# Package 'BioPred'

June 6, 2024

Type Package
Title An R Package for Biomarkers Analysis in Precision Medicine
Version 1.0.1
<b>Date</b> 2024-05-30
Maintainer Zihuan Liu <zihuan.liu@abbvie.com></zihuan.liu@abbvie.com>
Description
Provides functions for training extreme gradient boosting model using propensity score A-learning and weight-learning methods. For details on the A-learning and weight-learning methods, see Chen et al. (2017) <doi:10.1111 biom.12676="">.</doi:10.1111>
Encoding UTF-8
Language en
License GPL-3
<b>Imports</b> xgboost, pROC, ggplot2, PropCIs, survival, survminer, mgcv, onewaytests, car
RoxygenNote 7.3.1
Suggests knitr, rmarkdown, kableExtra
VignetteBuilder knitr
NeedsCompilation no
<b>Depends</b> R (>= 4.0.0)
LazyData true
Author Zihuan Liu [aut, cre], Yan Sun [aut], Xin Huang [aut]
Repository CRAN
<b>Date/Publication</b> 2024-06-06 16:50:09 UTC
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cat\_summary

Summarize Categorical Variables in Subgroup

# Description

This function provides a summary of categorical variables in a dataset.

# Usage

```
cat_summary(
  yvar,
  yname,
  xvars,
  xname.list,
  data,
  yvar.display = yvar,
  xvars.display = xvars
)
```

# Arguments

yvar Name of the variable for summary.

yname A vector of ordered y values.

xvars Names of the variables for grouping.

cdf\_plot 3

xname.list A list (same order as xvars) of ordered x values for each xvar.

data The dataset.

yvar.display Display name for yvar. xvars.display Display name for xvars.

#### Value

A list containing the contingency table, frequency table, and percentage table.

cdf\_plot CDF Plot for a biomarker

## **Description**

Cumulative Distribution Function (CDF) plot for a biomarker.

# Usage

```
cdf_plot(xvar, data, y.int = 5, xlim = NULL, xvar.display = xvar, group = NULL)
```

## **Arguments**

xvar The biomarker name.

data The dataset.

y. int Increasement interval on the y.

xlim cdf plot range for xvar, when NULL, c(min(x), max(x)) will be used.

xvar.display Display name of the biomarker.

group A separate CDF line will be plotted for each group.

# Value

A ggplot object representing the CDF inverse plot.

4 cut\_perf

	_
CHT	perf
Cut	DCII

Cutoff Performance Evaluation

# Description

This function evaluates the performance of a predictive model at a selected cutoff point.

# Usage

```
cut_perf(
  yvar,
  censorvar = NULL,
  xvar,
  cutoff,
  dir,
  xvars.adj = NULL,
  data,
  type,
  yvar.display = yvar,
  xvar.display = xvar
)
```

## **Arguments**

vvar	Response variable name.

censorvar Censoring variable name (0-censored, 1-event).

xvar Biomarker name.

cutoff Selected cutoff value.

dir Direction for desired subgroup (">", ">=", "<", "<=").

xvars.adj Other covariates to adjust when evaluating the performance.

data Data frame containing the variables.

type Type of analysis: "c" for continuous, "s" for survival, and "b" for binary.

yvar.display Display name of response variable. xvar.display Display name of biomarker variable.

#### Value

A list containing various performance metrics and optionally, plots.

eval\_metric\_bin 5

eval_metric_bin Evaluation Metrics for XGBoostSub_bin Model	
---	--

# **Description**

Function for evaluating XGBoostSub\_bin model performance.

#### Usage

```
eval_metric_bin(model, X_feature, y_label, pi, trt, Loss_type = "A_learning")
```

# Arguments

model The trained XGBoostSub\_bin model object.

X\_feature The input features matrix.

y\_label The input y matrix.

pi The propensity scores vector, which should range from 0 to 1, representing the

probability of assignment to treatment.

trt The treatment indicator vector. Should take values of 1 or -1, where 1 represents

the treatment group and -1 represents the control group.

Loss\_type Type of loss function to use: "A\_learning" or "Weight\_learning".

#### **Details**

eval\_metric: Function for Evaluating XGBoostSub\_bin Model Performance

This function evaluates the performance of an XGBoostSub\_bin model using a A-learning or weight-learning function.

