Package 'survivalmodels'

October 15, 2020

Title Models for Survival Analysis

Version 0.1.0

Description Implementations of classical and machine learning models for survival analysis, including deep neural networks via 'keras' and 'tensorflow'. Each model includes a separated fit and predict interface with consistent prediction types for predicting risk, survival probabilities, or survival distributions with 'distr6' <https://CRAN.R-project.org/package=distr6>. Models are either implemented from 'Python' via 'reticulate' <https://CRAN.R-project.org/package=reticulate>, from code in GitHub packages, or novel implementations using 'Rcpp' <https://CRAN.R-project.org/package=Rcpp>. Novel machine learning survival models wil be included in the package in near-future updates. Neural networks are implemented from the 'Python' package 'pycox' <https://github.com/havakv/pycox> and are detailed by Kvamme et al. (2019) <https://jmlr.org/papers/v20/18-424.html>. The 'Akritas' estimator is defined in Akritas (1994) <doi:10.1214/aos/1176325630>. 'DNNSurv' is defined in Zhao and Feng (2020) <arXiv:1908.02337>.

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URL https://github.com/RaphaelS1/survivalmodels/

BugReports https://github.com/RaphaelS1/survivalmodels/issues

Imports Rcpp (>= 1.0.5)

Suggests distr6 (>= 1.4.4), keras, pseudo, reticulate, survival, testthat

LinkingTo Rcpp

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R topics documented:

`survivalmodels-package`

survivalmodels: Models for Survival Analysis

Description

survivalmodels implements classical and machine learning models for survival analysis that either do not already exist in R or for more efficient implementations.

Author(s)

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See Also

Useful links:

- https://github.com/RaphaelS1/survivalmodels/
- Report bugs at https://github.com/RaphaelS1/survivalmodels/issues

akritas *Akritas Conditional Non-Parametric Survival Estimator*

Description

The Akritas survival estimator is a conditional nearest-neighbours approach to the more common Kaplan-Meier estimator. Common usage includes IPCW Survival models and measures, which do not assume that censoring is independent of the covariates.

Usage

```
akritas(
  formula = NULL,
  data = NULL,
  reverse = FALSE,
  time_variable = "time",
  status_variable = "status",
  x = NULL,
  y = NULL,
  ...
)
```

Arguments

formula	(formula(1)) Object specifying the model fit, left-hand-side of formula should describe a survival::Surv() object.
data	(data.frame(1)) Training data of data.frame like object, internally is coerced with stats::model.matrix().
reverse	(logical(1)) If TRUE fits estimator on censoring distribution, otherwise (default) survival distribution.
time_variable	(character(1)) Alternative method to call the function. Name of the 'time' variable, required if formula. or x and Y not given.
status_variable	(character(1)) Alternative method to call the function. Name of the 'status' variable, required if formula or x and Y not given.
x	(data.frame(1)) Alternative method to call the function. Required if formula, time_variable and status_variable not given. Data frame like object of features which is internally coerced with model.matrix.
y	([survival::Surv()]) Alternative method to call the function. Required if formula, time_variable and status_variable not given. Survival outcome of right-censored observations.

... ANY
Additional arguments, currently unused.

Details

This implementation uses a fit/predict interface to allow estimation on unseen data after fitting on training data. This is achieved by fitting the empirical CDF on the training data and applying this to the new data.

Value

An object inheriting from class `akritas`.

References

Akritas, M. G. (1994). Nearest Neighbor Estimation of a Bivariate Distribution Under Random Censoring. Ann. Statist., 22(3), 1299–1327. doi: 10.1214/aos/1176325630

Examples

```
if (requireNamespaces(c("distr6",  "survival"))) {
  library(survival)
  akritas(Surv(time, status) ~ ., data = rats[1:10, ])
}
```

`build_keras_net` *Build a Keras Multilayer Perceptron*

Description

Utility function to build a Keras MLP.

Usage

```
build_keras_net(
  n_in,
  n_out,
  nodes = c(32L, 32L),
  layer_pars = list(),
  activation = "linear",
  act_pars = list(),
  dropout = 0.1,
  batch_norm = TRUE,
  batch_pars = list()
)
```

Arguments

n_in	(integer(1)) Number of input features.
n_out	(integer(1)) Number of targets.
nodes	(numeric()) Hidden nodes in network, each element in vector represents number of hidden nodes in respective layer.
layer_pars	(list()) Arguments passed to keras::layer_dense.
activation	(character(1)) Activation function passed to keras::layer_activation. Default is linear.
act_pars	(list()) Parameters for activation function, see keras::layer_activation.
dropout	(numeric(1)) Optional dropout layer, if NULL then no dropout layer added otherwise either same dropout will be added to all layers.
batch_norm	(logical(1)) If TRUE (default) then batch normalisation is applied to all layers.
batch_pars	(list()) Parameters for batch normalisation, see keras::layer_batch_normalization.

Details

This function is a helper for R users with less Python experience. Currently it is limited to simple MLPs and with identical layers. More advanced networks will require manual creation with **keras**.

Examples

```
if (requireNamespaces("keras")) {
  build_keras_net(4L, 2L)

  build_keras_net(n_in = 4L, n_out = 2L, nodes = c(32L, 64L, 32L),
    activation = "elu", dropout = 0.4)
}
```

`build_pytorch_net` *Build a Pytorch Multilayer Perceptron*

Description

Utility function to build an MLP with a choice of activation function and weight initialization with optional dropout and batch normalization.

Usage

```
build_pytorch_net(
  n_in,
  n_out,
  nodes = c(32, 32),
  activation = "relu",
  act_pars = list(),
  dropout = 0.1,
  bias = TRUE,
  batch_norm = TRUE,
  batch_pars = list(eps = 1e-05, momentum = 0.1, affine = TRUE),
  init = "uniform",
  init_pars = list()
)
```

Arguments

n_in	(integer(1)) Number of input features.
n_out	(integer(1)) Number of targets.
nodes	(numeric()) Hidden nodes in network, each element in vector represents number of hidden nodes in respective layer.
activation	(character(1)|list()) Activation function, can either be a single character and the same function is used in all layers, or a list of length length(nodes). See get_pycox_activation for options.
act_pars	(list()) Passed to get_pycox_activation.
dropout	(numeric()) Optional dropout layer, if NULL then no dropout layer added otherwise either a single numeric which will be added to all layers or a vector of differing drop-out amounts.
bias	(logical(1)) If TRUE (default) then a bias parameter is added to all linear layers.
batch_norm	(logical(1)) If TRUE (default) then batch normalisation is applied to all layers.
batch_pars	(list()) Parameters for batch normalisation, see reticulate::py_help(torchnnBatchNorm1d).
init	(character(1)) Weight initialization method. See get_pycox_init for options.
init_pars	(list()) Passed to get_pycox_init.

Details

This function is a helper for R users with less Python experience. Currently it is limited to simple MLPs. More advanced networks will require manual creation with **reticulate**.

