

# Package ‘FunSurv’

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**Title** Modeling Time-to-Event Data with Functional Predictors

**Version** 1.0.0

**Description** A collection of methods for modeling time-to-event data using both functional and scalar predictors. It implements functional data analysis techniques for estimation and inference, allowing researchers to assess the impact of functional covariates on survival outcomes, including time-to-single event and recurrent event outcomes.

**Depends** R ( $\geq 3.5.0$ )

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**URL** <https://github.com/zifangkong/FunSurv>

**BugReports** <https://github.com/zifangkong/FunSurv/issues>

**Encoding** UTF-8

**RoxygenNote** 7.3.2

**Imports** MFPCA, MASS, funData, Matrix, ggplot2, methods, reda ( $\geq 0.5.0$ )

**NeedsCompilation** no

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ar1_cor	<i>Construct an AR(1) correlation matrix</i>
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**Description**

Construct an AR(1) correlation matrix

**Usage**

```
ar1_cor(n, rho)
```

**Arguments**

n	number of events for each subject
rho	autoregressive correlation

**Value**

A n by n matrix

**Examples**

```
## Generate AR(1) structure
ar1_cor(n = 5, rho = 0.3)

## first derivative of the AR(1) structure with respect to rho
dar1_cor.drho(n = 5, rho = 0.3)
```

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AR1_FRAILTY	<i>Fit a Functional Regression with Autoregressive frailTY (FRAILTY) model for Recurrent Event Data</i>
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**Description**

Jointly model longitudinal measurements and recurrent events, accommodating both scalar and functional predictors while capturing time-dependent correlations among events. The FRAILTY method employs a two-step estimation procedure. First, functional principal component analysis through conditional expectation (PACE) is applied to extract key temporal features from sparse, irregular longitudinal data. Second, the obtained scores are incorporated into a dynamic recurrent frailty model with an autoregressive structure to account for within-subject correlations across recurrent events. This function works only for univariate functional data.

**Usage**

```

ARI_FRAILTY(
  formula,
  sdat,
  fdat,
  para0 = c(0.5, 0.5),
  nbasis = 10,
  pve = 0.9,
  npc = NULL,
  makePD = FALSE,
  cov.weight.type = c("none", "counts"),
  iter.max = 50,
  eps = 1e-06
)

```

**Arguments**

formula	A formula, with the response on the left of a ~ operator being a <code>Recur</code> object as returned by function <code>Recur</code> in <b>reda</b> , and scalar covariates on the right.
sdat	A data frame containing subject IDs, time-to-event outcomes (starting time, end point, censoring time and event status), and scalar covariates
fdat	A data frame containing subject IDs, longitudinal measurements, and the corresponding time points for each measurement.
para0	A vector of initial values for $\theta^2$ and auto-regressive coefficient $\rho$ . Both default to 0.5.
nbasis	An integer, representing the number of B-spline basis functions used for estimation of the mean function and bivariate smoothing of the covariance surface. Defaults to 10 (cf. <code>fpca.sc</code> in <b>refund</b> ).
pve	A numeric value between 0 and 1, the proportion of variance explained: used to choose the number of principal components. Defaults to 0.9 (cf. <code>fpca.sc</code> in <b>refund</b> ).
npc	An integer, giving a prespecified value for the number of principal components. Defaults to NULL. If given, this overrides <code>pve</code> (cf. <code>fpca.sc</code> in <b>refund</b> ).
makePD	Logical: should positive definiteness be enforced for the covariance surface estimate? Defaults to FALSE (cf. <code>fpca.sc</code> in <b>refund</b> ).
cov.weight.type	The type of weighting used for the smooth covariance estimate. Defaults to "none", i.e. no weighting. Alternatively, "counts" (corresponds to <code>fpca.sc</code> in <b>refund</b> ) weights the pointwise estimates of the covariance function by the number of observation points.
iter.max	Maximum number of iterations for both inner iteration and outer iteration. Defaults to 50.
eps	Tolerance criteria for a possible infinite coefficient value. Defaults to 1e-6.

## Details

### Model specification:

Let  $T_{ij}$  denote the time of the  $j$ th event for subject  $i$ , and let  $C_i$  represent the censoring time. The observed event time, accounting for right censoring, is  $\tilde{T}_{ij} = \min(T_{ij}, C_i)$ , and  $\delta_{ij} = I(T_{ij} \leq C_i)$  serves as an indicator of whether the  $j$ th event for subject  $i$  is observed. The hazard function is specified as

$$h(t; \mathbf{Z}_i, X_i(\cdot)) = h_0(t - t_{i,j-1}) \exp(\eta_{ij}),$$

where  $h_0(\cdot)$  is the baseline hazard function, and  $\eta_{ij} = \boldsymbol{\alpha}^\top \mathbf{Z}_i + \int_{t_{i,j-1}}^t X_i(s) \beta(s) ds + v_{ij}$ . Here,  $t_{i,j-1}$  is the previous event time with  $t_{i0} = 0$ .  $\boldsymbol{\alpha}$  is the fixed effect parameter associated with the time-invariant covariates  $\mathbf{Z}_i$ , and  $\beta(t)$  is a time-varying coefficient that captures the effect of functional predictor  $X_i(t)$  on the hazard rate of recurrent events.

