.p = TRUE)

set.seed(123)
rasl(5)
rasla(5)

asl.optim(rasl(5000))

## End(Not run)
```

Cedric Convergence Utilities

Cedric Convergence Utilities

Description

This utility function provides convergence related functions by Cedric.

Usage

```
cubmultichain(cubmethod, reset.qr, seeds=NULL,
              teston=c("phi", "sphi"), swap=0, swapAt=0.05, monitor=NULL,
              min=0, max=160000, nchains=2, conv.thin=10,
              eps=0.1, ncores=2, ...)
```

```
cubsinglechain(cubmethod, frac1=0.1, frac2=0.5, reset.qr,
               seed=NULL, teston=c("phi", "sphi"), monitor=NULL,
               min=0, max=160000, conv.thin=10, eps=1, ...)
```

Arguments

cubmethod	String to choose method. Options are "cubfits", "cubappr", "cubpred"
reset.qr	recalculate QR decomposition matrix of covariance matrix until reset.qr samples are reached
swap	proportion of b matrix parameters to be swaped between convergence checks
swapAt	difference (L1-norm) between two consecutive convergence test leading to a swap in the b matrix
seeds	Vector of seed for random number generation
seed	Seed for random number generation
teston	Select data to test convergence on
monitor	A function to monitor the progress of the MCMC. The fuctions expects the result object and for cubmultichain an index i. (cubmultichain call: monitor(x,i), cubsinglechain call: monitor(x))
min	Minimum samples to be obtained. eps is ignored until number of samples reaches min
max	Maximum samples to be obtained. eps is ignored after max samples is obtained
eps	Convergence criterium
conv.thin	thinning of samples before performing convergence test
nchains	number of chains to run in parallel
ncores	number of cores to use for parallel execution of chains
frac1	fraction of samples at the beginning of set for Geweke test
frac2	fraction of samples at the end of set for Geweke test
...	named arguments for functions "cubfits", "cubappr" or "cubpred"

Details

under development

Value

under development

Author(s)

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

References

<https://github.com/clandere/cubfits/>

See Also

cubfits, cubappr, cubpred

Examples

```
## Not run:  
suppressMessages(library(cubfits, quietly = TRUE))
```

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

Cedric IO Utilities *Cedric IO Utilities*

Description

This utility function provides basic IO by Cedric.

Usage

```
readGenome(fn.genome, ex.sh.aa = 0, rm.first.aa = 0)  
  
normalizeDataSet(data)
```

Arguments

fn.genome	Fasta file with sequences
ex.sh.aa	Ignore sequences with a length less than ex.sh.aa. (After removal of the first rm.first.aa amino acids)
rm.first.aa	Remove the first rm.first.aa amino acids (after start codon)
data	Vector to be normalized. Means will be set to 1

Details

under development

Value

under development

Author(s)

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

References

<https://github.com/clandere/cubfits/>

See Also

under development

Examples

```
## Not run:
library(cubfits)
seq.string <- readGenome("my_genome.fasta", 150, 10)

## End(Not run)
```

Cedric Plot Utilities *Cedric Plot Utilities*

Description

This utility function provides basic plots by Cedric.

Usage

```
plotPTraces(pMat, ...)

plotExpectedPhiTrace(phiMat, ...)

plotCUB(reu13.df.obs, bMat = NULL, bVec = NULL, phi.bin,
        n.use.samples = 2000, main = "CUB", model.label = c("True Model"),
        model.lty = 1, weightedCenters = TRUE)

plotTraces(bMat, names.aa, param = c("logmu", "deltaeta", "deltat"),
           main = "AA parameter trace")
```


Arguments

<code>reu13.df.obs</code>	under development
<code>bVec</code>	a parameter vector
<code>phi.bin</code>	phi values to bin for comparison
<code>n.use.samples</code>	under development
<code>main</code>	Main name for plotTraces
<code>model.label</code>	Name of model
<code>model.lty</code>	line type for model
<code>weightedCenters</code>	if centers are weighted.
<code>names.aa</code>	List of amino acids used for estimation
<code>param</code>	select to plot parameter trace for either log(μ) values or delta t
<code>phiMat</code>	phi matrix from the output of "cubmultichain", "cubsinglechain", "cubfits", "cubappr", or "cubpred"
<code>bMat</code>	b matrix from the output of "cubmultichain", "cubsinglechain", "cubfits", "cubappr", or "cubpred"
<code>pMat</code>	p matrix from the output of "cubmultichain", "cubsinglechain", "cubfits", "cubappr", or "cubpred"
<code>...</code>	other plotting options

Details

under development

Value

under development

Author(s)

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

References

<https://github.com/clandere/cubfits/>

See Also

plot

Examples

```
## Not run:  
suppressMessages(library(cubfits, quietly = TRUE))  
  
## End(Not run)
```

Codon Adaptation Index*Function for Codon Adaptation Index (CAI)*

Description

Calculate the Codon Adaptation Index (CAI) for each gene. Used as a substitute for expression in cases of without expression measurements.

Usage

```
calc_cai_values(y, y.list, w = NULL)
```

Arguments

<code>y</code>	an object of format <code>y</code> .
<code>y.list</code>	an object of format <code>y.list</code> .
<code>w</code>	a specified relative frequency of synonymous codons.

Details

This function computes CAI for each gene. Typically, this method is completely based on entropy and information theory to estimate expression values of sequences according to their codon information.

If the input `w` is `NULL`, then empirical values are computed.

Value

A list with two named elements `CAI` and `w` are returned where `CAI` are CAI of input sequences (`y` and `y.list`) and `w` are the relative frequency used to computed those CAI's.

Author(s)

Wei-Chen Chen <wccsnow@gmail.com>.

References

Sharp P.M. and Li W.-H. "The codon Adaptation Index – a measure of directional synonymous codon usage bias, and its potential applications" *Nucleic Acids Res.* 15 (3): 1281-1295, 1987.

See Also

[calc_scuo_values\(\)](#), [calc_scu_values\(\)](#).

Examples

```

## Not run:
rm(list = ls())
library(cubfits, quietly = TRUE)

y <- ex.train$y
y.list <- convert.y.to.list(y)
CAI <- calc_cai_values(y, y.list)$CAI
plot(CAI, log10(ex.train$phi.Obs), main = "Expression vs CAI",
      xlab = "CAI", ylab = "Expression (log10)")

### Verify with the seqinr example.
library(seqinr, quietly = TRUE)
inputdatfile <- system.file("sequences/input.dat", package = "seqinr")
input <- read.fasta(file = inputdatfile, forcedNATolower = FALSE)
names(input)[65] <- paste(names(input)[65], ".1", sep = "") # name duplicated.
input <- input[order(names(input))]

### Convert to cubfits format.
seq.string <- convert.seq.data.to.string(input)
new.y <- gen.y(seq.string)
new.y.list <- convert.y.to.list(new.y)
ret <- calc_cai_values(new.y, new.y.list)

### Rebuild w.
w <- rep(1, 64)
names(w) <- codon.low2up(rownames(caitab))
for(i in 1:64){
  id <- which(names(ret$w) == names(w)[i])
  if(length(id) == 1){
    w[i] <- ret$w[id]
  }
}
CAI.res <- sapply(input, seqinr::cai, w = w)

### Plot.
plot(CAI.res, ret$CAI,
      main = "Comparison of seqinR and cubfits results",
      xlab = "CAI from seqinR", ylab = "CAI from cubfits", las = 1)
abline(c(0, 1))

## End(Not run)

```

Description

Default controls of **cubfits** include for models, optimizations, MCMC, plotting, global variables, etc.

Usage

