

Package ‘skipTrack’

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Title A Bayesian Hierarchical Model that Controls for Non-Adherence in Mobile Menstrual Cycle Tracking

Version 0.1.0

Description Implements a Bayesian hierarchical model designed to identify skips in mobile menstrual cycle self-tracking on mobile apps. Future developments will allow for the inclusion of covariates affecting cycle mean and regularity, as well as extra information regarding tracking non-adherence. Main methods to be outlined in a forthcoming paper, with alternative models from Li et al. (2022) <[doi:10.1093/jamia/ocab182](https://doi.org/10.1093/jamia/ocab182)>.

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URL <https://github.com/LukeDuttweiler/skipTrack>

BugReports <https://github.com/LukeDuttweiler/skipTrack/issues>

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gibbsStepLi

*Gibbs Step Li - One MCMC step for the Li Model***Description**

Gibbs Step Li - One MCMC step for the Li Model

Usage

gibbsStepLi(ijDat, iDat, kappa, gamma, alpha, beta, S, indFirst)

Arguments

<code>ijDat</code>	A data.frame with parameters at the individual-observation level: Individual, ys, lambdais, piis, ss.
<code>iDat</code>	A data.frame with parameters at the individual level: Individual, lambdas, pis.
<code>kappa</code>	Fixed value of hyperparameter kappa.
<code>gamma</code>	Fixed value of hyperparameter gamma.
<code>alpha</code>	Fixed value of hyperparameter alpha.
<code>beta</code>	Fixed value of hyperparamter beta.
<code>S</code>	Fixed input value S.
<code>indFirst</code>	A logical vector indicating the first occurrence of each individual.

Value

A list containing one MCMC draws for each parameter. Elements are:

ijDat A data.frame with updated parameters at the individual-observation level: Individual, ys, lambdais, piis, ss.

iDat A data.frame with updated parameters at the individual level: Individual, lambdas, pis.

kappa Fixed value of hyperparameter kappa.

gamma Fixed value of hyperparameter gamma.

alpha Fixed value of hyperparameter alpha.

beta Fixed value of hyperparamter beta.

S Fixed input value S.

indFirst A logical vector indicating the first occurrence of each individual.

References

Li, Kathy, et al. "A predictive model for next cycle start date that accounts for adherence in menstrual self-tracking." *Journal of the American Medical Informatics Association* 29.1 (2022): 3-11.

liInference

Perform hyperparameter inference assuming the model given in Li et al. (2022) on a cycle length dataset.

Description

This function performs hyperparameter inference on a given dataset of individuals and their tracked cycles, assuming the model specified in Li et al. (2022). Default starting values for hyperparameters and optimization tuning parameters are those given in Li et al.

Usage

```
liInference(
  Y,
  cluster,
  S = 10,
  startingParams = c(kappa = 180, gamma = 6, alpha = 2, beta = 20)
)
```

Arguments

Y A vector of observed cycle lengths.

cluster A vector indicating the individual cluster/group membership for each observation Y.

S Maximum number of possible skipped cycles (see Li et al. for details).

startingParams A vector of starting values for hyperparameters (default values from Li et al.).

Value

A list containing the results of hyperparameter inference.

References

Li, Kathy, et al. "A predictive model for next cycle start date that accounts for adherence in menstrual self-tracking." *Journal of the American Medical Informatics Association* 29.1 (2022): 3-11.

 likVec

Monte Carlo estimate of negative marginal log-likelihood of Li model

Description

This function calculates a Monte Carlo estimate of the negative marginal log-likelihood of the given hyperparameters for the generative model from Li et al. (2022). It samples M instances of the parameters from the given distributions and averages the the likelihoods, giving a marginal likelihood for the hyperparameters.

Usage

```
likVec(
  pars = c(kappa = 180, gamma = 6, alpha = 2, beta = 20),
  S = 10,
  M = 1000,
  cycleDat,
  verbose = FALSE,
  ...
)
```

Arguments

pars	Named numeric vector of hyperparameters containing the elements: kappa, gamma, alpha, beta. NOTE: MUST BE IN CORRECT ORDER. <ul style="list-style-type: none"> • kappa: Numeric value, shape parameter of Gamma distribution for Lambda_i. • gamma: Numeric value, rate parameter of Gamma distribution for Lambda_i. • alpha: Numeric value, shape1 parameter of Beta distribution for Pi_i. • beta: Numeric value, shape2 parameter of Beta distribution for Pi_i.
S	Integer, maximum number of allowed skips in the model.
M	Integer specifying the number of Monte Carlo iterations.
cycleDat	Data frame containing information about individuals and their tracked cycles.
verbose	Logical with default FALSE. If true, prints extra info while running.
...	Does nothing.

Value

Numeric value representing the Monte Carlo estimate of the negative marginal log-likelihood.

References

Li, Kathy, et al. "A predictive model for next cycle start date that accounts for adherence in menstrual self-tracking." *Journal of the American Medical Informatics Association* 29.1 (2022): 3-11.

liMCMC	<i>Runs MCMC algorithm for performing inference using the model from Li et al. (2022)</i>
--------	-------------------------------------------------------------------------------------------

Description

This function performs inference on cycle length data, assuming the model from Li et al. (2022). It is important to note that Li et al. does not actually use this algorithm as they target a particular analytic posterior predictive distribution, and solve directly. However, we are targeting a different posterior and thus use this MCMC to perform inference.