#### Value

Evaluation result of the XGBoostSub\_bin model.

eval_metric_con	Evaluation Metrics for XGBoostSub_con Model	
-----------------	---	--

# Description

Function for evaluating XGBoostSub\_con model performance.

```
eval_metric_con(model, X_feature, y_label, pi, trt, Loss_type = "A_learning")
```

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#### **Arguments**

model The trained XGBoostSub\_con model object.

X\_feature The input features matrix.

y\_label The input y matrix.

pi The propensity scores vector, which should range from 0 to 1, representing the

probability of assignment to treatment.

trt The treatment indicator vector. Should take values of 1 or -1, where 1 represents

the treatment group and -1 represents the control group.

Loss\_type Type of loss function to use: "A\_learning" or "Weight\_learning".

#### **Details**

eval\_metric: Function for Evaluating XGBoostSub\_con Model Performance

This function evaluates the performance of an XGBoostSub\_con model using a A-learning or weight-learning function.

#### Value

Evaluation result of the XGBoostSub\_con model.

eval\_metric\_sur

Evaluation Metrics for XGBoostSub\_sur Model

## **Description**

 $Function\ for\ evaluating\ XGBoostSub\_sur\ model\ performance.$ 

```
eval_metric_sur(
  model,
  X_feature,
  y_label,
  pi,
  trt,
  censor,
  Loss_type = "A_learning"
)
```

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## **Arguments**

model The trained XGBoostSub\_sur model object.

X\_feature The input features matrix.

y\_label The input y matrix.

pi The propensity scores vector, which should range from 0 to 1, representing the probability of assignment to treatment.

trt The treatment indicator vector. Should take values of 1 or -1, where 1 represents the treatment group and -1 represents the control group.

censor The censor status vector. Should take values of 1 or 0, where 1 represents censoring and 0 represents an observed event.

#### **Details**

Loss\_type

eval\_metric: Function for Evaluating XGBoostSub\_con Model Performance

This function evaluates the performance of an XGBoostSub\_con model using a A-learning or weight-learning function.

Type of loss function to use: "A\_learning" or "Weight\_learning".

#### Value

Evaluation result of the XGBoostSub\_sur model.

fixcut_bin	Fixed Cutoff Analysis for Individual Biomarker Associated with Binary Outcome Variables

#### **Description**

This function conducts fixed cutoff analysis for individual biomarker associated with binary outcome variables.

```
fixcut_bin(
  yvar,
  xvar,
  dir,
  cutoffs,
  data,
  method = "Fisher",
  yvar.display = yvar,
  xvar.display = xvar,
  vert.x = FALSE
)
```

fixcut\_con

#### Arguments

yvar Binary response variable name. 0 represents controls and 1 represents cases. xvar Biomarker name. dir Cutoff direction for the desired subgroup. Options are ">", ">=", "<", or "<=". cutoffs A vector of candidate cutoffs. data The dataset containing the variables. method Method for cutoff selection. Options are "Fisher", "Youden", "Conc.Prob", "Accuracy", or "Kappa". - "Fisher": Minimizes the Fisher test p-value. - "Youden": Maximizes the Youden index. - "Conc.Prob": Maximizes sensitivity \* specificity. - "Accuracy": Maximizes accuracy. - "Kappa": Maximizes Kappa coefficient. yvar.display Display name of the response variable. xvar.display Display name of the predictor variable.

#### Value

vert.x

A list containing statistical summaries, selected cutoff statistics, selected cutoff value, confusion matrix, and a ggplot object for visualization.

Whether to display the cutoff in a 90-degree angle when plotting (saves space).

fixcut\_con Fixed Cutoff Analysis for Individual Biomarker Associated with Continuous Outcome

# Description

This function conducts fixed cutoff analysis for individual biomarker associated with continuous outcome variables.

```
fixcut_con(
  yvar,
  xvar,
  dir,
  cutoffs,
  data,
  method = "t.test",
  yvar.display = yvar,
  xvar.display = xvar,
  vert.x = FALSE
)
```

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#### **Arguments**

yvar	Continuous response variable name.
xvar	Biomarker name.
dir	Cutoff direction for the desired subgroup. Options are ">", ">=", "<", or "<=".
cutoffs	A vector of candidate cutoffs.
data	The dataset containing the variables.
method	Method for cutoff selection. Currently only supports "t.test" "t.test": Minimizes the t-test p-value.
yvar.display	Display name of the response variable.
xvar.display	Display name of the predictor variable.
vert.x	Whether to display the cutoff in a 90-degree angle when plotting (saves space).