Examples

```
if (requireNamespaces("reticulate")) {
  build_pytorch_net(4L, 2L, nodes = c(32, 64, 32), activation = "selu")

  # pass parameters to activation and initializer functions
  build_pytorch_net(4L, 2L, activation = "elu", act_pars = list(alpha = 0.1),
  init  = "kaiming_uniform", init_pars = list(mode = "fan_out"))
}
```

---

coxtime *Cox-Time Survival Neural Network*

Description

Cox-Time fits a neural network based on the Cox PH with time-varying effects.

Usage

```
coxtime(
  formula = NULL,
  data = NULL,
  reverse = FALSE,
  time_variable = "time",
  status_variable = "status",
  x = NULL,
  y = NULL,
  frac = 0,
  standardize_time = FALSE,
  log_duration = FALSE,
  with_mean = TRUE,
  with_std = TRUE,
  activation = "relu",
  num_nodes = c(32L, 32L),
  batch_norm = TRUE,
  dropout = NULL,
  device = NULL,
  shrink = 0,
  early_stopping = FALSE,
  best_weights = FALSE,
  min_delta = 0,
```

```
  patience = 10L,
  batch_size = 256L,
  epochs = 1L,
  verbose = FALSE,
  num_workers = 0L,
  shuffle = TRUE,
  ...
)
```

Arguments

formula	(formula(1)) Object specifying the model fit, left-hand-side of formula should describe a survival::Surv() object.
data	(data.frame(1)) Training data of data.frame like object, internally is coerced with stats::model.matrix().
reverse	(logical(1)) If TRUE fits estimator on censoring distribution, otherwise (default) survival distribution.
time_variable	(character(1)) Alternative method to call the function. Name of the 'time' variable, required if formula. or x and Y not given.
status_variable	(character(1)) Alternative method to call the function. Name of the 'status' variable, required if formula or x and Y not given.
x	(data.frame(1)) Alternative method to call the function. Required if formula, time_variable and status_variable not given. Data frame like object of features which is internally coerced with model.matrix.
y	([survival::Surv()]) Alternative method to call the function. Required if formula, time_variable and status_variable not given. Survival outcome of right-censored observations.
frac	(numeric(1)) Fraction of data to use for validation dataset, default is 0 and therefore no separate validation dataset.
standardize_time	(logical(1)) If TRUE, the time outcome is standardized.
log_duration	(logical(1)) If TRUE and standardize_time is TRUE then time variable is log transformed.
with_mean	(logical(1)) If TRUE (default) and standardize_time is TRUE then time variable is centered.
with_std	(logical(1)) If TRUE (default) and standardize_time is TRUE then time variable is scaled to unit variance.

`activation` (character(1))
See get_pycox_activation.

`num_nodes, batch_norm, dropout`
(integer()/logical(1)/numeric(1))
See build_pytorch_net.

`device` (integer(1)|character(1))
Passed to `pycox.models.Coxtime`, specifies device to compute models on.

`shrink` (numeric(1))
Passed to `pycox.models.Coxtime`, shrinkage parameter for regularization.

`early_stopping, best_weights, min_delta, patience`
(logical(1)/logical(1)/numeric(1)/integer(1)
See get_pycox_callbacks.

`batch_size` (integer(1))
Passed to `pycox.models.Coxtime.fit`, elements in each batch.

`epochs` (integer(1))
Passed to `pycox.models.Coxtime.fit`, number of epochs.

`verbose` (logical(1))
Passed to `pycox.models.Coxtime.fit`, should information be displayed during fitting.

`num_workers` (integer(1))
Passed to `pycox.models.Coxtime.fit`, number of workers used in the dataloader.

`shuffle` (logical(1))
Passed to `pycox.models.Coxtime.fit`, should order of dataset be shuffled?

`...` ANY
Passed to get_pycox_optim.

Details

Implemented from the `pycox` Python package via **reticulate**. Calls `pycox.models.Coxtime`.

Value

An object inheriting from class `coxtime`.

An object of class `survivalmodel`.

References

Kvamme, H., Borgan, Ø., & Scheel, I. (2019). Time-to-event prediction with neural networks and Cox regression. Journal of Machine Learning Research, 20(129), 1–30.

Examples

```
if (requireNamespaces("reticulate")) {
  # all defaults
  coxtime(data = simsurvdata(50))
```

```
  # common parameters
  coxtime(data = simsurvdata(50), frac = 0.3, activation = "relu",
    num_nodes = c(4L, 8L, 4L, 2L), dropout = 0.1, early_stopping = TRUE, epochs = 100L,
    batch_size = 32L)
}
```

---

deephit *DeepHit Survival Neural Network*

Description

DeepHit fits a neural network based on the PMF of a discrete Cox model. This is the single (non-competing) event implementation.

Usage

```
deephit(
  formula = NULL,
  data = NULL,
  reverse = FALSE,
  time_variable = "time",
  status_variable = "status",
  x = NULL,
  y = NULL,
  frac = 0,
  cuts = 10,
  cutpoints = NULL,
  scheme = c("equidistant", "quantiles"),
  cut_min = 0,
  activation = "relu",
  custom_net = NULL,
  num_nodes = c(32L, 32L),
  batch_norm = TRUE,
  dropout = NULL,
  device = NULL,
  mod_alpha = 0.2,
  sigma = 0.1,
  early_stopping = FALSE,
  best_weights = FALSE,
  min_delta = 0,
  patience = 10L,
  batch_size = 256L,
  epochs = 1L,
  verbose = FALSE,
  num_workers = 0L,
```

```
  shuffle = TRUE,
  ...
)
```

Arguments

formula	(formula(1)) Object specifying the model fit, left-hand-side of formula should describe a survival::Surv() object.
data	(data.frame(1)) Training data of data.frame like object, internally is coerced with stats::model.matrix().
reverse	(logical(1)) If TRUE fits estimator on censoring distribution, otherwise (default) survival distribution.
time_variable	(character(1)) Alternative method to call the function. Name of the 'time' variable, required if formula. or x and Y not given.
status_variable	(character(1)) Alternative method to call the function. Name of the 'status' variable, required if formula or x and Y not given.
x	(data.frame(1)) Alternative method to call the function. Required if formula, time_variable and status_variable not given. Data frame like object of features which is internally coerced with model.matrix.
y	([survival::Surv()]) Alternative method to call the function. Required if formula, time_variable and status_variable not given. Survival outcome of right-censored observations.
frac	(numeric(1)) Fraction of data to use for validation dataset, default is 0 and therefore no separate validation dataset.
cuts	(integer(1)) If discretise is TRUE then determines number of cut-points for discretisation.
cutpoints	(numeric()) Alternative to cuts if discretise is true, provide exact cutpoints for discretisation. cuts is ignored if cutpoints is non-NULL.
scheme	(character(1)) Method of discretisation, either "equidistant" (default) or "quantiles". See reticulate::py_help(pycox$models$LogisticHazard$label_transform) for more detail.
cut_min	(integer(1)) Starting duration for discretisation, see reticulate::py_help(pycox$models$LogisticHazard$label for more detail.
activation	(character(1)) See get_pycox_activation.

`custom_net` `(torch.nn.modules.module.Module(1))`
Optional custom network built with build_pytorch_net, otherwise default architecture used. Note that if building a custom network the number of output channels depends on `cuts` or `cutpoints`.

