

# Package ‘LTRCtrees’

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**Type** Package

**Title** Survival Trees to Fit Left-Truncated and Right-Censored and Interval-Censored Survival Data

**Version** 1.1.0

**Description** Recursive partition algorithms designed for fitting survival tree with left-truncated and right censored (LTRC) data, as well as interval-censored data. The LTRC trees can also be used to fit survival tree with time-varying covariates.

**Imports** partykit (>= 1.2.0), rpart, survival, inum, icenReg

**Suggests** Formula, rpart.plot, knitr, rmarkdown, interval

**Depends** R (>= 3.2.0)

**License** GPL-3

**LazyData** TRUE

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**VignetteBuilder** knitr, rmarkdown

**RoxygenNote** 6.0.1

**NeedsCompilation** no

**Repository** CRAN

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## R topics documented:

extree_data . . . . .	2
ICtree . . . . .	3
LTRCART . . . . .	4
LTRCIT . . . . .	6
Pred.rpart . . . . .	8

<b>Index</b>	<b>10</b>
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extree_data	<i>Copy the partykit::extree_data function from partykit to avoid dependency issue</i>
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### Description

extree\_data imports partykit::extree\_data function

### Usage

```
extree_data(formula, data, subset, na.action = stats::na.pass, weights,
  offset, cluster, strata, scores = NULL, yx = c("none", "matrix"),
  ytype = c("vector", "data.frame", "matrix"), nmax = c(yx = Inf, z = Inf),
  ...)
```

### Arguments

formula	Same as the one in <a href="#">extree_data</a> , check <a href="#">extree_data</a> for usage
data	Same as the one in <a href="#">extree_data</a> , check <a href="#">extree_data</a> for usage
subset	Same as the one in <a href="#">extree_data</a> , check <a href="#">extree_data</a> for usage
na.action	Same as the one in <a href="#">extree_data</a> , check <a href="#">extree_data</a> for usage
weights	Same as the one in <a href="#">extree_data</a> , check <a href="#">extree_data</a> for usage
offset	Same as the one in <a href="#">extree_data</a> , check <a href="#">extree_data</a> for usage
cluster	Same as the one in <a href="#">extree_data</a> , check <a href="#">extree_data</a> for usage
strata	Same as the one in <a href="#">extree_data</a> , check <a href="#">extree_data</a> for usage
scores	Same as the one in <a href="#">extree_data</a> , check <a href="#">extree_data</a> for usage
yx	Same as the one in <a href="#">extree_data</a> , check <a href="#">extree_data</a> for usage
ytype	Same as the one in <a href="#">extree_data</a> , check <a href="#">extree_data</a> for usage
nmax	Same as the one in <a href="#">extree_data</a> , check <a href="#">extree_data</a> for usage
...	Same as the one in <a href="#">extree_data</a> , check <a href="#">extree_data</a> for usage

### Value

check [extree\\_data](#) for the return value

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`ICtree`*Fit a survival tree for interval-censored survival data*

---

## Description

Recursive partition for interval-censored survival data in a conditional inference framework.

## Usage

```
ICtree(Formula, data, Control = partykit::ctree_control())
```

## Arguments

Formula	A formula object, with the response be a <a href="#">Surv</a> object, with form <code>Surv(time1, time2, type="interval2")</code>
data	A data frame contains the variables named in Formula.
Control	A list of control parameters, see <a href="#">ctree_control</a>

## Details

ICtree returns a [party](#) object. This function extends the conditional inference survival tree algorithm in [ctree](#) to fit interval-censored survival data. This function itself not longer requires the interval package, but running the example below requires the interval package (for bcos data), which in turn requires the Icens package, which is not available on CRAN. To install the Icens package, enter the following commands

```
source("https://bioconductor.org/biocLite.R")
biocLite("Icens")
```

## Value

An object of class [party](#).

## References

Fu, W. and Simonoff, J.S. (2017). Survival trees for Interval Censored Survival data. *Statistics in medicine* 36 (30), 4831-4842

## Examples

```
library(interval)
library(LTRCtrees)
data(bcos)

## Fit ICtree survival tree
## make sure to attach survival package (by library(survival) ) before using Surv function
Ctree <- ICtree(Surv(left,right,type="interval2")~treatment, data = bcos)
```

```
## Plot the fitted tree
plot(Ctree)
```

---

 LTRCART

*Fit a relative risk survival tree for LTRC data*


---

## Description

LTRCART returns an [rpart](#) object. This function extends the survival tree algorithm in [rpart](#) to fit left-truncated and right censored (LTRC) data.

## Usage

```
LTRCART(formula, data, weights = NULL, subset = NULL, no.SE = 0,
        control = rpart::rpart.control(cp = 0.001))
```

## Arguments

formula	A formula object specifies the regression function, with the response be a <a href="#">Surv</a> object, with form <code>Surv(time1, time2, event)</code>
data	An optional data frame which contains the variables named in the formula.
weights	Optional case weights, same as in <a href="#">rpart</a>
subset	Optional expression saying that only a subset of the rows of the data should be used in the fit, same as in <a href="#">rpart</a>
no.SE	Number of standard errors used in pruning, with default value 0.
control	A list of control values used to control the <a href="#">rpart</a> algorithm, with default <code>cp = 0.001</code> . See <a href="#">rpart.control</a> for details.

