

# Package ‘mlsurvlnrs’

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**Title** R6-Based ML Survival Learners for 'mlexperiments'

**Version** 0.0.3

**Description** Enhances 'mlexperiments'

<<https://CRAN.R-project.org/package=mlexperiments>> with additional machine learning ('ML') learners for survival analysis. The package provides R6-based survival learners for the following algorithms: 'glmnet' <<https://CRAN.R-project.org/package=glmnet>>, 'ranger' <<https://CRAN.R-project.org/package=ranger>>, 'xgboost' <<https://CRAN.R-project.org/package=xgboost>>, and 'rpart' <<https://CRAN.R-project.org/package=rpart>>. These can be used directly with the 'mlexperiments' R package.

**License** GPL (>= 3)

**URL** <https://github.com/kapsner/mlsurvlnrs>

**BugReports** <https://github.com/kapsner/mlsurvlnrs/issues>

**Depends** R (>= 2.10)

**Imports** data.table, kdry, mlexperiments, mllrnrs, R6, stats

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**Author** Lorenz A. Kapsner [cre, aut, cph]  
(<<https://orcid.org/0000-0003-1866-860X>>)

**Maintainer** Lorenz A. Kapsner <[lorenz.kapsner@gmail.com](mailto:lorenz.kapsner@gmail.com)>

**Repository** CRAN

## R topics documented:

<i>c_index</i> . . . . .	2
LearnerSurvCoxPHCox . . . . .	3
LearnerSurvGlmnetCox . . . . .	5
LearnerSurvRangerCox . . . . .	7
LearnerSurvRpartCox . . . . .	9
LearnerSurvXgboostAft . . . . .	12
LearnerSurvXgboostCox . . . . .	14

<b>Index</b>	<b>18</b>
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<i>c_index</i>	<i>c_index</i>
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### Description

Calculate the Harrell's concordance index (C-index)

### Usage

```
c_index(ground_truth, predictions)
```

### Arguments

`ground_truth` A ‘survival::Surv‘ object with the ground truth.  
`predictions` A vector with predictions.

### Details

A wrapper function around [glmnet::Cindex()] for use with ‘mlexperiments’.

### See Also

[glmnet::Cindex()]

### Examples

```
set.seed(123)
gt <- survival::Surv(
  time = rnorm(100, 50, 15),
  event = sample(0:1, 100, TRUE)
)
preds <- rbeta(100, 2, 5)

c_index(gt, preds)
```

---

LearnerSurvCoxPHCox     *R6 Class to construct a Cox proportional hazards survival learner*

---

## Description

The ‘LearnerSurvCoxPHCox‘ class is the interface to perform a Cox regression with the ‘survival‘ R package for use with the ‘mlexperiments‘ package.

## Details

Can be used with \* [mlexperiments::MLCrossValidation]

## Super class

[mlexperiments::MLLearnerBase](#) -> LearnerSurvCoxPHCox

## Methods

### Public methods:

- [LearnerSurvCoxPHCox\\$new\(\)](#)
- [LearnerSurvCoxPHCox\\$clone\(\)](#)

**Method new():** Create a new ‘LearnerSurvCoxPHCox‘ object.

*Usage:*

`LearnerSurvCoxPHCox$new()`

*Returns:* A new ‘LearnerSurvCoxPHCox‘ R6 object.

*Examples:*

`LearnerSurvCoxPHCox$new()`

**Method clone():** The objects of this class are cloneable with this method.

*Usage:*

`LearnerSurvCoxPHCox$clone(deep = FALSE)`

*Arguments:*

`deep` Whether to make a deep clone.