Usage

```
liMCMC(
  Y,
  cluster,
  S,
  hyperparams = c(kappa = 180, gamma = 6, alpha = 2, beta = 20),
  initialParams = list(pi = c(1/3, 1/3, 1/3), lambdais = rep(30,
    length(unique(cycleDat$Individual))), piis = rep(0.2,
    length(unique(cycleDat$Individual))), ss = sample(0:S, nrow(cycleDat), replace =
    TRUE)),
  reps = 1000
)
```

Arguments

<code>Y</code>	A vector of observed cycle lengths.
<code>cluster</code>	A vector indicating the individual cluster/group membership for each observation <code>Y</code> .
<code>S</code>	Integer. The maximum number of skips to consider possible.
<code>hyperparams</code>	Named numeric vector of hyperparameters containing the elements: <code>kappa</code> , <code>gamma</code> , <code>alpha</code> , <code>beta</code> . NOTE: MUST BE IN CORRECT ORDER. <ul style="list-style-type: none"> • <code>kappa</code>: Numeric value, shape parameter of Gamma distribution for <code>Lambda_i</code>. • <code>gamma</code>: Numeric value, rate parameter of Gamma distribution for <code>Lambda_i</code>. • <code>alpha</code>: Numeric value, shape1 parameter of Beta distribution for <code>Pi_i</code>. • <code>beta</code>: Numeric value, shape2 parameter of Beta distribution for <code>Pi_i</code>.
<code>initialParams</code>	A list of initial parameter values for the MCMC algorithm. Default values are provided for <code>pi</code> , <code>lambdais</code> , <code>piis</code> , <code>ss</code> .
<code>reps</code>	The number of MCMC iterations (steps) to perform. Default is 1000.

Value

A list containing the MCMC draws for each parameter at each iteration. Each element in the list is itself a list containing:

ijDat A data.frame with updated parameters at the individual-observation level: `Individual`, `ys`, `lambdais`, `piis`, `ss`.

iDat A data.frame with updated parameters at the individual level: `Individual`, `lambdas`, `pis`.

kappa Fixed value of hyperparameter `kappa`.

gamma Fixed value of hyperparameter `gamma`.

alpha Fixed value of hyperparameter `alpha`.

beta Fixed value of hyperparameter `beta`.

S Fixed input value `S`.

indFirst A logical vector indicating the first occurrence of each individual.

References

Li, Kathy, et al. "A predictive model for next cycle start date that accounts for adherence in menstrual self-tracking." *Journal of the American Medical Informatics Association* 29.1 (2022): 3-11.

See Also

[gibbsStepLi](#)

liSim	<i>Simulate user tracked menstrual cycle data for an individual using the li model.</i>
-------	-----------------------------------------------------------------------------------------

Description

This function generates synthetic data for user tracked menstrual cycles for a single individual using the li model. For this model $\text{Beta0} = \log(30)$, and Gamma0 doesn't really make sense.

Usage

```
liSim(i, skipProb, maxCycles, trueBetas, trueGammas = NULL)
```

Arguments

<code>i</code>	Individual identifier. Character, numeric or integer.
<code>skipProb</code>	Vector, ignored for this model.
<code>maxCycles</code>	Integer, Maximum possible number of true cycles per tracked cycle.
<code>trueBetas</code>	Optional. True values for generated mean regression coefficients.
<code>trueGammas</code>	NULL, left for consistency. Will throw error if specified.

Value

'Individual' Individual identifiers.
'TrackedCycles' Tracked cycles.
'NumTrue' Number of true values.
'SkipProb' Individual's probability of skipping tracking a cycle
'Mean' Individual's mean values.
'Beta0' Beta0 true value.
'Gamma0' NA
'Z0' 1
'X0',..., 'XN' Covariate matrix for Mean, where N is the length of trueBetas.

See Also

[skipTrack.simulate](#)

mixSim	<i>Simulate user tracked menstrual cycle data for an individual using the mixture model.</i>
--------	----------------------------------------------------------------------------------------------

Description

This function generates synthetic data for user tracked menstrual cycles for a single individual using the mixture model. For this model Beta_0 is set to log(30) and Gamma_0 is set to 15, although for the skipTrack model this lacks interpretation.

Usage

```
mixSim(i, skipProb, maxCycles, trueBetas, trueGammas, overlap)
```

Arguments

i	Individual identifier. Character, numeric, or integer.
skipProb	Vector of probabilities for the number of true cycles per tracked cycle. For example, (.7, .2, .1) means that 70% of observed cycles will contain one true cycle, 20% will contain 2 true cycles, and 10% will contain 3 true cycles.
maxCycles	Maximum number of true cycles per tracked cycle.
trueBetas	Optional. True values for generated mean regression coefficients.
trueGammas	Optional. True values for the generated precision regression coefficients.
overlap	Optional. Number of (non-intercept) columns shared between X and Z. Columns are shared from left to right.

Value

'Individual' Individual identifiers.
'TrackedCycles' Tracked cycles.
'NumTrue' Number of true values.
'Mean' Individual's mean values.
'Beta0' Beta0 true value.
'Gamma0' Gamma0 true value.
'X0',..., 'XN' Covariate matrix for Mean, where N is the length of trueBetas.
'Z0',..., 'ZM' Covariate matrix for precision, where M is the length of trueGammas.