```
.cubfitsEnv
.CF.CT
.CF.CONF
.CF.GV
.CF.DP
.CF.OP
.CF.AC
.CF.PT
.CF.PARAM
.CO.CT
```

Format

All are in lists and contain several controlling options.

Details

See [init.function\(\)](#) for use cases of these objects.

- `.cubfitEnv` is a default environment to dynamically save functions and objects.
- `.CF.CT` is main controls of models. It currently includes

<code>model</code>	main models
<code>type.p</code>	proposal for hyper-parameters
<code>type.Phi</code>	proposal for Phi
<code>model.Phi</code>	prior of Phi
<code>init.Phi</code>	initial methods for Phi
<code>init.fit</code>	how is coefficient proposed
<code>parallel</code>	parallel functions
<code>adaptive</code>	method for adaptive MCMC

- `.CF.CONF` controls the initial and draw scaling. It currently includes

<code>scale.phi.Obs</code>	if phi were scaled to mean 1
<code>init.b.Scale</code>	initial b scale
<code>init.phi.Scale</code>	initial phi scale
<code>p.nclass</code>	number of classes if mixture phi
<code>b.DrawScale</code>	drawing scale for b if random walk
<code>p.DrawScale</code>	drawing scale for p if random walk
<code>phi.DrawScale</code>	random walk scale for phi
<code>phi.pred.DrawScale</code>	random walk scale for phi.pred
<code>sigma.Phi.DrawScale</code>	random walk scale for sigma.Phi
<code>bias.Phi.DrawScale</code>	random walk scale for bias.Phi
<code>estimate.bias.Phi</code>	if estimate bias of phi during MCMC
<code>compute.logL</code>	if compute logL in each iteration

- `.CF.GV` contains global variables for amino acids and codons. It currently includes

amino.acid	amino acids
amino.acid.3	amino acids
synonymous.codon	synonymous codons of amino acids
amino.acid.split	amino acid 'S' is split
amino.acid.split.3	amino acid 'S' is split
synonymous.codon.split	synonymous codons of split amino acid

- .CF.OP controls optimizations. It currently includes

optim.method	method for optim()
stable.min.exp	minimum exponent
stable.max.exp	maximum exponent
E.Phi	expected Phi
lower.optim	lower of derivative of logL(x)
upper.optim	upper of derivative of logL(x)
lower.integrate	lower of integration of L(x)
upper.integrate	upper of integration of L(x)

- .CF.DP is for dumping MCMC iterations. It currently includes

dump	if dumping within MCMC
iter	iterations per dumping
prefix.dump	path and file names of dumping
verbose	if verbose
iterThin	iterations to thin chain
report	iterations to report
report.proc	iterations to report proc.time()

- .CF.AC controls adaptive MCMC. It currently includes

renew.iter	per renewing iterations
target.accept.lower	target acceptant rate lower bound
target.accept.upper	target acceptant rate upper bound
scale.increase	increase scale size
scale.decrease	decrease scale size
sigma.lower	lower bound of relative scale size
sigma.upper	upper bound of relative scale size

- .CF.PT controls the plotting format. It currently includes

color color for codons.

- .CF.PARAM controls the parameters and hyperparameters of priors. It currently includes

phi.meanlog	mean of phi in loca scale
phi.sdlog	standard deviation of phi in loca scale

- .CO.CT controls the constrained optimization function. It currently includes

debug message printing level of debugging.

Author(s)

Wei-Chen Chen <wccsnow@gmail.com>.

References

<https://github.com/snoweye/cubfits/>

See Also

[init.function\(\)](#), [cubfits\(\)](#), [cubpred\(\)](#), [cubappr\(\)](#), and [mixnormerr.optim\(\)](#).

Examples

```
## Not run:
suppressMessages(library(cubfits, quietly = TRUE))

.CF.CT
.CF.CONF
.CF.DP
.CF.GV
.CF.OP
.CF.AC
.CF.PT
.CF.PARAM
.CO.CT

ls(.cubfitsEnv)
init.function()
ls(.cubfitsEnv)

## End(Not run)
```

Coverting Utility

Convert Data Frame to Other Formats

Description

These utility functions convert data of format divided by amino acids into list of format divided by ORFs, or convert data to other formats.

Usage

```
convert.reu13.df.to.list(reu13.df)
convert.y.to.list(y)
convert.n.to.list(n)
```

```
convert.y.to.scuo(y)
convert.seq.data.to.string(seq.data)

codon.low2up(x)
codon.up2low(x)

dna.low2up(x)
dna.up2low(x)

convert.b.to.bVec(b)
convert.bVec.to.b(bVec, aa.names, model = .CF.CT$model[1])
```

Arguments

reu13.df	a list of <code>reu13.df</code> data frames divided by amino acids.
y	a list of <code>y</code> data frames divided by amino acids.
n	a list of <code>n</code> vectors divided by amino acids.
seq.data	a vector of <code>seq.data</code> format.
x	a codon or dna string, such "ACG", "acg", or "A", "a".
b	a <code>b</code> object.
bVec	a <code>bVec</code> object.
aa.names	a vector contains amino acid names for analysis.
model	model fitted.

Details

`convert.reu13.df.to.list()`, `convert.y.to.list()`, and `convert.n.to.list()`: these utility functions take the inputs divided by amino acids and return the outputs divided by ORFs.

`convert.y.scuo()` converts `y` into `scuo` format.

`convert.seq.data.to.string()` converts `seq.data` into `seq.string` format.

`codon.low2up()` and `codon.up2low()` convert codon strings between lower or upper cases.

`convert.bVec.to.b()` and `convert.b.to.bVec()` convert objects `b` and `bVec`.

Value

All functions return the corresponding formats.

Author(s)

Wei-Chen Chen <wccsnow@gmail.com>.

References

<https://github.com/snoweye/cubfits/>

See Also

[AllDataFormats](#), [rearrange.n\(\)](#), [rearrange.reu13.df\(\)](#), [rearrange.y\(\)](#), and [read.seq\(\)](#).

Examples

```
## Not run:
suppressMessages(library(cubfits, quietly = TRUE))

reu13.list <- convert.reu13.df.to.list(ex.train$reu13.df)
y.list <- convert.y.to.list(ex.train$y)
n.list <- convert.n.to.list(ex.train$n)

scuo <- convert.y.to.scuo(ex.train$y)

seq.data <- read.seq(get.expath("seq_200.fasta"))
seq.string <- convert.seq.data.to.string(seq.data)

codon.low2up("acg")
codon.up2low("ACG")

dna.low2up(c("a", "c", "g"))
dna.up2low(c("A", "C", "G"))

## End(Not run)
```

CUB Model Approximation

Codon Usage Bias Approximation for ORFs without Expression

Description

This function provides codon usage bias approximation with observed ORFs but without any expressions.

Usage

```
cubappr(reu13.df.obs, phi.pred.Init, y, n,
        nIter = 1000,
        b.Init = NULL, init.b.Scale = .CF.CONF$init.b.Scale,
        b.DrawScale = .CF.CONF$b.DrawScale,
        b.RInit = NULL,
        p.Init = NULL, p.nclass = .CF.CONF$p.nclass,
        p.DrawScale = .CF.CONF$p.DrawScale,
        phi.pred.DrawScale = .CF.CONF$phi.pred.DrawScale,
        model = .CF.CT$model[1], model.Phi = .CF.CT$model.Phi[1],
        adaptive = .CF.CT$adaptive[1],
        verbose = .CF.DP$verbose,
        iterThin = .CF.DP$iterThin, report = .CF.DP$report)
```


Arguments

<code>reu13.df.obs</code>	a <code>reu13.df</code> object, ORFs information.
<code>phi.pred.Init</code>	a <code>phi.Obs</code> object, temporarily initial of expression without measurement errors.
<code>y</code>	a <code>y</code> object, codon counts.
<code>n</code>	a <code>n</code> object, total codon counts.
<code>nIter</code>	number of iterations after burn-in iterations.
<code>b.Init</code>	initial values for parameters <code>b</code> .
<code>init.b.Scale</code>	for initial <code>b</code> if <code>b.Init = NULL</code> .
<code>b.DrawScale</code>	scaling factor for adaptive MCMC with random walks when drawing new <code>b</code> .
<code>b.RInit</code>	initial values (in a list) for R matrices of parameters <code>b</code> yielding from QR decomposition of <code>vglm()</code> for the variance-covariance matrix of <code>b</code> .
<code>p.Init</code>	initial values for hyper-parameters.
<code>p.nclass</code>	number of components for <code>model.Phi = "logmixture"</code> .
<code>p.DrawScale</code>	scaling factor for adaptive MCMC with random walks when drawing new <code>sigma.Phi</code> .
<code>phi.pred.DrawScale</code>	scaling factor for adaptive MCMC with random walks when drawing new <code>Phi</code> of predicted set.
<code>model</code>	model to be fitted, currently "roc" only.
<code>model.Phi</code>	prior model for <code>Phi</code> , currently "lognormal".
<code>adaptive</code>	adaptive method of MCMC for proposing new <code>b</code> and <code>Phi</code> .
<code>verbose</code>	print iteration messages.
<code>iterThin</code>	thinning iterations.
<code>report</code>	number of iterations to report more information.

Details

Total number of MCMC iterations is `nIter + 1`, but the outputs may be thinned to `nIter / iterThin + 1` iterations.

Temporary result dumping may be controlled by `.CF.DP`.

Value

A list contains three big lists of MCMC traces including: `b.Mat` for mutation and selection coefficients of `b`, `p.Mat` for hyper-parameters, and `phi.Mat` for expected expression values `Phi`. All lists are of length `nIter / iterThin + 1` and each element contains the output of each iteration.

All lists also can be binded as trace matrices, such as via `do.call("rbind", b.Mat)` yielding a matrix of dimension number of iterations by number of parameters. Then, those traces can be analyzed further via other MCMC packages such as **coda**.

Note

Note that `phi.pred.Init` need to be normalized to mean 1.

`p.DrawScale` may cause scaling prior if adaptive MCMC is used, and it can result in non-exits of equilibrium distribution.

Author(s)

Wei-Chen Chen <wccsnow@gmail.com>.

References

<https://github.com/snoweye/cubfits/>

See Also

[DataIO](#), [DataConverting](#), [cubfits\(\)](#) and [cubpred\(\)](#).

Examples