## See Also

`[survival::coxph()]`

## Examples

```
# survival analysis

dataset <- survival::colon |>
  data.table::as.data.table() |>
  na.omit()
dataset <- dataset[get("etype") == 2, ]

seed <- 123
surv_cols <- c("status", "time", "rx")

feature_cols <- colnames(dataset)[3:(ncol(dataset) - 1)]

split_vector <- splitTools::multi_strata(
  df = dataset[, .SD, .SDcols = surv_cols],
  strategy = "kmeans",
  k = 4
)

train_x <- model.matrix(
  ~ -1 +.,
  dataset[, .SD, .SDcols = setdiff(feature_cols, surv_cols[1:2])]
)
train_y <- survival::Surv(
  event = (dataset[, get("status")] |>
    as.character() |>
    as.integer()),
  time = dataset[, get("time")],
  type = "right"
)

fold_list <- splitTools::create_folds(
  y = split_vector,
  k = 3,
  type = "stratified",
  seed = seed
)

surv_coxph_cox_optimizer <- mlexperiments::MLCrossValidation$new(
  learner = LearnerSurvCoxPHCox$new(),
  fold_list = fold_list,
  ncores = 1L,
  seed = seed
)
surv_coxph_cox_optimizer$performance_metric <- c_index

# set data
surv_coxph_cox_optimizer$set_data(
  x = train_x,
  y = train_y
)
```

```

surv_coxph_cox_optimizer$execute()

## -----
## Method `LearnerSurvCoxPHCox$new`
## -----


LearnerSurvCoxPHCox$new()

```

**LearnerSurvGlmnetCox** *R6 Class to construct a Glmnet survival learner for Cox regression*

## Description

The ‘LearnerSurvGlmnetCox‘ class is the interface to perform a Cox regression with the ‘glmnet‘ R package for use with the ‘mlexperiments‘ package.

## Details

Optimization metric: C-index Can be used with \* [mlexperiments::MLTuneParameters] \* [mlexperiments::MLCrossValidation] \* [mlexperiments::MLNestedCVs]

## Super class

`mlexperiments::MLLearnerBase` -> LearnerSurvGlmnetCox

## Methods

### Public methods:

- [LearnerSurvGlmnetCox\\$new\(\)](#)
- [LearnerSurvGlmnetCox\\$clone\(\)](#)

**Method new():** Create a new ‘LearnerSurvGlmnetCox‘ object.

*Usage:*

`LearnerSurvGlmnetCox$new()`

*Returns:* A new ‘LearnerSurvGlmnetCox‘ R6 object.

*Examples:*

`LearnerSurvGlmnetCox$new()`

**Method clone():** The objects of this class are cloneable with this method.

*Usage:*

`LearnerSurvGlmnetCox$clone(deep = FALSE)`

*Arguments:*

`deep` Whether to make a deep clone.

**See Also**

[`glmnet::glmnet()`], [`glmnet::cv.glmnet()`]

**Examples**

```
# survival analysis

dataset <- survival::colon |>
  data.table::as.data.table() |>
  na.omit()
dataset <- dataset[get("etype") == 2, ]

seed <- 123
surv_cols <- c("status", "time", "rx")

feature_cols <- colnames(dataset)[3:(ncol(dataset) - 1)]

param_list_glmnet <- expand.grid(
  alpha = seq(0, 1, .2)
)

ncores <- 2L

split_vector <- splitTools::multi_strata(
  df = dataset[, .SD, .SDcols = surv_cols],
  strategy = "kmeans",
  k = 4
)

train_x <- model.matrix(
  ~ -1 + .,
  dataset[, .SD, .SDcols = setdiff(feature_cols, surv_cols[1:2])]
)
train_y <- survival::Surv(
  event = (dataset[, get("status")] |>
    as.character() |>
    as.integer()),
  time = dataset[, get("time")],
  type = "right"
)

fold_list <- splitTools::create_folds(
  y = split_vector,
  k = 3,
  type = "stratified",
  seed = seed
)

surv_glmnet_cox_optimizer <- mlexperiments::MLCrossValidation$new(
  learner = LearnerSurvGlmnetCox$new(),
  fold_list = fold_list,
```

```

ncores = ncores,
seed = seed
)
surv_glmnet_cox_optimizer$learner_args <- list(
  alpha = 0.8,
  lambda = 0.002
)
surv_glmnet_cox_optimizer$performance_metric <- c_index

# set data
surv_glmnet_cox_optimizer$set_data(
  x = train_x,
  y = train_y
)

surv_glmnet_cox_optimizer$execute()

## -----
## Method `LearnerSurvGlmnetCox$new`
## -----
LearnerSurvGlmnetCox$new()