#### Value

A list containing statistical summaries, selected cutoff statistics, selected cutoff value, group statistics, and a ggplot object for visualization.

fixcut_sur Fixed Cutoff Analysis for Individual Biomarker Associated with Survival Outcome	~_
--	----

# Description

This function conducts fixed cutoff analysis for Individual Biomarker Associated with survival outcome variables.

```
fixcut_sur(
   yvar,
   censorvar,
   xvar,
   dir,
   cutoffs,
   data,
   method = "logrank",
   yvar.display = yvar,
   xvar.display = xvar,
   vert.x = FALSE
)
```

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#### **Arguments**

yvar Survival response variable name. Censoring variable. 0 indicates censored, 1 indicates an event. censorvar Biomarker name. xvar Cutoff direction for the desired subgroup. Options are ">", ">=", "<", or "<=". dir cutoffs A vector of candidate cutoffs. data The dataset containing the variables. Method for cutoff selection. Currently only supports "logrank". - "logrank": method Minimizes the logrank test p-value. yvar.display Display name of the response variable. xvar.display Display name of the predictor variable.

#### Value

vert.x

A list containing statistical summaries, selected cutoff statistics, selected cutoff value, group statistics, and a ggplot object for visualization.

Whether to display the cutoff in a 90-degree angle when plotting (saves space).

|--|

#### **Description**

Computes and plots the contrasts between treatment and control group based on a GAM for exploring the relationship be-tween treatment benefit and biomarker.

```
gam_ctr_plot(
 yvar,
  censorvar = NULL,
 xvar,
 xvars.adj = NULL,
  sxvars.adj = NULL,
  trtvar = NULL,
  type,
  data,
  k,
  title = "Group Contrast",
 ybreaks = NULL,
  xbreaks = NULL,
  rugcol.var = NULL,
  link.scale = TRUE,
  prt.sum = TRUE,
```

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```
prt.chk = FALSE,
  outlier.rm = FALSE
)
```

#### **Arguments**

yvar Response variable name.

censorvar Censoring variable name (0-censored, 1-event). Required if type is "s" (sur-

vival).

xvar Biomarker name.

xvars.adj Potential confounding variables to adjust for using linear terms.

sxvars.adj Potential confounding variables to adjust for using curves.

trtvar Treatment variable that the contrast will build upon (treatment-control).

type Type of response variable. Options are "c" for continuous, "s" for survival, and

"b" for binary response.

data The dataset containing the variables.

k Upper limit on the degrees of freedom associated with an s smooth. When this k

is too large, program will report error saying

title Title of the plot.

ybreaks Breaks on the y-axis. xbreaks Breaks on the x-axis.

rugcol.var Variable name that defines the color of the rug.

link.scale Whether to show the plot (y-axis) in the scale of the link function (linear predic-

tor).

prt.sum Whether to print summary or not. prt.chk Whether to print model diagnosis.

outlier.rm Whether to remove outliers based on 1.5IQR.

#### Value

A list containing the p-value table, summarized p-value table, s-value table, summarized s-value table, and the plot.

gam_plot	GAM Plot	
----------	----------	--

#### **Description**

Generates a generalized additive model (GAM) plot for exploring the relationship between a response variable and a biomarker.

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# Usage

```
gam_plot(
  yvar,
  censorvar = NULL,
  xvar,
  xvars.adj = NULL,
  sxvars.adj = NULL,
  type,
  data,
  k,
  pred.type = "iterms",
  link.scale = TRUE,
  title = "Trend Plot",
  ybreaks = NULL,
  xbreaks = NULL,
  rugcol.var = NULL,
  add.points = FALSE,
  prt.sum = TRUE,
  prt.chk = FALSE,
  outlier.rm = FALSE,
  newdat = NULL
)
```

# Arguments

yvar	Response variable name.
censorvar	Censoring variable name for survival analysis (0-censored, 1-event).
xvar	Biomarker name.
xvars.adj	Potential confounding variables to adjust for using linear terms.
sxvars.adj	Potential confounding variables to adjust for using curve terms.
type	"c" for continuous, "s" for survival, and "b" for binary response.
data	The dataset containing the variables.
k	Upper limit on the degrees of freedom associated with an s smooth.
pred.type	"iterms" for trend of xvar, "response" for Y at the original scale.
link.scale	Whether to show the plot in the scale of the link function.
title	Title of the plot.
ybreaks	Breaks on the y-axis.
xbreaks	Breaks on the x-axis.
rugcol.var	Variable name defining the color of the rug and points.
add.points	Whether to add data points to the plot.
prt.sum	Whether to print summary or not.
prt.chk	Whether to print model diagnosis.
outlier.rm	Whether to remove outliers based on 1.5IQR.
newdat	User-supplied customized data for prediction and plotting.

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#### Value

A list containing p-table, s-table, GAM summary, GAM check, and the plot.