See Also

[skipTrack.simulate](#)

plot.skipTrack.model *Plot skipTrack.model objects*

Description

Plot skipTrack.model objects

Usage

```
## S3 method for class 'skipTrack.model'
plot(x, ...)
```

Arguments

x skipTrack.model object from the function skipTrack.fit
 ... Needed for S3 consistency

Value

Invisible NULL. Prints plots from skipTrack.visualize

Examples

```
#Simulated data
simDat <- skipTrack.simulate(n = 100, skipProb = c(.7, .2, .1))

#Run model fit (should typically run with much more than 50 reps)
modFit <- skipTrack.fit(Y = simDat$Y, cluster = simDat$cluster, chains = 2, reps = 50)
plot(modFit)
```

postBeta *Draw from Posterior Distribution for Beta Parameters*

Description

In our model μ_i follows a normal distribution with mean $X_i^T \beta$ and precision ρ . Additionally we assume that β follows a mvnormal prior with mean 0 and precision $(\rho_{\beta})^{-1}$. This function draws from the posterior distribution of β under these assumptions.

Usage

```
postBeta(rhoBeta = 0.01, rho, Xi, muI)
```

Arguments

rhoBeta	A scalar representing the prior precision parameter for beta.
rho	A scalar representing the precision parameter.
Xi	A matrix of covariates, where each row represents an individual and each column represents a covariate.
muI	A vector where each element is the mean for individual i.

Details

This function assumes that Xi is a (num Individuals) x (dimension of beta) matrix of covariates.

Value

A vector representing a draw from the posterior distribution of beta parameters.

postCij	<i>Sample a vector of values from the full conditional posterior of the c_ij vector</i>
---------	-----------------------------------------------------------------------------------------

Description

In our model the data are drawn from $\text{LogN}(\mu_i + \log(c_{ij}), \tau_i)$. The prior for c_{ij} is a categorical prior with category probabilities π_1, \dots, π_k , and c_{ij} can take values $1, \dots, k$ where k is the length of π . This function samples from the full conditional posterior of all c_{ijs} , given vectors of equal length y_{ijs} , μ_{is} , τ_{is}

Usage

```
postCij(yijs, pi, muis, tauis)
```

Arguments

yijs	Numeric Vector, cycle lengths
pi	Numeric vector, must sum to 1. Sampled probabilities for c_ijs
muis	Numeric vector, log of sampled mean for all individuals yijs
tauis	Numeric vector > 0, sampled precision for all individuals yijs

Value

Integer vector

postGamma	<i>Perform a Metropolis-Hastings Step for Drawing a New Gamma</i>
-----------	-------------------------------------------------------------------

Description

Our model assumes that $\tau_i \sim \text{Gamma}(\alpha_i, \phi)$ where $\alpha_i/\phi = \theta_i$ and $g(\theta_i) = Z_i^T \Gamma$, where g is the log-link function. Because of this GLM formulation we cannot simply draw from a posterior here and instead use a Metropolis-Hastings step with proposal distribution $\text{propGamma} \sim \text{MVNormal}(\text{currentGamma}, \text{rhoGamma} * I)$

Usage

```
postGamma(tau_i, Zi, currentGamma, phi = 1, rhoGamma = 1000)
```

Arguments

tau_i	A vector of length num_Individuals representing precision parameters for individuals.
Zi	A matrix of covariates, where each row represents an individual and each column represents a covariate.
currentGamma	A vector (or matrix with 1 row) representing the current Gamma value.
phi	A scalar, the prior rate for tau_i.
rhoGamma	A scalar representing the proposal distribution precision parameter.

Details

This draw step assumes a log-link function for the Gamma GLM that we are fitting.

Value

A list containing the new Gamma value and the corresponding theta_i values.

postLambdai	<i>Compute random draw from full conditional posterior for lambda_i in Li algorithm.</i>
-------------	------------------------------------------------------------------------------------------

Description

This function calculates a random draw from the full conditional posterior distribution for λ_i in the Li algorithm, given the observed values y_{ij} , the indicators s_{ij} , and the prior hyperparameters priorK and priorG.

Usage

```
postLambdai(yij, sij, priorK, priorG)
```

Arguments

<code>yij</code>	Vector of observed values for individual i .
<code>sij</code>	Vector of cycle skip indicators for individual i .
<code>priorK</code>	Prior hyperparameter κ .
<code>priorG</code>	Prior hyperparameter γ .

Value

A random draw from the posterior distribution of $\lambda_{i,j}$.

<code>postMui</code>	<i>Sample a value from the full conditional posterior of $\mu_{i,j}$</i>
----------------------	-------------------------------------------------------------------------------------

Description

In our model the data are drawn from $\text{LogN}(\mu_{i,j} + \log(c_{i,j}), \tau_{i,j})$. The prior for $\mu_{i,j}$ is given as $N(x_{i,j}^T \beta, \rho)$. This function draws from the conditional posterior of $\mu_{i,j}$.