`num_nodes, batch_norm, dropout` `(integer()/logical(1)/numeric(1))`
See build_pytorch_net.

`device` `(integer(1)|character(1))`
Passed to `pycox.models.DeepHitSingle`, specifies device to compute models on.

`mod_alpha` `(numeric(1))`
Weighting in (0,1) for combining likelihood (L1) and rank loss (L2). See reference and `py_help(pycox$models$DeepHitSingle)` for more detail.

`sigma` `(numeric(1))`
From eta in rank loss (L2) of ref. See reference and `py_help(pycox$models$DeepHitSingle)` for more detail.

`early_stopping, best_weights, min_delta, patience`
(logical(1)/logical(1)/numeric(1)/integer(1)
See get_pycox_callbacks.

`batch_size` `(integer(1))`
Passed to `pycox.models.DeepHitSingle.fit`, elements in each batch.

`epochs` `(integer(1))`
Passed to `pycox.models.DeepHitSingle.fit`, number of epochs.

`verbose` `(logical(1))`
Passed to `pycox.models.DeepHitSingle.fit`, should information be displayed during fitting.

`num_workers` `(integer(1))`
Passed to `pycox.models.DeepHitSingle.fit`, number of workers used in the dataloader.

`shuffle` `(logical(1))`
Passed to `pycox.models.DeepHitSingle.fit`, should order of dataset be shuffled?

`...` `ANY`
Passed to get_pycox_optim.

Details

Implemented from the `pycox` Python package via **reticulate**. Calls `pycox.models.DeepHitSingle`.

Value

An object inheriting from class `deephit`.

An object of class `survivalmodel`.

References

Changhee Lee, William R Zame, Jinsung Yoon, and Mihaela van der Schaar. Deephit: A deep learning approach to survival analysis with competing risks. In Thirty-Second AAAI Conference on Artificial Intelligence, 2018. http://medianetlab.ee.ucla.edu/papers/AAAI_2018_DeepHit

Examples

```
if (requireNamespaces("reticulate")) {
  # all defaults
  deephit(data = simsurvdata(50))

  # common parameters
  deephit(data = simsurvdata(50), frac = 0.3, activation = "relu",
    num_nodes = c(4L, 8L, 4L, 2L), dropout = 0.1, early_stopping = TRUE, epochs = 100L,
    batch_size = 32L)
}
```

---

deepsurv *DeepSurv Survival Neural Network*

Description

DeepSurv neural fits a neural network based on the partial likelihood from a Cox PH.

Usage

```
deepsurv(
  formula = NULL,
  data = NULL,
  reverse = FALSE,
  time_variable = "time",
  status_variable = "status",
  x = NULL,
  y = NULL,
  frac = 0,
  activation = "relu",
  num_nodes = c(32L, 32L),
  batch_norm = TRUE,
  dropout = NULL,
  device = NULL,
  early_stopping = FALSE,
  best_weights = FALSE,
  min_delta = 0,
  patience = 10L,
  batch_size = 256L,
```

```
  epochs = 1L,
  verbose = FALSE,
  num_workers = 0L,
  shuffle = TRUE,
  ...
)
```

Arguments

formula	(formula(1)) Object specifying the model fit, left-hand-side of formula should describe a survival::Surv() object.
data	(data.frame(1)) Training data of data.frame like object, internally is coerced with stats::model.matrix().
reverse	(logical(1)) If TRUE fits estimator on censoring distribution, otherwise (default) survival distribution.
time_variable	(character(1)) Alternative method to call the function. Name of the 'time' variable, required if formula. or x and Y not given.
status_variable	(character(1)) Alternative method to call the function. Name of the 'status' variable, required if formula or x and Y not given.
x	(data.frame(1)) Alternative method to call the function. Required if formula, time_variable and status_variable not given. Data frame like object of features which is internally coerced with model.matrix.
y	([survival::Surv()]) Alternative method to call the function. Required if formula, time_variable and status_variable not given. Survival outcome of right-censored observations.
frac	(numeric(1)) Fraction of data to use for validation dataset, default is 0 and therefore no separate validation dataset.
activation	(character(1)) See get_pycox_activation.
num_nodes, batch_norm, dropout	(integer()/logical(1)/numeric(1)) See build_pytorch_net.
device	(integer(1)|character(1)) Passed to pycox.models.CoxPH, specifies device to compute models on.
early_stopping, best_weights, min_delta, patience	(logical(1)/logical(1)/numeric(1)/integer(1) See get_pycox_callbacks.
batch_size	(integer(1)) Passed to pycox.models.CoxPH.fit, elements in each batch.

epochs (integer(1))
Passed to pycox.models.CoxPH.fit, number of epochs.

verbose (logical(1))
Passed to pycox.models.CoxPH.fit, should information be displayed during fitting.

num_workers (integer(1))
Passed to pycox.models.CoxPH.fit, number of workers used in the dataloader.

shuffle (logical(1))
Passed to pycox.models.CoxPH.fit, should order of dataset be shuffled?

... ANY
Passed to get_pycox_optim.

Details

Implemented from the pycox Python package via **reticulate**. Calls pycox.models.CoxPH.

Value

An object inheriting from class deepsurv.

An object of class survivalmodel.

References

Katzman, J. L., Shaham, U., Cloninger, A., Bates, J., Jiang, T., & Kluger, Y. (2018). DeepSurv: personalized treatment recommender system using a Cox proportional hazards deep neural network. BMC Medical Research Methodology, 18(1), 24. https://doi.org/10.1186/s12874-018-0482-1

Examples

```
if (requireNamespaces("reticulate")) {
  # all defaults
  deepsurv(data = simsurvdata(50))

  # common parameters
  deepsurv(data = simsurvdata(50), frac = 0.3, activation = "relu",
    num_nodes = c(4L, 8L, 4L, 2L), dropout = 0.1, early_stopping = TRUE, epochs = 100L,
    batch_size = 32L)
}
```

dnnsurv *DNNSurv Neural Network for Conditional Survival Probabilities*

Description

DNNSurv neural fits a neural network based on pseudo-conditional survival probabilities.

Usage

```
dnnsurv(
  formula = NULL,
  data = NULL,
  reverse = FALSE,
  time_variable = "time",
  status_variable = "status",
  x = NULL,
  y = NULL,
  cutpoints = NULL,
  cuts = 5L,
  custom_model = NULL,
  loss_weights = NULL,
  weighted_metrics = NULL,
  optimizer = "adam",
  early_stopping = FALSE,
  min_delta = 0,
  patience = 0L,
  verbose = 0L,
  baseline = NULL,
  restore_best_weights = FALSE,
  batch_size = 32L,
  epochs = 10L,
  validation_split = 0,
  shuffle = TRUE,
  sample_weight = NULL,
  initial_epoch = 0L,
  steps_per_epoch = NULL,
  validation_steps = NULL,
  ...
)
```

Arguments

formula	(formula(1)) Object specifying the model fit, left-hand-side of formula should describe a survival::Surv() object.
data	(data.frame(1)) Training data of data.frame like object, internally is coerced with stats::model.matrix().