```
## Not run:
suppressMessages(library(cubfits, quietly = TRUE))

demo(roc.appr, 'cubfits', ask = F, echo = F)

## End(Not run)
```

CUB Model Fits

Codon Usage Bias Fits for Observed ORFs and Expression

Description

This function provides codon usage bias fits with observed ORFs and expressions which possibly contains measurement errors.

Usage

```
cubfits(reu13.df.obs, phi.Obs, y, n,
        nIter = 1000,
        b.Init = NULL, init.b.Scale = .CF.CONF$init.b.Scale,
        b.DrawScale = .CF.CONF$b.DrawScale,
        b.RInit = NULL,
        p.Init = NULL, p.nclass = .CF.CONF$p.nclass,
        p.DrawScale = .CF.CONF$p.DrawScale,
        phi.Init = NULL, init.phi.Scale = .CF.CONF$init.phi.Scale,
        phi.DrawScale = .CF.CONF$phi.DrawScale,
        model = .CF.CT$model[1], model.Phi = .CF.CT$model.Phi[1],
        adaptive = .CF.CT$adaptive[1],
        verbose = .CF.DP$verbose,
        iterThin = .CF.DP$iterThin, report = .CF.DP$report)
```

Arguments

<code>reu13.df.obs</code>	a <code>reu13.df</code> object, ORFs information.
<code>phi.Obs</code>	a <code>phi.Obs</code> object, expression with measurement errors.
<code>y</code>	a <code>y</code> object, codon counts.
<code>n</code>	a <code>n</code> object, total codon counts.
<code>nIter</code>	number of iterations after burn-in iterations.
<code>b.Init</code>	initial values for parameters <code>b</code> .
<code>init.b.Scale</code>	for initial <code>b</code> if <code>b.Init = NULL</code> .
<code>b.DrawScale</code>	scaling factor for adaptive MCMC with random walks when drawing new <code>b</code> .
<code>b.RInit</code>	initial values (in a list) for R matrices of parameters <code>b</code> yielding from QR decomposition of <code>vglm()</code> for the variance-covariance matrix of <code>b</code> .
<code>p.Init</code>	initial values for hyper-parameters.
<code>p.nclass</code>	number of components for <code>model.Phi = "logmixture"</code> .
<code>p.DrawScale</code>	scaling factor for adaptive MCMC with random walks when drawing new <code>sigma.Phi</code> .
<code>phi.Init</code>	initial values for <code>Phi</code> .
<code>init.phi.Scale</code>	for initial <code>phi</code> if <code>phi.Init = NULL</code> .
<code>phi.DrawScale</code>	scaling factor for adaptive MCMC with random walks when drawing new <code>Phi</code> .
<code>model</code>	model to be fitted, currently "roc" only.
<code>model.Phi</code>	prior model for <code>Phi</code> , currently "lognormal".
<code>adaptive</code>	adaptive method of MCMC for proposing new <code>b</code> and <code>Phi</code> .
<code>verbose</code>	print iteration messages.
<code>iterThin</code>	thinning iterations.
<code>report</code>	number of iterations to report more information.

Details

This function correctly and carefully implements a combining version of Shah and Gilchrist (2011) and Wallace et al. (2013).

Total number of MCMC iterations is `nIter + 1`, but the outputs may be thinned to `nIter / iterThin + 1` iterations.

Temporary result dumping may be controlled by `.CF.DP`.

Value

A list contains three big lists of MCMC traces including: `b.Mat` for mutation and selection coefficients of `b`, `p.Mat` for hyper-parameters, and `phi.Mat` for expected expression values `Phi`. All lists are of length `nIter / iterThin + 1` and each element contains the output of each iteration.

All lists also can be binded as trace matrices, such as via `do.call("rbind", b.Mat)` yielding a matrix of dimension number of iterations by number of parameters. Then, those traces can be analyzed further via other MCMC packages such as **coda**.

Note

Note that `phi.Init` need to be normalized to mean 1.

`p.DrawScale` may cause scaling prior if adaptive MCMC is used, and it can result in non-exits of equilibrium distribution.

Author(s)

Wei-Chen Chen <wccsnow@gmail.com>.

References

<https://github.com/snoweye/cubfits/>

Shah P. and Gilchrist M.A. “Explaining complex codon usage patterns with selection for translational efficiency, mutation bias, and genetic drift” *Proc Natl Acad Sci USA* (2011) 108:10231–10236.

Wallace E.W.J., Airoidi E.M., and Drummond D.A. “Estimating Selection on Synonymous Codon Usage from Noisy Experimental Data” *Mol Biol Evol* (2013) 30(6):1438–1453.

See Also

[DataIO](#), [DataConverting](#), [cubappr\(\)](#) and [cubpred\(\)](#).

Examples

```
## Not run:
suppressMessages(library(cubfits, quietly = TRUE))

demo(roc.train, 'cubfits', ask = F, echo = F)

## End(Not run)
```

CUB Model Prediction *Codon Usage Bias Prediction for Observed ORFs*

Description

This function provides codon usage bias fits of training set which has observed ORFs and expressions possibly containing measurement errors, and provides predictions of testing set which has other observed ORFs but without expression.

Usage

```

cubpred(reu13.df.obs, phi.Obs, y, n,
        reu13.df.pred, y.pred, n.pred,
        nIter = 1000,
        b.Init = NULL, init.b.Scale = .CF.CONF$init.b.Scale,
        b.DrawScale = .CF.CONF$b.DrawScale,
        b.RInit = NULL,
        p.Init = NULL, p.nclass = .CF.CONF$p.nclass,
        p.DrawScale = .CF.CONF$p.DrawScale,
        phi.Init = NULL, init.phi.Scale = .CF.CONF$init.phi.Scale,
        phi.DrawScale = .CF.CONF$phi.DrawScale,
        phi.pred.Init = NULL,
        phi.pred.DrawScale = .CF.CONF$phi.pred.DrawScale,
        model = .CF.CT$model[1], model.Phi = .CF.CT$model.Phi[1],
        adaptive = .CF.CT$adaptive[1],
        verbose = .CF.DP$verbose,
        iterThin = .CF.DP$iterThin, report = .CF.DP$report)

```

Arguments

<code>reu13.df.obs</code>	a <code>reu13.df</code> to be trained.
<code>phi.Obs</code>	a <code>phi.Obs</code> to be trained.
<code>y</code>	a <code>y</code> to be trained.
<code>n</code>	a <code>n</code> to be trained.
<code>reu13.df.pred</code>	a <code>reu13.df</code> to be predicted.
<code>y.pred</code>	a <code>y</code> to be predicted.
<code>n.pred</code>	a <code>n</code> to be predicted.
<code>nIter</code>	number of iterations after burn-in iterations.
<code>b.Init</code>	initial values for parameters <code>b</code> .
<code>init.b.Scale</code>	for initial <code>b</code> if <code>b.Init = NULL</code> .
<code>b.DrawScale</code>	scaling factor for adaptive MCMC with random walks when drawing new <code>b</code> .
<code>b.RInit</code>	initial values (in a list) for R matrices of parameters <code>b</code> yielding from QR decomposition of <code>vglm()</code> for the variance-covariance matrix of <code>b</code> .
<code>p.Init</code>	initial values for hyper-parameters.
<code>p.nclass</code>	number of components for <code>model.Phi = "logmixture"</code> .
<code>p.DrawScale</code>	scaling factor for adaptive MCMC with random walks when drawing new <code>sigma.Phi</code> .
<code>phi.Init</code>	initial values for <code>Phi</code> .
<code>init.phi.Scale</code>	for initial <code>phi</code> if <code>phi.Init = NULL</code> .
<code>phi.DrawScale</code>	scaling factor for adaptive MCMC with random walks when drawing new <code>Phi</code> .
<code>phi.pred.Init</code>	initial values for <code>Phi</code> of predicted set.
<code>phi.pred.DrawScale</code>	as <code>phi.DrawScale</code> but for predicted set.

<code>model</code>	model to be fitted, currently "roc" only.
<code>model.Phi</code>	prior model for Phi, currently "lognormal".
<code>adaptive</code>	adaptive method of MCMC for proposing new <code>b</code> and Phi.
<code>verbose</code>	print iteration messages.
<code>iterThin</code>	thinning iterations.
<code>report</code>	number of iterations to report more information.

Details

This function correctly and carefully implements an extension of Shah and Gilchrist (2011) and Wallace et al. (2013).

Total number of MCMC iterations is `nIter + 1`, but the outputs may be thinned to `nIter / iterThin + 1` iterations.

Temporary result dumping may be controlled by `.CF.DP`.

Value

A list contains four big lists of MCMC traces including: `b.Mat` for mutation and selection coefficients of `b`, `p.Mat` for hyper-parameters, `phi.Mat` for expected expression values Phi, and `phi.pred.Mat` for predictive expression values Phi. All lists have `nIter / iterThin + 1` elements, and each element contains the output of each iteration.

All lists also can be binded as trace matrices, such as via `do.call("rbind", b.Mat)` yielding a matrix of dimension number of iterations by number of parameters. Then, those traces can be analyzed further via other MCMC packages such as **coda**.

Note

Note that `phi.Init` and `phi.pred.Init` need to be normalized to mean 1.

`p.DrawScale` may cause scaling prior if adaptive MCMC is used, and it can result in non-exits of equilibrium distribution.

Author(s)

Wei-Chen Chen <wccsnow@gmail.com>.

References

<https://github.com/snoweye/cubfits/>

Shah P. and Gilchrist M.A. "Explaining complex codon usage patterns with selection for translational efficiency, mutation bias, and genetic drift" *Proc Natl Acad Sci USA* (2011) 108:10231–10236.

See Also

[DataIO](#), [DataConverting](#), [cubfits\(\)](#) and [cubappr\(\)](#).

Examples

