Package 'FMCensSkewReg'

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Title Finite Mixture of Censored Regression Models with Skewed Distributions

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Description Provides an implementation of finite mixture regression models for censored data under four distributional families: Normal (FM-NCR), Student-t (FM-TCR), Skew-Normal (FM-SNCR), and Skew-t (FM-STCR). The package enables flexible modeling of skewness and heavy tails often observed in real-world data, while explicitly accounting for censoring. Functions are included for parameter estimation via the EM algorithm, computation of standard errors, and model comparison criteria such as AIC, BIC, and EDC.
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Config/testthat/edition 3
R topics documented:
EM.skewCens.mixR
Index

2 EM.skewCens.mixR

EM.skewCens.mixR

EM Algorithm for Finite Mixture Censored Regression

Description

Fits finite mixture censored regression models under four families: Normal ("Normal"), Student-t ("T"), Skew-Normal ("SN"), and Skew-t ("ST").

Usage

```
EM.skewCens.mixR(
  cc,
  у,
  х,
  Abetas = NULL,
  sigma2 = NULL,
  shape = NULL,
  pii = NULL,
  nu = NULL,
  g = NULL,
  get.init = TRUE,
  criteria = TRUE,
  group = FALSE,
  family = "Normal",
  error = 1e-05,
  iter.max = 100,
  obs.prob = FALSE,
  kmeans.param = NULL,
  aitken = TRUE,
  IM = TRUE
)
```

Arguments

сс	Integer vector of length n; censoring indicator (1 = censored, 0 = observed).
у	Numeric response vector (univariate).
x	Numeric design matrix (n x p); include intercept column if needed.
Abetas	Optional initial regression coefficient matrix (p x g).
sigma2	Optional initial variance(s), length g.
shape	Optional initial skewness parameter(s), length g (used in SN/ST).
pii	Optional initial mixing proportions, length g, must sum to 1.
nu	Degrees of freedom for T/ST models (scalar).
g	Number of mixture components ($g \ge 1$). Required if get.init = TRUE.
get.init	Logical; if TRUE, k-means-based initialization is used.
criteria	Logical; if TRUE, returns AIC/BIC/EDC.
group	Logical; if TRUE, returns hard cluster labels.
family	One of "Normal", "T", "SN", "ST".

EM.skewCens.mixR 3

error Convergence tolerance for EM.
iter.max Maximum number of EM iterations.

obs.prob Logical; if TRUE, returns posterior membership matrix.

kmeans.param Optional list for kmeans init.

aitken Logical; use Aitken acceleration for convergence monitoring.

IM Logical; if TRUE, compute (robust) standard errors via information matrix.

Details

Left-censoring is indicated by cc[i] = 1 and replacing y[i] by the censoring point. The routine supports Normal, t, Skew-Normal, and Skew-t families with finite mixtures.

Value

A list with elements:

Abetas Estimated regression coefficients (p x g).

sigma2 Estimated variances (length g).

shape Estimated skewness parameters (length g; SN/ST).

pii Estimated mixing proportions (length g).

sd Standard errors (if IM=TRUE).

nu Estimated/used degrees of freedom (T/ST).

loglik Final log-likelihood.

loglikT Log-likelihood trace over iterations.
aic,bic,edc Information criteria (if criteria=TRUE).

iter Number of EM iterations.

n Sample size.

group Hard labels (if group=TRUE).

Examples

```
set.seed(1)
n <- 150
X <- cbind(1, runif(n), rnorm(n))</pre>
pi <- c(0.6, 0.4); nu <- 4
b1 \leftarrow c(0.5, 1.0, -1.0); sigma1 \leftarrow 1; shape1 \leftarrow 2
b2 \leftarrow c(1.0, -0.5, 0.5); sigma2 \leftarrow 2; shape2 \leftarrow 3
mu1 <- drop(X %*% b1); mu2 <- drop(X %*% b2)
draw1 <- function(i){</pre>
  a1 <- list(mu=mu1[i], sigma2=sigma1, shape=shape1, nu=nu)
  a2 <- list(mu=mu2[i], sigma2=sigma2, shape=shape2, nu=nu)</pre>
  mixsmsn::rmix(1, pi, "Skew.t", list(a1,a2), cluster=FALSE)
y0 <- vapply(seq_len(n), draw1, numeric(1))</pre>
cutoff <- unname(stats::quantile(y0, 0.20))</pre>
cc <- as.integer(y0 <= cutoff)</pre>
y <- ifelse(cc == 1, cutoff, y0)
fit <- EM.skewCens.mixR(cc, y, X, g=2, family="Normal", iter.max=50)</pre>
fit$loglik
```

Index

EM.skewCens.mixR, 2