## Value

An object of class `rpart`. See [rpart.object](#).

## References

Fu, W. and Simonoff, J.S. (2017). Survival trees for left-truncated and right-censored data, with application to time-varying covariate data. *Biostatistics* 18 (2), 352-369.

## Examples

```
## The Assay of serum free light chain data in survival package
## Adjust data & clean data
library(survival)
library(LTRCtrees)
Data <- flchain
Data <- Data[!is.na(Data$creatinine),]
Data$End <- Data$age + Data$futime/365
```

```

DATA <- Data[Data$End > Data$age,]
names(DATA)[6] <- "FLC"

## Setup training set and test set
Train = DATA[1:500,]
Test = DATA[1000:1020,]

## Fit LTRCART survival tree
## make sure to attach survival package (by library(survival) ) before using Surv function
LTRCART.obj <- LTRCART(Surv(age, End, death) ~ sex + FLC + creatinine, Train)

## Putting Surv(End, death) in formula would result an error message
## since LTRCART is expecting Surv(time1, time2, event)

## Plot the fitted tree
library(rpart.plot)
rpart.plot(LTRCART.obj)

## Plot as partykit::party object
library(partykit)
plot(as.party(LTRCART.obj))

## Plot as partykit::party object with survival curves on terminal nodes
LTRCART.obj.party <- as.party(LTRCART.obj)
LTRCART.obj.party$fitted[["(response)"]]<- Surv(Train$age, Train$End, Train$death)
plot(LTRCART.obj.party)

## Predict relative risk on test set
LTRCART.pred <- predict(LTRCART.obj, newdata = Test)

#####
##### Survival tree with time-varying covariates #####
#####
## The pbcseq dataset of survival package
library(survival)
## Create the start-stop-event triplet needed for coxph and LTRC trees
first <- with(pbcseq, c(TRUE, diff(id) !=0)) #first id for each subject
last <- c(first[-1], TRUE) #last id
time1 <- with(pbcseq, ifelse(first, 0, day))
time2 <- with(pbcseq, ifelse(last, futime, c(day[-1], 0)))
event <- with(pbcseq, ifelse(last, status, 0))
event <- 1*(event==2)

pbcseq$time1 <- time1
pbcseq$time2 <- time2
pbcseq$event <- event
## Fit the Cox model and LTRCART tree with time-varying covariates
fit.cox <- coxph(Surv(time1, time2, event) ~ age + sex + log(bili), pbcseq)
LTRCART.fit <- LTRCART(Surv(time1, time2, event) ~ age + sex + log(bili), pbcseq)
rpart.plot(LTRCART.fit)

```

```

### transform the wide format data into long format data using tmerge function
### from survival function
## Stanford Heart Transplant data
jasa$subject <- 1:nrow(jasa)

tdata <- with(jasa, data.frame(subject = subject,
                             futime= pmax(.5, fu.date - accept.dt),
                             txtime= ifelse(tx.date== fu.date,
                                             (tx.date -accept.dt) -.5,
                                             (tx.date - accept.dt)),
                             fustat = fustat))

sdata <- tmerge(jasa, tdata, id=subject,death = event(futime, fustat),
               trt = tdc(txtime), options= list(idname="subject"))

sdata$age <- sdata$age - 48

sdata$year <- as.numeric(sdata$accept.dt - as.Date("1967-10-01"))/365.25

Cox.fit <- coxph(Surv(tstart, tstop, death) ~ age+ surgery, data= sdata)
LTRCART.fit <- LTRCART(Surv(tstart, tstop, death) ~ age + transplant, data = sdata)
rpart.plot(LTRCART.fit)

```

---

LTRCIT

*Fit a conditional inference survival tree for LTRC data*


---

## Description

LTRCIT returns a [party](#) object. This function extends the conditional inference survival tree algorithm in [ctree](#) to fit left-truncated and right censored (LTRC) data.

## Usage

```
LTRCIT(Formula, data, Control = partykit::ctree_control())
```

## Arguments

Formula	A formula object, with the response be a <a href="#">Surv</a> object, with form <code>Surv(time1, time2, event)</code>
data	A data frame contains the variables named in formula.
Control	A list of control parameters, see <a href="#">ctree_control</a>

## Value

An object of class [party](#).

## References

Fu, W. and Simonoff, J.S.(2017). Survival trees for left-truncated and right-censored data, with application to time-varying covariate data. *Biostatistics* 18 (2), 352-369.

