

Package ‘fmrs’

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Type Package

Title Variable Selection in Finite Mixture of AFT Regression and FMR Models

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Description The package obtains parameter estimation, i.e., maximum likelihood estimators (MLE), via the Expectation-Maximization (EM) algorithm for the Finite Mixture of Regression (FMR) models with Normal distribution, and MLE for the Finite Mixture of Accelerated Failure Time Regression (FMAFTR) subject to right censoring with Log-Normal and Weibull distributions via the EM algorithm and the Newton-Raphson algorithm (for Weibull distribution).
More importantly, the package obtains the maximum penalized likelihood (MPLE) for both FMR and FMAFTR models (collectively called FMRs). A component-wise tuning parameter selection based on a component-wise BIC is implemented in the package.
Furthermore, this package provides Ridge Regression and Elastic Net.

Depends R (>= 4.3.0)

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fmrs-package	<i>Variable Selection in Finite Mixture of AFT Regression and FMR Models</i>
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Description

The package obtains parameter estimation, i.e., maximum likelihood estimators (MLE), via the Expectation-Maximization (EM) algorithm for the Finite Mixture of Regression (FMR) models with Normal distribution, and MLE for the Finite Mixture of Accelerated Failure Time Regression (FMAFTR) subject to right censoring with Log-Normal and Weibull distributions via the EM algorithm and the Newton-Raphson algorithm (for Weibull distribution). More importantly, the package obtains the maximum penalized likelihood (MPLE) for both FMR and FMAFTR models (collectively called FMRs). A component-wise tuning parameter selection based on a component-wise BIC is implemented in the package. Furthermore, this package provides Ridge Regression and Elastic Net. The `fmrs.mle` method provides MLE for FMRs models. The `fmrs.tunsel` method provides component-wise tuning parameters. The `fmrs.varsel` method provides variable selection for FMRs models.

fmrs methods

`fmrs.mle`, `fmrs.tunsel`, `fmrs.varsel`, `fmrs.gendata`.

fmrs objects

`fmrsfit-class`, `fmrstunpar-class`

BIC

BIC method

Description

Provides the estimated BIC of an FMRs model from an `fmrsfit-class`

Usage

```
BIC(object, ...)

## S4 method for signature 'fmrsfit'
BIC(object, ...)
```

Arguments

<code>object</code>	An <code>fmrsfit-class</code>
<code>...</code>	Other possible arguments

Value

A numeric value

Author(s)

Farhad Shokoohi <shokoohi@icloud.com>

Examples

```
set.seed(1980)
K = 2
D = 10
n = 500
sig = c(1, 1)
piM = c(0.4, 0.6)
r1 = 0.5
coeff1 = c( 2,  2, -1, -2, 1, 2, 0, 0,  0, 0,  0)
coeff2 = c(-1, -1,  1,  2, 0, 0, 0, 0, -1, 2, -2)
Um = 40

dat <- fmrs.gendata(nObs = n, nComp = K, nCov = D, coeff = c(coeff1, coeff2),
```

```

dispersion = sig, mixProp = piM, rho = r1, umax = Um, disFamily = 'lnorm')

res.mle <- fmrs.mle(y = dat$y, x = dat$x, delta = dat$delta, nComp = K,
disFamily = 'lnorm', initCoeff = rnorm(K*D+K), initDispersion = rep(1, K),
initmixProp = rep(1/K, K))

BIC(res.mle)

```

coefficients

coefficients method

Description

Provides the estimated regression coefficients from the fitted FMRs model from an [fmrsfit-class](#)

Usage

```

coefficients(object, ...)

## S4 method for signature 'fmrsfit'
coefficients(object, ...)

```

Arguments

object	An fmrsfit-class
...	Other possible arguments

Value

A numeric array of dimension-(nCov+1)-nComp

Author(s)

Farhad Shokoohi <shokoohi@icloud.com>

Examples

```

set.seed(1980)
K = 2
D = 10
n = 500
sig = c(1, 1)
piM = c(0.4, 0.6)
r1 = 0.5
coeff1 = c( 2,  2, -1, -2, 1, 2, 0, 0,  0, 0,  0)
coeff2 = c(-1, -1,  1,  2, 0, 0, 0, 0, -1, 2, -2)
Um = 40

dat <- fmrs.gendata(nObs = n, nComp = K, nCov = D, coeff = c(coeff1, coeff2),

```

```

dispersion = sig, mixProp = piM, rho = r1, umax = Um, disFamily = 'lnorm')

res.mle <- fmrs.mle(y = dat$y, x = dat$x, delta = dat$delta, nComp = K,
disFamily = 'lnorm', initCoeff = rnorm(K*D+K), initDispersion = rep(1, K),
initmixProp = rep(1/K, K))

coefficients(res.mle)

```

dispersion

*dispersion method***Description**

Provides the estimated dispersions of the fitted FMRs model from an [fmrsfit-class](#)

Usage

```

dispersion(object, ...)

## S4 method for signature 'fmrsfit'
dispersion(object, ...)

