

Package ‘uplifteval’

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Type Package

Title Uplift Model Evaluation with Plots and Metrics

Version 0.1.0

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Description Provides a variety of plots and metrics to evaluate

uplift models including the 'R uplift' package's Qini metric and Qini plot,
a port of the 'python pylift' module's plotting function, and
an alternative plot (in beta) useful for continuous outcomes.

Background: Radcliffe (2007) <<https://pdfs.semanticscholar.org/147b/32f3d56566c8654a9999c5477dded233328e.pdf>>.

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URL <https://github.com/ras44/uplifteval>

BugReports <https://github.com/ras44/uplifteval/issues>

Encoding UTF-8

LazyData true

Suggests testthat, knitr, rmarkdown, grf, tweedie, qpdf

Depends R (>= 3.0.0),

Imports ggplot2, whisker, gridExtra, dplyr

RoxygenNote 6.1.1

VignetteBuilder knitr

NeedsCompilation no

Repository CRAN

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<i>new_PlUpliftEval</i>	<i>A non-S3 "constructor" function that returns a list representing a pylift uplift eval object, PlUpliftEval. This object contains metrics and can be used to generate plots.</i>
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Description

A non-S3 "constructor" function that returns a list representing a pylift uplift eval object, PlUpliftEval. This object contains metrics and can be used to generate plots.

Usage

```
new_PlUpliftEval(treatment = integer(), outcome = integer(),
                  prediction = numeric(), p = "infer", n_bins = 20)
```

Arguments

<i>treatment</i>	numeric vector of treatment identifiers
<i>outcome</i>	numeric vector of outcomes
<i>prediction</i>	numeric vector of uplift predictions
<i>p</i>	optional "infer", numeric, numeric vector representing treatment propensities
<i>n_bins</i>	integer number of bins on x-axis; default 20

Value

a list representing a pylift uplift eval object

<i>plot_uplift</i>	<i>Creates an uplift plot of cumulative differential treatment/control outcomes versus model score. Also provides a selection of metrics: max uplift as pct of total control outcome, optimum users targeted and optimum score targeting range.</i>
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Description

Creates an uplift plot of cumulative differential treatment/control outcomes versus model score. Also provides a selection of metrics: max uplift as pct of total control outcome, optimum users targeted and optimum score targeting range.

Usage

```
plot_uplift(p1, W, Y, ns = min(table(W)), n_bs = 1, W_label = W,
            p0 = rep(0, length(p1)), balanced = TRUE, replace = TRUE,
            x_interval = 0.1, ...)
```

Arguments

p1	numeric vector of uplift predictions; can also be predicted outcomes for treated case (in this case p0 should contain predicted outcomes for the control case)
W	binary vector 1,0 of treatment assignments
Y	numeric vector of responses
ns	integer number of samples per bootstrap iteration; default min(table(W))
n_bs	integer number of bootstrap iterations
W_label	optional labels for the treatment options (default W)
p0	optional numeric vector of predicted outcomes for control case
balanced	optional boolean whether to sample equal proportions from treatment and control cases; default TRUE
replace	optional boolean whether to use replacement when sampling; default TRUE
x_interval	optional numeric the interval with which to split the additional arguments (unused) x-axis
...	

Examples

```
set.seed(0)
rl <- function(x){
  round(1/(1+exp(-x)))
}
n <- 2000; p <- 3
beta <- -0.5
X <- matrix(rnorm(n*p), n, p)
W <- rbinom(n, 1, 0.5)
Y <- rl(pmax(beta+X[,1], 0) * W + X[,2])
p1 <- 1/(1+exp(-(beta+X[,1])))
plot_uplift(p1, W, Y, n_bs=20, x_interval = 0.05, balanced = TRUE)

set.seed(0)
n <- 2000; p <- 3
beta <- -0.5
X <- matrix(rnorm(n*p), n, p)
W <- rbinom(n, 1, 0.5)
Y <- pmax(beta+X[,1], 0) * W + X[,2]
p1 <- 1/(1+exp(-(beta+X[,1])))
plot_uplift(p1, W, Y, n_bs=20, x_interval = 0.05, balanced = TRUE)
```

```

library(grf)
set.seed(123)

rl <- function(x){
  round(1/(1+exp(-x)))
}

n = 2000; p = 10
X = matrix(rnorm(n*p), n, p)
W = rbinom(n, 1, 0.2)
Y = rl(X[,1]) * W - rl(X[,3]) * W + rnorm(n)
tau.forest = causal_forest(X, Y, W)
tau.hat = predict(tau.forest, X)
plot_uplift(tau.hat$predictions, W, Y, n_bs=20, x_interval = 0.05, balanced = FALSE)
plot_uplift(tau.hat$predictions, W, Y, n_bs=20, x_interval = 0.05, balanced = TRUE)

```

`plot_uplift_guelman` A direct copy of Leo Guelman's `uplift::qini` function available in the R `uplift` package at commit 95965272e71c312623c95c439fb0b84f95c185b7: <https://github.com/cran/uplift/blob/95965272e71c312623c95c439fb0b84f95c185b7/R/qini.R#L5>

