

# Package ‘Meanimiles’

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**Title** Estimation of Meanimiles

**Version** 0.1.0

**Description** Provides a comprehensive suite of estimation tools for meanimiles, a general class of (risk) functionals. This package includes nonparametric estimators for univariate meanimile evaluation, copula-based estimation for portfolio risk aggregation (full parametric, semiparametric, and nonparametric), and novel estimators for meanimiles in regression settings. Following the articles D. Debrauwer, I. Gijbels, and K. Herrmann (2025)  [<doi:10.1214/25-EJS2391>](https://doi.org/10.1214/25-EJS2391), D. Debrauwer and I. Gijbels (2026)  [<doi:10.1007/s00184-026-01022-9>](https://doi.org/10.1007/s00184-026-01022-9).

**License** GPL (>= 3)

**Encoding** UTF-8

**RoxygenNote** 7.3.3

**URL** <https://github.com/DieterDebrauwer/Meanimiles>

**BugReports** <https://github.com/DieterDebrauwer/Meanimiles/issues>

**Depends** R (>= 4.5.0)

**Imports** copula (>= 1.1.7), fitdistrplus (>= 1.2.6), stats

**Suggests** knitr, rmarkdown

**NeedsCompilation** no

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ConditionalCDF	<i>Estimate conditional cumulative distribution function</i>
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### Description

Computes a doubly smoothed Nadaraya-Watson estimator for the conditional cumulative distribution function.

### Usage

ConditionalCDF(Y\_eval, X\_eval, Y\_train, X\_train, h\_y, h\_x)

### Arguments

Y_eval	A numeric vector of Y values where the CDF should be evaluated.
X_eval	A numeric matrix of X values where the CDF should be evaluated.
Y_train	A numeric vector of training Y values.
X_train	A numeric matrix of training X values.
h_y	A numeric scalar representing the bandwidth for the response variable Y.
h_x	A numeric vector representing the bandwidths for the covariates X.

### Value

A numeric vector of estimated CDF probabilities.

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MeanimileEstimator      *Estimate a univariate meanimile*


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### Description

Calculates the empirical meanimile for a given (univariate) data sample, weight function, and loss derivative.

### Usage

```
MeanimileEstimator(
  d_tau,
  tau,
  lossDerivative,
  delta = NULL,
  sample,
  grid_size = 1000
)
```

### Arguments

d_tau	A function representing the density function of the distributional weight function $D_{\tau}$ .
tau	A numeric risk level/index (e.g., 0.95). Argument of d_tau function.
lossDerivative	Function defining the derivative of the loss function.
delta	Optional numeric parameter for the loss function (default:NULL).
sample	A numeric vector of univariate data.
grid_size	An integer defining the resolution of the fallback grid (default: 1000).

### Value

A numeric scalar representing the estimated meanimile.

### References

D. Debrauwer, I. Gijbels, and K. Herrmann. (2026) On a general class of functionals: Statistical inference and application to risk measures *Electronic Journal of Statistics* doi:<https://doi.org/10.1214/25-EJS2391>

### Examples

```
set.seed(123)
x <- rnorm(100)
d_tau_expectile <- function(u, tau) {
  1
}
lossDerivativeExpectiles <- function(x, c, delta) {
```

```

-2 * ifelse(x > c, delta, 1 - delta) * (x - c)
}
MeanimileEstimator(d_tau_expectile, tau = 0, lossDerivativeExpectiles,
  delta = 0.95, sample = x)

```

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MeanimileLinearRegression

*Estimate linear conditional meanimile*


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### Description

Estimates the parameters of a linear conditional meanimile model using a two-fold sample splitting procedure to mitigate bias from the conditional CDF estimation. This function acts as a unified wrapper, allowing the user to seamlessly switch between the square loss formulation and the general loss formulation.

### Usage

```

MeanimileLinearRegression(
  X,
  Y,
  d_tau,
  tau,
  h_x,
  h_y,
  loss_type = c("square", "general"),
  lossDerivative = NULL,
  delta = NULL
)

```

### Arguments

X	A numeric matrix of covariates (n x d).
Y	A numeric vector of the response variable.
d_tau	A function representing the density function of the distributional weight function D_tau.
tau	A numeric risk level/index. Argument of d_tau function.
h_x	A numeric vector of bandwidths for the covariates X.
h_y	A numeric scalar bandwidth for the response variable Y.
loss_type	A character string specifying the loss formulation. Must be either "square" (default) or "general".
lossDerivative	A function defining the derivative of the loss function. Required only if loss_type = "general".
delta	Optional numeric parameter for the loss function. Used only if loss_type = "general".

**Value**

A numeric vector of estimated regression coefficients (including the intercept).

**Examples**

```

set.seed(123)
n <- 400
X <- matrix(rnorm(n * 2), ncol = 2)
Y <- 0.9 * X[, 1] + 0.5 * X[, 2] + rnorm(n)
d_tau_minvar <- function(u, tau) {
  (tau + 1) * u^tau
}
lossDerivativeExpectiles <- function(x, c, delta) {
  -2 * ifelse(x > c, delta, 1 - delta) * (x - c)
}

# 1. Square Loss
MeanimileLinearRegression(X, Y,
  d_tau = d_tau_minvar, tau = 2,
  h_x = c(0.4, 0.4), h_y = 0.4, loss_type = "square"
)

# 2. General Loss
MeanimileLinearRegression(X, Y,
  d_tau = d_tau_minvar, tau = 2,
  h_x = c(0.4, 0.4), h_y = 0.4, loss_type = "general",
  lossDerivative = lossDerivativeExpectiles, delta = 0.9
)

```

---

MeanimileNonparametricRegression

*Estimate nonparametric conditional meanimile*

---

**Description**

Estimates a conditional meanimile in nonparametric way. This function acts as a unified wrapper, allowing the user to seamlessly switch between the square loss formulation and the general loss formulation.