`reverse` (`logical(1)`)
If `TRUE` fits estimator on censoring distribution, otherwise (default) survival distribution.

`time_variable` (`character(1)`)
Alternative method to call the function. Name of the 'time' variable, required if `formula`. or `x` and `Y` not given.

`status_variable` (`character(1)`)
Alternative method to call the function. Name of the 'status' variable, required if `formula` or `x` and `Y` not given.

`x` (`data.frame(1)`)
Alternative method to call the function. Required if formula, time_variable and `status_variable` not given. Data frame like object of features which is internally coerced with `model.matrix`.

`y` ([survival::Surv()])
Alternative method to call the function. Required if formula, time_variable and `status_variable` not given. Survival outcome of right-censored observations.

`cutpoints` (`numeric()`)
Points at which to cut survival time into discrete points.

`cuts` (`integer(1)`)
If `cutpoints` not provided then number of equally spaced points at which to cut survival time.

`custom_model` (`keras.engine.training.Model(1)`)
Optional custom architecture built with build_keras_net or directly with **keras**. Output layer should be of length `1` input is number of features plus number of cuts.

`loss_weights, weighted_metrics`
See keras::compile.keras.engine.training.Model.

`optimizer` (`character(1)`)
See get_keras_optimizer.

`early_stopping` (`logical(1)`)
If `TRUE` then early stopping callback is included.

`min_delta, patience, baseline, restore_best_weights`
See keras::callback_early_stopping.

`verbose` (`integer(1)`)
Level of verbosity for printing, `0` or `1`.

`batch_size, epochs, validation_split, shuffle, sample_weight, initial_epoch, steps_per_epoch, validatio`
See keras::fit.keras.engine.training.Model. # nolint

`...` `ANY`
Passed to get_keras_optimizer.

Details

Code for generating the conditional probabilities and pre-processing data is taken from https://github.com/lilizhaoUM/DNNSurv.

Value

An object of class `survivalmodel`.

References

Zhao, L., & Feng, D. (2020). DNNSurv: Deep Neural Networks for Survival Analysis Using Pseudo Values. https://arxiv.org/abs/1908.02337

Examples

```
if (requireNamespaces(c("keras", "pseudo")))
  # all defaults
  dnnsurv(data = simsurvdata(10))

  # setting common parameters
  dnnsurv(time_variable = "time", status_variable = "status", data = simsurvdata(10),
          early_stopping = TRUE, epochs = 100L, validation_split = 0.3)
```

---

get_keras_optimizer *Get Keras Optimizer*

Description

Utility function to construct optimiser from **keras**, primarily for internal use.

Usage

```
get_keras_optimizer(
  optimizer = "adam",
  lr = 0.02,
  beta_1 = 0.9,
  beta_2 = 0.999,
  epsilon = NULL,
  decay = 0,
  clipnorm = NULL,
  clipvalue = NULL,
  schedule_decay = 0.004,
  momentum = 0,
  nesterov = FALSE
)
```

Arguments

`optimizer`	`(character(1))` Optimizer to construct, see details for those available. Default is `"adam"`.
`lr`	`(numeric(1))` Passed to all optimizers except `adadelta` and `adagrad`.
`beta_1, beta_2, epsilon`	`(numeric(1))` Passed to `adamax`, `adam`, and `nadam`.
`decay`	`(numeric(1))` Passed to `adamax`, `adam`, and `sgd`.
`clipnorm, clipvalue`	`(numeric(1))` Passed to `adamax`, `adam`, `nadam`, and `sgd`.
`schedule_decay`	`(numeric(1))` Passed to `nadam`.
`momentum`	`(numeric(1))` Passed to `sgd`.
`nesterov`	`(logical(1))` Passed to `sgd`.

Details

Implemented optimizers are

- `"adadelta"`
  keras::optimizer_adadelta
- `"adagrad"`
  keras::optimizer_adagrad
- `"adamax"`
  keras::optimizer_adamax
- `"adam"`
  keras::optimizer_adam
- `"nadam"`
  keras::optimizer_nadam
- `"rmsprop"`
  keras::optimizer_rmsprop
- `"sgd"`
  keras::optimizer_sgd

Examples

```
if (requireNamespaces("keras")) {
  get_keras_optimizer()
```

```
  get_keras_optimizer(optimizer = "adamax", decay = 0.1, lr = 0.01)
}
```

`get_pycox_activation` *Get Pytorch Activation Function*

Description

Helper function to return a class or constructed object for pytorch activation function from `torch.nn.modules.activation`.

Usage

```
get_pycox_activation(
  activation = "relu",
  construct = TRUE,
  alpha = 1,
  dim = NULL,
  lambd = 0.5,
  min_val = -1,
  max_val = 1,
  negative_slope = 0.01,
  num_parameters = 1L,
  init = 0.25,
  lower = 1/8,
  upper = 1/3,
  beta = 1,
  threshold = 20,
  value = 20
)
```

Arguments

`activation`	`(character(1))` Activation function method, see details for list of implemented methods.
`construct`	`(logical(1))` If `TRUE` (default) returns constructed object, otherwise a class.
`alpha`	`(numeric(1))` Passed to `celu` and `elu`.
`dim`	`(integer(1))` Passed to `glu`, `logsoftmax`, `softmax`, and `softmin`.
`lambd`	`(numeric(1))` Passed to `hardshrink` and `softshrink`.
`min_val, max_val`	`(numeric(1))` Passed to `hardtanh`.

`negative_slope` (`numeric(1)`)
Passed to `leakyrelu`.

`num_parameters` (`integer(1)`)
Passed to `prelu`.

`init` (`numeric(1)`)
Passed to `prelu`.

`lower, upper` (`numeric(1)`)
Passed to `rrelu`.

`beta` (`numeric(1)`)
Passed to `softplus`.

`threshold` (`numeric(1)`)
Passed to `softplus` and `threshold`.

`value` (`numeric(1)`)
Passed to `threshold`.

Details

Implemented methods (with help pages) are

- `"celu"`
  `reticulate::py_help(torch$nn$modules$activation$CELU)`
- `"elu"`
  `reticulate::py_help(torch$nn$modules$activation$ELU)`
- `"gelu"`
  `reticulate::py_help(torch$nn$modules$activation$GELU)`
- `"glu"`
  `reticulate::py_help(torch$nn$modules$activation$GLU)`
- `"hardshrink"`
  `reticulate::py_help(torch$nn$modules$activation$Hardshrink)`
- `"hardsigmoid"`
  `reticulate::py_help(torch$nn$modules$activation$Hardsigmoid)`
- `"hardswish"`
  `reticulate::py_help(torch$nn$modules$activation$Hardswish)`
- `"hardtanh"`
  `reticulate::py_help(torch$nn$modules$activation$Hardtanh)`
- `"relu6"`
  `reticulate::py_help(torch$nn$modules$activation$ReLU6)`
- `"leakyrelu"`
  `reticulate::py_help(torch$nn$modules$activation$LeakyReLU)`
- `"logsigmoid"`
  `reticulate::py_help(torch$nn$modules$activation$LogSigmoid)`
- `"logsoftmax"`
  `reticulate::py_help(torch$nn$modules$activation$LogSoftmax)`