```

Arguments

object	An fmrsfit-class
...	Other possible arguments

Value

A numeric array of dimension-(nCov+1)-nComp

Author(s)

Farhad Shokoohi <shokoohi@icloud.com>

Examples

```

set.seed(1980)
K = 2
D = 10
n = 500
sig = c(1, 1)
piM = c(0.4, 0.6)
r1 = 0.5
coeff1 = c( 2,  2, -1, -2, 1, 2, 0, 0,  0, 0,  0)
coeff2 = c(-1, -1,  1,  2, 0, 0, 0, 0, -1, 2, -2)
Um = 40

dat <- fmrs.gendata(nObs = n, nComp = K, nCov = D, coeff = c(coeff1, coeff2),

```

```

dispersion = sig, mixProp = piM, rho = r1, umax = Um, disFamily = 'lnorm')

res.mle <- fmrs.mle(y = dat$y, x = dat$x, delta = dat$delta, nComp = K,
disFamily = 'lnorm', initCoeff = rnorm(K*D+K), initDispersion = rep(1, K),
initmixProp = rep(1/K, K))

dispersion(res.mle)

```

fitted

*fitted method***Description**

Provides the fitted response of the fitted FMRs model from an [fmrsfit-class](#)

Usage

```

fitted(object, ...)

## S4 method for signature 'fmrsfit'
fitted(object, ...)

```

Arguments

object	An fmrsfit-class
...	Other possible arguments

Value

A numeric array of dimension-nObs-nComp

Author(s)

Farhad Shokoohi <shokoohi@icloud.com>

Examples

```

set.seed(1980)
K = 2
D = 10
n = 500
sig = c(1, 1)
piM = c(0.4, 0.6)
r1 = 0.5
coeff1 = c( 2,  2, -1, -2, 1, 2, 0, 0,  0, 0,  0)
coeff2 = c(-1, -1,  1,  2, 0, 0, 0, 0, -1, 2, -2)
Um = 40

dat <- fmrs.gendata(nObs = n, nComp = K, nCov = D, coeff = c(coeff1, coeff2),

```

```

dispersion = sig, mixProp = piM, rho = r1, umax = Um, disFamily = 'lnorm')

res.mle <- fmrs.mle(y = dat$y, x = dat$x, delta = dat$delta, nComp = K,
disFamily = 'lnorm', initCoeff = rnorm(K*D+K), initDispersion = rep(1, K),
initmixProp = rep(1/K, K))

head(fitted(res.mle))

```

fmrs.gendata

fmrs.gendata method

Description

Generates a data set from Finite Mixture of AFT regression models or Finite Mixture of Regression models under the specified setting.

Usage

```

fmrs.gendata(nObs, nComp, nCov, coeff, dispersion, mixProp, rho, umax, ...)

## S4 method for signature 'ANY'
fmrs.gendata(
  nObs,
  nComp,
  nCov,
  coeff,
  dispersion,
  mixProp,
  rho,
  umax,
  disFamily = "lnorm"
)

```

Arguments

nObs	A numeric value represents sample size
nComp	A numeric value represents the order mixture in FMRs
nCov	A numeric value represents the number of covariates in design matrix
coeff	A vector of all regression coefficients including intercepts. It must be a vector of length $nComp * (nCov + 1)$.
dispersion	A vector of positive values for dispersion parameters of sub-distributions in FMRs models
mixProp	A vector of mixing proportions which their sum must be one
rho	A numeric value in $[-1, 1]$ which represents the correlation between covariates of design matrix

umax	A numeric value represents the upper bound in Uniform distribution for censoring
...	Other possible options
disFamily	A sub-distribution family. The options are 'norm' for FMR models, 'lnorm' for mixture of AFT regression models with Log-Normal sub-distributions, 'weibull' for mixture of AFT regression models with Weibull sub-distributions

Value

A list including response, covariates and censoring variables

Author(s)

Farhad Shokoohi <shokoohi@icloud.com>

See Also

Other lnorm, norm, weibull: [fmrs.mle\(\)](#), [fmrs.tunsel\(\)](#), [fmrs.varsel\(\)](#)

Examples

```
set.seed(1980)
K = 2
D = 10
n = 500
REP = 500
sig = c(1, 1)
piM = c(0.4, 0.6)
r1 = 0.5
coeff1 = c( 2,  2, -1, -2, 1, 2, 0, 0,  0, 0,  0)
coeff2 = c(-1, -1,  1,  2, 0, 0, 0, 0, -1, 2, -2)
Um = 40

dat <- fmrs.gendata(nObs = n, nComp = K, nCov = D, coeff = c(coeff1, coeff2),
dispersion = sig, mixProp = piM, rho = r1, umax = Um, disFamily = 'lnorm')
```

<code>fmrs.mle</code>	<i>fmrs.mle method</i>
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Description

Provides MLE for Finite Mixture of Accelerated Failure Time Regression Models or Finite Mixture of Regression Models. It also provides Ridge Regression.

