

# Package: mixqr (via r-universe)

June 25, 2026

**Type** Package

**Title** Extensible Finite Mixtures of Quantile and Expectile Regressions

**Version** 0.2.0

**Description** An extensible expectation-maximization (EM) framework for finite mixtures of quantile regressions (clusterwise / mixture-of-experts quantile regression). A single EM substrate with an engine/extension contract carries a family of capabilities: the core free-weight mixture of Wu and Yao (2016) [doi:10.1016/j.csda.2014.04.014](https://doi.org/10.1016/j.csda.2014.04.014) -- a fast asymmetric-Laplace path and the nonparametric kernel-density EM with components constrained to have their tau-quantile equal to zero (Hall and Presnell 1999 device); expectile and M-quantile component-loss families (Newey and Powell 1987; Breckling and Chambers 1988); component-specific penalized variable selection (SCAD / adaptive-LASSO, the quantile analogue of Khalili and Chen 2007); and joint multi-quantile estimation with a shared latent classification and non-crossing component curves. Provides classification-aware standard errors (sparsity and stochastic-EM multiple imputation), multi-start estimation, component-count selection, and prediction. The companion package 'mixqrgate' adds location-varying gating.

**License** MIT + file LICENSE

**Encoding** UTF-8

**LazyData** true

**Depends** R (>= 4.1)

**Imports** quantreg, stats, graphics, utils

**Suggests** ggplot2, rqPen, testthat (>= 3.0.0), knitr, rmarkdown

**RoxygenNote** 7.3.3

**Config/testthat/edition** 3

**Config/Needs/website** pkgdown

**VignetteBuilder** knitr

**URL** <https://github.com/kvenkita/mixqr>,  
<https://kvenkita.github.io/mixqr/>

**BugReports** <https://github.com/kvenkita/mixqr/issues>

**NeedsCompilation** no

**Author** Kailas Venkitasubramanian [aut, cre, cph]

**Maintainer** Kailas Venkitasubramanian <kailasv@gmail.com>

**Repository** <https://cran.r-universe.dev>

**Date/Publication** 2026-06-25 18:29:07 UTC

**RemoteUrl** <https://github.com/cran/mixqr>

**RemoteRef** HEAD

**RemoteSha** d9c629744b446cd9fd7717c6f140cc713180cba5

## Contents

coef.mixqr . . . . .	3
confint.mixqr . . . . .	3
engine . . . . .	4
fitted.mixqr . . . . .	5
get_mixqr_engine . . . . .	5
list_mixqr_engines . . . . .	6
logLik.mixqr . . . . .	6
mixqr . . . . .	7
mixqr_control . . . . .	9
mixqr_nc . . . . .	10
mixqr_pen . . . . .	11
mixqr_select . . . . .	12
mixqr_vcontrol . . . . .	13
plot.mixqr . . . . .	14
plot.mixqr_multitau . . . . .	15
predict.mixqr . . . . .	15
predict.mixqr_multitau . . . . .	16
register_mixqr_engine . . . . .	17
selectedVars . . . . .	17
sim_mixqr_cross . . . . .	18
sim_mixqr2 . . . . .	18
sim_mixqr3 . . . . .	19
summary.mixqr . . . . .	19
vcov.mixqr . . . . .	20

<b>Index</b>	<b>21</b>
--------------	-----------

---

coef.mixqr	<i>Component coefficients of a mixqr fit</i>
------------	--

---

**Description**

Component coefficients of a mixqr fit

**Usage**

```
## S3 method for class 'mixqr'
coef(object, include_pi = FALSE, ...)
```

**Arguments**

object	A "mixqr" object.
include_pi	If TRUE, return a list with both beta and the mixing probabilities pi.
...	Unused.

**Value**

A (p+1) x m coefficient matrix (one column per component), or a list if include\_pi = TRUE.

---

confint.mixqr	<i>Confidence intervals for a mixqr fit</i>
---------------	---

---

**Description**

Wald intervals from the fitted standard errors. Requires variance = "sparsity" or "stochEM"; "stochEM" is recommended (sparsity SEs are classification-conditional and understate uncertainty).