- "prelu"
  reticulate::py_help(torchnnmodules$activation$PReLU)
- "rrelu"
  reticulate::py_help(torchnnmodules$activation$RReLU)
- "relu"
  reticulate::py_help(torchnnmodules$activation$ReLU)
- "selu"
  reticulate::py_help(torchnnmodules$activation$SELU)
- "sigmoid"
  reticulate::py_help(torchnnmodules$activation$Sigmoid)
- "softmax"
  reticulate::py_help(torchnnmodules$activation$Softmax)
- "softmax2d"
  reticulate::py_help(torchnnmodules$activation$Softmax2d)
- "softmin"
  reticulate::py_help(torchnnmodules$activation$Softmin)
- "softplus"
  reticulate::py_help(torchnnmodules$activation$Softplus)
- "softshrink"
  reticulate::py_help(torchnnmodules$activation$Softshrink)
- "softsign"
  reticulate::py_help(torchnnmodules$activation$Softsign)
- "tanh"
  reticulate::py_help(torchnnmodules$activation$Tanh)
- "tanhshrink"
  reticulate::py_help(torchnnmodules$activation$Tanhshrink)
- "threshold"
  reticulate::py_help(torchnnmodules$activation$Threshold)

Examples

```
if (requireNamespaces("reticulate")) {
  #' # returns constructed objects
  get_pycox_activation(activation = "relu", construct = TRUE)

  # returns class
  get_pycox_activation(activation = "selu", construct = FALSE)
}
```

get_pycox_callbacks *Get Torchtuples Callbacks*

Description

Helper function to return torchtuples callbacks from `torchtuples.callbacks`.

Usage

```
get_pycox_callbacks(
  early_stopping = FALSE,
  best_weights = FALSE,
  min_delta = 0,
  patience = 10L
)
```

Arguments

`early_stopping`	`(logical(1))` If `TRUE` then constructs torchtuples.callbacks,EarlyStopping.
`best_weights`	`(logical(1))` If `TRUE` then returns `torchtuples.callbacks.BestWeights`. Ignored if `early_stopping` is `TRUE`.
`min_delta`	`(numeric(1))` Passed to `torchtuples.callbacks.EarlyStopping`.
`patience`	`(integer(1))` Passed to `torchtuples.callbacks.EarlyStopping`.

Examples

```
if (requireNamespaces("reticulate")) {
  get_pycox_callbacks(early_stopping = TRUE)

  get_pycox_callbacks(best_weights = TRUE)
}
```

`get_pycox_init` *Get Pytorch Weight Initialization Method*

Description

Helper function to return a character string with a populated pytorch weight initializer method from `torch.nn.init`. Used in build_pytorch_net to define a weighting function.

Usage

```
get_pycox_init(
  init = "uniform",
  a = 0,
  b = 1,
  mean = 0,
  std = 1,
  val,
  gain = 1,
  mode = c("fan_in", "fan_out"),
  non_linearity = c("leaky_relu", "relu")
)
```

Arguments

`init`	`(character(1))` Initialization method, see details for list of implemented methods.
`a`	`(numeric(1))` Passed to `uniform`, `kaiming_uniform`, and `kaiming_normal`.
`b`	`(numeric(1))` Passed to `uniform`.
`mean, std`	`(numeric(1))` Passed to `normal`.
`val`	`(numeric(1))` Passed to `constant`.
`gain`	`(numeric(1))` Passed to `xavier_uniform`, `xavier_normal`, and `orthogonal`.
`mode`	`(character(1))` Passed to `kaiming_uniform` and `kaiming_normal`, one of `fan_in` (default) and `fan_out`.
`non_linearity`	`(character(1))` Passed to `kaiming_uniform` and `kaiming_normal`, one of `leaky_relu` (default) and `relu`.

Details

Implemented methods (with help pages) are

- "uniform"
  reticulate::py_help(torchnninit$uniform_)
- "normal"
  reticulate::py_help(torchnninit$normal_)
- "constant"
  reticulate::py_help(torchnninit$constant_)
- "xavier_uniform"
  reticulate::py_help(torchnninit$xavier_uniform_)
- "xavier_normal"
  reticulate::py_help(torchnninit$xavier_normal_)
- "kaiming_uniform"
  reticulate::py_help(torchnninit$kaiming_uniform_)
- "kaiming_normal"
  reticulate::py_help(torchnninit$kaiming_normal_)
- "orthogonal"
  reticulate::py_help(torchnninit$orthogonal_)

Examples

```
if (requireNamespaces("reticulate")) {
  get_pycox_init(init = "uniform")

  get_pycox_init(init = "kaiming_uniform", a = 0, mode = "fan_out")
}
```

---

get_pycox_optim *Get Pytorch Optimizer*

Description

Helper function to return a constructed pytorch optimizer from `torch.optim`.

Usage

```
get_pycox_optim(
  optimizer = "adam",
  net,
  rho = 0.9,
  eps = 1e-08,
```

```
  lr = 1,
  weight_decay = 0,
  learning_rate = 0.01,
  lr_decay = 0,
  betas = c(0.9, 0.999),
  amsgrad = FALSE,
  lambd = 1e-04,
  alpha = 0.75,
  t0 = 1e+06,
  momentum = 0,
  centered = TRUE,
  etas = c(0.5, 1.2),
  step_sizes = c(1e-06, 50),
  dampening = 0,
  nesterov = FALSE
)
```

Arguments

`optimizer`	`(character(1))` Optimizer, see details for list of implemented methods.
`net`	`(torch.nn.modules.module.Module)` Network architecture, can be built from build_pytorch_net.
`rho, lr, lr_decay`	`(numeric(1))` Passed to `adadelta`.
`eps`	`(numeric(1))` Passed to all methods except `asgd`, `rprop`, and `sgd`.
`weight_decay`	`(numeric(1))` Passed to all methods except `rprop` and `sparse_adam`.
`learning_rate`	`(numeric(1))` Passed to all methods except `adadelta`.
`betas`	`(numeric(2))` Passed to `adam`, `adamax`, `adamw`, and `sparse_adam`.
`amsgrad`	`(logical(1))` Passed to `adam` and `adamw`.
`lambd, t0`	`(numeric(1))` Passed to `asgd`.
`alpha`	`(numeric(1))` Passed to `asgd` and `rmsprop`.
`momentum`	`(numeric(1))` Passed to `rmsprop` and `sgd`.
`centered`	`(logical(1))` Passed to `rmsprop`.

`etas, step_sizes` (numeric(2))
Passed to rprop.

`dampening` (numeric(1))
Passed to sgd.

`nesterov` (logical(1))
Passed to sgd.