Usage

```
fmrs.mle(y, delta, x, nComp, ...)
```

```
## S4 method for signature 'ANY'
```

```
fmrs.mle(
  y,
  delta,
  x,
  nComp = 2,
  disFamily = "lnorm",
  initCoeff,
  initDispersion,
  initmixProp,
  lambRidge = 0,
  nIterEM = 400,
  nIterNR = 2,
  conveps = 1e-08,
  convepsEM = 1e-08,
  convepsNR = 1e-08,
  NRpor = 2,
  activeset
)
```

Arguments

y	Responses (observations)
delta	Censoring indicator vector
x	Design matrix (covariates)
nComp	Order (Number of components) of mixture model
...	Other possible options
disFamily	A sub-distribution family. The options are 'norm' for FMR models, 'lnorm' for mixture of AFT regression models with Log-Normal sub-distributions, 'weibull' for mixture of AFT regression models with Weibull sub-distributions
initCoeff	Vector of initial values for regression coefficients including intercepts
initDispersion	Vector of initial values for standard deviations
initmixProp	Vector of initial values for proportion of components
lambRidge	A positive value for tuning parameter in Ridge Regression or Elastic Net
nIterEM	Maximum number of iterations for EM algorithm
nIterNR	Maximum number of iterations for Newton-Raphson algorithm
conveps	A positive value for avoiding NaN in computing divisions
convepsEM	A positive value for threshold of convergence in EM algorithm
convepsNR	A positive value for threshold of convergence in Newton-Raphson algorithm
NRpor	A positive integer for maximum number of searches in NR algorithm
activeset	A matrix of zero-one that shows which intercepts and covariates are active in the fitted fmrs model

Details

Finite mixture of AFT regression models are represented as follows. Let X be the survival time with non-negative values, and $\mathbf{z} = (z_1, \dots, z_d)^\top$ be a d -dimensional vector of covariates that may have an effect on X . If the survival time is subject to right censoring, then the observed response time is $T = \min\{Y, C\}$, where $Y = \log X$, C is logarithm of the censoring time and $\delta = I_{\{y < c\}}$ is the censoring indicator. We say that $V = (T, \delta, \mathbf{z})$ follows a finite mixture of AFT regression models of order K if the conditional density of (T, δ) given $\mathbf{z}$ has the form

$$f(t, \delta; \mathbf{z}, \Psi) = \sum_{k=1}^K \pi_k [f_Y(t; \theta_k(\mathbf{z}), \sigma_k)]^\delta [S_Y(t; \theta_k(\mathbf{z}), \sigma_k)]^{1-\delta} [f_C(t)]^{1-\delta} [S_C(t)]^\delta$$

where $f_Y(\cdot)$ and $S_Y(\cdot)$ are respectively the density and survival functions of Y , $f_C(\cdot)$ and $S_C(\cdot)$ are respectively the density and survival functions of C ; and $\theta_k(\mathbf{z}) = h(\beta_{0k} + \mathbf{z}^\top \boldsymbol{\beta}_k)$ for a known link function $h(\cdot)$, $\Psi = (\pi_1, \dots, \pi_K, \beta_{01}, \dots, \beta_{0K}, \boldsymbol{\beta}_1, \dots, \boldsymbol{\beta}_K, \sigma_1, \dots, \sigma_K)^\top$ with $\boldsymbol{\beta}_k = (\beta_{k1}, \beta_{k2}, \dots, \beta_{kd})^\top$ and $0 < \pi_k < 1$ with $\sum_{k=1}^K \pi_k = 1$. The log-likelihood of a sample of size n is formed as

$$\ell_n(\Psi) = \sum_{i=1}^n \log \sum_{k=1}^K \pi_k [f_Y(t_i, \theta_k(\mathbf{z}_i), \sigma_k)]^{\delta_i} [S_Y(t_i, \theta_k(\mathbf{z}_i), \sigma_k)]^{1-\delta_i}.$$

Note that we assume the censoring distribution is non-informative and hence won't play any role in the estimation process. We use EM and Newton-Raphson algorithms in our method to find the maximizer of above Log-Likelihood.

Value

An `fmrsfit-class` that includes parameter estimates of the specified FMRs model

Author(s)

Farhad Shokoohi <shokoohi@icloud.com>

References

Shokoohi, F., Khalili, A., Asgharian, M. and Lin, S. (2016 submitted) Variable Selection in Mixture of Survival Models for Biomedical Genomic Studies

See Also

Other Inorm, norm, weibull: `fmrs.gendata()`, `fmrs.tunsel()`, `fmrs.varsel()`

Examples

```
set.seed(1980)
K = 2
D = 10
n = 500
sig = c(1, 1)
piM = c(0.4, 0.6)
```

```

r1 = 0.5
coeff1 = c( 2,  2, -1, -2, 1, 2, 0, 0,  0, 0,  0)
coeff2 = c(-1, -1,  1,  2, 0, 0, 0, 0, -1, 2, -2)
Um = 40

dat <- fmrs.gendata(nObs = n, nComp = K, nCov = D, coeff = c(coeff1, coeff2),
dispersion = sig, mixProp = piM, rho = r1, umax = Um, disFamily = 'lnorm')

res.mle <- fmrs.mle(y = dat$y, x = dat$x, delta = dat$delta, nComp = K,
disFamily = 'lnorm', initCoeff = rnorm(K*D+K), initDispersion = rep(1, K),
initmixProp = rep(1/K, K))

summary(res.mle)

```

fmrs.tunsel

fmrs.tunsel method

Description

Provides component-wise tuning parameters using BIC for Finite Mixture of Accelerated Failure Time Regression Models and Finite Mixture of Regression Models.