```

LearnerSurvRangerCox R6 Class to construct a Ranger survival learner for Cox regression

## Description

The ‘LearnerSurvRangerCox‘ class is the interface to perform a Cox regression with the ‘ranger‘ R package for use with the ‘mlexperiments‘ package.

## Details

Optimization metric: C-index Can be used with \* [mlexperiments::MLTuneParameters] \* [mlexperiments::MLCrossValidation] \* [mlexperiments::MLNestedCVs]

## Super class

[mlexperiments::MLLearnerBase](#) -> LearnerSurvRangerCox

## Methods

### Public methods:

- [LearnerSurvRangerCox\\$new\(\)](#)
- [LearnerSurvRangerCox\\$clone\(\)](#)

**Method new():** Create a new ‘LearnerSurvRangerCox‘ object.

*Usage:*

```
LearnerSurvRangerCox$new()

Returns: A new ‘LearnerSurvRangerCox‘ R6 object.

Examples:
```

```
LearnerSurvRangerCox$new()
```

**Method clone():** The objects of this class are cloneable with this method.

*Usage:*

```
LearnerSurvRangerCox$clone(deep = FALSE)
```

*Arguments:*

deep Whether to make a deep clone.

## See Also

[ranger::ranger()]

## Examples

```
# survival analysis

dataset <- survival::colon |>
  data.table::as.data.table() |>
  na.omit()
dataset <- dataset[get("etype") == 2, ]

seed <- 123
surv_cols <- c("status", "time", "rx")

feature_cols <- colnames(dataset)[3:(ncol(dataset) - 1)]

param_list_ranger <- expand.grid(
  sample.fraction = seq(0.6, 1, .2),
  min.node.size = seq(1, 5, 4),
  mtry = seq(2, 6, 2),
  num.trees = c(5L, 10L),
  max.depth = seq(1, 5, 4)
)
ncores <- 2L

split_vector <- splitTools::multi_strata(
  df = dataset[, .SD, .SDcols = surv_cols],
  strategy = "kmeans",
  k = 4
)

train_x <- model.matrix(
  ~ -1 +.,
  dataset[, .SD, .SDcols = setdiff(feature_cols, surv_cols[1:2])]
)
```

```

train_y <- survival::Surv(
  event = (dataset[, get("status")] |>
    as.character() |>
    as.integer()),
  time = dataset[, get("time")],
  type = "right"
)

fold_list <- splitTools::create_folds(
  y = split_vector,
  k = 3,
  type = "stratified",
  seed = seed
)

surv_ranger_cox_optimizer <- mlexperiments::MLCrossValidation$new(
  learner = LearnerSurvRangerCox$new(),
  fold_list = fold_list,
  ncores = ncores,
  seed = seed
)
surv_ranger_cox_optimizer$learner_args <- as.list(
  data.table::data.table(param_list_ranger[1, ], stringsAsFactors = FALSE)
)
surv_ranger_cox_optimizer$performance_metric <- c_index

# set data
surv_ranger_cox_optimizer$set_data(
  x = train_x,
  y = train_y
)

surv_ranger_cox_optimizer$execute()

## -----
## Method `LearnerSurvRangerCox$new`
## -----
```

LearnerSurvRangerCox\$new()

LearnerSurvRpartCox    *LearnerSurvRpartCox R6 class*

## Description

This learner is a wrapper around [rpart::rpart()] in order to fit recursive partitioning and regression trees with survival data.