```
get_subgroup_results Get Subgroup Results
```

## **Description**

This function predicts the treatment assignment for each patient based on a cutoff value.

# Usage

```
get_subgroup_results(model, X_feature, subgroup_label = NULL, cutoff = 0.5)
```

# **Arguments**

model The trained XGBoost-based subgroup model.

X\_feature The data matrix containing patient features.

subgroup\_label (Optional) The subgroup labels. In real-world data, this information is typically

unknown and only available in simulated data. If provided, the prediction accu-

racy will also be returned.

cutoff The cutoff value for treatment assignment, defaulted to 0.5.

#### Value

A data frame containing each subject and assigned treatment (1 for treatment, 0 for control). If subgroup labels are provided, it also returns the prediction accuracy of the subgroup labels.

```
predictive_biomarker_imp
```

Plot Predictive Biomarker Importance based on XGBoost-based Subgroup Model

# **Description**

This function calculates and plots the importance of biomarkers in a trained XGBoostSub\_con, XGBoostSub\_bin or XGBoostSub\_sur model.

#### Usage

```
predictive_biomarker_imp(model)
```

#### **Arguments**

model The trained XGBoost-based model.

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#### Value

A barplot showing the biomarker importance.

roc\_bin AUC ROC Table for Biomarkers Associated with Binary Outcomes

## Description

Computes the area under the receiver operating characteristic (ROC) curve for Biomarkers Associated with Binary Outcomes, and returns the results as a table.

# Usage

```
roc_bin(yvar, xvars, dirs, data, yvar.display = yvar, xvars.display = xvars)
```

# **Arguments**

yvar	Binary response variable name,	where 0 represents controls	and 1 represents

cases.

xvars A vector of biomarker names.

dirs A vector of directions for the biomarkers. Options are "auto", ">", or "<". -

"auto" (default): automatically determines in which group the median is higher and takes the direction accordingly. - ">": indicates that the biomarkers for the control group are higher than those for the case group (controls > t >= cases). - "<": indicates that the biomarkers for the control group are lower or equal to

those for the case group (controls < t <= cases).

data The dataset containing the variables.

yvar.display Display name for the binary response variable.

xvars.display Display names for the biomarkers.

# Value

A table containing the AUC values for each biomarker.

roc\_bin\_plot

roc_bin_plot	ROC Plot Biomarkers Associated with Binary Outcomes

# Description

Generates ROC plots for different biomarkers associated with binary outcomes.

# Usage

```
roc_bin_plot(
  yvar,
  xvars,
  dirs,
  data,
  yvar.display = yvar,
  xvars.display = xvars
)
```

# Arguments

yvar	Binary response variable name, where 0 represents controls and 1 represents cases.
xvars	A vector of biomarker names.
dirs	A vector of directions for the biomarkers. Options are "auto", ">", or "<" "auto" (default): automatically determines in which group the median is higher and takes the direction accordingly ">" indicates that the biomarkers for the control group are higher than those for the case group (controls > $t >= cases$ ) "<" indicates that the biomarkers for the control group are lower or equal to those for the case group (controls < $t <= cases$ ).
data	The dataset containing the variables.
yvar.display	Display name for the binary response variable.

# Value

xvars.display

ROC plots for different biomarkers associated with binary outcomes.

Display names for the biomarkers.

subgrp\_perf

scat_cont_plot	Scatter Plot for a Biomarker Associated with Continuous Outcome

# **Description**

Generates a scatter plot for exploring the relationship between a continuous response variable and a biomarker variable.

# Usage