Usage

```
fmrs.tunsel(y, delta, x, nComp, ...)
```

```
## S4 method for signature 'ANY'
```

```

fmrs.tunsel(
  y,
  delta,
  x,
  nComp,
  disFamily = "lnorm",
  initCoeff,
  initDispersion,
  initmixProp,
  penFamily = "lasso",
  lambRidge = 0,
  nIterEM = 400,
  nIterNR = 2,
  conveps = 1e-08,
  convepsEM = 1e-08,
  convepsNR = 1e-08,
  NRpor = 2,
  gamMixPor = 1,
  activeset,
  lambMCP,
  lambSICA,

```

```

    cutpoint = 0.05,
    LambMin = 0.01,
    LambMax = 1,
    nLamb = 100
)

```

Arguments

y	Responses (observations)
delta	Censoring indicator vector
x	Design matrix (covariates)
nComp	Order (Number of components) of mixture model
...	Other possible options
disFamily	A sub-distribution family. The options are 'norm' for FMR models, 'lnorm' for mixture of AFT regression models with Log-Normal sub-distributions, 'weibull' for mixture of AFT regression models with Weibull sub-distributions,
initCoeff	Vector of initial values for regression coefficients including intercepts
initDispersion	Vector of initial values for standard deviations
initmixProp	Vector of initial values for proportion of components
penFamily	Penalty name that is used in variable selection method. The available options are 'lasso', 'adplasso', 'mcp', 'scad', 'sica' and 'hard'.
lambRidge	A positive value for tuning parameter in Ridge Regression or Elastic Net
nIterEM	Maximum number of iterations for EM algorithm
nIterNR	Maximum number of iterations for Newton-Raphson algorithm
conveps	A positive value for avoiding NaN in computing divisions
convepsEM	A positive value for threshold of convergence in EM algorithm
convepsNR	A positive value for threshold of convergence in NR algorithm
NRpor	A positive interger for maximum number of searches in NR algorithm
gamMixPor	Proportion of mixing parameters in the penalty. The value must be in the interval [0,1]. If gamMixPor = 0, the penalty structure is no longer mixture.
activeset	A matrix of zero-one that shows which intercepts and covariates are active in the fitted fmrs model
lambMCP	A positive numbers for mcp's extra tuning parameter
lambSICA	A positive numbers for sica's extra tuning parameter
cutpoint	A positive value for shrinking small values of parameter estimations in the EM algorithm toward zero
LambMin	A positive value for minimum value of tuning parameter
LambMax	A positive value for maximum value of tuning parameter
nLamb	A positive value for number of tuning parameter

Details

The maximizer of penalized Log-Likelihood depends on selecting a set of good tuning parameters which is a rather thorny issue. We choose a value in an equally spaced set of values in $(0, \lambda_{max})$ for a pre-specified λ_{max} that maximize the component-wise BIC,

$$\hat{\lambda}_k = \operatorname{argmax}_{\lambda_k} BIC_k(\lambda_k) = \operatorname{argmax}_{\lambda_k} \left\{ \ell_{k,n}^c(\hat{\Psi}_{\lambda_k,k}) - |d_{\lambda_k,k}| \log(n) \right\},$$

where $d_{\lambda_k,k} = \{j : \hat{\beta}_{\lambda_k,kj} \neq 0, j = 1, \dots, d\}$ is the active set excluding the intercept and $|d_{\lambda_k,k}|$ is its size. This approach is much faster than using an nComp by nComp grid to select the set λ to maximize the penallized Log-Likelihood.

Value

An `fmrstunpar-class` that includes component-wise tuning parameter estimates that can be used in variable selection procedure.

Author(s)

Farhad Shokoohi <shokoohi@icloud.com>

References

Shokoohi, F., Khalili, A., Asgharian, M. and Lin, S. (2016 submitted) Variable Selection in Mixture of Survival Models for Biomedical Genomic Studies

See Also

Other lnorm, norm, weibull: `fmrs.gendata()`, `fmrs.mle()`, `fmrs.varsel()`

Examples

```
set.seed(1980)
K = 2
D = 10
n = 500
sig = c(1, 1)
piM = c(0.4, 0.6)
r1 = 0.5
coeff1 = c( 2,  2, -1, -2, 1, 2, 0, 0, 0, 0, 0)
coeff2 = c(-1, -1, 1, 2, 0, 0, 0, 0, -1, 2, -2)
Um = 40

dat <- fmrs.gendata(nObs = n, nComp = K, nCov = D, coeff = c(coeff1, coeff2),
  dispersion = sig, mixProp = piM, rho = r1, umax = Um, disFamily = 'lnorm')

res.mle <- fmrs.mle(y = dat$y, x = dat$x, delta = dat$delta, nComp = K,
  disFamily = 'lnorm', initCoeff = rnorm(K*D+K), initDispersion = rep(1, K),
  initmixProp = rep(1/K, K))

res.lam <- fmrs.tunsel(y = dat$y, x = dat$x, delta = dat$delta, nComp = K,
  disFamily = 'lnorm', initCoeff = c(coefficients(res.mle)),
```

```

initDispersion = dispersion(res.mle), initmixProp = mixProp(res.mle),
penFamily = 'adplasso')

show(res.lam)

```

fmrs.varsel

fmrs.varsel method

Description

Provides variable selection and penalized MLE for Finite Mixture of Accelerated Failure Time Regression (FMAFTR) Models and Finite Mixture of Regression (FMR) Models. It also provide Ridge Regression and Elastic Net.

