

# Package ‘uni.survival.tree’

July 22, 2025

**Type** Package

**Title** A Survival Tree Based on Stabilized Score Tests for High-dimensional Covariates

**Version** 1.5

**Author** Takeshi Emura and Wei-Chern Hsu

**Maintainer** Takeshi Emura <takeshiemura@gmail.com>

## Description

A classification (decision) tree is constructed from survival data with high-dimensional covariates. The method is a robust version of the logrank tree, where the variance is stabilized. The main function `uni.tree` returns a classification tree for a given survival dataset. The inner nodes (splitting criterion) are selected by minimizing the P-value of the two-sample score tests. The decision of declaring terminal nodes (stopping criterion) is the P-value threshold given by an argument (specified by user). This tree construction algorithm is proposed by Emura et al. (2021, in review).

**License** GPL-3

**Encoding** UTF-8

**LazyData** true

**RoxygenNote** 7.1.0

**Depends** survival,compound.Cox

**NeedsCompilation** no

**Repository** CRAN

**Date/Publication** 2021-03-22 06:40:02 UTC

## Contents

feature.selected . . . . .	2
KM.split . . . . .	2
risk.classification . . . . .	3
uni.logrank . . . . .	4
uni.tree . . . . .	5
X.pathway_discrete.balanced . . . . .	6
X.pathway_discrete.imbalanced . . . . .	7

**Index****8**


---

feature.selected	<i>The names of features that are selected in a tree</i>
------------------	----------------------------------------------------------

---

**Description**

The function returns the names of features (covariates) that are selected as the internal nodes of a tree. Only the names of the covariates are shown by excluding the cut-off values.

**Usage**

```
feature.selected(tree)
```

**Arguments**

tree :an object made from the "uni.tree" function

**Details**

The outputs show important features for predicting survival outcomes.

**Value**

An array of characters that are the names from those covariates selected in the tree

**Examples**

```
data(Lung,package="compound.Cox")
train_Lung=Lung[which(Lung["train"]==TRUE),] #select training data
t.vec=train_Lung[,1]
d.vec=train_Lung[,2]
x.mat=train_Lung[,-c(1,2,3)]
res=uni.tree(t.vec,d.vec,x.mat,P.value=0.01,d0=0.01,S.plot=FALSE,score=TRUE)
feature.selected(res)
```

---

KM.split

*Kaplan-Meier estimator of binary splitting*


---

**Description**

Given a cut-off-point and selected covariate, return the survival curve for binary classification and the P-value of two sample log-rank test.

**Usage**

```
KM.split(t.vec, d.vec, X.mat, x.name, cutoff)
```

**Arguments**

t.vec	:Vector of survival times (time to either death or censoring)
d.vec	:Vector of censoring indicators (1=death, 0=censoring)
X.mat	:n by p matrix of covariates, where n is the sample size and p is the number of covariates
x.name	:the name of covariate
cutoff	:cut-off-point

**Value**

P-value of two sample logrank test and a plot of two KM estimates

**Examples**

```
data(Lung,package="compound.Cox")
train_Lung=Lung[which(Lung["train"]==TRUE),] #select training data
t.vec=train_Lung[,1]
d.vec=train_Lung[,2]
x.mat=train_Lung[,-c(1,2,3)]
KM.split(t.vec,d.vec,x.mat,x.name="ANXA5",cutoff=1)
```

---

risk.classification    *The risk ranks of the samples predicted by a tree*

---

**Description**

The function returns the ranks (1=the lowest risk, 2=the 2nd lowest risk, ..., k=the highest risk) predicted for the samples.

**Usage**

```
risk.classification(tree, X.mat)
```

**Arguments**

tree	:an object made from the "uni.tree" function
X.mat	:n by p matrix of covariates from the samples, where n is the sample size and p is the number of covariates

**Details**

If the tree has k terminal nodes, then the response 1 represents the lowest risk and k represents the highest risk.

**Value**

A vector of integers, 1, 2, ..., k, that represent the ranks predicted for the samples.

**Examples**

```

data(Lung,package="compound.Cox")
train_Lung=Lung[which(Lung[,"train"]==TRUE),] #select training data
t.vec=train_Lung[,1]
d.vec=train_Lung[,2]
x.mat=train_Lung[,-c(1,2,3)]
res=uni.tree(t.vec,d.vec,x.mat,P.value=0.01,d0=0.01,S.plot=FALSE,score=TRUE)
risk.classification(res,x.mat)

```

---

uni.logrank

*Univariate binary splits by the logrank test*


---

**Description**

The output is the summary of significance tests for binary splits, where the cut-off values are optimized for each covariate.

**Usage**

```
uni.logrank(t.vec, d.vec, X.mat)
```

**Arguments**

t.vec	:Vector of survival times (time to either death or censoring)
d.vec	:Vector of censoring indicators (1=death, 0=censoring)
X.mat	:n by p matrix of covariates, where n is the sample size and p is the number of covariates

**Details**

The output can be used to construct a logrank tree.

**Value**

A dataframe containing:

Pvalue: the P-value of the two-sample logrank test, where the cut-off value is optimized

cut\_off\_point: the optimal cut-off values of the binary splits given a feature

left.sample.size: the sample size of a left child node

right.sample.size: the sample size of a right child node

**Examples**

```

data(Lung,package="compound.Cox")
train_Lung=Lung[which(Lung[,"train"]==TRUE),] #select training data
t.vec=train_Lung[,1]
d.vec=train_Lung[,2]
x.mat=train_Lung[,-c(1,2,3)]
uni.logrank(t.vec,d.vec,x.mat)

```

---

`uni.tree`*A survival tree based on stabilized score tests*

---

### Description

This function returns a classification (decision) tree for a given survival dataset. The decision of making inner nodes (splitting criterion) is based on the univariate score tests. The decision of declaring terminal nodes (stopping criterion) is the P-value threshold given by an argument. This tree construction algorithm is proposed by Emura et al. (2021).