Usage

```
postMui(yij, cij, tau_i, xib, rho)
```

Arguments

<code>yij</code>	Numeric vector, cycle lengths for a single individual
<code>cij</code>	Positive Integer vector, a sampled vector of $\text{length}(yij)$ where the corresponding values in <code>cij</code> indicate a sampled number of TRUE cycles in each cycle length given by <code>yij</code>
<code>tau_i</code>	Numeric > 0, A sampled precision for the <code>yij</code> s
<code>xib</code>	Numeric, result of multiplying $x_{i,j}^T \beta$ (single value, not vector)
<code>rho</code>	Numeric > 0, sampled prior precision of $\mu_{i,j}$

Details

Additionally, note that in order to vectorize the remainder of the MCMC algorithm this function returns the sampled value repeated for $\text{length}(yij)$

Value

Numeric vector, repeated sampled value of $\text{length}(yij)$

postPhi	<i>Metropolis-Hastings step to draw a new value for phi.</i>
---------	--------------------------------------------------------------

Description

In our model the data are drawn from $\text{LogN}(\mu_i + \log(c_{ij}), \tau_i)$. The prior for τ_i is given as $\text{Gamma}(\theta_{\text{phi}}, \phi)$. This function uses a MH step to draw a new sample of ϕ . Proposal distribution is $\text{Gamma}(\text{currentPhi}, \rho_{\text{phi}})$. Note that we parameterize with RATE, not SCALE.

Usage

```
postPhi(tau_i, theta_i, currentPhi, rhoPhi = 1000)
```

Arguments

tau_i	Numeric vector, individuals precisions.
theta_i	Numeric vector. individuals precisions means (estimate)
currentPhi	Previous draw of phi
rhoPhi	Proposal rate for gamma distribution that draws proposal for phi, default is 1000.

Value

Numeric, new draw of phi

postPi	<i>Sample a value from the full conditional posterior of pi</i>
--------	-----------------------------------------------------------------

Description

In our model the data are drawn from $\text{LogN}(\mu_i + \log(c_{ij}), \tau_i)$. The prior for c_{ij} is a categorical prior with category probabilities π_1, \dots, π_k , and c_{ij} can take values $1, \dots, k$ where k is the length of π . This function samples from the posterior of $\pi = \pi_1, \dots, \pi_k$, assuming that π follows $\text{Dirichlet}(\text{priorAlphas})$.

Usage

```
postPi(ci, priorAlphas)
```

Arguments

ci	Integer vector, all of the sampled c_{ij} values for all individuals
priorAlphas	Numeric vector, prior dirichlet parameters for π

Value

Numeric vector

postPii *Compute M-H draw for pi_i in Li algorithm*

Description

This performs a Metropolis-Hastings draw for π_i , assuming s_{ij} follows a truncated geometric distribution with parameters π_i and S . The proposal distribution for π_i is $\text{Beta}(\alpha, \beta)$.

Usage

```
postPii(sij, currentPii, priorA, priorB, S)
```

Arguments

sij	Vector of cycle skip indicators for individual i
currentPii	Current value of π_i
priorA	Hyperparameter alpha.
priorB	Hyperparameter beta.
S	Maximum number of skips allowed in algorithm

Value

Draw for π_i , repeated for the number of observations from individual i

postRho *Sample a value from the full conditional posterior of rho*

Description

In our model the data are drawn from $\text{LogN}(\mu_i + \log(c_{ij}), \tau_i)$. The prior for μ_i is given as $\text{N}(\mu, \rho)$. This function draws from the conditional posterior of ρ , given that the prior on ρ is a uniform prior on the standard deviation.

Usage

```
postRho(muI, xib)
```

Arguments

muI	Numeric vector, log of individuals mean values.
xib	Numeric vector, result of $X \sim \text{Beta}$, same length as muI.

Value

Numeric

postSij	<i>Compute random draw from full conditional posterior for s_{ij} in Li algorithm.</i>
---------	------------------------------------------------------------------------------------------------

Description

This function calculates a random draw from the full conditional posterior distribution for s_{ij} in the Li algorithm, given the observed values y_{ijs} , the parameter π_i , the $\lambda_{i,j}$ value, and the truncation level S .

Usage

```
postSij(yijs, pii, lambdai, S)
```

Arguments

<code>yijs</code>	Vector of observed values for s_{ij} .
<code>pii</code>	Probability parameter π_i .
<code>lambdai</code>	Value of $\lambda_{i,j}$.
<code>S</code>	Truncation level.

Value

A random draw from the posterior distribution of s_{ij} .

postTauI	<i>Sample a value from the full conditional posterior of tau_i</i>
----------	------------------------------------------------------------------------------

Description

In our model the data are drawn from $\text{LogN}(\mu_i + \log(c_{ij}), \tau_i)$. The prior for τ_i is given as $\text{Gamma}(\theta_i \cdot \phi, \phi)$. This function draws from the conditional posterior of τ_i . Note that we parameterize with RATE, not SCALE.