Details

Implemented methods (with help pages) are

- "adadelta"
  `reticulate::py_help(torch$optim$Adadelta)`
- "adagrad"
  `reticulate::py_help(torch$optim$Adagrad)`
- "adam"
  `reticulate::py_help(torch$optim$Adam)`
- "adamax"
  `reticulate::py_help(torch$optim$Adamax)`
- "adamw"
  `reticulate::py_help(torch$optim$AdamW)`
- "asgd"
  `reticulate::py_help(torch$optim$ASGD)`
- "rmsprop"
  `reticulate::py_help(torch$optim$RMSprop)`
- "rprop"
  `reticulate::py_help(torch$optim$Rprop)`
- "sgd"
  `reticulate::py_help(torch$optim$SGD)`
- "sparse_adam"
  `reticulate::py_help(torch$optim$SparseAdam)`

---

install_keras *Install Keras and Tensorflow*

Description

Stripped back version of keras::install_keras.

Usage

```
install_keras(
  method = "auto",
  conda = "auto",
  pip = FALSE,
  install_tensorflow = FALSE
)
```

Arguments

method, conda, pip	See reticulate::py_install.
install_tensorflow	If TRUE installs the dependency tensorflow package as well.

`install_pycox` *Install Pycox With Reticulate*

Description

Installs the python 'pycox' package via reticulate.

Usage

```
install_pycox(
  method = "auto",
  conda = "auto",
  pip = FALSE,
  install_torch = FALSE
)
```

Arguments

method, conda, pip	See reticulate::py_install.
install_torch	If TRUE installs the dependency torch package as well.

install_torch *Install Torch With Reticulate*

Description

Installs the python 'torch' package via reticulate.

Usage

```
install_torch(method = "auto", conda = "auto", pip = FALSE)
```

Arguments

`method, conda, pip`
: See reticulate::py_install

loghaz *Logistic-Hazard Survival Neural Network*

Description

Logistic-Hazard fits a discrete neural network based on a cross-entropy loss and predictions of a discrete hazard function, also known as Nnet-Survival.

Usage

```
loghaz(
  formula = NULL,
  data = NULL,
  reverse = FALSE,
  time_variable = "time",
  status_variable = "status",
  x = NULL,
  y = NULL,
  frac = 0,
  cuts = 10,
  cutpoints = NULL,
  scheme = c("equidistant", "quantiles"),
  cut_min = 0,
  activation = "relu",
  custom_net = NULL,
  num_nodes = c(32L, 32L),
  batch_norm = TRUE,
  dropout = NULL,
  device = NULL,
  early_stopping = FALSE,
```

```
  best_weights = FALSE,
  min_delta = 0,
  patience = 10L,
  batch_size = 256L,
  epochs = 1L,
  verbose = FALSE,
  num_workers = 0L,
  shuffle = TRUE,
  ...
)
```

Arguments

formula	(formula(1)) Object specifying the model fit, left-hand-side of formula should describe a survival::Surv() object.
data	(data.frame(1)) Training data of data.frame like object, internally is coerced with stats::model.matrix().
reverse	(logical(1)) If TRUE fits estimator on censoring distribution, otherwise (default) survival distribution.
time_variable	(character(1)) Alternative method to call the function. Name of the 'time' variable, required if formula. or x and Y not given.
status_variable	(character(1)) Alternative method to call the function. Name of the 'status' variable, required if formula or x and Y not given.
x	(data.frame(1)) Alternative method to call the function. Required if formula, time_variable and status_variable not given. Data frame like object of features which is internally coerced with model.matrix.
y	([survival::Surv()]) Alternative method to call the function. Required if formula, time_variable and status_variable not given. Survival outcome of right-censored observations.
frac	(numeric(1)) Fraction of data to use for validation dataset, default is 0 and therefore no separate validation dataset.
cuts	(integer(1)) If discretise is TRUE then determines number of cut-points for discretisation.
cutpoints	(numeric()) Alternative to cuts if discretise is true, provide exact cutpoints for discretisation. cuts is ignored if cutpoints is non-NULL.
scheme	(character(1)) Method of discretisation, either "equidistant" (default) or "quantiles". See reticulate::py_help(pycox$models$LogisticHazard$label_transform) for more detail.

`cut_min` (integer(1))
Starting duration for discretisation, see `reticulate::py_help(pycox$models$LogisticHazard$label` for more detail.

`activation` (character(1))
See get_pycox_activation.

`custom_net` (torch.nn.modules.module.Module(1))
Optional custom network built with build_pytorch_net, otherwise default architecture used. Note that if building a custom network the number of output channels depends on `cuts` or `cutpoints`.

`num_nodes, batch_norm, dropout`
(integer()/logical(1)/numeric(1))
See build_pytorch_net.

`device` (integer(1)|character(1))
Passed to `pycox.models.LogisticHazard`, specifies device to compute models on.

`early_stopping, best_weights, min_delta, patience`
(logical(1)/logical(1)/numeric(1)/integer(1)
See get_pycox_callbacks.

`batch_size` (integer(1))
Passed to `pycox.models.LogisticHazard.fit`, elements in each batch.

`epochs` (integer(1))
Passed to `pycox.models.LogisticHazard.fit`, number of epochs.

`verbose` (logical(1))
Passed to `pycox.models.LogisticHazard.fit`, should information be displayed during fitting.

`num_workers` (integer(1))
Passed to `pycox.models.LogisticHazard.fit`, number of workers used in the dataloader.

`shuffle` (logical(1))
Passed to `pycox.models.LogisticHazard.fit`, should order of dataset be shuffled?

`...` ANY
Passed to get_pycox_optim.

Details

Implemented from the `pycox` Python package via **reticulate**. Calls `pycox.models.LogisticHazard`.

Value

An object inheriting from class `loghaz`.

An object of class `survivalmodel`.

References

Gensheimer, M. F., & Narasimhan, B. (2018). A Simple Discrete-Time Survival Model for Neural Networks, 1–17. https://doi.org/arXiv:1805.00917v3

Kvamme, H., & Borgan, Ø. (2019). Continuous and discrete-time survival prediction with neural networks. https://doi.org/arXiv:1910.06724.

Examples

```
if (requireNamespaces("reticulate")) {
  # all defaults
  loghaz(data = simsurvdata(50))

  # common parameters
  loghaz(data = simsurvdata(50), frac = 0.3, activation = "relu",
    num_nodes = c(4L, 8L, 4L, 2L), dropout = 0.1, early_stopping = TRUE, epochs = 100L,
    batch_size = 32L)
}
```

---

pchazard *PC-Hazard Survival Neural Network*

Description

Logistic-Hazard fits a discrete neural network based on a cross-entropy loss and predictions of a discrete hazard function, also known as Nnet-Survival.