## Details

Optimization metric: C-index \* Can be used with \* [mlexperiments::MLTuneParameters] \* [mlexperiments::MLCrossValidation] \* [mlexperiments::MLNestedCV]

Implemented methods: \* '\$fit' To fit the model. \* '\$predict' To predict new data with the model. \* '\$cross\_validation' To perform a grid search (hyperparameter optimization). \* '\$bayesian\_scoring\_function' To perform a Bayesian hyperparameter optimization.

Parameters that are specified with 'parameter\_grid' and / or 'learner\_args' are forwarded to 'rpart''s argument 'control' (see [rpart::rpart.control()] for further details).

## Super class

`mlexperiments::MLLearnerBase` -> LearnerSurvRpartCox

## Methods

### Public methods:

- [LearnerSurvRpartCox\\$new\(\)](#)
- [LearnerSurvRpartCox\\$clone\(\)](#)

**Method** `new()`: Create a new 'LearnerSurvRpartCox' object.

*Usage:*

`LearnerSurvRpartCox$new()`

*Details:* This learner is a wrapper around [rpart::rpart()] in order to fit recursive partitioning and regression trees with survival data.

*Examples:*

`LearnerSurvRpartCox$new()`

**Method** `clone()`: The objects of this class are cloneable with this method.

*Usage:*

`LearnerSurvRpartCox$clone(deep = FALSE)`

*Arguments:*

`deep` Whether to make a deep clone.

## See Also

[rpart::rpart()], [c\_index()], [rpart::rpart.control()]  
[ rpart::rpart()], [c\_index()],

## Examples

```
# survival analysis

dataset <- survival::colon |>
  data.table::as.data.table() |>
  na.omit()
```

```
dataset <- dataset[get("etype") == 2, ]  
  
seed <- 123  
surv_cols <- c("status", "time", "rx")  
  
feature_cols <- colnames(dataset)[3:(ncol(dataset) - 1)]  
  
ncores <- 2L  
  
split_vector <- splitTools::multi_strata(  
  df = dataset[, .SD, .SDcols = surv_cols],  
  strategy = "kmeans",  
  k = 4  
)  
  
train_x <- model.matrix(  
  ~ -1 + .,  
  dataset[, .SD, .SDcols = setdiff(feature_cols, surv_cols[1:2])]  
)  
train_y <- survival::Surv(  
  event = (dataset[, get("status")] |>  
    as.character() |>  
    as.integer()),  
  time = dataset[, get("time")],  
  type = "right"  
)  
  
fold_list <- splitTools::create_folds(  
  y = split_vector,  
  k = 3,  
  type = "stratified",  
  seed = seed  
)  
  
surv_rpart_optimizer <- mlexperiments::MLCrossValidation$new(  
  learner = LearnerSurvRpartCox$new(),  
  fold_list = fold_list,  
  ncores = ncores,  
  seed = seed  
)  
surv_rpart_optimizer$learner_args <- list(  
  minsplit = 10L,  
  maxdepth = 20L,  
  cp = 0.03,  
  method = "exp"  
)  
surv_rpart_optimizer$performance_metric <- c_index  
  
# set data  
surv_rpart_optimizer$set_data(  
  x = train_x,  
  y = train_y  
)
```

```

surv_rpart_optimizer$execute()

## -----
## Method `LearnerSurvRpartCox$new`
## -----


LearnerSurvRpartCox$new()

```

**LearnerSurvXgboostAft** *R6 Class to construct a Xgboost survival learner for accelerated failure time models*

## Description

The ‘LearnerSurvXgboostAft’ class is the interface to accelerated failure time models with the ‘xgboost’ R package for use with the ‘mlexperiments’ package.