Usage

```
postTauI(yij, cij, mui, thetai, phi = 1)
```

Arguments

<code>yij</code>	Numeric vector, cycle lengths for a single individual
<code>cij</code>	Positive Integer vector, a sampled vector of length(<code>yij</code>) where the corresponding values in <code>cij</code> indicate a sampled number of TRUE cycles in each cycle length given by <code>yij</code>
<code>mui</code>	Numeric, log of sampled mean of this individual's y_{ijs}
<code>thetai</code>	Numeric, mean of prior (gamma) distribution on τ_i
<code>phi</code>	Numeric, rate for TauI prior

Details

Additionally, note that in order to vectorize the remainder of the MCMC algorithm this function returns the sampled value repeated for `length(yij)`

Value

Numeric vector, repeated sampled value of `length(yij)`

```
print.skipTrack.model Print skipTrack.model to console
```

Description

Print `skipTrack.model` to console

Usage

```
## S3 method for class 'skipTrack.model'  
print(x, ...)
```

Arguments

<code>x</code>	<code>skipTrack.model</code> object from the function <code>skipTrack.fit</code>
<code>...</code>	Needed for S3 consistency

Value

Invisible NULL. Prints info about `skipTrack.model` object

Examples

```
#Simulated data  
simDat <- skipTrack.simulate(n = 100, skipProb = c(.7, .2, .1))  
  
#Run model fit (should typically run with much more than 50 reps)  
modFit <- skipTrack.fit(Y = simDat$Y, cluster = simDat$cluster, chains = 2, reps = 50)  
print(modFit)
```

sampleStep	<i>Perform a single step of the MCMC sampling process for skipTrack</i>
------------	-------------------------------------------------------------------------

Description

This function performs a single step of the Markov Chain Monte Carlo (MCMC) algorithm to update parameters in a hierarchical model used for identifying skips in menstrual cycle tracking.

Usage

```
sampleStep(
  ijDat,
  iDat,
  rho,
  pi,
  Xi,
  Zi,
  Beta,
  Gamma,
  priorAlphas,
  indFirst,
  rhoBeta,
  rhoGamma,
  phi,
  rhoPhi,
  fixedSkips
)
```

Arguments

ijDat	A data.frame with individual-observation level parameters: Individual, ys, cijs, muis, tauis.
iDat	A data.frame with individual level parameters: Individual, mus, taus, thetas.
rho	Updated value of the global parameter rho.
pi	Updated value of the global parameter pi.
Xi	A matrix (numIndividuals x length(Beta)) of covariates for cycle length mean. Default is a vector of 1's. NOTE THE DIFFERENCE (from skipTrack.MCMC) IN EXPECTED DIMENSION OF X
Zi	A matrix (numIndividuals x length(Gamma)) of covariates for cycle length precision. Default is a vector of 1's. NOTE THE DIFFERENCE (from skipTrack.MCMC) IN EXPECTED DIMENSION OF Z
Beta	Matrix (1 x length(Beta)) of coefficients for cycle length mean.
Gamma	Matrix of (1 x length(Gamma)) coefficients for cycle length precision.
priorAlphas	Vector of prior alpha values for updating pi.

<code>indFirst</code>	A logical vector indicating the first occurrence of each individual.
<code>rhoBeta</code>	Updated value of the global parameter rhoBeta.
<code>rhoGamma</code>	Value of the proposal parameter rhoGamma.
<code>phi</code>	Value of the parameter phi.
<code>rhoPhi</code>	Value of the proposal parameter rhoPhi.
<code>fixedSkips</code>	Logical. If TRUE cycle skip information (cijs) is not updated in sample steps and the inputs are instead assumed to be true.

Value

A list containing updated parameters after performing a single MCMC step. The list includes:

ijDat A data.frame with updated parameters at the individual-observation level: Individual, ys, cijs, muis, tauis.

iDat A data.frame with updated parameters at the individual level: Individual, mus, taus, thetas.

rho Updated value of the global parameter rho.

pi Updated value of the global parameter pi.

Xi Matrix of covariates for cycle length mean.

Zi Matrix of covariates for cycle length precision.

Beta Updated matrix of coefficients for cycle length mean.

Gamma Updated matrix of coefficients for cycle length precision.

priorAlphas Vector of prior alpha values for updating pi.

indFirst A logical vector indicating the first occurrence of each individual.

rhoBeta Hyperprior parameter rhoBeta, used to update Beta.

rhoGamma Value of the proposal parameter rhoGamma.

phi Updated value of the parameter phi.

rhoPhi Value of the proposal parameter rhoPhi.

fixedSkips Logical. Indicates if skips were fixed.

skipTrack.diagnostics *skipTrack MCMC Diagnostics*

Description

Takes model results from skipTrack.fit and uses genMCMCDiag to get generalized mcmc diagnostics

Usage

```
skipTrack.diagnostics(
  stFit,
  param = c("rho", "phi", "Betas", "Gammas", "muis", "tauis", "cijs"),
  method = NULL,
  ...
)
```

Arguments

stFit	A list of MCMC results from the skipTrack.fit function.
param	A character string specifying the parameter for which diagnostics are to be calculated. Must be one of: 'rho', 'phi', 'Betas', 'Gammas', 'muis', 'tauis', or 'cijs'.
method	An optional parameter specifying the method for calculating diagnostics. See package genMCMCDiag for details. Default is NULL.
...	Arguments passed on to genMCMCDiag::genDiagnostic
diagnostics	A character vector or list of diagnostic functions to be evaluated. Options include 'traceplot', 'ess', 'gelmanRubin', or custom functions. See details.
distance	Function for evaluating distance between MCMC draws if required by 'method'. This should be a pairwise distance function that operates on elements of the chains from mhDraws. Note that the lanfear and ts methods ALWAYS require a distance function.
verbose	If TRUE, informative messages are displayed.