Usage

```
pchazard(
  formula = NULL,
  data = NULL,
  reverse = FALSE,
  time_variable = "time",
  status_variable = "status",
  x = NULL,
  y = NULL,
  frac = 0,
  cuts = 10,
  cutpoints = NULL,
  scheme = c("equidistant", "quantiles"),
  cut_min = 0,
  activation = "relu",
  custom_net = NULL,
  num_nodes = c(32L, 32L),
```

```
  batch_norm = TRUE,
  reduction = c("mean", "none", "sum"),
  dropout = NULL,
  device = NULL,
  early_stopping = FALSE,
  best_weights = FALSE,
  min_delta = 0,
  patience = 10L,
  batch_size = 256L,
  epochs = 1L,
  verbose = FALSE,
  num_workers = 0L,
  shuffle = TRUE,
  ...
)
```

Arguments

formula	(formula(1)) Object specifying the model fit, left-hand-side of formula should describe a survival::Surv() object.
data	(data.frame(1)) Training data of data.frame like object, internally is coerced with stats::model.matrix().
reverse	(logical(1)) If TRUE fits estimator on censoring distribution, otherwise (default) survival distribution.
time_variable	(character(1)) Alternative method to call the function. Name of the 'time' variable, required if formula. or x and Y not given.
status_variable	(character(1)) Alternative method to call the function. Name of the 'status' variable, required if formula or x and Y not given.
x	(data.frame(1)) Alternative method to call the function. Required if formula, time_variable and status_variable not given. Data frame like object of features which is internally coerced with model.matrix.
y	([survival::Surv()]) Alternative method to call the function. Required if formula, time_variable and status_variable not given. Survival outcome of right-censored observations.
frac	(numeric(1)) Fraction of data to use for validation dataset, default is 0 and therefore no separate validation dataset.
cuts	(integer(1)) If discretise is TRUE then determines number of cut-points for discretisation.

`cutpoints` (numeric())
Alternative to `cuts` if `discretise` is true, provide exact cutpoints for discretisation. `cuts` is ignored if `cutpoints` is non-NULL.

`scheme` (character(1))
Method of discretisation, either "equidistant" (default) or "quantiles". See `reticulate::py_help(pycox$models$LogisticHazard$label_transform)` for more detail.

`cut_min` (integer(1))
Starting duration for discretisation, see `reticulate::py_help(pycox$models$LogisticHazard$label` for more detail.

`activation` (character(1))
See get_pycox_activation.

`custom_net` (torch.nn.modules.module.Module(1))
Optional custom network built with build_pytorch_net, otherwise default architecture used. Note that if building a custom network the number of output channels depends on `cuts` or `cutpoints`.

`num_nodes, batch_norm, dropout`
(integer()/logical(1)/numeric(1))
See build_pytorch_net.

`reduction` (character(1))
How to reduce the loss, see to `reticulate::py_help(pycox$models$loss$NLLPCHazardLoss)`.

`device` (integer(1)|character(1))
Passed to `pycox.models.PCHazard`, specifies device to compute models on.

`early_stopping, best_weights, min_delta, patience`
(logical(1)/logical(1)/numeric(1)/integer(1)
See get_pycox_callbacks.

`batch_size` (integer(1))
Passed to `pycox.models.PCHazard.fit`, elements in each batch.

`epochs` (integer(1))
Passed to `pycox.models.PCHazard.fit`, number of epochs.

`verbose` (logical(1))
Passed to `pycox.models.PCHazard.fit`, should information be displayed during fitting.

`num_workers` (integer(1))
Passed to `pycox.models.PCHazard.fit`, number of workers used in the dataloader.

`shuffle` (logical(1))
Passed to `pycox.models.PCHazard.fit`, should order of dataset be shuffled?

`...` ANY
Passed to get_pycox_optim.

Details

Implemented from the `pycox` Python package via **reticulate**. Calls `pycox.models.PCHazard`.

Value

An object inheriting from class pchazard.

An object of class survivalmodel.

References

Kvamme, H., & Borgan, Ø. (2019). Continuous and discrete-time survival prediction with neural networks. https://doi.org/arXiv:1910.06724.

Examples

```
if (requireNamespaces("reticulate")) {
  # all defaults
  pchazard(data = simsurvdata(50))

  # common parameters
  pchazard(data = simsurvdata(50), frac = 0.3, activation = "relu",
    num_nodes = c(4L, 8L, 4L, 2L), dropout = 0.1, early_stopping = TRUE, epochs = 100L,
    batch_size = 32L)
}
```

---

predict.akritas *Predict method for Akritas Estimator*

Description

Predicted values from a fitted Akritas estimator.

Usage

```
## S3 method for class 'akritas'
predict(
  object,
  newdata,
  times = NULL,
  lambda = 0.5,
  type = c("survival", "risk", "all"),
  distr6 = FALSE,
  ...
)
```

Arguments

object	(akritas(1)) Object of class inheriting from "akritas".
newdata	(data.frame(1)) Testing data of data.frame like object, internally is coerced with stats::model.matrix(). If missing then training data from fitted object is used.
times	(numeric()) Times at which to evaluate the estimator. If NULL (default) then evaluated at all unique times in the training set.
lambda	(numeric(1)) Bandwidth parameter for uniform smoothing kernel in nearest neighbours estimation. The default value of 0.5 is arbitrary and should be chosen by the user.
type	(numeric(1)) Type of predicted value. Choices are survival probabilities over all time-points in training data ("survival") or a relative risk ranking ("risk"), which is the mean cumulative hazard function over all time-points, or both ("all").
distr6	(logical(1)) If FALSE (default) and type is "survival" or "all" returns matrix of survival probabilities, otherwise returns a distr6::VectorDistribution().
...	ANY Currently ignored.

Details

This implementation uses a fit/predict interface to allow estimation on unseen data after fitting on training data. This is achieved by fitting the empirical CDF on the training data and applying this to the new data.

Value

A numeric if type = "risk", a distr6::VectorDistribution() (if distr6 = TRUE) and type = "survival"; a matrix if (distr6 = FALSE) and type = "survival" where entries are survival probabilities with rows of observations and columns are time-points; or a list combining above if type = "all".

References

Akritas, M. G. (1994). Nearest Neighbor Estimation of a Bivariate Distribution Under Random Censoring. Ann. Statist., 22(3), 1299–1327. doi: 10.1214/aos/1176325630

Examples

```
if (requireNamespaces(c("distr6", "survival"))) {

library(survival)

train <- 1:10
```

```
test <- 11:20
fit <- akritas(Surv(time, status) ~ ., data = rats[train, ])
predict(fit, newdata = rats[test, ])

# when lambda = 1, identical to Kaplan-Meier
fit <- akritas(Surv(time, status) ~ ., data = rats[1:100, ])
predict_akritas <- predict(fit, newdata = rats[1:100, ], lambda = 1)[1, ]
predict_km <- survfit(Surv(time, status) ~ 1, data = rats[1:100, ])$surv
all(predict_akritas == predict_km)

# Use distr6 = TRUE to return a distribution
predict_distr <- predict(fit, newdata = rats[test, ], distr6 = TRUE)
predict_distr$survival(100)

# Return a relative risk ranking with type = "risk"
predict(fit, newdata = rats[test, ], type = "risk")

# Or survival probabilities and a rank
predict(fit, newdata = rats[test, ], type = "all", distr6 = TRUE)
}
```

`predict.dnnsurv` *Predict Method for DNNSurv*

Description

Predicted values from a fitted object of class dnnsurv.