```
scat_cont_plot(
  yvar,
  xvar,
  data,
  ybreaks = NULL,
  xbreaks = NULL,
  yvar.display = yvar,
  xvar.display = xvar
)
```

## **Arguments**

yvar Continuous response variable name.

xvar biomarker name.

data The dataset containing the variables.

ybreaks Breaks on the y-axis.

xbreaks Breaks on the x-axis.

yvar.display Display name for the response variable.

xvar.display Display name for the biomarker variable.

# Value

A list containing correlation coefficients, scatter plot, slope, and intercept.

subgrp_perf	Subgroup Performance Evaluation for Prognostic Cases

# Description

This function evaluates subgroup performance based on different types of response variables.

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#### Usage

```
subgrp_perf(
  yvar,
  censorvar = NULL,
  grpvar,
  grpname,
  xvars.adj = NULL,
  data,
  type,
  yvar.display = yvar,
  grpvar.display = grpvar
)
```

#### **Arguments**

yvar The response variable name.

censorvar (Optional) The censoring variable name (0-censored, 1-event).

grpvar The subgroup variable name.

grpname A vector of ordered subgroup names (values in the column of grpvar).

xvars.adj (Optional) Other covariates to adjust when evaluating the performance.

data The dataset containing the variables.

type The type of response variable: "c" for continuous, "s" for survival, and "b" for

binary.

yvar.display Display name of the response variable.

grpvar.display Display name of the group variable.

# Value

A list containing subgroup performance results including logrank p-value, median and mean survival, Cox model p-value, ANOVA p-value, and more based on the specified response variable type.

subgrp\_perf\_pred Subgroup Performance Evaluation for Predictive Cases

## Description

This function evaluates the performance of subgroups based on different types of response variables in predictive cases.

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#### Usage

```
subgrp_perf_pred(
  yvar,
  censorvar = NULL,
  grpvar,
  grpname,
  trtvar,
  trtname,
  xvars.adj = NULL,
  data,
  type,
  yvar.display = yvar,
  grpvar.display = grpvar,
  trtvar.display = trtvar
)
```

#### **Arguments**

yvar Response variable name.

censorvar Censoring variable name (0-censored, 1-event).

grpvar Subgroup variable name.

grpname A vector of ordered subgroup names (values in the column of grpvar).

trtvar Treatment variable name.

trtname A vector of ordered treatment names (values in the column of trtvar).

xvars.adj Other covariates to adjust when evaluating the performance.

data The dataset.

type "c" for continuous; "s" for "survival", and "b" for binary.

yvar.display Display name of the response variable. grpvar.display Display name of the group variable. trtvar.display Display name of the treatment variable.

## Value

A list containing the comparison results, group results, and possibly a plot.

tutorial_data

# Description

A dataset containing sample data for demonstrating the functionalities of the BioPred package.

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#### Usage

```
data(tutorial_data)
```

#### **Format**

A data frame with the following columns:

- x1 Numeric. A biomarker variable.
- x2 Numeric. A biomarker variable.
- **x3** Numeric. A biomarker variable.
- x4 Numeric. A biomarker variable.
- x5 Numeric. A biomarker variable.
- **x6** Numeric. A biomarker variable.
- x7 Numeric. A biomarker variable.
- x8 Numeric. A biomarker variable.
- x9 Numeric. A biomarker variable.
- x10 Numeric. A biomarker variable.
- **y.con** Numeric. A continuous outcome variable.
- **y.bin** Binary. A binary outcome variable, where 0 represents one class and 1 represents another class.
- y.time Numeric. The time in months, used for survival analysis.
- **y.event** Binary. Event indicator variable, where 0 indicates censoring and 1 indicates the event of interest occurred.
- **subgroup\_label** Binary. Ground truth of subgroup label. In real-world scenarios, this information is typically unavailable.
- **treatment** Binary. Treatment indicator variable, where 0 represents control and 1 represents treatment.
- **treatment\_categorical** Factor. A categorical version of the treatment variable, with levels "Placebo" and "Treatment".
- risk\_category Factor.

#### **Details**

This dataset is used to illustrate various functions within the BioPred package, including predictive modeling and subgroup analysis. The columns represent different types of data typically encountered in clinical studies.

## **Examples**

```
data(tutorial_data)
head(tutorial_data)
```

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XGBoostSub_bin	XGBoost Model with Modified Loss Function for Subgroup Identifica- tion with Binary Outcomes

# Description

Function for training XGBoost model with customized loss function for binary outcomes

# Usage