Details

If the parameter is 'rho' or 'phi' (the univariate parameters), the function extracts the specified parameter from the MCMC results and calculates diagnostics using the genDiagnostic function with the standard method. If the parameter is any of the other available options, the function extracts the corresponding values and calculates diagnostics using the genDiagnostic function with the specified or default method ('lanfear') and hammingDist as the distance function.

Details on the genDiagnostic function can be found in the genMCMCDiag package.

Value

A mcmcDiag object of MCMC diagnostics for the specified parameter

See Also

[genDiagnostic](#), [skipTrack.fit](#)

Examples

```
#Simulated data
simDat <- skipTrack.simulate(n = 100, skipProb = c(.7, .2, .1))

#Run model fit (should typically run with much more than 50 reps)
modFit <- skipTrack.fit(Y = simDat$Y, cluster = simDat$cluster, chains = 2, reps = 50)

#Get diagnostics for cijS
skipTrack.diagnostics(modFit, 'cijS')
```

<code>skipTrack.fit</code>	<i>Fits the skipTrack Model using 1 or more MCMC chains</i>
----------------------------	-------------------------------------------------------------

Description

This function fits the model using multiple instances of `skipTrack.MCMC`, either in parallel or sequentially.

Usage

```
skipTrack.fit(
  Y,
  cluster,
  X = matrix(1, nrow = length(cluster)),
  Z = matrix(1, nrow = length(cluster)),
  numSkips = 10,
  reps = 1000,
  chains,
  useParallel = FALSE,
  ...
)
```

Arguments

<code>Y</code>	A vector of observed cycle lengths.
<code>cluster</code>	A vector indicating the individual cluster/group membership for each observation <code>Y</code> .
<code>X</code>	A matrix (<code>length(Y) x length(Beta)</code>) of covariates for cycle length mean. Default is a vector of 1's.
<code>Z</code>	A matrix (<code>length(Y) x length(Gamma)</code>) of covariates for cycle length precision. Default is a vector of 1's.
<code>numSkips</code>	The maximum number of skips to allow. Default is 10.
<code>reps</code>	The number of MCMC iterations (steps) to perform. Default is 1000.
<code>chains</code>	Number of chains to run.
<code>useParallel</code>	Logical, indicating whether to use parallel processing, as supported by <code>doParallel</code> . Default is FALSE.
<code>...</code>	Arguments passed on to skipTrack.MCMC
<code>fixedSkips</code>	If TRUE cycle skip information (<code>cijs</code>) is not updated in sample steps and the inputs are instead assumed to be true.
<code>initialParams</code>	A list of initial parameter values for the MCMC algorithm. Default values are provided for <code>pi</code> , <code>muis</code> , <code>tauis</code> , <code>rho</code> , <code>cijs</code> , <code>alphas</code> , <code>Beta</code> , <code>Gamma</code> , <code>phi</code> , <code>rhoBeta</code> , <code>rhoGamma</code> , and <code>rhoPhi</code> .
<code>verbose</code>	logical. If true progress bars and additional info are printed to the console.

Value

A list containing the results of skipTrack.MCMC for each chain.

See Also

[skipTrack.MCMC](#)

Examples

```
#Simulated data
simDat <- skipTrack.simulate(n = 100, skipProb = c(.7, .2, .1))

#Run model fit (should typically run with much more than 50 reps)
modFit <- skipTrack.fit(Y = simDat$Y, cluster = simDat$cluster, chains = 2, reps = 50)
modFit
```

skipTrack.MCMC

Perform one chain of MCMC sampling for the skipTrack model.

Description

This function runs a single Markov Chain Monte Carlo (MCMC) chain to update parameters in the skipTrack hierarchical model.

Usage

```
skipTrack.MCMC(
  Y,
  cluster,
  X = matrix(1, nrow = length(cluster)),
  Z = matrix(1, nrow = length(cluster)),
  numSkips = 10,
  reps = 1000,
  fixedSkips = FALSE,
  initialParams = list(pi = rep(1/(numSkips + 1), numSkips + 1), nuis = rep(log(30),
    length(unique(cluster))), tauis = rep(5, length(unique(cluster))), rho = 1, cijs =
    sample(1:3, length(Y), replace = TRUE), alphas = rep(1, numSkips + 1), Beta =
    matrix(rep(0, ncol(as.matrix(X))), 1), Gamma = matrix(rep(0, ncol(as.matrix(Z))), 1),
    rhoBeta = 0.01, rhoGamma = 1000, phi = 0.01, rhoPhi = 1000),
  verbose = FALSE
)
```

Arguments

<code>Y</code>	A vector of observed cycle lengths.
<code>cluster</code>	A vector indicating the individual cluster/group membership for each observation <code>Y</code> .
<code>X</code>	A matrix ($\text{length}(Y) \times \text{length}(\text{Beta})$) of covariates for cycle length mean. Default is a vector of 1's.
<code>Z</code>	A matrix ($\text{length}(Y) \times \text{length}(\text{Gamma})$) of covariates for cycle length precision. Default is a vector of 1's.
<code>numSkips</code>	The maximum number of skips to allow. Default is 10.
<code>reps</code>	The number of MCMC iterations (steps) to perform. Default is 1000.
<code>fixedSkips</code>	If TRUE cycle skip information (<code>cijs</code>) is not updated in sample steps and the inputs are instead assumed to be true.
<code>initialParams</code>	A list of initial parameter values for the MCMC algorithm. Default values are provided for <code>pi</code> , <code>muis</code> , <code>tauis</code> , <code>rho</code> , <code>cijs</code> , <code>alphas</code> , <code>Beta</code> , <code>Gamma</code> , <code>phi</code> , <code>rhoBeta</code> , <code>rhoGamma</code> , and <code>rhoPhi</code> .
<code>verbose</code>	logical. If true progress bars and additional info are printed to the console.