## Details

Optimization metric: needs to be specified with the learner parameter ‘eval\_metric’. Can be used with \* [mlexperiments::MLTuneParameters] \* [mlexperiments::MLCrossValidation] \* [mlexperiments::MLNestedCVs] Also see the official xgboost documentation on [aft models]([https://xgboost.readthedocs.io/en/stable/survival\\_analysis.html](https://xgboost.readthedocs.io/en/stable/survival_analysis.html))

## Super classes

`mlexperiments::MLLearnerBase -> mllrnrs::LearnerXgboost -> LearnerSurvXgboostAft`

## Methods

### Public methods:

- `LearnerSurvXgboostAft$new()`
- `LearnerSurvXgboostAft$clone()`

**Method new():** Create a new ‘LearnerSurvXgboostAft’ object.

*Usage:*

`LearnerSurvXgboostAft$new(metric_optimization_higher_better)`

*Arguments:*

`metric_optimization_higher_better` A logical. Defines the direction of the optimization metric used throughout the hyperparameter optimization.

*Returns:* A new ‘LearnerSurvXgboostAft’ R6 object.

*Examples:*

```
LearnerSurvXgboostAft$new(metric_optimization_higher_better = FALSE)
```

**Method clone():** The objects of this class are cloneable with this method.

*Usage:*

```
LearnerSurvXgboostAft$clone(deep = FALSE)
```

*Arguments:*

deep Whether to make a deep clone.

## See Also

[xgboost::xgb.train()], [xgboost::xgb.cv()]

## Examples

```
# execution time >2.5 sec
# survival analysis

dataset <- survival::colon |>
  data.table::as.data.table() |>
  na.omit()
dataset <- dataset[get("etype") == 2, ]

seed <- 123
surv_cols <- c("status", "time", "rx")

feature_cols <- colnames(dataset)[3:(ncol(dataset) - 1)]

param_list_xgboost <- expand.grid(
  objective = "survival:aft",
  eval_metric = "aft-nloglik",
  subsample = seq(0.6, 1, .2),
  colsample_bytree = seq(0.6, 1, .2),
  min_child_weight = seq(1, 5, 4),
  learning_rate = c(0.1, 0.2),
  max_depth = seq(1, 5, 4)
)
ncores <- 2L

split_vector <- splitTools::multi_strata(
  df = dataset[, .SD, .SDcols = surv_cols],
  strategy = "kmeans",
  k = 4
)

train_x <- model.matrix(
  ~ -1 +.,
  dataset[, .SD, .SDcols = setdiff(feature_cols, surv_cols[1:2])])
)
train_y <- survival::Surv(
  event = (dataset[, get("status")]) |>
```

```

        as.character() |>
        as.integer())),
  time = dataset[, get("time")],
  type = "right"
)

fold_list <- splitTools::create_folds(
  y = split_vector,
  k = 3,
  type = "stratified",
  seed = seed
)

surv_xgboost_aft_optimizer <- mlexperiments::MLCrossValidation$new(
  learner = LearnerSurvXgboostAft$new(
    metric_optimization_higher_better = FALSE
  ),
  fold_list = fold_list,
  ncores = ncores,
  seed = seed
)
surv_xgboost_aft_optimizer$learner_args <- c(as.list(
  data.table::data.table(param_list_xgboost[1, ], stringsAsFactors = FALSE)
),
nrounds = 45L
)
surv_xgboost_aft_optimizer$performance_metric <- c_index

# set data
surv_xgboost_aft_optimizer$set_data(
  x = train_x,
  y = train_y
)

surv_xgboost_aft_optimizer$execute()

## -----
## Method `LearnerSurvXgboostAft$new`
## -----


LearnerSurvXgboostAft$new(metric_optimization_higher_better = FALSE)

```

## Description

The ‘LearnerSurvXgboostCox‘ class is the interface to perform a Cox regression with the ‘xgboost‘ R package for use with the ‘mlexperiments‘ package.