Usage

```
## S3 method for class 'dnnsurv'
predict(
  object,
  newdata,
  batch_size = 32L,
  verbose = 0L,
  steps = NULL,
  callbacks = NULL,
  type = c("survival", "risk", "all"),
  distr6 = FALSE,
  ...
)
```

Arguments

`object` `(dnnsurv(1))`
Object of class inheriting from `"dnnsurv"`.

newdata (data.frame(1))
Testing data of data.frame like object, internally is coerced with stats::model.matrix(). If missing then training data from fitted object is used.

batch_size (integer(1))
Passed to keras::predict.keras.engine.training.Model, elements in each batch.

verbose (integer(1))
Level of verbosity for printing, 0 or 1.

steps (integer(1))
Number of batches before evaluation finished, see keras::predict.keras.engine.training.Model.

callbacks (list())
Optional callbacks to apply during prediction.

type (numeric(1))
Type of predicted value. Choices are survival probabilities over all time-points in training data ("survival") or a relative risk ranking ("risk"), which is the mean cumulative hazard function over all time-points, or both ("all").

distr6 (logical(1))
If FALSE (default) and type is "survival" or "all" returns matrix of survival probabilities, otherwise returns a distr6::VectorDistribution().

... ANY
Currently ignored.

Value

A numeric if type = "risk", a distr6::VectorDistribution() (if distr6 = TRUE) and type = "survival"; a matrix if (distr6 = FALSE) and type = "survival" where entries are survival probabilities with rows of observations and columns are time-points; or a list combining above if type = "all".

Examples

```
if (requireNamespaces(c("keras", "pseudo")))
  fit <- dnnsurv(data = simsurvdata(10))

  # predict survival matrix and relative risks
  predict(fit, simsurvdata(10), type = "all")

  # return as distribution
  if (requireNamespaces("distr6")) {
    predict(fit, simsurvdata(10), distr6 = TRUE)
  }
```

predict.pycox *Predict Method for pycox Neural Networks*

Description

Predicted values from a fitted pycox ANN.

Usage

```
## S3 method for class 'pycox'
predict(
  object,
  newdata,
  batch_size = 256L,
  num_workers = 0L,
  interpolate = FALSE,
  inter_scheme = c("const_hazard", "const_pdf"),
  sub = 10L,
  type = c("survival", "risk", "all"),
  distr6 = FALSE,
  ...
)
```

Arguments

object	(pycox(1)) Object of class inheriting from "pycox".
newdata	(data.frame(1)) Testing data of data.frame like object, internally is coerced with stats::model.matrix(). If missing then training data from fitted object is used.
batch_size	(integer(1)) Passed to pycox.models.X.fit, elements in each batch.
num_workers	(integer(1)) Passed to pycox.models.X.fit, number of workers used in the dataloader.
interpolate	(logical(1)) For models deephit and loghaz, should predictions be linearly interpolated? Ignored for other models.
inter_scheme	(character(1)) If interpolate is TRUE then the scheme for interpolation, see reticulate::py_help(py_help(pycox$m for further details.
sub	(integer(1)) If interpolate is TRUE or model is loghaz, number of sub-divisions for inter- polation. See reticulate::py_help(py_help(pycox$models$DeepHitSingle$interpolate))‘ for further details.

type	(numeric(1)) Type of predicted value. Choices are survival probabilities over all time-points in training data ("survival") or a relative risk ranking ("risk"), which is the mean cumulative hazard function over all time-points, or both ("all").
distr6	(logical(1)) If FALSE (default) and type is "survival" or "all" returns matrix of survival probabilities, otherwise returns a distr6::VectorDistribution().
...	ANY Currently ignored.

Value

A numeric if type = "risk", a distr6::VectorDistribution() (if distr6 = TRUE) and type = "survival"; a matrix if (distr6 = FALSE) and type = "survival" where entries are survival probabilities with rows of observations and columns are time-points; or a list combining above if type = "all".

Examples

```
if (requireNamespaces("reticulate")) {
  fit <- coxtime(data = simsurvdata(50))

  # predict survival matrix and relative risks
  predict(fit, simsurvdata(10), type = "all")

  # return as distribution
  if (requireNamespaces("distr6")) {
    predict(fit, simsurvdata(10), distr6 = TRUE)
  }
}
```

---

pycox_prepare_train_data

Prepare Data for Pycox Model Training

Description

Utility function to prepare data for training in a Pycox model. Generally used internally only.

Usage

```
pycox_prepare_train_data(
  x_train,
  y_train,
  frac = 0,
```

```
  standardize_time = FALSE,
  log_duration = FALSE,
  with_mean = TRUE,
  with_std = TRUE,
  discretise = FALSE,
  cuts = 10L,
  cutpoints = NULL,
  scheme = c("equidistant", "quantiles"),
  cut_min = 0L,
  model = c("coxtime", "deepsurv", "deephit", "loghaz", "pchazard")
)
```

Arguments

x_train	(matrix(1)) Training covariates.
y_train	(matrix(1)) Training outcomes.
frac	(numeric(1)) Fraction of data to use for validation dataset, default is 0 and therefore no separate validation dataset.
standardize_time	(logical(1)) If TRUE, the time outcome to be standardized. For use with coxtime.
log_duration	(logical(1)) If TRUE and standardize_time is TRUE then time variable is log transformed.
with_mean	(logical(1)) If TRUE (default) and standardize_time is TRUE then time variable is centered.
with_std	(logical(1)) If TRUE (default) and standardize_time is TRUE then time variable is scaled to unit variance.
discretise	(logical(1)) If TRUE then time is discretised. For use with the models deephit, pchazard, and loghaz.
cuts	(integer(1)) If discretise is TRUE then determines number of cut-points for discretisation.
cutpoints	(numeric()) Alternative to cuts if discretise is true, provide exact cutpoints for discretisation. cuts is ignored if cutpoints is non-NULL.
scheme	(character(1)) Method of discretisation, either "equidistant" (default) or "quantiles". See reticulate::py_help(pycox$models$LogisticHazard$label_transform).
cut_min	(integer(1)) Starting duration for discretisation, see reticulate::py_help(pycox$models$LogisticHazard$label
model	(character(1)) Corresponding pycox model.

requireNamespaces *Vectorised Logical requireNamespace*

Description

Vectorises the requireNamespace function and returns TRUE if all packages, x, are available and FALSE otherwise.

Usage

```
requireNamespaces(x)
```

Arguments

x	(character()) string naming the packages/name spaces to load.

Examples

```
requireNamespaces("survivalmodels")
requireNamespaces(c("survivalmodels", "distr6"))

# test for some fake package
requireNamespaces(c("survivalmodels", "distr6", "111"))
```

simsurvdata *Simulate Survival Data*

Description

Function for simulating survival data.

Usage

```
simsurvdata(n = 100, trt = 2, age = 2, sex = 1.5, cutoff = 20)
```

Arguments

n	(integer(1)) Number of samples
trt, age, sex	(numeric(1)) Coefficients for covariates.
cutoff	(numeric(1)) Cutoff for determining Type I censoring.

Details

Currently limited to three covariates, Weibull survival times, and Type I censoring. This will be expanded to a flexible simulation function in future updates. For now the function is primarily limited to helping function examples.

Value

`data.frame()`

Examples

```
simsurvdata()
```

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