Value

A list containing the MCMC draws for each parameter at each iteration. Each element in the list is itself a list containing:

ijDat A data.frame with updated parameters at the individual-observation level: `Individual`, `ys`, `cijs`, `muis`, `tauis`.

iDat A data.frame with updated parameters at the individual level: `Individual`, `mus`, `taus`, `thetas`.

rho Updated value of the global parameter `rho`.

pi Updated value of the global parameter `pi`.

Xi Matrix of covariates for cycle length mean.

Zi Matrix of covariates for cycle length precision.

Beta Updated matrix of coefficients for cycle length mean.

Gamma Updated matrix of coefficients for cycle length precision.

priorAlphas Vector of prior alpha values for updating `pi`.

indFirst A logical vector indicating the first occurrence of each individual.

rhoBeta Hyperprior parameter `rhoBeta`, used to update `Beta`.

rhoGamma Value of the proposal parameter `rhoGamma`.

phi Updated value of the parameter `phi`.

rhoPhi Value of the proposal parameter `rhoPhi`.

fixedSkips Logical. Indicates if skips were fixed.

See Also

[sampleStep](#)

skipTrack.results *Get tables of Inference results from skipTrack.fit*

Description

This function calculates inference results on Betas, Gammas, and cijs based on the provided MCMC results. It returns summaries such as credible intervals for Betas, Gammas, wald-type confidence intervals for cijs, and Gelman-Rubin diagnostics for all 3. Note that true values and coverage are included in the output if trueVals is supplied, but otherwise not.

Usage

```
skipTrack.results(stFit, trueVals = NULL, burnIn = 750)
```

Arguments

stFit	Object result of skipTrack.fit function.
trueVals	Optional named list containing true values for Betas, Gammas, and cijs. (Also can use output of skipTrack.simulate)
burnIn	Number of MCMC iterations to discard as burn-in per chain.

Value

A list containing the following elements:

Betas	data.frame with 95% credible intervals and (if trueVals is supplied) true values for Betas and Coverage tag.
Gammas	data.frame with 95% credible intervals and (if trueVals is supplied) true values for Gammas and Coverage tag.
cijs	data.frame with Wald-type 95% confidence intervals and (if trueVals is supplied) true values for cijs and Coverage tags.
Diagnostics	data.frame with ess and gelman-rubin diagnostics from genMCMCDiag package, for parameter sets 'Betas', 'Gammas' and 'cijs'.

Examples

```
#Simulated data
simDat <- skipTrack.simulate(n = 100, skipProb = c(.7, .2, .1))

#Run model fit (should typically run with much more than 50 reps)
modFit <- skipTrack.fit(Y = simDat$Y, cluster = simDat$cluster, chains = 2, reps = 50)
modFit

# If using simulated data (which includes access to ground truth):
#
skipTrack.results(modFit, trueVals = simDat, burnIn = 25)
#Recommended burnIn with real data is at least 750
```

```
#
# If not using simulated data:
#
skipTrack.results(modFit, burnIn = 25)
#Recommended burnIn with real data is at least 750
```

skipTrack.simulate *Simulate user-tracked menstrual cycle data for multiple individuals*

Description

This function generates synthetic data for user-tracked menstrual cycles given a generative model, skip probabilities, maximum cycles and covariates (depending on the model). It supports built-in models ('skipTrack', 'li', 'mixture') and custom models written as functions.

Usage

```
skipTrack.simulate(
  n,
  model = c("skipTrack", "li", "mixture"),
  skipProb = NULL,
  maxCycles = length(skipProb),
  trueBetas = NULL,
  trueGammas = NULL,
  overlap = 0
)
```

Arguments

n	Number of individuals to simulate data for.
model	model for data simulation. Can be a character ('skipTrack', 'li', 'mixture') or a custom function.
skipProb	Vector of probabilities for number of true cycles per tracked cycle. For example, (.7, .2, .1) means that 70% of observed cycles will contain one true cycle, 20% will contain 2 true cycles and 10% will contain 3 true cycles. Default is NULL. If model == 'li', skipProb values are set and user input will be ignored.
maxCycles	Maximum number of cycles for generating skip cycles. Default is the length of skipProb. If model == 'li', this must be specified; if model == 'skipTrack' or 'mixture', leave as default.
trueBetas	Optional. True values for the mean regression coefficients (not counting intercept which is automatic based on the model).
trueGammas	Optional. True values for the precision regression coefficients (not counting intercept which is automatic based on the model). Precision covariates not available for model == 'li'.
overlap	Optional. Number of (non-intercept) columns shared between X and Z. Columns are shared from left to right.

Value

A list containing:

'Y' Tracked cycles from the simulated data.

'cluster' Individual identifiers from the simulated data.

'X' Covariate matrix for Betas (mean cycle length).

'Z' Covariate matrix for Gammas (regularity).

'Beta' True beta coefficients.

'Gamma' True gamma coefficients.

