

Package ‘LTRCtrees’

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Type Package

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

Version 0.1.5

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

Imports partykit, rpart, survival

Suggests Formula, rpart.plot, knitr, rmarkdown

Depends R (>= 3.2.0)

License GPL-3

LazyData TRUE

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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 Surv 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 rpart
subset	Optional expression saying that only a subset of the rows of the data should be used in the fit, same as in rpart
no.SE	Number of standard errors used in pruning, with default value 0.
control	A list of control values used to control the rpart algorithm, with default <code>cp = 0.001</code> . See rpart.control for details.

Value

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

References

Fu, W. and Simonoff, J.S. (2016). Survival trees for left-truncated and right-censored data, with application to time-varying covariate data. [arXiv:1606.03033 \[stat.ME\]](#)

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
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 an [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 Surv 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 ctree_control

Value

An object of class [party](#).

References

Fu, W. and Simonoff, J.S.(2016). Survival trees for left-truncated and right-censored data, with application to time-varying covariate data. arXiv:1606.03033 [stat.ME]

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

```

 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

<code>formula</code>	A formula used to fit the survival tree. The response is a Surv 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 rpart is called internally.
<code>train</code>	Training set
<code>test</code>	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
```

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