Usage

```
fmrs.varsel(y, delta, x, nComp, ...)
```

```
## S4 method for signature 'ANY'
```

```

fmrs.varsel(
  y,
  delta,
  x,
  nComp,
  disFamily = "lnorm",
  initCoeff,
  initDispersion,
  initmixProp,
  penFamily = "lasso",
  lambPen,
  lambRidge = 0,
  nIterEM = 2000,
  nIterNR = 2,
  conveps = 1e-08,
  convepsEM = 1e-08,
  convepsNR = 1e-08,
  NRpor = 2,
  gamMixPor = 1,
  activeset,
  lambMCP,
  lambSICA,
  cutpoint = 0.05
)

```

Arguments

y Responses (observations)

delta	Censoring indicators
x	Design matrix (covariates)
nComp	Order (Number of components) of mixture model
...	Other possible options
disFamily	A sub-distribution family. The options are 'norm' for FMR models, 'lnorm' for mixture of AFT regression models with Log-Normal sub-distributions, 'weibull' for mixture of AFT regression models with Weibull sub-distributions
initCoeff	Vector of initial values for regression coefficients including intercepts
initDispersion	Vector of initial values for standard deviations
initmixProp	Vector of initial values for proportion of components
penFamily	Penalty name that is used in variable selection method The available options are 'lasso', 'adplasso', 'mcp', 'scad', 'sica' and 'hard'.
lambPen	A vector of positive numbers for tuning parameters
lambRidge	A positive value for tuning parameter in Ridge Regression or Elastic Net
nIterEM	Maximum number of iterations for EM algorithm
nIterNR	Maximum number of iterations for Newton-Raphson algorithm
conveps	A positive value for avoiding NaN in computing divisions
convepsEM	A positive value for threshold of convergence in EM algorithm
convepsNR	A positive value for threshold of convergence in NR algorithm
NRpor	A positive interger for maximum number of searches in NR algorithm
gamMixPor	Proportion of mixing parameters in the penalty. The value must be in the interval [0,1]. If gamMixPor = 0, the penalty structure is no longer mixture.
activeset	A matrix of zero-one that shows which intercepts and covariates are active in the fitted fmrs model
lambMCP	A positive numbers for mcp's extra tuning parameter
lambSICA	A positive numbers for sica's extra tuning parameter
cutpoint	A positive value for shrinking small values of parameter estimations in the EM algorithm toward zero

Details

The penalized likelihood of a finite mixture of AFT regression models is written as

$$\tilde{\ell}_n(\Psi) = \ell_n(\Psi) - \mathbf{p}_{\lambda_n}(\Psi)$$

where

$$\mathbf{p}_{\lambda_n}(\Psi) = \sum_{k=1}^K \pi_k^\alpha \left\{ \sum_{j=1}^d p_{\lambda_{n,k}}(\beta_{kj}) \right\}.$$

In the M step of EM algorithm the function

$$\tilde{Q}(\Psi, \Psi^{(m)}) = \sum_{k=1}^K \tilde{Q}_k(\Psi_k, \Psi_k^{(m)}) = \sum_{k=1}^K \left[Q_k(\Psi_k, \Psi_k^{(m)}) - \pi_k^\alpha \left\{ \sum_{j=1}^d p_{\lambda_{n,k}}(\beta_{kj}) \right\} \right]$$

is maximized. Since the penalty function is singular at origin, we use a local quadratic approximation (LQA) for the penalty as follows,

$$\mathbf{p}_{k,\lambda_n}^*(\boldsymbol{\beta}, \boldsymbol{\beta}^{(m)}) = (\pi_k^{(m)})^\alpha \sum_{j=1}^d \left\{ p_{\lambda_n,k}(\beta_{kj}^{(m)}) + \frac{p'_{\lambda_n,k}(\beta_{kj}^{(m)})}{2\beta_{kj}^{(m)}} (\beta_{kj}^2 - \beta_{kj}^{(m)2}) \right\}.$$

Therefore maximizing Q is equivalent to maximizing the function

$$Q^*(\boldsymbol{\Psi}, \boldsymbol{\Psi}^{(m)}) = \sum_{k=1}^K Q_k^*(\boldsymbol{\Psi}_k, \boldsymbol{\Psi}_k^{(m)}) = \sum_{k=1}^K \left[Q_k(\boldsymbol{\Psi}_k, \boldsymbol{\Psi}_k^{(m)}) - \mathbf{p}_{k,\lambda_n}^*(\boldsymbol{\beta}, \boldsymbol{\beta}^{(m)}) \right].$$

In case of Log-Normal sub-distributions, the maximizers of Q_k functions are as follows. Given the data and current estimates of parameters, the maximizers are

$$\boldsymbol{\beta}_k^{(m+1)} = (\mathbf{z}' \boldsymbol{\tau}_k^{(m)} \mathbf{z} + \varpi_k(\boldsymbol{\beta}_{kj}^{(m)}))^{-1} \mathbf{z}' \boldsymbol{\tau}_k^{(m)} T_k^{(m)},$$

where $\varpi_k(\boldsymbol{\beta}_{kj}^{(m)}) = \text{diag} \left(\left(\pi_k^{(m+1)} \right)^\alpha \frac{p'_{\lambda_n,k}(\boldsymbol{\beta}_{kj}^{(m)})}{\boldsymbol{\beta}_{kj}^{(m)}} \right)$ and $\sigma_k^{(m+1)}$ is equal to

$$\sigma_k^{(m+1)} = \sqrt{\frac{\sum_{i=1}^n \tau_{ik}^{(m)} (t_{ik}^{(m)} - \mathbf{z}_i \boldsymbol{\beta}_k^{(m)})^2}{\sum_{i=1}^n \tau_{ik}^{(m)} [\delta_i + (1 - \delta_i) \{A(w_{ik}^{(m)})[A(w_{ik}^{(m)}) - w_{ik}^{(m)}]\}]}.$$