**Usage**

```
## S3 method for class 'mixqr'
confint(object, parm, level = 0.95, ...)
```

**Arguments**

object	A "mixqr" object.
parm	Optional subset of parameter names.
level	Confidence level. Default 0.95.
...	Unused.

**Value**

A matrix of lower/upper bounds.

---

engine

*Engine ethanol-combustion data (Brinkman 1981)*

---

### Description

Measurements from a single-cylinder automobile test engine burning ethanol, exhibiting two homogeneous regimes when the equivalence ratio is regressed on nitrous-oxide concentration – the engine example of Wu & Yao (2016, Fig. 5). Derived from the public-domain ethanol data redistributed in the `lattice` package (Brinkman 1981).

### Usage

`engine`

### Format

A data frame with 88 rows and 3 columns:

**equivalence** equivalence ratio: richness of the air/ethanol mix (response).

**nox** concentration of nitrous oxide (NO and NO<sub>2</sub>) in the exhaust, normalized by engine work (predictor).

**compression** compression ratio of the engine.

### Source

Brinkman, N. D. (1981). Ethanol fuel – a single-cylinder engine study of efficiency and exhaust emissions. *SAE Transactions* 90, 1410–1427. Redistributed via the `lattice` package ethanol dataset.

### References

Wu, Q. and Yao, W. (2016). Mixtures of quantile regressions. *CSDA* 93, 162–176.

### Examples

```
fit <- mixqr(equivalence ~ nox, data = engine, tau = 0.5, m = 2)
summary(fit)
```

---

fitted.mixqr	<i>Fitted values, residuals and number of observations</i>
--------------	--

---

**Description**

fitted() returns the classified conditional tau-quantile for each observation; residuals() the corresponding residuals (or, with type = "component", the full  $n \times m$  residual matrix); nobs() the sample size.

**Usage**

```
## S3 method for class 'mixqr'
fitted(object, ...)

## S3 method for class 'mixqr'
residuals(object, type = c("hard", "component"), ...)

## S3 method for class 'mixqr'
nobs(object, ...)
```

**Arguments**

object	A "mixqr" object.
...	Unused.
type	For residuals: "hard" (classified) or "component".

**Value**

A numeric vector or matrix.

---

get_mixqr_engine	<i>Retrieve a registered mixqr engine constructor</i>
------------------	---

---

**Description**

Retrieve a registered mixqr engine constructor

**Usage**

```
get_mixqr_engine(name)
```

**Arguments**

name	Character engine name.
------	------------------------

**Value**

The constructor function.

---

list_mixqr_engines	<i>List registered mixqr engines</i>
--------------------	--------------------------------------

---

**Description**

List registered mixqr engines

**Usage**

```
list_mixqr_engines()
```

**Value**

Character vector of engine names.

---

logLik.mixqr	<i>Log-likelihood, AIC and BIC of a mixqr fit</i>
--------------	---

---

**Description**

Available for the parametric ALD engine (a genuine working likelihood); NA for the semiparametric kdEM engine, which has no global likelihood.

**Usage**

```
## S3 method for class 'mixqr'
logLik(object, ...)
```

```
## S3 method for class 'mixqr'
AIC(object, ..., k = 2)
```

```
## S3 method for class 'mixqr'
BIC(object, ...)
```

**Arguments**

object	A "mixqr" object.
...	Unused.
k	The penalty per parameter for AIC (default 2).

**Value**

A "logLik" object (logLik) or a numeric criterion (AIC, BIC).

**Description**

Estimates a finite mixture of tau-quantile regressions (clusterwise quantile regression) at a single quantile level. Two engines are available: a fast parametric asymmetric-Laplace mixture ("ald", genuine likelihood and AIC/BIC) and the kernel-density EM of Wu & Yao (2016) ("kdEM", non-parametric component error densities constrained to have their tau-th quantile = 0).