'NumTrue' Number of true cycles in each tracked cycle. Order matches Y.

'Underlying' Subset of the simulated data containing individual level information. For 'skipTrack' - individual mean and precision for log(cycle lengths), for 'li' - individual mean for cycle lengths, for 'mixture' - individual mean for cycle lengths

References

Li, Kathy, et al. "A predictive model for next cycle start date that accounts for adherence in menstrual self-tracking." *Journal of the American Medical Informatics Association* 29.1 (2022): 3-11.

See Also

[stSim](#), [liSim](#), [mixSim](#)

Examples

```
# Example simulation from the SkipTrack model
resultSt <- skipTrack.simulate(1000, model = 'skipTrack', skipProb = c(.7, .2, .1))
hist(resultSt$Y, breaks = 5:200)
```

```
# Example simulation from the Li model
resultLi <- skipTrack.simulate(1000, model = 'li', maxCycles = 3)
hist(resultLi$Y, breaks = 5:200)
```

```
#Example simulation from the mixture model
resultMix <- skipTrack.simulate(1000, model = 'mixture', skipProb = c(.7, .2, .1))
hist(resultMix$Y, breaks = 5:200)
```

skipTrack.visualize *Visualize Results from skipTrack.fit*

Description

This function takes results from skipTrack.fit and produces several helpful visualizations.

Usage

```
skipTrack.visualize(stFit)
```

Arguments

stFit A list containing MCMC results obtained from skipTrack.fit.

Value

A list of three ggplot2 objects:

- cijOverLength - Scatter plot of estimated Cij values against reported cycle length.
- cijOverTaus - Scatter plot of estimated Cij values against estimated individual precisions, colored by cycle length.
- cijDens - Density plot of Y values overlaid with a density plot of Y values separated by estimated cij value.

See Also

[skipTrack.fit](#) for generating MCMC results.

Examples

```
#Simulated data
simDat <- skipTrack.simulate(n = 100, skipProb = c(.7, .2, .1))

#Run model fit (should typically run with much more than 50 reps)
modFit <- skipTrack.fit(Y = simDat$Y, cluster = simDat$cluster, chains = 2, reps = 50)

#Visualize results
skipTrack.visualize(modFit)
```

```
str.skipTrack.model    Report skipTrack.model structure to console
```

Description

Report skipTrack.model structure to console

Usage

```
## S3 method for class 'skipTrack.model'
str(object, ...)
```

Arguments

object skipTrack.model object from the function skipTrack.fit
 ... To match other str calls

Value

Invisible NULL. Prints structure of skipTrack.model object

Examples

```
#Simulated data
simDat <- skipTrack.simulate(n = 100, skipProb = c(.7, .2, .1))

#Run model fit (should typically run with much more than 50 reps)
modFit <- skipTrack.fit(Y = simDat$Y, cluster = simDat$cluster, chains = 2, reps = 50)
str(modFit)
```

stSim	<i>Simulate user tracked menstrual cycle data for an individual, based on the skipTrack model.</i>
-------	----------------------------------------------------------------------------------------------------

Description

This function generates synthetic data for user tracked menstrual cycles for a single individual. For this model $\text{Beta}_0 = \log(30)$, $\text{Gamma}_0 = 5.5$, and $\text{phi} = .01$.

Usage

```
stSim(i, skipProb, maxCycles, trueBetas, trueGammas, overlap)
```

Arguments

i	Individual identifier. Character, numeric or integer.
skipProb	Vector of probabilities for number of true cycles per tracked cycle. For example, (.7, .2, .1) means that 70% of observed cycles will contain one true cycle, 20% will contain 2 true cycles and 10% will contain 3 true cycles.
maxCycles	Maximum number of true cycles per tracked cycle. Ignored for this model.
trueBetas	Optional. True values for the mean regression coefficients (not counting intercept which is automatic based on the model).
trueGammas	Optional. True values for the precision regression coefficients (not counting intercept which is automatic based on the model).
overlap	Optional. Number of (non-intercept) columns shared between X and Z. Columns are shared from left to right.

Value

'**Individual**' Individual identifiers.
 '**TrackedCycles**' Tracked cycles.
 '**NumTrue**' Number of true values.
 '**LogMean**' Individual's mean of $\log(Y)$.
 '**LogPrec**' Individual's precision of $\log(Y)$.
 '**Beta0**' Beta0 true value.
 '**Gamma0**' Gamma0 true value.
 '**X0**', ..., '**XN**' Covariate matrix for Mean, where N is the length of trueBetas.
 '**Z0**', ..., '**ZM**' Covariate matrix for precision, where M is the length of trueGammas.

See Also

[skipTrack.simulate](#)

summary.skipTrack.model

Report skipTrack.model results to console

Description

Report skipTrack.model results to console

Usage

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

Arguments

object	skipTrack.model object from the function skipTrack.fit
...	arguments passed to skipTrack.results

Value

Invisible skipTrack.results. Prints info about skipTrack.model object

Examples

```
#Simulated data
simDat <- skipTrack.simulate(n = 100, skipProb = c(.7, .2, .1))

#Run model fit (should typically run with much more than 50 reps)
modFit <- skipTrack.fit(Y = simDat$Y, cluster = simDat$cluster, chains = 2, reps = 50)
summary(modFit, burnIn = 25) #Recommended burnIn with real data is at least 750
```

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