```
XGBoostSub_bin(
  X_data,
  y_data,
  trt,
  pi,
  Loss_type = "A_learning",
  params = list(),
  nrounds = 50,
  disable_default_eval_metric = 1,
  verbose = TRUE
)
```

## **Arguments**

X_data	The input features matrix.	
y_data	The input y matrix.	
trt	The treatment indicator vector. Should take values of 1 or -1, where 1 represents the treatment group and -1 represents the control group.	
pi	The propensity scores vector, which should range from 0 to 1, representing the probability of assignment to treatment.	
Loss_type	Type of loss function to use: "A_learning" or "Weight_learning".	
params	A list of additional parameters for the xgb.train function.	
nrounds	Number of boosting rounds. Default is 50.	
disable_default_eval_metric		
	If 1, default evaluation metric will be disabled.	
verbose	Logical. If TRUE, training progress will be printed; if FALSE, no progress will be printed.	

# **Details**

XGBoostSub\_bin: Function for Training XGBoost Model with Customized Loss Function for binary outcomes

This function trains an XGBoost model using a customized loss function based on the A-learning and weight-learning.

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This function requires the 'xgboost' library. Make sure to install and load the 'xgboost' library before using this function.

After running this function, the returned model can be used like a regular xgboost model.

# Value

Trained XGBoostSub\_bin model.

XGBoostSub_con	XGBoost Model with Modified Loss Function for Subgroup Identification with Continuous Outcomes

# Description

Function for training XGBoost model with customized loss function for continuous outcomes

## Usage

```
XGBoostSub_con(
  X_data,
  y_data,
  trt,
  pi,
  Loss_type = "A_learning",
  params = list(),
  nrounds = 50,
  disable_default_eval_metric = 1,
  verbose = TRUE
)
```

#### **Arguments**

X_data	The input features matrix.	
y_data	The input y matrix.	
trt	The treatment indicator vector. Should take values of 1 or -1, where 1 represents the treatment group and -1 represents the control group.	
pi	The propensity scores vector, which should range from 0 to 1, representing the probability of assignment to treatment.	
Loss_type	Type of loss function to use: "A_learning" or "Weight_learning".	
params	A list of additional parameters for the xgb.train function.	
nrounds	Number of boosting rounds. Default is 50.	
disable_default_eval_metric		
	If 1, default evaluation metric will be disabled.	
verbose	Logical. If TRUE, training progress will be printed; if FALSE, no progress will be printed.	

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#### **Details**

XGBoostSub\_con: Function for Training XGBoost Model with Customized Loss Function for continuous outcomes

This function trains an XGBoost model using a customized loss function based on the A-learning and weight-learning.

This function requires the 'xgboost' library. Make sure to install and load the 'xgboost' library before using this function.

After running this function, the returned model can be used like a regular xgboost model.

#### Value

Trained XGBoostSub\_con model.

XGBoostSub\_sur XGBoost Model with Modified Loss Function for Subgroup Identification with Survival Outcomes

## **Description**

Function for training XGBoost model with customized loss function for survival outcomes

# Usage

```
XGBoostSub_sur(
   X_data,
   y_data,
   trt,
   pi,
   censor,
   Loss_type = "Weight_learning",
   params = list(),
   nrounds = 50,
   disable_default_eval_metric = 1,
   verbose = TRUE
)
```

#### **Arguments**

X_data	The input features matrix.
y_data	The input y matrix.
trt	The treatment indicator vector. Should take values of 1 or -1, where 1 represents the treatment group and -1 represents the control group.
pi	The propensity scores vector, which should range from 0 to 1, representing the probability of assignment to treatment.

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censor The censor status vector. Should take values of 1 or 0, where 1 represents cen-

soring and 0 represents an observed event.

Loss\_type Type of loss function to use: "A\_learning" or "Weight\_learning".

params A list of additional parameters for the xgb.train function.

nrounds Number of boosting rounds. Default is 50.

disable\_default\_eval\_metric

If 1, default evaluation metric will be disabled.

verbose Logical. If TRUE, training progress will be printed; if FALSE, no progress will

be printed.

#### **Details**

XGBoostSub\_sur: Function for Training XGBoost Model with Customized Loss Function for survival outcomes

This function trains an XGBoost model using a customized loss function based on the A-learning and weight-learning.

This function requires the 'xgboost' library. Make sure to install and load the 'xgboost' library before using this function.

#### Value

Trained XGBoostSub\_sur model.

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