**Usage**

```

mixqr(
  formula,
  data,
  tau = 0.5,
  m = 2L,
  family = c("quantile", "expectile", "mquantile"),
  engine = "ald",
  error_density = c("unequal", "equal"),
  init = c("ald", "kmeans", "random", "manual"),
  nstart = 20L,
  control = mixqr_control(),
  weights = NULL,
  manual_init = NULL,
  variance = c("none", "sparsity", "stochEM"),
  vcontrol = mixqr_vcontrol(),
  ...
)

```

**Arguments**

formula	A model formula $y \sim x_1 + x_2$ (intercept implied).
data	A data frame.
tau	Quantile level in (0, 1). Default 0.5.
m	Number of mixture components ( $\geq 1$ ). Default 2.
family	Component-loss family: "quantile" (default; check loss, the Wu & Yao model), "expectile" (asymmetric least squares, Newey & Powell 1987 – a smooth, crossing-free location device), or "mquantile" (asymmetric Huber, Breckling & Chambers 1988 – a robust expectile/quantile blend). A non-quantile family selects the matching engine.
engine	"ald" (default) or "kdEM", or the name of a custom engine registered via <a href="#">register_mixqr_engine()</a> (e.g. "expectile", "mquantile").
error_density	For "kdEM": "unequal" (per-component densities, eq. 2.5) or "equal" (pooled density, eq. 2.8).

<code>init</code>	Initialisation strategy: "ald" (default; ALD pre-fit seeds the kdEM engine), "kmeans", "random", or "manual".
<code>nstart</code>	Number of multi-start initialisations (the mixture likelihood is multimodal). Default 20.
<code>control</code>	A <code>mixqr_control()</code> list.
<code>weights</code>	Optional prior observation weights.
<code>manual_init</code>	Optional $n \times m$ responsibility matrix for <code>init = "manual"</code> .
<code>variance</code>	Standard-error method: "none" (default), "sparsity" (eq. 3.3) or "stochEM" (Algorithm 3.1 multiple imputation).
<code>vcontrol</code>	A <code>mixqr_vcontrol()</code> list for <code>variance = "stochEM"</code> .
<code>...</code>	Reserved for engine extensions; currently unused.

### Value

An object of class "mixqr".

### Bias under asymmetric errors (Wu & Yao sec.6)

The semiparametric (kdEM) estimator solves estimating equations whose score  $I(a \leq \theta) - \tau$  is not orthogonal to the nuisance tangent space of the unknown error densities, so it can be BIASED when component error densities are asymmetric and the clusters have imbalanced overlap (Wu & Yao 2016, sec.6, Fig.6). Watch `fit$diagnostics$overlap` (responsibility entropy; larger = more overlap = more bias risk) and cross-check against the parametric ald engine. Well-separated clusters and (near-)symmetric errors are the safe regime.

### Standard errors

Sparsity SEs (`variance = "sparsity"`, eq.3.3) are CONDITIONAL on the fitted classification and understate uncertainty (they are flagged as such by `summary()`). Use `variance = "stochEM"` (the stochastic-EM multiple-imputation estimator, Algorithm 3.1) for inference; it propagates classification and mixing uncertainty.

### References

Wu, Q. and Yao, W. (2016). Mixtures of quantile regressions. *CSDA* 93, 162–176.

### Examples

```
set.seed(1)
d <- sim_mixqr2(n = 300)
fit <- mixqr(y ~ x, data = d, tau = 0.5, m = 2, engine = "ald", nstart = 10)
fit
coef(fit)
```

---

mixqr_control	<i>Control parameters for mixqr()</i>
---------------	---------------------------------------

---

## Description

Control parameters for `mixqr()`

## Usage

```

mixqr_control(
  tol = 1e-06,
  maxit = 500L,
  reltol = TRUE,
  bandwidth = NULL,
  kde_grid = 512L,
  label_order = "slope",
  distinct_tol = 0.001,
  trace = FALSE,
  seed = NULL
)

