

# Package: aftPenCDA (via r-universe)

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**Title** Estimating Penalized AFT Models via Coordinate Descent

**Version** 0.1.1

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**Description** Provides penalized accelerated failure time (AFT) model estimation for right-censored and partly interval-censored survival data using induced smoothing and coordinate descent algorithms. Supported penalties include broken adaptive ridge (BAR), LASSO, adaptive LASSO, and SCAD. Core estimation routines are implemented in 'C++' via 'Rcpp' and 'RcppArmadillo' for computational efficiency. The methodology is related to Zeng and Lin (2008) <doi:10.1093/biostatistics/kxm034>, Xu et al. (2010) <doi:10.1002/sim.2576>, Dai et al. (2018) <doi:10.1016/j.jmva.2018.08.007>, and Choi et al. (2025) <doi:10.48550/arXiv.2503.11268>.

**URL** <https://github.com/seonsy/aftPenCDA>

**BugReports** <https://github.com/seonsy/aftPenCDA/issues>

**License** GPL-3

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**Imports** Rcpp, stats

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

**VignetteBuilder** knitr

**Config/testthat/edition** 3

**Depends** R (>= 3.5)

**LazyData** true

**NeedsCompilation** yes

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aftpen	<i>Penalized AFT estimation for right-censored data</i>
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## Description

Fits a penalized accelerated failure time (AFT) model for right-censored survival data using induced smoothing and a penalized coordinate descent algorithm. Supported penalties include BAR, LASSO, adaptive LASSO, and SCAD.

## Usage

```
aftpen(
  dt,
  lambda,
  se,
  type = c("BAR", "LASSO", "ALASSO", "SCAD"),
  r = 3.7,
  eps = 1e-08,
  max.iter = 100
)
```

## Arguments

dt	A data frame whose first two columns are y and d, where y is the observed survival or censoring time and d is the event indicator (1 for observed event and 0 for right-censoring). The remaining columns are treated as covariates.
lambda	A nonnegative tuning parameter controlling the amount of penalization.
se	A character string specifying the variance estimation method. <b>"CF"</b> Closed-form (plug-in) variance estimator based on the analytic expression of the estimating function (see Equation (6)).

	"ZL" Perturbation-based variance estimator using the resampling approach of Zeng and Lin (2008).
type	Penalty type. One of "BAR", "LASSO", "ALASSO", or "SCAD".
r	A positive tuning constant used in the SCAD penalty. Ignored unless type = "SCAD". The default is 3.7.
eps	Convergence tolerance for the outer penalized coordinate descent iterations. The default is 1e-8.
max.iter	Maximum number of iterations for the outer penalized coordinate descent algorithm. The default is 100.

### Details

The function first calls the Rcpp backend `is_aft_cpp()` to obtain an initial estimator together with gradient and Hessian information. A Cholesky-based transformation is then applied, followed by coordinate-wise penalized updates.

For type = "BAR", the update uses the internal `BAR_threshold()` operator. For type = "LASSO", "ALASSO", and "SCAD", soft-thresholding-based updates are used.

### Value

A list containing the following components:

- beta: final coefficient estimate on the original scale.

### Examples

```
n = 100
p = 10
beta0 = c(rep(1,3),rep(0,7))
x = matrix(rnorm(n * p), n, p)
T = exp(x%*%beta0 + rnorm(n))
C = rexp(n, rate = exp(-2))
d = 1*(T<C)
y = pmin(T,C)
dt = data.frame(y,d,x)
fit <- aftpen(dt, lambda = 0.1, se = "CF", type = "BAR")
fit$beta
```

### Description

Fits a penalized accelerated failure time (AFT) model for partly interval censored survival data using induced smoothing and a penalized coordinate descent algorithm. Supported penalties include BAR, LASSO, adaptive LASSO, and SCAD.

**Usage**

```
aftpen_pic(
  dt,
  lambda,
  se,
  type = c("BAR", "LASSO", "ALASSO", "SCAD"),
  r = 3.7,
  eps = 1e-08,
  max.iter = 100
)
```

**Arguments**

dt	A data frame containing PIC survival data. It must include L, R, delta, and id, where L and R define the observation interval, delta indicates whether the failure time is exactly observed (1) or censored (0), and id is the cluster identifier. The remaining columns are treated as covariates.
lambda	A nonnegative tuning parameter controlling the amount of penalization.
se	A character string specifying the variance estimation method. <b>"CF"</b> Closed-form (analytic plug-in) variance estimator based on the estimating function. <b>"ZL"</b> Perturbation-resampling variance estimator following Zeng and Lin (2008).
type	Penalty type. One of "BAR", "LASSO", "ALASSO", or "SCAD".
r	A positive tuning constant used in the SCAD penalty. Ignored unless type = "SCAD". The default is 3.7.
eps	Convergence tolerance for the outer penalized coordinate descent iterations. The default is 1e-8.
max.iter	Maximum number of iterations for the outer penalized coordinate descent algorithm. The default is 100.