```
## Not run:
suppressMessages(library(cubfits, quietly = TRUE))

demo(roc.pred, 'cubfits', ask = F, echo = F)

## End(Not run)
```

Data Formats

*Data Formats***Description**

Data formats used in **cubfits**.

Format

All are in simple formats as S3 default lists or data frames.

Details

- **Format b:**
A named list *A* contains amino acids. Each element of the list *A*[[*i*]] is a list of elements coefficients (coefficients of log(μ) and Δ .t), *coef.mat* (matrix format of coefficients), and *R* (covariance matrix of coefficients). Note that coefficients and *R* are typically as in the output of *vglm()* of **VGAM** package. Also, *coef.mat* and *R* may miss in some cases. e.g. *A*[[*i*]]\$*coef.mat* is the regression beta matrix of *i*-th amino acid.
- **Format bVec:**
A vector simply contains all coefficients of a *b* object *A*. Note that this is probably only used inside MCMC or the output of *vglm()* of **VGAM** package. e.g. `do.call("c", lapply(A, function(x) x$coefficients))`.
- **Format n:**
A named list *A* contains amino acids. Each element of the list *A*[[*i*]] is a vector containing total codon counts. e.g. *A*[[*i*]][*j*] is for *j*-th ORF of *i*-th amino acid *names(A)*[[*i*]].
- **Format n.list:**
A named list *A* contains ORFs. Each element of the list *A*[[*i*]] is a named list of amino acid containing total count. e.g. *A*[[*i*]][[*j*]] contains total count of *j*-th amino acid in *i*-th ORF.
- **Format phi.df:**
A data frame *A* contains two columns ORF and *phi.value*. e.g. *A*[*i*,] is for *i*-th ORF.
- **Format reu13.df:**
A named list *A* contains amino acids. Each element is a data frame summarizing ORF and expression. The data frame has four to five columns including ORF, *phi* (expression), *Pos* (amino acid position), *Codon* (synonymous codon), and *Codon.id* (synonymous codon id, for

computing only). Note that `Codon.id` may miss in some cases.
 e.g. `A[[i]][17,]` is the 17-th recode of *i*-th amino acid.

- Format `reu13.list`:
 A named list `A` contains ORFs. Each element is a named list `A[[i]]` contains amino acids. Each element of nested list `A[[i]][[j]]` is a position vector of synonymous codon.
 e.g. `A[[i]][[j]][k]` is the *k*-th synonymous codon position of *j*-th amino acid in the *i*-th ORF.
- Format `scuo`:
 A data frame of 8 named columns includes AA (amino acid), ORF, C1, ..., C6 where *C**'s are for codon counts.
- Format `seq.string`:
 Default outputs of `read.fasta()` of **seqinr** package. A named list `A` contains ORFs. Each element of the list is a long string of a ORF.
 e.g. `A[[i]][1]` or `A[[i]]` is the sequence of *i*-th ORF.
- Format `seq.data`:
 Converted from `seq.string` format. A named list `A` contains ORFs. Each element of the list `A[[i]]` is a string vector. Each element of the vector is a codon string.
 e.g. `A[[i]][j]` is *i*-th ORF and *j*-th codon.
- Format `phi.Obs`:
 A named vector `A` of observed expression values and possibly with measurement errors.
 e.g. `A[i]` is the observed ϕ value of *i*-th ORF.
- Format `y`:
 A named list `A` contains amino acids. Each element of the list `A[[i]]` is a matrix where ORFs are in row and synonymous codons are in column. The element of the matrix contains codon counts.
 e.g. `A[[i]][j,k]` is the count for *i*-th amino acid, *j*-th ORF, and *k*-th synonymous codon.
- Format `y.list`:
 A named list `A` contains ORFs. Each element of the list `A[[i]]` is a named list `A[[i]][[j]]` contains amino acids. The element of amino acids list is a codon count vector.
 e.g. `A[[i]][[j]][k]` is the count for *i*-th ORF, *j*-th amino acid, and *k*-th synonymous codon.

Author(s)

Wei-Chen Chen <wccsnow@gmail.com>.

References

<https://github.com/snoweye/cubfits/>

Description

Examples of toy data to test and demonstrate **cubfits**.

Usage

```
b.Init  
ex.test  
ex.train
```

Format

All are in list formats.

Details

`b.Init` contains two sets (`roc` and `rocse`) of initial coefficients including mutation and selection parameters for 3 amino acids 'A', 'C', and 'D' in `matrix` format. Both sets are in `b` format.

`ex.train` contains a training set of 100 sequences including 3 `reu13.df` (codon counts in `reu13` data frame format divided by amino acids), 3 `y` (codon counts in simplified data frame format divided by amino acids), 3 `n` (total amino acid counts in vector format divided by amino acids), and `phi.0bs` (observed phi values in vector format).

`ex.test` contains a testing set of the other 100 sequences in the same format of `ex.train`.

Author(s)

Wei-Chen Chen <wccsnow@gmail.com>.

References

<https://github.com/snoweye/cubfits/>

See Also

`init.function()`, `cubfits()`, `cubpred()`, and `cubappr()`.

Examples

```
## Not run:  
suppressMessages(library(cubfits, quietly = TRUE))  
  
str(b.Init)  
str(ex.test)  
str(ex.train)  
  
## End(Not run)
```

Description

This generic function estimates Phi (expression value) either by posterior mean (PM) or by maximum likelihood estimator (MLE) depending on options set by `init.function()`.

Usage

```
estimatePhi(fitlist, reu13.list, y.list, n.list,
  E.Phi = .CF.OP$E.Phi, lower.optim = .CF.OP$lower.optim,
  upper.optim = .CF.OP$upper.optim,
  lower.integrate = .CF.OP$lower.integrate,
  upper.integrate = .CF.OP$upper.integrate, control = list())
```

Arguments

<code>fitlist</code>	an object of format <code>b</code> .
<code>reu13.list</code>	an object of format <code>reu13.list</code> .
<code>y.list</code>	an object of format <code>y.list</code> .
<code>n.list</code>	an object of format <code>n.list</code> .
<code>E.Phi</code>	potential expected value of Phi.
<code>lower.optim</code>	lower bound to <code>optim()</code> .
<code>upper.optim</code>	upper bound to <code>optim()</code> .
<code>lower.integrate</code>	lower bound to <code>integrate()</code> .
<code>upper.integrate</code>	upper bound to <code>integrate()</code> .
<code>control</code>	control options to <code>optim()</code> .

Details

`estimatePhi()` is a generic function first initialized by `init.function()`, then it estimates Phi accordingly. By default, `.CF.CT$init.Phi` sets the method PM for the posterior mean.

PM uses a flat prior and `integrate()` to estimate Phi. While, MLE uses `optim()` to estimate Phi which may have boundary solutions for some sequences.

Value

Estimated Phi for every sequence is returned.

Author(s)

Wei-Chen Chen <wccsnow@gmail.com>.

References

<https://github.com/snoweye/cubfits/>

See Also

`init.function()` and `fitMultinom()`.

Examples

```
## Not run:
suppressMessages(library(cubfits, quietly = TRUE))
set.seed(1234)

# Convert data.
reu13.list <- convert.reu13.df.to.list(ex.test$reu13.df)
y.list <- convert.y.to.list(ex.test$y)
n.list <- convert.n.to.list(ex.test$n)

# Get phi.pred.Init
init.function(model = "roc")
fitlist <- fitMultinom(ex.train$reu13.df, ex.train$phi.Obs, ex.train$y, ex.train$n)
phi.pred.Init <- estimatePhi(fitlist, reu13.list, y.list, n.list,
                             E.Phi = median(ex.test$phi.Obs),
                             lower.optim = min(ex.test$phi.Obs) * 0.9,
                             upper.optim = max(ex.test$phi.Obs) * 1.1)

## End(Not run)
```

Fit Multinomial

Fit Multinomial Model (Generic)

Description

This generic function estimates **b** (mutation (log(mu))) and selection (Delta.t) parameters) depending on options set by `init.function()`.

Usage