```

## Arguments

tol	Convergence tolerance on the summed absolute parameter change (Wu & Yao 2016, p. 164). Default 1e-6.
maxit	Maximum EM iterations. Default 500.
reltol	Logical; if TRUE use the relative criterion $\sum  d\theta  / (\sum  \theta  + \text{eps}) < \text{tol}$ (recommended because pi and beta differ in scale). Default TRUE.
bandwidth	Optional numeric KDE bandwidth override (kdEM engine). If NULL, Silverman's rule $1.06 * \text{sd} * n^{(-1/5)}$ is used per component.
kde_grid	Number of grid points for stored KDE evaluation / plotting. Default 512.
label_order	Label-switching constraint: "slope" (ascending first slope coordinate, default – aligned with the distinct-slope identifiability of Wu & Yao Thm 2.1), "intercept", or "slope:k" for the k-th slope.
distinct_tol	RELATIVE threshold below which component slope vectors are flagged as near-coincident (near-violation of Wu & Yao Thm 2.1), measured as $\ b_a - b_b\  / (\ b_a\  + \ b_b\ )$ . Default 1e-3.
trace	Logical; print iteration progress. Default FALSE.
seed	Optional integer RNG seed for reproducible multi-start.

## Value

A list of class "mixqr\_control".

---

mixqr_nc	<i>Fit a joint multi-tau mixture of quantile regressions (non-crossing, shared labels)</i>
----------	--

---

### Description

Estimates a finite mixture of quantile regressions at a *vector* of quantile levels  $\tau$  jointly, with one latent classification shared across all levels (a coupled E-step that pools the per-level component likelihoods) and guaranteed non-crossing of the within-component quantile curves. This closes the two problems Wu & Yao (2016, sec.5) leave open: within-component crossing and cross-tau classification ambiguity.

### Usage

```

mixqr_nc(
  formula,
  data,
  m = 2L,
  tau = c(0.25, 0.5, 0.75),
  noncrossing = c("rearrange", "none"),
  class_coupling = c("pool", "median"),
  nstart = 10L,
  control = mixqr_control(),
  weights = NULL
)

```

### Arguments

`formula`, `data`, `m`, `nstart`, `control`, `weights`  
 As in `mixqr()`.

`tau` Increasing vector of quantile levels (length  $\geq 2$ ).

`noncrossing` "rearrange" (default; monotone-rearrange the fitted per-tau quantiles within each component, Chernozhukov et al. 2010) or "none" (legacy per-tau fits, which may cross – for diagnosis/comparison).

`class_coupling` How the shared E-step pools over  $\tau$ : "pool" (default; geometric mean, weights  $1/L$ ) or "median" (classify on the median level only).

### Value

An object of class `c("mixqr_multitau", "mixqr")` with a coefficient array `coefficients` ( $[p+1, m, L]$ ), the shared posterior/classification, `mix_prop`, the  $\tau$  grid, and a crossing diagnostic (raw crossings, 0 after rearrangement).

### References

Wu, Q. and Yao, W. (2016). Mixtures of quantile regressions. *CSDA* 93, 162–176. Chernozhukov, V., Fernandez-Val, I. and Galichon, A. (2010). Quantile and probability curves without crossing. *Econometrica* 78, 1093–1125.

**Examples**

```
d <- sim_mixqr_cross(n = 160, seed = 1)
fit <- mixqr_nc(y ~ x, data = d, m = 2, tau = c(0.1, 0.25, 0.5, 0.75, 0.9))
fit$crossing      # raw crossings -> 0 after rearrangement
predict(fit)[1, , ] # non-crossing quantiles for observation 1
```

---

mixqr\_pen

*Fit a penalized (variable-selecting) mixture of quantile regressions*


---

**Description**

Component-specific penalized variable selection for the `mixqr()` model: each latent component gets its own sparse slope vector, a covariate active in one quantile-regression cluster and dropped in another. The weighted check-loss M-step gains a SCAD / adaptive-LASSO / LASSO / MCP penalty (intercept free); the penalty strength  $\lambda$  is selected by a mixture BIC over a path, and components whose mixing weight falls below  $\pi_{\min}$  are pruned.