## Details

Optimization metric: needs to be specified with the learner parameter ‘eval\_metric‘. Can be used with \* [mlexperiments::MLTuneParameters] \* [mlexperiments::MLCrossValidation] \* [mlexperiments::MLNestedCVs]

## Super classes

```
mlexperiments::MLLearnerBase -> mllrnrs::LearnerXgboost -> LearnerSurvXgboostCox
```

## Methods

### Public methods:

- `LearnerSurvXgboostCox$new()`
- `LearnerSurvXgboostCox$clone()`

**Method new():** Create a new ‘LearnerSurvXgboostCox‘ object.

*Usage:*

```
LearnerSurvXgboostCox$new(metric_optimization_higher_better)
```

*Arguments:*

`metric_optimization_higher_better` A logical. Defines the direction of the optimization metric used throughout the hyperparameter optimization.

*Returns:* A new ‘LearnerSurvXgboostCox‘ R6 object.

*Examples:*

```
LearnerSurvXgboostCox$new(metric_optimization_higher_better = FALSE)
```

**Method clone():** The objects of this class are cloneable with this method.

*Usage:*

```
LearnerSurvXgboostCox$clone(deep = FALSE)
```

*Arguments:*

`deep` Whether to make a deep clone.

## See Also

[`xgboost::xgb.train()`], [`xgboost::xgb.cv()`]

## Examples

```

# execution time >2.5 sec
# survival analysis

dataset <- survival::colon |>
  data.table::as.data.table() |>
  na.omit()
dataset <- dataset[get("etype") == 2, ]

seed <- 123
surv_cols <- c("status", "time", "rx")

feature_cols <- colnames(dataset)[3:(ncol(dataset) - 1)]

param_list_xgboost <- expand.grid(
  objective = "survival:cox",
  eval_metric = "cox-nloglik",
  subsample = seq(0.6, 1, .2),
  colsample_bytree = seq(0.6, 1, .2),
  min_child_weight = seq(1, 5, 4),
  learning_rate = c(0.1, 0.2),
  max_depth = seq(1, 5, 4)
)
ncores <- 2L

split_vector <- splitTools::multi_strata(
  df = dataset[, .SD, .SDcols = surv_cols],
  strategy = "kmeans",
  k = 4
)

train_x <- model.matrix(
  ~ -1 + .,
  dataset[, .SD, .SDcols = setdiff(feature_cols, surv_cols[1:2])]
)
train_y <- survival::Surv(
  event = (dataset[, get("status")] |>
    as.character() |>
    as.integer()),
  time = dataset[, get("time")],
  type = "right"
)

fold_list <- splitTools::create_folds(
  y = split_vector,
  k = 3,
  type = "stratified",
  seed = seed
)

surv_xgboost_cox_optimizer <- mlexperiments::MLCrossValidation$new(
  learner = LearnerSurvXgboostCox$new(

```

```
    metric_optimization_higher_better = FALSE
),
fold_list = fold_list,
ncores = ncores,
seed = seed
)
surv_xgboost_cox_optimizer$learner_args <- c(as.list(
  data.table::data.table(param_list_xgboost[1, ], stringsAsFactors = FALSE)
),
nrounds = 45L
)
surv_xgboost_cox_optimizer$performance_metric <- c_index

# set data
surv_xgboost_cox_optimizer$set_data(
  x = train_x,
  y = train_y
)
surv_xgboost_cox_optimizer$execute()

## -----
## Method `LearnerSurvXgboostCox$new`
## -----
```

LearnerSurvXgboostCox\$new(metric\_optimization\_higher\_better = FALSE)

# Index

c\_index, [2](#)  
LearnerSurvCoxPHCox, [3](#)  
LearnerSurvGlmnetCox, [5](#)  
LearnerSurvRangerCox, [7](#)  
LearnerSurvRpartCox, [9](#)  
LearnerSurvXgboostAft, [12](#)  
LearnerSurvXgboostCox, [14](#)  
  
mlexperiments::MLLearnerBase, [3](#), [5](#), [7](#), [10](#),  
[12](#), [15](#)  
mllrnrs::LearnerXgboost, [12](#), [15](#)