For the Weibull distribution, on the other hand, we have $\tilde{\boldsymbol{\Psi}}_k^{(m+1)} = \tilde{\boldsymbol{\Psi}}_k^{(m)} - 0.5^\kappa \left[H_k^{p,(m)} \right]^{-1} I_k^{p,(m)}$, where $H_k^p = H_k + h(\boldsymbol{\Psi}_k)$ is the penalized version of hessian matrix and $I_k^p = I_k + h(\boldsymbol{\Psi}_k) \boldsymbol{\Psi}_k$ is the penalized version of vector of first derivatives evaluated at $\tilde{\boldsymbol{\Psi}}_k^{(m)}$.

Value

`fmrsfit-class`

Author(s)

Farhad Shokoohi <shokoohi@icloud.com>

References

Shokoohi, F., Khalili, A., Asgharian, M. and Lin, S. (2016 submitted) Variable Selection in Mixture of Survival Models for Biomedical Genomic Studies

See Also

Other Inorm, norm, weibull: `fmrs.gendata()`, `fmrs.mle()`, `fmrs.tunsel()`

Examples

```

set.seed(1980)
K = 2
D = 10
n = 500
sig = c(1, 1)
piM = c(0.4, 0.6)
r1 = 0.5
coeff1 = c( 2,  2, -1, -2, 1, 2, 0, 0,  0, 0,  0)
coeff2 = c(-1, -1,  1,  2, 0, 0, 0, 0, -1, 2, -2)
Um = 40

dat <- fmrs.gendata(nObs = n, nComp = K, nCov = D, coeff = c(coeff1, coeff2),
  dispersion = sig, mixProp = piM, rho = r1, umax = Um, disFamily = 'lnorm')

res.mle <- fmrs.mle(y = dat$y, x = dat$x, delta = dat$delta, nComp = K,
  disFamily = 'lnorm', initCoeff = rnorm(K*D+K), initDispersion = rep(1, K),
  initmixProp = rep(1/K, K))

res.lam <- fmrs.tunsel(y = dat$y, x = dat$x, delta = dat$delta,
  nComp = ncomp(res.mle), disFamily = 'lnorm',
  initCoeff = c(coefficients(res.mle)), initDispersion = dispersion(res.mle),
  initmixProp = mixProp(res.mle), penFamily = 'adplasso')

res.var <- fmrs.varsel(y = dat$y, x = dat$x, delta = dat$delta,
  nComp = ncomp(res.mle), disFamily = 'lnorm',
  initCoeff = c(coefficients(res.mle)), initDispersion = dispersion(res.mle),
  initmixProp = mixProp(res.mle), penFamily = 'adplasso',
  lambPen = slot(res.lam, 'lambPen'))

summary(res.var)

```

fmrsfit-class

An S4 class to represent a fitted FMRs model

Description

is an S4 class represents a fitted of FMRs model resulted from running `fmrs.mle` or `fmrs.varsel`

Slots

y A length-nobs numeric vector
delta A length-nobs numeric vector
x A dimension-nobs-ncov numeric matrix
nobs A length-one numeric vector
ncov A length-one numeric vector
ncomp A length-one numeric vector

coefficients A length-(ncov+1)-ncomp numeric matrix
 dispersion A length-ncomp numeric vector
 mixProp A length-ncomp numeric vector
 logLik A length-one numeric vector
 BIC A length-one numeric vector
 nIterEMconv A length-one numeric vector
 disFamily A length-one character vector
 penFamily A length-one character vector
 lambPen A length-ncomp numeric vector
 lambRidge A length-one numeric vector
 MCPGam A length-one numeric vector
 SICAGam A length-one numeric vector
 model A length-one character vector
 fitted A dimension-nobs-ncomp numeric matrix
 residuals A dimension-nobs-ncomp numeric matrix
 weights A dimension-nobs-ncomp numeric matrix
 activeset A dimension-ncov+1-ncomp 0-1 matrix
 selectedset A dimension-ncov-ncomp 0-1 matrix

fmrstunpar-class

An S4 class to represent estimated optimal lambdas

Description

An S4 class to represent estimated optimal lambdas resulted from running [fmrs.tunsel](#)

Slots

ncov A length-one numeric vector
 ncomp A length-one numeric vector
 lambPen A dimension-one-ncomp numeric array
 MCPGam A length-one numeric vector
 SICAGam A length-one numeric vector
 disFamily A length-one character vector
 penFamily A length-one character vector
 lambRidge A length-one numeric vector
 model A length-one character vector
 activeset A dimension-nobs-ncomp 0-1 matrix

logLik	<i>logLik method</i>
--------	----------------------

Description

Provides the estimated logLikelihood of an FMRs model from an [fmrsfit-class](#)

Usage

```
logLik(object, ...)

## S4 method for signature 'fmrsfit'
logLik(object, ...)
```

Arguments

object	An fmrsfit-class
...	Other possible arguments

Value

A numeric value

Author(s)