```
fitMultinom(reu13.df, phi, y, n, phi.new = NULL, coefstart = NULL)
```

Arguments

<code>reu13.df</code>	an object of format <code>reu13.df</code> .
<code>phi</code>	an object of format <code>phi.Obs</code> .
<code>y</code>	an object of format <code>y</code> .
<code>n</code>	an object of format <code>n</code> .
<code>phi.new</code>	an object of format <code>phi.Obs</code> for MCMC only.
<code>coefstart</code>	initial value for b (mutation (log(mu))) and selection (Delta.t) parameters) only used in <code>vglm()</code> .

Details

`fitMultinom()` fits a multinomial logistic regression via vector generalized linear model fitting, `vglm()`. By default, for each amino acids, the last codon (order by characters) is assumed as a based line, and other codons are compared to the based line relatively.

In MCMC, `phi.new` are new proposed expression values and used to propose new `b`. The `coef.start` is used to avoid randomization of estimating `b` in `vglm()`, and speed up computation.

Value

A list of format `b` is returned which are modified from the returns of `vglm()`. Mainly, it includes `b$coefficient` (parameters in vector), `b$coef.mat` (parameters in matrix), and `b$R` (covariance matrix of parameters, `*R*` matrix in QR decomposition).

Author(s)

Wei-Chen Chen <wccsnow@gmail.com>.

References

<https://github.com/snoweye/cubfits/>

Shah P. and Gilchrist M.A. “Explaining complex codon usage patterns with selection for translational efficiency, mutation bias, and genetic drift” *Proc Natl Acad Sci USA* (2011) 108:10231–10236.

See Also

[init.function\(\)](#) and [estimatePhi\(\)](#).

Examples

```
## Not run:
suppressMessages(library(cubfits, quietly = TRUE))
set.seed(1234)

# Convert data.
reu13.list <- convert.reu13.df.to.list(ex.test$reu13.df)
y.list <- convert.y.to.list(ex.test$y)
n.list <- convert.n.to.list(ex.test$n)

# Get phi.pred.Init
init.function(model = "roc")
fitlist <- fitMultinom(ex.train$reu13.df, ex.train$phi.Obs, ex.train$y, ex.train$n)
phi.pred.Init <- estimatePhi(fitlist, reu13.list, y.list, n.list,
                           E.Phi = median(ex.test$phi.Obs),
                           lower.optim = min(ex.test$phi.Obs) * 0.9,
                           upper.optim = max(ex.test$phi.Obs) * 1.1)

## End(Not run)
```

Description

These utility functions generate and summarize sequence strings into several useful formats such as [reu13.df](#), [y](#), and [n](#), etc.

Usage

```
gen.reu13.df(seq.string, phi.df = NULL, aa.names = .CF.GV$amino.acid,
             split.S = TRUE, drop.X = TRUE, drop.MW = TRUE,
             drop.1st.codon = TRUE)
gen.y(seq.string, aa.names = .CF.GV$amino.acid,
      split.S = TRUE, drop.X = TRUE, drop.MW = TRUE)
gen.n(seq.string, aa.names = .CF.GV$amino.acid,
      split.S = TRUE, drop.X = TRUE, drop.MW = TRUE)

gen.reu13.list(seq.string, aa.names = .CF.GV$amino.acid,
              split.S = TRUE, drop.X = TRUE, drop.MW = TRUE,
              drop.1st.codon = TRUE)
gen.phi.Obs(phi.df)
gen.scuo(seq.string, aa.names = .CF.GV$amino.acid,
         split.S = TRUE, drop.X = TRUE, drop.MW = TRUE)
```

Arguments

<code>seq.string</code>	a list of sequence strings.
<code>phi.df</code>	a phi.df object returned from read.phi.df() .
<code>aa.names</code>	a vector contains amino acid names for analysis.
<code>split.S</code>	split amino acid 'S' if any.
<code>drop.X</code>	drop amino acid 'X' if any.
<code>drop.MW</code>	drop amino acid 'M' and 'W' if any.
<code>drop.1st.codon</code>	if drop the first codon.

Details

These functions mainly take inputs of sequence strings [seq.string](#) or [phi.df](#) and turn them into corresponding format.

Value

The outputs are data structure in corresponding formats. See [AllDataFormats](#) for details.

Author(s)

Wei-Chen Chen <wccsnow@gmail.com>.

References

<https://github.com/snoweye/cubfits/>

See Also

[AllDataFormats](#), [read.seq\(\)](#), [read.phi.df\(\)](#), and [convert.seq.data.to.string\(\)](#).

Examples

```
## Not run:
suppressMessages(library(cubfits, quietly = TRUE))

seq.data <- read.seq(get.expath("seq_200.fasta"))
phi.df <- read.phi.df(get.expath("phi_200.tsv"))
aa.names <- c("A", "C", "D")

# Read in from FASTA file.
seq.string <- convert.seq.data.to.string(seq.data)
reu13.df <- gen.reu13.df(seq.string, phi.df, aa.names)
reu13.list.new <- gen.reu13.list(seq.string, aa.names)
y <- gen.y(seq.string, aa.names)
n <- gen.n(seq.string, aa.names)
scuo <- gen.scuo(seq.string, aa.names)

# Convert to list format.
reu13.list <- convert.reu13.df.to.list(reu13.df)
y.list <- convert.y.to.list(y)
n.list <- convert.n.to.list(n)

## End(Not run)
```

Initial Generic Functions

Initial Generic Functions of Codon Usage Bias Fits

Description

Initial generic functions for model fitting/approximation/prediction of **cubfits**.

Usage

```
init.function(model = .CF.CT$model[1],
              type.p = .CF.CT$type.p[1],
              type.Phi = .CF.CT$type.Phi[1],
              model.Phi = .CF.CT$model.Phi[1],
```

```

init.Phi = .CF.CT$init.Phi[1],
init.fit = .CF.CT$init.fit[1],
parallel = .CF.CT$parallel[1],
adaptive = .CF.CT$adaptive[1])

```

Arguments

<code>model</code>	main fitted model.
<code>type.p</code>	proposal method for hyper-parameters.
<code>type.Phi</code>	proposal method for Phi (true expression values).
<code>model.Phi</code>	prior of Phi.
<code>init.Phi</code>	initial methods for Phi.
<code>init.fit</code>	how is coefficient initialed in <code>vglm()</code> of VGAM .
<code>parallel</code>	parallel functions.
<code>adaptive</code>	method for adaptive MCMC.

Details

This function mainly takes the options, find the according generic functions, and assign those functions to `.cubfitsEnv`. Those generic functions can be executed accordingly later within functions for MCMC or multinomial logistic regression such as `cubfits()`, `cubappr()`, and `cubpred()`. By default, those options are provided by `.CF.CT` which also leaves rooms for extensions of more complicated models and further optimizations.

It is supposed to call this function before running any MCMC or multinomial logistic regression. This function may affect `cubfits()`, `cubpred()`, `cubappr()`, `estimatePhi()`, and `fitMultinom()`.

- `model` is the main fitting model, currently only `roc` is fully supported.
- `type.p` is for proposing hyper-parameters in Gibb sampler. Currently, `lognormal_fix` is suggested where mean 1 is fixed for log normal distribution. Conjugated prior and flat prior exist and are easily available in this step
- `type.Phi` is for proposing Phi (expression values) in the random walk chain updates. Only, `RW_Norm` is supported. Usually, the acceptance ratio can be adapted within 25% and 50% controlled by `.CF.AC` if `adaptive = simple`.
- `model.Phi` is for the distribution of Phi. Typically, log normal distribution `lognormal` is assumed.
- `init.Phi` is a way to initial Phi. Posterior mean PM is recommended which avoid boundary values.
- `init.fit` is a way of initial coefficients to fit mutation and selection coefficients ($\log \mu$ and Δt or ω) in `vglm()`. Option `current` means the `b` ($\log(\mu)$ and Δt) of current MCMC iteration is the initial values, while `random` means `vglm()` provides the initial values.
- `parallel` is a way of parallel methods to speed up code. `lapply` means `lapply()` is used and no `parallel`; `mclapply` means `mclapply()` of **parallel** is used and good for shared memory machines; `task.pull` means `task.pull()` of **pbDMPI** is used and good for heterogeneous machines; `pbDLapply` means `pbDLapply()` of **pbDMPI** is used and good for homogeneous machines. Among those, `task.pull` is tested thoroughly and is the most reliable and efficient method.

- adaptive is a way for adaptive MCMC that propose better mixing distributions for random walks of Phi. The simple method is suggested and only the proposal distribution of Phi (`type.Phi = RW_Norm`) is adjusted gradually.

Value

Return an invisible object which is a list contain all generic functions according to the input options. All functions are also assigned in the `.cubfitsEnv` for later evaluations called by MCMC or multinomial logistic regression.

Note

Note that all options are taken default values from the global control object `.CF.CT`, so one can utilize/alter the object's values to adjust those affected functions.

Note that `phi.Obs` should be scaled to mean 1 before applying to MCMC.

Author(s)

Wei-Chen Chen <wccsnow@gmail.com>.

References

<https://github.com/snoweye/cubfits/>

See Also

`.CF.CT`, `.CF.CT`, `cubfits()`, `cubpred()`, and `cubappr()`.

Examples