### Usage

```
uni.tree(  
  t.vec,  
  d.vec,  
  X.mat,  
  P.value = 0.01,  
  d0 = 0.01,  
  S.plot = FALSE,  
  score = TRUE  
)
```

### Arguments

<code>t.vec</code>	:Vector of survival times (time to either death or censoring)
<code>d.vec</code>	:Vector of censoring indicators (1=death, 0=censoring)
<code>X.mat</code>	:n by p matrix of covariates (features), where n is the sample size and p is the number of covariates
<code>P.value</code>	:the threshold of P-value for stop splitting (stopping criterion)
<code>d0</code>	:A positive constant to stabilize the variance of score statistics (Witten & Tibshirani 2010)
<code>S.plot</code>	:call for plot the KM estimator for each split
<code>score</code>	:TRUE = score test (Emura et al. 2019); FALSE = log-rank test

### Details

In order to stabilize the univariate score tests, a small value "d0" is added to the variance of the score statistics (Witten and Tibshirani 2010). `d0=0` corresponds to the logrank test. To perform a large number of the score tests, the "compound.Cox" packages (Emura et al.2019) is applied with `d0` as a option.

### Value

A nested list describing a classification tree, consisting of inner nodes and terminal node.

**References**

- Emura T, Hsu WC, Chou WC (2021). A survival tree based on stabilized score tests for high-dimensional covariates, in review
- Emura T, Matsui S, Chen HY (2019). compound.Cox: Univariate Feature Selection and Compound Covariate for Predicting Survival, *Computer Methods and Programs in Biomedicine* 168: 21-37.
- Witten DM, Tibshirani R (2010) Survival analysis with high-dimensional covariates. *Stat Method Med Res* 19:29-51

**Examples**

```
data(Lung,package="compound.Cox")
train_Lung=Lung[which(Lung["train"]==TRUE),] #select training data
t.vec=train_Lung[,1]
d.vec=train_Lung[,2]
x.mat=train_Lung[,-c(1,2,3)]
uni.tree(t.vec,d.vec,x.mat,P.value=0.01,d0=0.01,S.plot=FALSE,score=TRUE)
```

---

`X.pathway_discrete.balanced`

*Generate a matrix of gene expressions (discrete version of `X.pathway()`) against to Emura (2012)) in the presence of gene pathways*

---

**Description**

Generate a matrix of gene expressions in the presence of gene pathways, we first produce the matrix by `X.pathway`(Emura et al. 2012), then we change each value to 1 ~ 4 depend on the quantile.

**Usage**

```
X.pathway_discrete.balanced(n, p, q1, q2, rho1 = 0.5, rho2 = 0.5)
```

**Arguments**

<code>n</code>	:the number of individuals (sample size)
<code>p</code>	:the number of genes
<code>q1</code>	:the number of genes in the first pathway
<code>q2</code>	:the number of genes in the second pathway
<code>rho1</code>	:the correlation coefficient for the first pathway
<code>rho2</code>	:the correlation coefficient for the second pathway

**Value**

X n by p matrix of gene expressions

## References

Emura T, Chen YH, Chen HY (2012). Survival Prediction Based on Compound Covariate under Cox Proportional Hazard Models. PLoS ONE 7(10): e47627. doi:10.1371/journal.pone.0047627

## Examples

```
## generate 6 gene expressions from 10 individuals
X.pathway_discrete.balanced(n=10,p=6,q1=2,q2=2,rho1=0.5,rho2=0.5)
```

---

```
X.pathway_discrete.imbalanced
```

*Generate a matrix of unbalance gene expressions (discrete version of X.pathway() against to Emura (2012)) in the presence of gene pathways*

---

## Description

Generate a matrix of gene expressions in the presence of gene pathways, we first produce the matrix by X.pathway(Emura et al. 2012), then we change each value to 1 ~ 3 depend on the quantile and randomly replace a element to 4 in the last p-(q1+q2) column for each row.

## Usage

```
X.pathway_discrete.imbalanced(n, p, q1, q2, rho1 = 0.5, rho2 = 0.5)
```

## Arguments

n	:the number of individuals (sample size)
p	:the number of genes
q1	:the number of genes in the first pathway
q2	:the number of genes in the second pathway
rho1	:the correlation coefficient for the first pathway
rho2	:the correlation coefficient for the second pathway

## Value

X n by p matrix of gene expressions

## References

Emura T, Chen YH, Chen HY (2012). Survival Prediction Based on Compound Covariate under Cox Proportional Hazard Models. PLoS ONE 7(10): e47627. doi:10.1371/journal.pone.0047627

## Examples

```
## generate 6 gene expressions from 10 individuals
X.pathway_discrete.imbalanced(n=10,p=6,q1=2,q2=2,rho1=0.5,rho2=0.5)
```

# Index

`feature.selected`, [2](#)

`KM.split`, [2](#)

`risk.classification`, [3](#)

`uni.logrank`, [4](#)

`uni.tree`, [5](#)

`X.pathway_discrete.balanced`, [6](#)

`X.pathway_discrete.imbalanced`, [7](#)