Farhad Shokoohi <shokoohi@icloud.com>

Examples

```
set.seed(1980)
K = 2
D = 10
n = 500
sig = c(1, 1)
piM = c(0.4, 0.6)
r1 = 0.5
coeff1 = c( 2,  2, -1, -2, 1, 2, 0, 0,  0, 0,  0)
coeff2 = c(-1, -1,  1,  2, 0, 0, 0, 0, -1, 2, -2)
Um = 40

dat <- fmrs.gendata(nObs = n, nComp = K, nCov = D, coeff = c(coeff1, coeff2),
  dispersion = sig, mixProp = piM, rho = r1, umax = Um, disFamily = 'lnorm')

res.mle <- fmrs.mle(y = dat$y, x = dat$x, delta = dat$delta, nComp = K,
  disFamily = 'lnorm', initCoeff = rnorm(K*D+K), initDispersion = rep(1, K),
  initmixProp = rep(1/K, K))

logLik(res.mle)
```

mixProp

*mixProp method***Description**

Provides the estimated mixing proportions of an FMRs model form an [fmrsfit-class](#)

Usage

```
mixProp(object, ...)

## S4 method for signature 'fmrsfit'
mixProp(object, ...)
```

Arguments

object An [fmrsfit-class](#)
 ... Other possible arguments

Value

A numeric vector of length-nComp

Author(s)

Farhad Shokoohi <shokoohi@icloud.com>

Examples

```
set.seed(1980)
K = 2
D = 10
n = 500
sig = c(1, 1)
piM = c(0.4, 0.6)
r1 = 0.5
coeff1 = c( 2,  2, -1, -2, 1, 2, 0, 0,  0, 0,  0)
coeff2 = c(-1, -1,  1,  2, 0, 0, 0, 0, -1, 2, -2)
Um = 40

dat <- fmrs.gendata(nObs = n, nComp = K, nCov = D, coeff = c(coeff1, coeff2),
  dispersion = sig, mixProp = piM, rho = r1, umax = Um, disFamily = 'lnorm')

res.mle <- fmrs.mle(y = dat$y, x = dat$x, delta = dat$delta, nComp = K,
  disFamily = 'lnorm', initCoeff = rnorm(K*D+K), initDispersion = rep(1, K),
  initmixProp = rep(1/K, K))

mixProp(res.mle)
```

ncomp	<i>ncomp method</i>
-------	---------------------

Description

Provides the order of an FMRs model from an [fmrsfit-class](#)

Usage

```
ncomp(object, ...)
```

S4 method for signature 'fmrsfit'

```
ncomp(object, ...)
```

Arguments

object	An fmrsfit-class
...	Other possible arguments

Value

An integer value

Author(s)

Farhad Shokoohi <shokoohi@icloud.com>

Examples

```
set.seed(1980)
K = 2
D = 10
n = 500
sig = c(1, 1)
piM = c(0.4, 0.6)
r1 = 0.5
coeff1 = c( 2,  2, -1, -2, 1, 2, 0, 0,  0, 0,  0)
coeff2 = c(-1, -1,  1,  2, 0, 0, 0, 0, -1, 2, -2)
Um = 40

dat <- fmrs.gendata(nObs = n, nComp = K, nCov = D, coeff = c(coeff1, coeff2),
  dispersion = sig, mixProp = piM, rho = r1, umax = Um, disFamily = 'lnorm')

res.mle <- fmrs.mle(y = dat$y, x = dat$x, delta = dat$delta, nComp = K,
  disFamily = 'lnorm', initCoeff = rnorm(K*D+K), initDispersion = rep(1, K),
  initmixProp = rep(1/K, K))

ncomp(res.mle)
```

ncov

*ncov method***Description**

Provides the number of covariates of an FMRs model from an [fmrsfit-class](#)

Usage

```
ncov(object, ...)

## S4 method for signature 'fmrsfit'
ncov(object, ...)
```

Arguments

object An [fmrsfit-class](#)
... Other possible arguments

Value

An integer value

Author(s)

Farhad Shokoohi <shokoohi@icloud.com>

Examples

```
set.seed(1980)
K = 2
D = 10
n = 500
sig = c(1, 1)
piM = c(0.4, 0.6)
r1 = 0.5
coeff1 = c( 2,  2, -1, -2, 1, 2, 0, 0, 0, 0, 0)
coeff2 = c(-1, -1,  1,  2, 0, 0, 0, 0, -1, 2, -2)
Um = 40

dat <- fmrs.gendata(nObs = n, nComp = K, nCov = D, coeff = c(coeff1, coeff2),
dispersion = sig, mixProp = piM, rho = r1, umax = Um, disFamily = 'lnorm')

res.mle <- fmrs.mle(y = dat$y, x = dat$x, delta = dat$delta, nComp = K,
disFamily = 'lnorm', initCoeff = rnorm(K*D+K), initDispersion = rep(1, K),
initmixProp = rep(1/K, K))

ncov(res.mle)
```

nobs	<i>nobs method</i>
------	--------------------

Description

Provides the number of observations in an FMRs model from an [fmrsfit-class](#)

Usage

```
nobs(object, ...)

## S4 method for signature 'fmrsfit'
nobs(object, ...)
```

Arguments

object	An fmrsfit-class
...	Other possible arguments

Value

An integer value

Author(s)