```
## Not run:
suppressMessages(library(cubfits, quietly = TRUE))
set.seed(1234)

# Convert data.
reu13.list <- convert.reu13.df.to.list(ex.test$reu13.df)
y.list <- convert.y.to.list(ex.test$y)
n.list <- convert.n.to.list(ex.test$n)

# Get phi.pred.Init
init.function(model = "roc")
fitlist <- fitMultinom(ex.train$reu13.df, ex.train$phi.Obs, ex.train$y,
                      ex.train$n)
phi.pred.Init <- estimatePhi(fitlist, reu13.list, y.list, n.list,
                            E.Phi = median(ex.test$phi.Obs),
                            lower.optim = min(ex.test$phi.Obs) * 0.9,
                            upper.optim = max(ex.test$phi.Obs) * 1.1)

## End(Not run)
```

Input and Output Utility*Input and Output Utility*

Description

These utility functions read and write data of FASTA and phi.df formats.

Usage

```
read.seq(file.name, forcedDNAtolower = FALSE, convertDNAtoupper = TRUE)
write.seq(seq.data, file.name)

read.phi.df(file.name, header = TRUE, sep = "\t", quote = "")
write.phi.df(phi.df, file.name)

get.expath(file.name, path.root = "./ex_data/", pkg = "cubfits")
```

Arguments

file.name	a file name to read or write.
forcedDNAtolower	an option passed to read.fasta() of seqinr package.
convertDNAtoupper	force everything in upper case.
header	an option passed to read.table().
sep	an option passed to read.table().
quote	an option passed to read.table().
seq.data	a seq.data object.
phi.df	a phi.df object.
path.root	root path for the file name relatively to the pkg.
pkg	package name for the path of root.

Details

read.seq() and write.seq() typically read and write FASTA files (DNA ORFs or sequences).

read.phi.df() and write.phi.df() typically read and write phi.df files (expression values of ORFs or sequences).

get.expath() is only for demonstration returning a full path to the file.

Value

read.seq() returns an object of [seq.data](#) format which can be converted to [seq.string](#) format later via [convert.seq.data.to.string\(\)](#).

read.phi.df() returns an object of [phi.df](#) format which contains expression values.

Author(s)

Wei-Chen Chen <wccsnow@gmail.com>.

References

<https://github.com/snoweye/cubfits/>

See Also

[convert.seq.data.to.string\(\)](#).

Examples

```
## Not run:
suppressMessages(library(cubfits, quietly = TRUE))

seq.data <- read.seq(get.expath("seq_200.fasta"))
phi.df <- read.phi.df(get.expath("phi_200.tsv"))
aa.names <- c("A", "C", "D")

# Read in from FASTA file.
seq.string <- convert.seq.data.to.string(seq.data)

## End(Not run)
```

Mixed Normal Optimization

Mixed Normal Optimization

Description

Constrained optimization for mixed normal in 1D and typically for 2 components.

Usage

```
mixnormerr.optim(X, K = 2, param = NULL)
dmixnormerr(x, param)
```

Arguments

X	a gene expression data matrix of dimension $N \times R$ which has N genes and R replicates.
K	number of components to fit.
x	vector of quantiles.
param	parameters of <code>mixnormerr</code> , typically the element <code>param</code> of the <code>mixnormerr.optim()</code> returning object.

Details

The function `mixnormerr.optim()` maximizes likelihood using `constrOptim()` based on the gene expression data X (usually in log scale) for N genes and R replicates (NA is allowed). The likelihood of each gene expression is a $K = 2$ component mixed normal distribution ($\sum_k p_k N(\mu_k, \sigma_k^2 + \sigma_e^2)$) with measurement errors of the replicates ($N(0, \sigma_e^2)$).

The `sigma_k^2` is as the error of random component and the `sigma_e^2` is as the error of fixed component. Both are within a mixture model of two normal distributions.

The function `dmixnormerr()` computes the density of the mixed normal distribution.

`param` is a parameter list and contains five elements: `K` for number of components, `prop` for proportions, `mu` for centers of components, `sigma2` for variance of components, and `sigma2.e` for variance of measurement errors.

Value

`mixnormerr.optim()` returns a list containing three main elements `param` is the final results (MLEs), `param.start` is the starting parameters, and `optim.ret` is the original returns of `constrOptim()`.

Note

This function is limited for small K . An equivalent EM algorithm should be done in a more stable way for large K .

Author(s)

Wei-Chen Chen <wccsnow@gmail.com>.

References

<https://github.com/snoweye/cubfits/>

See Also

`print.mixnormerr()`, `simu.mixnormerr()`.

Examples

```
## Not run:
suppressMessages(library(cubfits, quietly = TRUE))

### Get individual of phi.Obs.
GM <- apply(yassour[, -1], 1, function(x) exp(mean(log(x[x != 0]))))
phi.Obs.all <- yassour[, -1] / sum(GM) * 15000
phi.Obs.all[phi.Obs.all == 0] <- NA

### Run optimization.
X <- log(as.matrix(phi.Obs.all))
param.init <- list(K = 2, prop = c(0.95, 0.05), mu = c(-0.59, 3.11),
                  sigma2 = c(1.40, 0.59), sigma2.e = 0.03)
ret <- mixnormerr.optim(X, K = 2, param = param.init)
```

```
print(ret)

## End(Not run)
```

 Plotbin

Plot Binning Results

Description

Plot binning results to visualize the effects of mutation and selection along with expression levels empirically.

Usage

```
prop.bin.roc(reu13.df, phi.Obs = NULL, nclass = 20, bin.class = NULL,
            weightedCenters = TRUE, logBins = FALSE)
```

```
plotbin(ret.bin, ret.model = NULL, main = NULL,
        xlab = "Production Rate (log10)", ylab = "Proportion",
        xlim = NULL, lty = 1, x.log10 = TRUE, stderr = FALSE, ...)
```

Arguments

reu13.df	a reu13.df object.
phi.Obs	a phi.Obs object.
nclass	number of binning classes across the range of <code>phi.Obs</code> .
bin.class	binning proportion, e.g. <code>c(0, seq(0.05, 0.95, length = nclass), 1)</code> .
ret.bin	binning results from prop.bin.roc() .
weightedCenters	if centers are weighted.
logBins	if use log scale for bin.
ret.model	model results from prop.model.roc() .
main	an option passed to <code>plot()</code> .
xlab	an option passed to <code>plot()</code> .
ylab	an option passed to <code>plot()</code> .
xlim	range of X-axis.
lty	line type if <code>ret.model</code> is provided.
x.log10	<code>log10()</code> transformation of X-axis.
stderr	plot stand error instead of stand deviation.
...	options passed to <code>plot()</code> .

Details

The function `plotbin()` plots the binning results `ret.bin` returned from `prop.bin.roc()`. Fitted curves may be added if `ret.model` is provided which can be obtained from `prop.model.roc()`.

`plotaddmodel()` can append model later if `ret.model` is not provided to `plotbin()`.

Currently, only ROC model is supported. Colors are controlled by `.CF.PT`.

Value

A binning plot is drawn.

Author(s)

Wei-Chen Chen <wccsnow@gmail.com>.

References

<https://github.com/snoweye/cubfits/>

See Also

`plotmodel()` and `prop.model.roc()`.

Examples

```
## Not run:
demo(plotbin, 'cubfits', ask = F, echo = F)

## End(Not run)
```

Plotmodel

Plot Fitted Models

Description

Plot model results to visualize the effects of mutation and selection along with expression levels. The model can be fitted by MCMC or multinomial logistic regression.

Usage

```
prop.model.roc(b.Init, phi.Obs.lim = c(0.01, 10), phi.Obs.scale = 1,
              nclass = 40, x.log10 = TRUE)
```

```
plotmodel(ret.model, main = NULL,
          xlab = "Production Rate (log10)", ylab = "Proportion",
          xlim = NULL, lty = 1, x.log10 = TRUE, ...)
```