**Usage**

```

MeanimileNonparametricRegression(
  X,
  Y,
  X0,
  d_tau,
  tau,
  h_x_loc,

```

```

    h_x_cdf,
    h_y_cdf,
    loss_type = c("square", "general"),
    lossDerivative = NULL,
    delta = NULL
  )

```

### Arguments

X	A numeric matrix of covariates (n x d).
Y	A numeric vector of the response variable.
X0	A numeric matrix of evaluation points (m x d) where the curve should be estimated.
d_tau	A function representing the density function of the distributional weight function D_tau.
tau	A numeric risk level/index. Argument of d_tau function.
h_x_loc	A numeric vector of local spatial bandwidths for the covariates X.
h_x_cdf	A numeric vector of bandwidths for the covariates X used in the CDF estimation.
h_y_cdf	A numeric scalar bandwidth for the response variable Y used in the CDF estimation.
loss_type	Character string specifying the loss formulation ("square" or "general").
lossDerivative	A function defining the derivative of the loss function. Required for "general".
delta	Optional numeric parameter for the loss function. Only for "general".

### Value

A numeric vector of length m, containing the estimated meanimile predictions at each point in X0.

### Examples

```

set.seed(123)
n <- 300

# 1. Simulate 1D Non-linear Data (Sine Wave)
X <- matrix(runif(n, -2, 2), ncol = 1)
Y <- sin(X[, 1] * 1.5) + rnorm(n, sd = 0.4)

# 2. Define Evaluation Grid
X0 <- matrix(seq(-2, 2, length.out = 50), ncol = 1)

# 3. Define Functions
d_tau_minvar <- function(u, tau) {
  (tau + 1) * u^tau
}
loss_deriv_expectile <- function(x, c, delta) {
  -2 * ifelse(x > c, delta, 1 - delta) * (x - c)
}

```

```
# 4. Fit Square Loss Nonparametric
preds_sq <- MeanimileNonparametricRegression(
  X = X, Y = Y, X0 = X0, d_tau = d_tau_minvar, tau = 2,
  h_x_loc = 0.4, h_x_cdf = 0.4, h_y_cdf = 0.4, loss_type = "square"
)

# 5. Fit General Loss Nonparametric (Expectile)
preds_gen <- MeanimileNonparametricRegression(
  X = X, Y = Y, X0 = X0, d_tau = d_tau_minvar, tau = 2,
  h_x_loc = 0.4, h_x_cdf = 0.4, h_y_cdf = 0.4, loss_type = "general",
  lossDerivative = loss_deriv_expectile, delta = 0.9
)
```

---

PortfolioMeanimile      *Estimate portfolio meanimile*

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## Description

Estimates the risk of an aggregated portfolio using the meanimile framework. Supports parametric and nonparametric modeling of marginal distributions and copula.

## Usage

```
PortfolioMeanimile(
  data,
  weights,
  d_tau,
  tau,
  lossDerivative,
  delta = NULL,
  copula_obj = NULL,
  margin_dists = NULL,
  N = 1e+05,
  grid_size = 1000
)
```

## Arguments

<code>data</code>	A numeric matrix or data.frame (n observations x d assets).
<code>weights</code>	A numeric vector of portfolio weights summing to 1.
<code>d_tau</code>	A function representing the density function of the distributional weight function <code>D_tau</code> .
<code>tau</code>	A numeric risk level/index (e.g., 0.95). Argument of <code>d_tau</code> function.
<code>lossDerivative</code>	Function defining the derivative of the loss function.
<code>delta</code>	Optional numeric parameter for the loss function (default:NULL).

<code>copula_obj</code>	An unfitted 'copula' object (e.g., <code>copula::gumbelCopula(dim=3)</code> ). If NULL, an Empirical Beta Copula is used.
<code>margin_dists</code>	A character vector of base R distribution root names for <code>fitdistrplus::fitdist</code> (e.g., <code>c("norm", "exp", "weibull")</code> ). If NULL, empirical margins are used.
<code>N</code>	Integer. Number of Monte Carlo simulations to draw (default: 100 000).
<code>grid_size</code>	An integer defining the resolution of the fallback grid (default: 1000).

**Value**

A numeric scalar representing the estimated meanimile.

**References**

D. Debrauwer and I. Gijbels (2026) Copula-based estimation of meanimiles of aggregated risks. *Metrika* doi:<https://doi.org/10.1007/s00184-026-01022-9>

**Examples**

```
gumbel.cop <- copula::gumbelCopula(10 / 9, dim = 3)
myMvdX <- copula::mvdc(
  copula = gumbel.cop, margins = c("norm", "norm", "exp"),
  paramMargins = list(
    list(mean = 0, sd = 1),
    list(mean = 0, sd = 1),
    list(rate = 1)
  )
)
X <- copula::rMvdc(500, myMvdX)

d_tau_expectile <- function(u, tau) {
  1
}
lossDerivativeExpectiles <- function(x, c, delta) {
  -2 * ifelse(x > c, delta, 1 - delta) * (x - c)
}
gumbelCopula <- copula::gumbelCopula(dim = 3)
margin_dists <- c("norm", "norm", "exp")
weights <- c(1 / 3, 1 / 3, 1 / 3)
PortfolioMeanimile(
  data = X, weights, d_tau_expectile, tau = 0, lossDerivativeExpectiles, delta = 0.95,
  copula_obj = gumbelCopula, margin_dists = margin_dists, N = 100000, grid_size = 1000
)
```

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