Farhad Shokoohi <shokoohi@icloud.com>

Examples

```
set.seed(1980)
K = 2
D = 10
n = 500
sig = c(1, 1)
piM = c(0.4, 0.6)
r1 = 0.5
coeff1 = c( 2,  2, -1, -2, 1, 2, 0, 0,  0, 0,  0)
coeff2 = c(-1, -1,  1,  2, 0, 0, 0, 0, -1, 2, -2)
Um = 40

dat <- fmrs.gendata(nObs = n, nComp = K, nCov = D, coeff = c(coeff1, coeff2),
  dispersion = sig, mixProp = piM, rho = r1, umax = Um, disFamily = 'lnorm')

res.mle <- fmrs.mle(y = dat$y, x = dat$x, delta = dat$delta, nComp = K,
  disFamily = 'lnorm', initCoeff = rnorm(K*D+K), initDispersion = rep(1, K),
  initmixProp = rep(1/K, K))

nobs(res.mle)
```

residuals

*residuals method***Description**

Provides the residuals of the fitted FMRs model from an [fmrsfit-class](#)

Usage

```
residuals(object, ...)

## S4 method for signature 'fmrsfit'
residuals(object, ...)
```

Arguments

object An [fmrsfit-class](#)
... Other possible arguments

Value

A numeric array of dimension-nObs-nComp

Author(s)

Farhad Shokoohi <shokoohi@icloud.com>

Examples

```
set.seed(1980)
K = 2
D = 10
n = 500
sig = c(1, 1)
piM = c(0.4, 0.6)
r1 = 0.5
coeff1 = c( 2,  2, -1, -2, 1, 2, 0, 0,  0, 0,  0)
coeff2 = c(-1, -1,  1,  2, 0, 0, 0, 0, -1, 2, -2)
Um = 40

dat <- fmrs.gendata(nObs = n, nComp = K, nCov = D, coeff = c(coeff1, coeff2),
  dispersion = sig, mixProp = piM, rho = r1, umax = Um, disFamily = 'lnorm')

res.mle <- fmrs.mle(y = dat$y, x = dat$x, delta = dat$delta, nComp = K,
  disFamily = 'lnorm', initCoeff = rnorm(K*D+K), initDispersion = rep(1, K),
  initmixProp = rep(1/K, K))

head(residuals(res.mle))
```

summary

*summary method***Description**

Displays the fitted FMRs model by showing the estimated coefficients, dispersions and mixing proportions

Display the selected component-wise tuning parameters

Usage

```
summary(object, ...)
```

```
summary(object, ...)
```

```
## S4 method for signature 'fmrsfit'
```

```
summary(object, ...)
```

```
## S4 method for signature 'fmrstunpar'
```

```
summary(object, ...)
```

Arguments

object An *fmrsfit-class* or *fmrstunpar-class*

... Other possible arguments

Value

Summary of the fitted FMRs model

Summary of the selected component-wise tuning parameters

Author(s)

Farhad Shokoohi <shokoohi@icloud.com>

Examples

```
set.seed(1980)
K = 2
D = 10
n = 500
sig = c(1, 1)
piM = c(0.4, 0.6)
r1 = 0.5
coeff1 = c( 2,  2, -1, -2, 1, 2, 0, 0,  0, 0,  0)
coeff2 = c(-1, -1,  1,  2, 0, 0, 0, 0, -1, 2, -2)
Um = 40
```

```

dat <- fmrs.gendata(nObs = n, nComp = K, nCov = D, coeff = c(coeff1, coeff2),
dispersion = sig, mixProp = piM, rho = r1, umax = Um, disFamily = 'lnorm')

res.mle <- fmrs.mle(y = dat$y, x = dat$x, delta = dat$delta, nComp = K,
disFamily = 'lnorm', initCoeff = rnorm(K*D+K), initDispersion = rep(1, K),
initmixProp = rep(1/K, K))

summary(res.mle)
res.lam <- fmrs.tunsel(y = dat$y, x = dat$x, delta = dat$delta,
nComp = K, disFamily = 'lnorm', initCoeff = c(coefficients(res.mle)),
initDispersion = dispersion(res.mle), initmixProp = mixProp(res.mle),
penFamily = 'adplasso')

summary(res.lam)

```

weights

weights method

Description

Provides the weights of fitted observations for each observation under all components of an FMRs model

Usage

```

weights(object, ...)

## S4 method for signature 'fmrsfit'
weights(object, ...)

```

Arguments

object	An <code>fmrsfit-class</code>
...	Other possible arguments

Value

A numeric array of dimension-nObs-nComp

Author(s)

Farhad Shokoohi <shokoohi@icloud.com>

Examples

```
set.seed(1980)
K = 2
D = 10
n = 500
sig = c(1, 1)
piM = c(0.4, 0.6)
r1 = 0.5
coeff1 = c( 2,  2, -1, -2, 1, 2, 0, 0,  0, 0,  0)
coeff2 = c(-1, -1,  1,  2, 0, 0, 0, 0, -1, 2, -2)
Um = 40

dat <- fmrs.gendata(nObs = n, nComp = K, nCov = D, coeff = c(coeff1, coeff2),
dispersion = sig, mixProp = piM, rho = r1, umax = Um, disFamily = 'lnorm')

res.mle <- fmrs.mle(y = dat$y, x = dat$x, delta = dat$delta, nComp = K,
disFamily = 'lnorm', initCoeff = rnorm(K*D+K), initDispersion = rep(1, K),
initmixProp = rep(1/K, K))

head(weights(res.mle))
```

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