```
plotaddmodel(ret.model, lty, u.codon = NULL, color = NULL,
             x.log10 = TRUE)
```

Arguments

<code>b.Init</code>	a <code>b</code> object.
<code>phi.Obs.lim</code>	range of <code>phi.Obs</code> .
<code>phi.Obs.scale</code>	optional scaling factor.
<code>nclass</code>	number of binning classes across the range of <code>phi.Obs</code> .
<code>x.log10</code>	<code>log10()</code> transformation of X-axis.
<code>ret.model</code>	model results from <code>prop.model.roc()</code> .
<code>main</code>	an option passed to <code>plot()</code> .
<code>xlab</code>	an option passed to <code>plot()</code> .
<code>ylab</code>	an option passed to <code>plot()</code> .
<code>xlim</code>	range of X-axis.
<code>lty</code>	line type.
<code>u.codon</code>	unique synonymous codon names.
<code>color</code>	a color vector for unique codon, typically returns of the internal function <code>get.color()</code> .
<code>...</code>	options passed to <code>plot()</code> .

Details

The function `plotmodel()` plots the fitted curves obtained from `prop.model.roc()`.

The function `plotaddmodel()` can append model curves to a binning plot provided unique synonymous codons and colors are given. This function is nearly for an internal call within `plotmodel()`, but is exported and useful for workflow.

Currently, only ROC model is supported. Colors are controlled by `.CF.PT`.

Value

A fitted curve plot is drawn.

Author(s)

Wei-Chen Chen <wccsnow@gmail.com>.

References

<https://github.com/snoweye/cubfits/>

See Also

`plotbin()`, `prop.bin.roc()`, and `prop.model.roc()`.

Examples

```
## Not run:
demo(plotbin, 'cubfits', ask = F, echo = F)

## End(Not run)
```

Description

This utility function provides a basic plot of production rates.

Usage

```
plotprxy(x, y, x.ci = NULL, y.ci = NULL,
         log10.x = TRUE, log10.y = TRUE,
         add.lm = TRUE, add.one.to.one = TRUE, weights = NULL,
         add.legend = TRUE,
         xlim = NULL, ylim = NULL,
         xlab = "Predicted Production Rate (log10)",
         ylab = "Observed Production Rate (log10)",
         main = NULL)
```

Arguments

x	expression values.
y	expression values, of the same length of x.
x.ci	confidence interval of x, of dimension length{x} * 2, for outliers labeling.
y.ci	confidence interval of y, of dimension length{y} * 2, for outliers labeling.
log10.x	log10() and mean transformation of x axis.
log10.y	log10() and mean transformation of y axis.
add.lm	if add lm() fit.
add.one.to.one	if add one-to-one line.
weights	weights to lm().
add.legend	if add default legend.
xlim	limits of x-axis.
ylim	limits of y-axis.
xlab	an option passed to plot().
ylab	an option passed to plot().
main	an option passed to plot().

Details

As the usual X-Y plot where x and y are expression values.

If add.lm = TRUE and weights are given, then both ordinary and weighted least squares results will be plotted.

Value

A scatter plot with a fitted `lm()` line and R squared value.

Author(s)

Wei-Chen Chen <wccsnow@gmail.com>.

References

<https://github.com/snoweye/cubfits/>

See Also

[plotbin\(\)](#) and [plotmodel\(\)](#).

Examples

```
## Not run:
suppressMessages(library(cubfits, quietly = TRUE))

y.scuo <- convert.y.to.scuo(ex.train$y)
SCU0 <- calc_scuo_values(y.scuo)$SCU0
plotprxy(ex.train$phi.obs, SCU0)

## End(Not run)
```

Posterior Results of Yassour2009

Posterior Results of Yassour 2009 Yeast Experiment Dataset

Description

Output summarized from MCMC posterior results analyzing Yassour 2009 data.

Usage

```
yassour.PM.fits
yassour.PM.appr
yassour.info
```

Format

These are list's containing several posterior means: `E.Phi` for expected expression, `b.InitList.roc` for parameters, `AA.prob` for proportion of amino acids, `sigmaW` for standard error of measure errors, and `gene.length` for gene length.

Details

`yassour.PM.fits` and `yassour.PM.appr` are the MCMC output of with/without observed expression, respectively. Both contain posterior means of expected expressions and coefficient parameters: `E.Phi` and `b.InitList.roc` are scaled results such that each MCMC iteration has mean 1 at `E.Phi`. `yassour.info` contains sequences information (Yeast): `AA.prob` and `gene.length` are summarized from corresponding genes in the analysis.

Note that some of genes may not have good quality of expression or sequence information, so those genes are dropped from `yassour` dataset.

References

<https://github.com/snoweye/cubfits/>

See Also

[yassour](#)

Examples

```
## Not run:
str(yassour.PM.fits)
str(yassour.PM.appr)
str(yassour.PM.info)

## End(Not run)
```

Print

Functions for Printing Objects According to Classes

Description

A Class `mixnormerr` is declared in `cubfits`, and this is the function to print and summary objects.

Usage

```
## S3 method for class 'mixnormerr'
print(x, digits = max(4, getOption("digits") - 3), ...)
```

Arguments

<code>x</code>	an object with the class attributes.
<code>digits</code>	for printing out numbers.
<code>...</code>	other possible options.

Details

This is an useful function for summarizing and debugging.

Value

The results will cat or print on the STDOUT by default.

Author(s)

Wei-Chen Chen <wccsnow@gmail.com>.

References

<https://github.com/snoweye/cubfits/>

See Also

[mixnormerr.optim\(\)](#).

Examples

```
## Not run:
suppressMessages(library(cubfits, quietly = TRUE))

### Get individual of phi.Obs.
GM <- apply(yassour[, -1], 1, function(x) exp(mean(log(x[x != 0]))))
phi.Obs.all <- yassour[, -1] / sum(GM) * 15000
phi.Obs.all[phi.Obs.all == 0] <- NA

### Run optimization.
X <- log(as.matrix(phi.Obs.all))
param.init <- list(K = 2, prop = c(0.95, 0.05), mu = c(-0.59, 3.11),
                  sigma2 = c(1.40, 0.59), sigma2.e = 0.03)
ret <- mixnormerr.optim(X, K = 2, param = param.init)
print(ret)

## End(Not run)
```

Randomize SCUO Index *Generate Randomized SCUO Index*

Description

Generate randomized SCUO indices in log normal distribution, but provided original unchanged SCUO order.

Usage

```
scuo.random(SCUO, phi.Obs = NULL, meanlog = .CF.PARAM$phi.meanlog,
            sdlog = .CF.PARAM$phi.sdlog)
```

Arguments

SCUO	SCUO index returned from <code>calc_scuo_values()</code> .
<code>phi.Obs</code>	optional object of format <code>phi.Obs</code> .
<code>meanlog</code>	mean of log normal distribution.
<code>sdlog</code>	std of log normal distribution.

Details

This function takes SCUO indices (outputs of `calc_scuo_values()`) computes the rank of them, generates log normal random variables, and replaces SCUO indices by those variables in the same rank orders. Typically, these random variables are used to replace expression values when either no expression is observed or for the purpose of model validation.

If `phi.Obs` is provided, the mean and std of $\log(\text{phi.Obs})$ are used for log normal random variables. Otherwise, `meanlog` and `sdlog` are used.

The default `meanlog` and `sdlog` was estimated from `yassour` dataset.

Value

A vector of log normal random variables is returned.

Author(s)

Wei-Chen Chen <wccsnow@gmail.com>.

References

<https://github.com/snoweye/cubfits/>

See Also

`calc_scuo_values()`, `yassour`.

Examples

```
## Not run:
suppressMessages(library(cubfits, quietly = TRUE))

### example dataset.
y.scuo <- convert.y.to.scuo(ex.train$y)
SCUO <- calc_scuo_values(y.scuo)$SCUO
plotprxy(ex.train$phi.Obs, SCUO)

### yassour dataset.
GM <- apply(yassour[, -1], 1, function(x) exp(mean(log(x[x != 0]))))
phi.Obs <- GM / sum(GM) * 15000
mean(log(phi.Obs))
sd(log(phi.Obs))
ret <- scuo.random(SCUO, meanlog = -0.441473, sdlog = 1.393285)
plotprxy(ret, SCUO)
```

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

Rearrangment Utility *Rearrange Data Structure by ORF Names*

Description

These utility functions rearrange data in the order of ORF names.

Usage

```
rearrange.reu13.df(reu13.df)
rearrange.y(y)
rearrange.n(n)
rearrange.phi.Obs(phi.Obs)
```

Arguments

<code>reu13.df</code>	a list of <code>reu13.df</code> data frames divided by amino acids.
<code>y</code>	a list of <code>y</code> data frames divided by amino acids.
<code>n</code>	a list of <code>n</code> vectors divided by amino acids.
<code>phi.Obs</code>	a vector of <code>phi.Obs</code> format.

Details

These utility functions take inputs and return ordered outputs. It is necessary to rearrange data in a right order of ORF names which avoids subsetting data frame within MCMC and improve performance.

Value

The outputs are in the same format of inputs except the order of data is sorted by ORF names.

Author(s)

Wei-Chen Chen <wccsnow@gmail.com>.

References

<https://github.com/snoweye/cubfits/>

See Also

[AllDataFormats](#), [convert.n.to.list\(\)](#), [convert.reu13.df.to.list\(\)](#), and [convert.y.to.list\(\)](#).

Examples