**Usage**

```
mixqr_pen(
  formula,
  data,
  tau = 0.5,
  m = 2L,
  penalty = c("SCAD", "aLASSO", "LASSO", "MCP"),
  lambda = NULL,
  nlambda = 40L,
  a = 3.7,
  alasso_gamma = 1,
  penalty_factor = NULL,
  pi_min = 0.02,
  nstart = 10L,
  control = mixqr_control(),
  weights = NULL
)
```

**Arguments**

formula, data, tau, m, nstart, control, weights

As in `mixqr()`.

penalty One of "SCAD" (default), "aLASSO", "LASSO", "MCP". "aLASSO" is implemented as a LASSO with data-driven penalty factors  $1 / |b_{\text{pilot}}|^{\gamma}$  from a responsibility-weighted unpenalised pilot.

lambda Optional fixed penalty (skips selection). If NULL, a path of nlambda values is built and tuned by BIC.

nlambda	Path length when lambda is NULL. Default 40.
a	SCAD/MCP concavity. Default 3.7.
alasso_gamma	Adaptive-LASSO weight exponent (default 1).
penalty_factor	Optional length-p per-slope multipliers (0 never penalises a covariate). NULL penalises all slopes equally.
pi_min	Components with mixing weight below this are pruned. Default 0.02.

### Value

A "mixqr" object with an extra \$selection slot (active covariates per component, chosen lambda, the BIC path, and df).

### References

Khalili, A. and Chen, J. (2007). Variable selection in finite mixture of regression models. *JASA* 102, 1025–1038. Sherwood, B., Li, S. and Maidman, A. (2025). rqPen. *R Journal*.

### Examples

```
set.seed(1)
n <- 250; p <- 8; x <- matrix(rnorm(n * p), n)
colnames(x) <- paste0("x", 1:p)
z <- rbinom(n, 1, 0.5)
y <- ifelse(z == 0, 1 + 2 * x[, 1], -1 + 2 * x[, 2]) + rnorm(n)
d <- data.frame(y = y, x)
fit <- mixqr_pen(y ~ ., data = d, tau = 0.5, m = 2, penalty = "SCAD")
selectedVars(fit)
```

---

mixqr\_select

*Select the number of mixture components*

---

### Description

Fits mixqr over a range of component counts and scores them by an information criterion or cross-validated check loss.

### Usage

```
mixqr_select(
  formula,
  data,
  tau = 0.5,
  m = 1:4,
  criterion = c("BIC", "AIC", "cv"),
  engine = "ald",
  error_density = c("unequal", "equal"),
```

```

    folds = 5L,
    weights = NULL,
    nstart = 20L,
    control = mixqr_control()
  )

```

### Arguments

formula, data, tau, weights	Passed to <code>mixqr()</code> .
m	Integer vector of component counts to try. Default 1:4.
criterion	"BIC" (default), "AIC", or "cv".
engine	Engine to use ("ald" or "kdEM"). For "AIC"/"BIC" the score is always computed with the ALD working likelihood.
error_density	For the kdEM engine.
folds	Number of CV folds when criterion = "cv". Default 5.
nstart	Multi-starts per fit.
control	A <code>mixqr_control()</code> list.

### Details

For criterion `%in% c("AIC", "BIC")` the score uses the parametric ALD working-likelihood (the semiparametric kdEM engine has no global likelihood; Wu & Yao 2016, p. 164). Note that AIC/BIC for the *number of mixture components* is heuristic: testing  $m$  vs  $m+1$  places a component on the parameter-space boundary ( $\pi_{i,j} = 0$ ), so the usual penalty asymptotics do not hold (McLachlan & Peel 2000, ch. 6; Chen, Chen & Kalbfleisch 2001). The "cv" criterion (cross-validated held-out check loss) avoids the likelihood entirely and works for either engine.

### Value

A list with `table` (data frame of scores by  $m$ ), `best` (chosen  $m$ ), `criterion`, and `fit` (the chosen model refit on all data).

---

<code>mixqr_vcontrol</code>	<i>Control parameters for stochastic-EM variance estimation (Algorithm 3.1)</i>
-----------------------------	---

---

### Description

Control parameters for stochastic-EM variance estimation (Algorithm 3.1)

### Usage

```
mixqr_vcontrol(B = 500L, burnin = 50L, thin = 1L, seed = NULL)
```

**Arguments**

B	Number of imputation draws. Default 500.
burnin	Burn-in draws discarded before accumulation. Default 50.
thin	Thinning interval. Default 1.
seed	Optional integer RNG seed.

**Value**

A list of class "mixqr\_vcontrol".

---

plot.mixqr	<i>Plot a mixqr fit</i>
------------	-------------------------

---

**Description**

Draws the data coloured by hard classification with the component quantile-regression lines, and a panel of the estimated component error densities (mirroring Wu & Yao Figs. 1, 3–5).