## Value

A funsurv object containing the following components:

beta	Estimation of coefficients of scalar covariates and FPC scores. Including estimated values, standard errors, and p-values
beta_vcov	Estimated variance-covariance of the estimates of beta
eAR	Estimation of variance components ( $\theta^2$ and $\rho$ )
eAR_vcov	Estimated variance of estimates of $\theta^2$ and $\rho$
frailties	Estimated frailty terms (random effects)
basesurv	Estimated baseline survival probability
time	Time points associated with baseline survival probability
FPC	Functional principal components

## See Also

[Recur](#)

[PACE](#)

## Examples

```
data(simDat)

fit <- AR1_FRAILITY(Recur(t_start %to% t_stop, id, status) ~ z1,
                    sdat = sdat, fdat = fdat, iter.max = 30)

summary(fit)
```

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basesurv	<i>A function to obtain the baseline survival function</i>
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**Description**

A function to obtain the baseline survival function

**Usage**

```
basesurv(object)
```

**Arguments**

object            A funsurv object

**Value**

A data frame including time and baseline survival

**Examples**

```
data(simDat)

fit <- AR1_FRAILITY(Recur(t_start %to% t_stop, id, status) ~ z1,
                    sdat = sdat, fdat = fdat, iter.max = 30)

plot(fit, what = "basesurv")
```

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dar1_cor.drho	<i>First derivative of AR(1) correlation matrix with respect to the auto-regressive coefficient</i>
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**Description**

First derivative of AR(1) correlation matrix with respect to the auto-regressive coefficient

**Usage**

```
dar1_cor.drho(n, rho)
```

**Arguments**

n                    number of events for each subject  
rho                   autoregressive correlation

**Value**

A n by n inverse matrix

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plot.funsurv *Plot method for 'funsurv' objects*

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### Description

Plot method for 'funsurv' objects

### Usage

```
## S3 method for class 'funsurv'
plot(x, what = c("beta", "fpc", "basesurv"), ...)
```

### Arguments

x	A funsurv object
what	A character string specifying what to be plotted. Use what = "beta" to plot the estimated $\beta(t)$ . Use what = "fpc" to plot the functional principal components associated with the the longitudinal measurements. Use what = "basesurv" to plot the baseline survival probabilities.
...	additional graphical parameters to be passed to methods.

### Value

A ggplot object ...

### Examples

```
data(simDat)
fit <- AR1_FRAILITY(Recur(t_start %to% t_stop, id, status) ~ z1,
                    sdat = sdat, fdat = fdat, iter.max = 30)
plot(fit, what = "beta")
plot(fit, what = "fpc")
plot(fit, what = "basesurv")
```

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simDat *Simulated datasets for demonstration*

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### Description

The dataset was generated based on the proposed model  $h(t; \mathbf{Z}_i, X_i(\cdot)) = h_0(t - t_{i,j-1}) \exp(\eta_{ij})$ , where  $h_0(\cdot)$  is the baseline hazard function generated from a Weibull distribution.  $\eta_{ij} = \boldsymbol{\alpha}^\top \mathbf{Z}_i + \int_{t_{i,j-1}}^t X_i(s) \beta(s) ds + v_{ij}$ .  $\boldsymbol{\alpha}$  is the fixed effect parameter associated with the time-invariant covariates  $\mathbf{Z}_i$ , and  $\beta(t)$  is a time-varying coefficient that captures the effect of functional predictor  $X_i(t)$  on the hazard rate of recurrent events. The simulated dataset is organized into two data frames: a survival data frame (sdat) and a functional data frame (fdat). The variables in each data frame are listed below:

**Usage**

`data(simDat)`

**Format**

A list with two data frame:

**sdat** Survival data; a data frame with xxx rows and xxx variables:

**id** Subjects identification

**event** A sequence of the number of events per subject

**t\_start** Event starting time

**t\_end** Event end time

**censoring\_time** Event censoring time

**status** Event status: status=1 if event is observed and status=0 if event is censored

**z1** A univariate scalar covariates. Can be extended to multiple scalar covariates

**fdat** Functional data; a data frame with xxx rows and xxx variables:

**id** Subjects identification

**time** Time points for each longitudinal measurement

**x** Longitudinal measurements at distinct time points

**Source**

Simulated data

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