```
## Not run:
suppressMessages(library(cubfits, quietly = TRUE))

reu13.df <- rearrange.reu13.df(ex.train$reu13.df)
y <- rearrange.y(ex.train$y)
n <- rearrange.n(ex.train$n)
phi.Obs <- rearrange.phi.Obs(ex.train$phi.Obs)

## End(Not run)
```

SCUO Index

Function for Synonymous Codon Usage Order (SCUO) Index

Description

Calculate the Synonymous Codon Usage Order (SCUO) index for each gene. Used as a substitute for expression in cases of without expression measurements.

Usage

```
calc_scuo_values(codon.counts)
```

Arguments

codon.counts an object of format `scuo`.

Details

This function computes SCUO index for each gene. Typically, this method is completely based on entropy and information theory to estimate expression values of sequences according to their codon information.

Value

SCUO indices are returned.

Author(s)

Drew Schmidt.

References

<https://www.tandfonline.com/doi/abs/10.1080/03081070500502967>

Wan X.-F., Zhou J., Xu D. “CodonO: a new informatics method for measuring synonymous codon usage bias within and across genomes” International Journal of General Systems Vol. 35, Iss. 1, 2006.

See Also

[scuo.random\(\)](#), [calc_cai_values\(\)](#), [calc_scu_values\(\)](#).

Examples

```
## Not run:
suppressMessages(library(cubfits, quietly = TRUE))

y.scuo <- convert.y.to.scuo(ex.train$y)
SCUO <- calc_scuo_values(y.scuo)$SCUO
plotprxy(ex.train$phi.Obs, SCUO, ylab = "SCUO (log10)")

## End(Not run)
```

Selection on Codon Usage

Function for Selection on Codon Usage (SCU)

Description

Calculate the average translational selection per transcript include mSCU and SCU (if gene expression is provided) for each gene.

Usage

```
calc_scu_values(b, y.list, phi.Obs = NULL)
```

Arguments

<code>b</code>	an object of format b .
<code>y.list</code>	an object of format y.list .
<code>phi.Obs</code>	an object of format phi.Obs , for SCU only.

Details

This function computes SCU and mSCU for each gene. Typically, this method is completely based on estimated parameters of mutation and selection such as outputs of MCMC or [fitMultinom\(\)](#).

Value

A list with two named elements SCU and mSCU are returned.

Author(s)

Wei-Chen Chen <wccsnow@gmail.com>.

References

Wallace E.W.J., Airoidi E.M., and Drummond D.A. “Estimating Selection on Synonymous Codon Usage from Noisy Experimental Data” *Mol Biol Evol* (2013) 30(6):1438–1453.

See Also

[calc_scuo_values\(\)](#), [calc_cai_values\(\)](#).

Examples

```
## Not run:
library(cubfits, quietly = TRUE)

b <- b.Init$roc
phi.Obs <- ex.train$phi.Obs
y <- ex.train$y
y.list <- convert.y.to.list(y)
mSCU <- calc_scu_values(b, y.list, phi.Obs)$mSCU
plot(mSCU, log10(phi.Obs), main = "Expression vs mSCU",
     xlab = "mSCU", ylab = "Expression (log10)")

### Compare with CAI with weights seqinr::cubtab$sc.
library(seqinr, quietly = TRUE)
w <- caitab$sc
names(w) <- codon.low2up(rownames(caitab))
CAI <- calc_cai_values(y, y.list, w = w)$CAI

plot(mSCU, CAI, main = "CAI vs mSCU",
     xlab = "mSCU", ylab = "CAI")

## End(Not run)
```

Description

These utility functions generate data for simulation studies including fake ORFs and expression values.

Usage

```
simu.orf(n, b.Init, phi.Obs = NULL, AA.prob = NULL, orf.length = NULL,
         orf.names = NULL, model = .CF.CT$model)
simu.phi.Obs(Phi, sigmaW.lim = 1, bias.Phi = 0)
simu.mixnormerr(n, param)
```

Arguments

<code>n</code>	number of ORFs or sequences.
<code>b.Init</code>	parameters of mutation and selection of format <code>b</code> .
<code>phi.Obs</code>	an object of format <code>phi.Obs</code> .
<code>AA.prob</code>	proportion of amino acids.
<code>orf.length</code>	lengths of ORFs.
<code>orf.names</code>	names of ORFs.
<code>model</code>	model to be simulated.
<code>Phi</code>	expression values (potentially true expression).
<code>sigmaW.lim</code>	std of measurement errors (between <code>Phi</code> and <code>phi.Obs</code>).
<code>bias.Phi</code>	bias (in log scale) for observed <code>phi</code> .
<code>param</code>	as in <code>dmixnormerr()</code>

Details

`simu.orf()` generates ORFs or sequences based on the `b.Init` and `phi.Obs`.

If `phi.Obs` is omitted, then standard log normal random variables are instead).

If `AA.prob` is omitted, then uniform proportion is assigned.

If `orf.length` is omitted, then 10 to 20 codons are randomly assigned.

If `orf.names` is omitted, then "ORF1" to "ORFn" are assigned.

`simu.phi.Obs()` generates `phi.Obs` by adding normal random errors to `Phi`, and errors have mean 0 and standard deviation `sigmaW.lim`.

`simu.mixnormerr()` generates `Phi` according to the `param`, and adds normal random errors to `Phi`.

Value

`simu.orf()` returns a list of format `seq.data`.

`simu.phi.Obs()` returns a vector of format `phi.Obs`.

`simu.mixnormerr()` returns a list contains three vectors of length `n`: one for expected gene expression `Phi`, one for observed gene expression `phi.Obs`, and one for the component id `id.K`.

Author(s)

Wei-Chen Chen <wccsnow@gmail.com>.

References

<https://github.com/snoweye/cubfits/>

See Also

`read.seq()`, `read.phi.df()`, `write.seq()`, `write.phi.df()`, and `mixnormerr.optim()`.

Examples

```
## Not run:
suppressMessages(library(cubfits, quietly = TRUE))
set.seed(1234)

# Generate sequences.
da.roc <- simu.orf(length(ex.train$phi.Obs), b.Init$roc,
                   phi.Obs = ex.train$phi.Obs, model = "roc")
names(da.roc) <- names(ex.train$phi.Obs)
write.fasta(da.roc, names(da.roc), "toy_roc.fasta")

## End(Not run)
```

Yassour2009

Yassour 2009 Yeast Experiment Dataset

Description

Experiments and data are obtained from Yassour et. al. (2009).

Usage

```
yassour
```

Format

A data.frame contains 6303 rows and 5 columns: ORF is for gene names in character, and YPD0.1, YPD0.2, YPD15.1, and YPD15.2 are gene expressions in positive double corresponding to 4 controlled Yeast experiments.

Details

The original data are available as the URL of the section of Source next. As the section of Examples next, data are selected from SD3.xls and reordered by ORF.

For further analysis, the Examples section also provides how to convert them to phi.Obs values either in geometric means or individually.

Source

<https://www.pnas.org/content/early/2009/02/10/0812841106>

https://www.pnas.org/highwire/filestream/598612/field_highwire_adjunct_files/3/SD3.xls

Yassour M, Kaplan T, Fraser HB, Levin JZ, Pfiffner J, Adiconis X, Schroth G, Luo S, Khrebtukova I, Gnirke A, Nusbaum C, Thompson DA, Friedman N, Regev A. (2009) "Ab initio construction of a eukaryotic transcriptome by massively parallel mRNA sequencing." Proc Natl Acad Sci USA 106(9):3264-9. [PMID:19208812]

References

Wallace E.W.J., Airoidi E.M., and Drummond D.A. “Estimating Selection on Synonymous Codon Usage from Noisy Experimental Data” *Mol Biol Evol* (2013) 30(6):1438–1453.

Examples

```
## Not run:
### SD3.xls is available from the URL provided in the References.
da <- read.table("SD3.xls", header = TRUE, sep = "\t", quote = "",
                stringsAsFactors = FALSE)

### Select ORF, YPD0.1, YPD0.2, YPD15.1, YPD15.2.
da <- da[, c(1, 8, 9, 10, 11)]
colnames(da) <- c("ORF", "YPD0.1", "YPD0.2", "YPD15.1", "YPD15.2")

### Drop inappropriate values (NaN, NA, Inf, -Inf, and 0).
tmp <- da[, 2:5]
id.tmp <- rowSums(is.finite(as.matrix(tmp)) & tmp != 0) >= 3
tmp <- da[id.tmp, 1:5]
yassour <- tmp[order(tmp$ORF),] # cubfits::yassour

### Get geometric mean of phi.Obs and scaling similar to Wallace (2013).
GM <- apply(yassour[, -1], 1, function(x) exp(mean(log(x[x != 0]))))
phi.Obs <- GM / sum(GM) * 15000

### Get individual of phi.Obs.
GM <- apply(yassour[, -1], 1, function(x) exp(mean(log(x[x != 0]))))
phi.Obs.all <- yassour[, -1] / sum(GM) * 15000
phi.Obs.all[phi.Obs.all == 0] <- NA

## End(Not run)
```

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