**Details**

The input data `dt` are assumed to arise from clustered partly interval-censored survival data with informative cluster sizes.

Specifically, observations are grouped into clusters, where each cluster shares a latent frailty variable that affects both the failure times and the cluster size. As a result, the number of observations within each cluster is not fixed but depends on the underlying frailty, leading to an informative cluster size structure.

For each subject, the failure time follows an accelerated failure time (AFT) model, and the observed data consist of an interval  $(L, R)$  together with an indicator `delta`. When  $L = R$  (i.e., `delta = 1`), the observation is exact; otherwise (`delta = 0`), the observation is censored and may correspond to left-censoring, right-censoring, or interval-censoring depending on the relationship between the true failure time and the inspection times.

The function first calls the Rcpp backend `is_aftpic_cpp()` to obtain an initial estimator together with gradient and Hessian information. A Cholesky-based transformation is then applied, followed by coordinate-wise penalized updates.

For type = "BAR", the update uses the internal BAR\_threshold() operator. For "LASSO", "ALASSO", and "SCAD", soft-thresholding-based updates are used.

### Value

A list containing the following components:

- beta: final coefficient estimate on the original scale.

### Examples

```
set.seed(1)

## simplified generator for clustered partly interval-censored data
n <- 50
p <- 2
beta0 <- c(1, 1)
clu_rate <- 0.5
exactrates <- 0.8
left <- 0.001
right <- 0.01

## cluster-level frailty and informative cluster sizes
eta <- 1 / clu_rate
v <- rgamma(n, shape = eta, rate = eta)
m <- ifelse(v > median(v), 5, 3)
id <- rep(seq_len(n), m)
vi <- rep(v, m)

## subject-level covariates and failure times
N <- sum(m)
x <- matrix(rnorm(N * p), ncol = p)
T <- as.vector(exp(x %*% beta0 + vi * log(rexp(N))))

## build (L, R, delta)
L <- R <- delta <- numeric(N)
index <- rbinom(N, 1, exactrates)

for (i in seq_len(N)) {
  if (index[i] == 1) {
    L[i] <- T[i]
    R[i] <- T[i]
    delta[i] <- 1
  } else {
    U <- cumsum(c(1e-8, runif(10, left, right)))
    LL <- U[-length(U)]
    RR <- U[-1]

    if (T[i] < min(LL)) {
      L[i] <- 1e-8
      R[i] <- min(LL)
      delta[i] <- 0
    } else if (T[i] > max(RR)) {
```

```

    L[i] <- max(RR)
    R[i] <- 1e8
    delta[i] <- 0
  } else {
    idd <- which(T[i] > LL & T[i] < RR)
    if (length(idd) == 1) {
      L[i] <- LL[idd]
      R[i] <- RR[idd]
      delta[i] <- 0
    } else {
      L[i] <- T[i]
      R[i] <- T[i]
      delta[i] <- 1
    }
  }
}
}
}

dt <- data.frame(
  L = L, R = R, delta = delta, id = id,
  x1 = x[, 1], x2 = x[, 2]
)

fit <- aftpen_pic(dt, lambda = 0.001, se = "CF", type = "BAR")
fit$beta

```

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simdat\_pic

*Simulated clustered partly interval-censored data*


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### Description

Example dataset for penalized AFT model fitting with clustered partly interval-censored survival data.

### Usage

```
simdat_pic
```

### Format

A data frame with 6 variables:

**L** Left endpoint of observation interval

**R** Right endpoint of observation interval

**delta** Exact observation indicator (1 = exact, 0 = interval)

**id** Cluster identifier

**x1** First covariate

**x2** Second covariate

**Source**

Simulated data

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simdat_rc	<i>Simulated right-censored data</i>
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**Description**

Example dataset for penalized AFT model fitting with right-censored survival data.

**Usage**

simdat\_rc

**Format**

A data frame with 100 rows and 12 variables:

**y** Observed survival or censoring time

**d** Censoring indicator (1 = event observed, 0 = censored)

**X1** First covariate

**X2** Second covariate

**X3** Third covariate

**X4** Fourth covariate

**X5** Fifth covariate

**X6** Sixth covariate

**X7** Seventh covariate

**X8** Eighth covariate

**X9** Ninth covariate

**X10** Tenth covariate

**Source**

Simulated data

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