**Usage**

```
## S3 method for class 'mixqr'
plot(x, which = c("both", "fit", "density"), ...)
```

**Arguments**

x	A "mixqr" object.
which	"both" (default), "fit", or "density".
...	Passed to <code>graphics::plot()</code> .

**Value**

Invisibly x.

---

plot.mixqr\_multitau     *Plot the non-crossing component quantile curves of a multi-tau fit*

---

### Description

Overlays, for each component, the (rearranged, non-crossing) fitted quantile curves at every tau against the first covariate.

### Usage

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

### Arguments

x                    A `mixqr_nc()` fit.  
 ...                  Passed to `graphics::matplot()`.

### Value

x, invisibly.

---

predict.mixqr             *Predict from a mixqr fit*

---

### Description

Predict from a mixqr fit

### Usage

```
## S3 method for class 'mixqr'
predict(
  object,
  newdata = NULL,
  type = c("quantile", "quantile_byclass", "posterior", "class", "density"),
  ...
)
```

**Arguments**

object	A "mixqr" object.
newdata	Optional data frame. If omitted, the training data are used.
type	One of "quantile" (per-component conditional tau-quantiles, an $n \times m$ matrix), "quantile_byclass" (the conditional tau-quantile of each point's most likely component, a vector), "posterior" (responsibilities $p_{ij}$ ), "class" (hard labels), or "density" (the component error-density objects).
...	Unused.

**Value**

A matrix, vector, or list depending on type.

**Classifying genuinely new x**

"posterior", "class", and "quantile\_byclass" require the RESPONSE in newdata, because component membership is inferred from the residual densities. This core therefore cannot assign a class to a covariate vector  $x$  alone; covariate-based gating (membership as a function of  $x$ ) is provided by the location-varying-gating extension (QMM sub-project 03).

---

predict.mixqr\_multitau

*Predict non-crossing component quantiles from a multi-tau fit*

---

**Description**

Predict non-crossing component quantiles from a multi-tau fit

**Usage**

```
## S3 method for class 'mixqr_multitau'
predict(object, newdata = NULL, ...)
```

**Arguments**

object	A <code>mixqr_nc()</code> fit.
newdata	Optional data frame; defaults to the training design.
...	Unused.

**Value**

An  $[n, m, L]$  array of fitted quantiles. With `noncrossing = "rearrange"` each component's  $L$  quantiles are sorted to be nondecreasing in tau (Chernozhukov et al. 2010), guaranteeing no crossing.

---

register\_mixqr\_engine *Register a mixqr EM engine*

---

### Description

An engine is a constructor function(`tau, m, control`) returning a list of closures with the FROZEN signatures documented in `mixqr_engine_contract()`. Sub-projects 03–06 of the QMM suite ship engines this way without forking the driver.

### Usage

```
register_mixqr_engine(name, constructor)
```

### Arguments

name	Character engine name (e.g. "ald", "kdEM").
constructor	Function ( <code>tau, m, control</code> ) -> <code>mixqr_engine</code> (a list of closures: <code>estep, mstep_pi, mstep_beta, update_density, loglik, npar</code> , plus name).

### Value

Invisibly TRUE.

---

selectedVars *Active (selected) covariates per component of a penalized mixqr fit*

---

### Description

Active (selected) covariates per component of a penalized mixqr fit

### Usage

```
selectedVars(object)
```

### Arguments

object	A <code>mixqr_pen()</code> fit.
--------	---------------------------------

### Value

A named list (one entry per component) of selected covariate names.

---

sim_mixqr_cross	<i>Simulate a two-component mixture whose per-tau fits cross (for O4 validation)</i>
-----------------	--

---

### Description

One component is well-behaved; the other has a near-flat slope and strong heteroscedasticity (error scale growing across the design), so its independently fitted per-tau quantile lines cross in finite samples – exactly the within-component crossing Wu & Yao (2016, sec.5) flag as common "when the sample size is small." Use to demonstrate that noncrossing = "none" crosses while "rearrange" does not. (A valid location-scale model has non-crossing true quantiles; crossing here is the finite-sample estimation artefact the method repairs.)

