

# Package ‘fmrs’

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**Type** Package

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

**Version** 1.14.0

**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)

**Imports** methods, survival, stats

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**License** GPL-3

**VignetteBuilder** knitr

**BugReports** <https://github.com/shokoohi/fmrs/issues>

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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](#)

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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**

```
object      An fmrsfit-class
...         Other possible arguments
```

**Value**

A numeric value

**Author(s)**

Farhad Shokoohi <[shokoohi@icloud.com](mailto: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 <a href="#">fmrsfit-class</a>
...	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	<i>dispersion method</i>
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---

### 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 <a href="#">fmrsfit-class</a>
...	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 <a href="#">fmrsfit-class</a>
...	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')
```

---

fmrs.mle

*fmrs.mle method*

---

**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  $(T, \delta)$  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

<code>y</code>	Responses (observations)
<code>delta</code>	Censoring indicator vector
<code>x</code>	Design matrix (covariates)
<code>nComp</code>	Order (Number of components) of mixture model
<code>...</code>	Other possible options
<code>disFamily</code>	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,
<code>initCoeff</code>	Vector of initial values for regression coefficients including intercepts
<code>initDispersion</code>	Vector of initial values for standard deviations
<code>initmixProp</code>	Vector of initial values for proportion of components
<code>penFamily</code>	Penalty name that is used in variable selection method. The available options are 'lasso', 'adplasso', 'mcp', 'scad', 'sica' and 'hard'.
<code>lambRidge</code>	A positive value for tuning parameter in Ridge Regression or Elastic Net
<code>nIterEM</code>	Maximum number of iterations for EM algorithm
<code>nIterNR</code>	Maximum number of iterations for Newton-Raphson algorithm
<code>conveps</code>	A positive value for avoiding NaN in computing divisions
<code>convepsEM</code>	A positive value for threshold of convergence in EM algorithm
<code>convepsNR</code>	A positive value for threshold of convergence in NR algorithm
<code>NRpor</code>	A positive interger for maximum number of searches in NR algorithm
<code>gamMixPor</code>	Proportion of mixing parameters in the penalty. The value must be in the interval [0,1]. If <code>gamMixPor = 0</code> , the penalty structure is no longer mixture.
<code>activeset</code>	A matrix of zero-one that shows which intercepts and covariates are active in the fitted fmrs model
<code>lambMCP</code>	A positive numbers for mcp's extra tuning parameter
<code>lambSICA</code>	A positive numbers for sica's extra tuning parameter
<code>cutpoint</code>	A positive value for shrinking small values of parameter estimations in the EM algorithm toward zero
<code>LambMin</code>	A positive value for minimum value of tuning parameter
<code>LambMax</code>	A positive value for maximum value of tuning parameter
<code>nLamb</code>	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  $\lambda_{max}$  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  $\lambda$  to maximize the penalized 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`: `fms.gendata()`, `fms.mle()`, `fms.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 <- fms.gendata(nObs = n, nComp = K, nCov = D, coeff = c(coeff1, coeff2),
dispersion = sig, mixProp = piM, rho = r1, umax = Um, disFamily = 'lnorm')

res.mle <- fms.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 <- fms.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

$$\boldsymbol{\beta}_k^{(m+1)} = (\mathbf{z}' \boldsymbol{\tau}_k^{(m)} \mathbf{z} + \varpi_k(\boldsymbol{\beta}_{kj}^{(m)}))^{-1} \mathbf{z}' \boldsymbol{\tau}_k^{(m)} \mathbf{T}_k^{(m)},$$

where  $\varpi_k(\boldsymbol{\beta}_{kj}^{(m)}) = \text{diag} \left( (\pi_k^{(m+1)})^\alpha \frac{p'_{\lambda_n,k}(\beta_{kj}^{(m)})}{\beta_{kj}^{(m)}} \right)$  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  $\tilde{\boldsymbol{\Psi}}_k^{(m+1)} = \tilde{\boldsymbol{\Psi}}_k^{(m)} - 0.5^\kappa [H_k^{p,(m)}]^{-1} I_k^{p,(m)}$ , 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 <a href="#">fmrsfit-class</a>
...	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 <a href="#">fmrsfit-class</a>
...	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 <a href="#">fmrsfit-class</a>
...	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 `fmrfit-class`

### Usage

```
ncov(object, ...)
```

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

### Arguments

<code>object</code>	An <code>fmrfit-class</code>
<code>...</code>	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 <- fmrsgendata(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 <a href="#">fmrsfit-class</a>
...	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 <a href="#">fmrsfit-class</a>
...	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

Displays 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</code> -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(weights(res.mle))
```

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