## Examples

```
## The Assay of serum free light chain data in survival package
## Adjust data & clean data
library(survival)
library(LTRCtrees)
Data <- flchain
Data <- Data[!is.na(Data$creatinine),]
Data$End <- Data$age + Data$futime/365
DATA <- Data[Data$End > Data$age,]
names(DATA)[6] <- "FLC"

## Setup training set and test set
Train = DATA[1:500,]
Test = DATA[1000:1020,]

## Fit LTRCIT survival tree
## make sure to attach survival package (by library(survival) ) before using Surv function
LTRCIT.obj <- LTRCIT(Surv(age, End, death) ~ sex + FLC + creatinine, Train)
plot(LTRCIT.obj)

## Putting Surv(End, death) in formula would result an error message
## since LTRCIT is expecting Surv(time1, time2, event)

## Note that LTRCIT.obj is an object of class party
## predict median survival time on test data
LTRCIT.pred <- predict(LTRCIT.obj, newdata = Test, type = "response")

## predict Kaplan Meier survival curve on test data,
## return a list of survfit objects -- the predicted KM curves
LTRCIT.pred <- predict(LTRCIT.obj, newdata = Test, type = "prob")

#####
##### Survival tree with time-varying covariates #####
#####
## The pbcseq dataset of survival package
library(survival)
## Create the start-stop-event triplet needed for coxph and LTRC trees
first <- with(pbcseq, c(TRUE, diff(id) !=0)) #first id for each subject
last <- c(first[-1], TRUE) #last id
time1 <- with(pbcseq, ifelse(first, 0, day))
time2 <- with(pbcseq, ifelse(last, futime, c(day[-1], 0)))
event <- with(pbcseq, ifelse(last, status, 0))
event <- 1*(event==2)

pbcseq$time1 <- time1
pbcseq$time2 <- time2
pbcseq$event <- event
```

```

pbcseq = pbcseq[1:1000,] ## fit on subset of the data to save fitting time
## Fit the Cox model and LTRCIT tree with time-varying covariates
fit.cox <- coxph(Surv(time1, time2, event) ~ age + sex + log(bili), pbcseq)
LTRCIT.fit <- LTRCIT(Surv(time1, time2, event) ~ age + sex + log(bili), pbcseq)
plot(LTRCIT.fit)

## transform the wide format data into long format data using tmerge function
## from survival function
## Stanford Heart Transplant data
jasa$subject <- 1:nrow(jasa)

tdata <- with(jasa, data.frame(subject = subject,
                             futime= pmax(.5, fu.date - accept.dt),
                             txtime= ifelse(tx.date== fu.date,
                                             (tx.date -accept.dt) - .5,
                                             (tx.date - accept.dt)),
                             fustat = fustat))

sdata <- tmerge(jasa, tdata, id=subject,death = event(futime, fustat),
               trt = tdc(txtime), options= list(idname="subject"))

sdata$age <- sdata$age - 48

sdata$year <- as.numeric(sdata$accept.dt - as.Date("1967-10-01"))/365.25

Cox.fit <- coxph(Surv(tstart, tstop, death) ~ age+ surgery, data= sdata)
LTRCIT.fit <- LTRCIT(Surv(tstart, tstop, death) ~ age + transplant, data = sdata)
plot(LTRCIT.fit)

```

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Pred.rpart

*Prediction function for [rpart](#) object*

---

## Description

The output of LTRCART is an [rpart](#) object, and as a result the usual [predict](#) function on such an object returns the predicted relative risk on the test set. `Pred.rpart` returns the predicted Kaplan-Meier curves and median survival times on the test set, which in some circumstances might be desirable in practice. Note that this function can be applied to any [rpart](#) survival tree object, not just one produced by LTRCART

## Usage

```
Pred.rpart(formula, train, test)
```

## Arguments

formula	A formula used to fit the survival tree. The response is a <a href="#">Surv</a> object. If it has the form <code>Surv(time1, time2, event)</code> , then LTRCART is called internally; if response has the form <code>Surv(time, event)</code> , then the <a href="#">rpart</a> is called internally.
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train	Training set
test	Test set

**Value**

A list of predicted KM curves and median survival times.

**Examples**

```
## The Assay of serum free light chain data in survival package
## Adjust data & clean data
library(survival)
library(LTRCtrees)
Data <- flchain
Data <- Data[!is.na(Data$creatinine),]
Data$End <- Data$age + Data$futime/365
DATA <- Data[Data$End > Data$age,]
names(DATA)[6] <- "FLC"

## Setup training set and test set
Train = DATA[1:500,]
Test = DATA[1000:1020,]

## Predict median survival time and Kaplan Meier survival curve
## on test data using Pred.rpart
LTRCART.pred <- Pred.rpart(Surv(age, End, death) ~ sex + FLC + creatinine, Train, Test)
LTRCART.pred$KMcurves ## list of predicted KM curves
LTRCART.pred$Medians ## vector of predicted median survival time
```

# Index

ctree, [3](#), [6](#)  
ctree\_control, [3](#), [6](#)  
  
extree\_data, [2](#), [2](#)  
  
ICtree, [3](#)  
  
LTRCART, [4](#)  
LTRCIT, [6](#)  
  
party, [3](#), [6](#)  
Pred.rpart, [8](#)  
predict, [8](#)  
  
rpart, [4](#), [8](#)  
rpart.control, [4](#)  
rpart.object, [4](#), [8](#)  
  
Surv, [3](#), [4](#), [6](#), [8](#)