### Usage

```
sim_mixqr_cross(n = 160, seed = NULL)
```

### Arguments

n	Sample size (default 160; crossing is a small-sample phenomenon).
seed	Optional RNG seed.

### Value

A data frame with columns y, x, and the true label z.

---

sim_mixqr2	<i>Simulate the Wu &amp; Yao two-component design (eq. 4.1-4.2)</i>
------------	---

---

### Description

$Y = 10 - 10x + e$  (component 1) or  $Y = -10 + 10x + e$  (component 2), with  $\pi_i = (0.5, 0.5)$ ,  $x \sim U(0, 1)$ , and a bimodal error  $e \sim 0.5 N(-1, 1^2) + 0.5 N(2, 2^2)$  (median 0; median regression,  $\tau = 0.5$ ).

### Usage

```
sim_mixqr2(n = 300L, pi1 = 0.5)
```

### Arguments

n	Sample size.
pi1	Probability of component 1.

### Value

A data frame with columns x, y, z (true component label).

---

sim_mixqr3	<i>Simulate the Wu &amp; Yao three-component design (eq. 4.3-4.4)</i>
------------	---

---

**Description**

Three components with  $\pi = (1/3, 1/3, 1/3)$ , covariates  $x_1, x_2 \sim U(0, 1)$ , means  $-20 x_1 - 20 x_2$ ,  $0, 20 x_1 + 20 x_2$ , and shifted-lognormal errors  $\exp(N(1, \sigma^2)) - \exp(1)$  with  $\sigma^2$  in  $\{1, 0.5, 0.25\}$  (each median 0).

**Usage**

```
sim_mixqr3(n = 300L)
```

**Arguments**

n                      Sample size.

**Value**

A data frame with columns  $x_1, x_2, y, z$ .

---

summary.mixqr	<i>Summarize a mixqr fit</i>
---------------	------------------------------

---

**Description**

Prints, per component, the coefficient estimates with standard errors, z-values and p-values (when a variance method was used), the mixing probabilities, and the separability / overlap diagnostics. Sparsity standard errors are flagged as classification-conditional.

**Usage**

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

**Arguments**

object                  A "mixqr" object.  
...                      Unused.

**Value**

An object of class "summary.mixqr".

---

`vcov.mixqr`*Variance-covariance matrix of a mixqr fit*

---

**Description**

The estimated parameter covariance (requires a fitted variance method).

**Usage**

```
## S3 method for class 'mixqr'  
vcov(object, ...)
```

**Arguments**

<code>object</code>	A "mixqr" object.
<code>...</code>	Unused.

**Value**

The variance-covariance matrix.

# Index

- \* **datasets**
  - engine, 4
  
- AIC.mixqr (logLik.mixqr), 6
  
- BIC.mixqr (logLik.mixqr), 6
  
- coef.mixqr, 3
- confint.mixqr, 3
  
- engine, 4
  
- fitted.mixqr, 5
  
- get\_mixqr\_engine, 5
- graphics::matplot(), 15
- graphics::plot(), 14
  
- list\_mixqr\_engines, 6
- logLik.mixqr, 6
  
- mixqr, 7
- mixqr(), 9–11, 13
- mixqr\_control, 9
- mixqr\_control(), 8, 13
- mixqr\_engine\_contract(), 17
- mixqr\_nc, 10
- mixqr\_nc(), 15, 16
- mixqr\_pen, 11
- mixqr\_pen(), 17
- mixqr\_select, 12
- mixqr\_vcontrol, 13
- mixqr\_vcontrol(), 8
  
- nobs.mixqr (fitted.mixqr), 5
  
- plot.mixqr, 14
- plot.mixqr\_multitau, 15
- predict.mixqr, 15
- predict.mixqr\_multitau, 16
  
- register\_mixqr\_engine, 17
  
- register\_mixqr\_engine(), 7
- residuals.mixqr (fitted.mixqr), 5
  
- selectedVars, 17
- sim\_mixqr2, 18
- sim\_mixqr3, 19
- sim\_mixqr\_cross, 18
- summary.mixqr, 19
  
- vcov.mixqr, 20