Description

A direct copy of Leo Guelman's `uplift::qini` function available in the R `uplift` package at commit 95965272e71c312623c95c439fb0b84f95c185b7: <https://github.com/cran/uplift/blob/95965272e71c312623c95c439fb0b84f95c185b7/R/qini.R#L5>

Usage

```
plot_uplift_guelman(p1, W, Y, p0 = rep(0, length(p1)), plotit = TRUE,
                     direction = 1, groups = 10)
```

Arguments

<code>p1</code>	vector of numeric uplift predictions. Some uplift models produce two predictions: if-treated and if-control. In this case, if-treated predictions can be provided as <code>p1</code> , and if-control predictions can be provided as <code>p0</code> .
<code>W</code>	vector of 0,1 treatment indicators
<code>Y</code>	vector of 0,1 outcomes
<code>p0</code>	vector of numeric control predictions (default 0)
<code>plotit</code>	boolean plot the Qini chart
<code>direction</code>	1: calculate the differential response as <code>p1-p0</code> , 2: <code>p0-p1</code>
<code>groups</code>	5, 10, or 20: the number of quantiles in which to divide the population

Examples

```

set.seed(0)
rl <- function(x){
  round(1/(1+exp(-x)))
}
n <- 2000; p <- 3
beta <- -0.5
X <- matrix(rnorm(n*p), n, p)
W <- rbinom(n, 1, 0.5)
Y <- rl(pmax(beta+X[,1], 0) * W + X[,2])
p1 <- 1/(1+exp(-(beta+X[,1])))
plot_uplift_guelman(p1, W, Y, groups=10, plotit=TRUE)

library(grf)
set.seed(123)

alpha <- 0.1
n <- 1000
W <- rbinom(n, 1, 0.5)
Y <- W
p1 <- Y + alpha*rnorm(n)
plot_uplift_guelman(p1, W, Y, groups=10)

rl <- function(x){
  round(1/(1+exp(-x)))
}
n <- 2000; p = 10
X <- matrix(rnorm(n*p), n, p)
W <- rbinom(n, 1, 0.2)
Y <- rl(X[,1]) * W - rl(X[,3]) * W + rnorm(n)
tau.forest <- causal_forest(X, Y, W)
tau.hat <- predict(tau.forest, X)
plot_uplift_guelman(tau.hat$predictions, W, Y)

```

`plUpliftEval`

A helper for the new _PlUpEval function that validates the treatment, outcome, prediction, p, and n_bins arguments.

Description

A helper for the new _PlUpEval function that validates the treatment, outcome, prediction, p, and n_bins arguments.

Usage

```
plUpliftEval(treatment, outcome, prediction, p = "infer", n_bins = 20)
```

Arguments

treatment	numeric vector of treatment identifiers
outcome	numeric vector of outcomes
prediction	numeric vector of uplift predictions
p	optional "infer", numeric, numeric vector representing treatment propensities
n_bins	integer number of bins on x-axis; default 20

Value

a list representing a pylift uplift eval object

Examples

```
set.seed(0)
rl <- function(x){
  round(1/(1+exp(-x)))
}
n <- 2000; p <- 3
beta <- -0.5
X <- matrix(rnorm(n*p), n, p)
W <- rbinom(n, 1, 0.5)
Y <- rl(pmax(beta+X[,1], 0) * W + X[,2])
p1 <- 1/(1+exp(-(beta+X[,1])))
plUpliftEval(W, Y, p1)

library(grf)
set.seed(123)

rl <- function(x){
  round(1/(1+exp(-x)))
}
n <- 2000; p <- 10
X <- matrix(rnorm(n*p), n, p)
W <- rbinom(n, 1, 0.2)
Y <- rl(X[,1]) * W - rl(X[,3]) * W + rnorm(n)
tau.forest <- causal_forest(X, Y, W)
tau.hat <- predict(tau.forest, X)
plue <- plUpliftEval(W, Y, tau.hat$predictions)
plue
```

<code>pl_plot</code>	A port of pylift's plot function (https://github.com/wayfair/pylift) as of commit: https://github.com/wayfair/pylift/tree/bb69692388b1fe085001c3ba7edf6dd81d888353 <i>pylift: Plots the different kinds of percentage-targeted curves.</i>
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Description

A port of pylift's plot function (<https://github.com/wayfair/pylift>) as of commit: <https://github.com/wayfair/pylift/tree/bb69692388b1fe085001c3ba7edf6dd81d888353>
pylift: Plots the different kinds of percentage-targeted curves.

Usage

```
pl_plot(plue, plot_type = "cgains", n_bins = 20,
       show_theoretical_max = FALSE, show_practical_max = FALSE,
       show_random_selection = TRUE, show_no_dogs = FALSE, ...)
```

Arguments

<code>plue</code>	the result of a call to the plUpliftEval constructor
<code>plot_type</code>	string, optional Either 'qini', 'aqini', 'uplift', 'cuplift', or 'balance'. 'aqini' refers to an adjusted qini plot, 'cuplift' gives a cumulative uplift plot. 'balance' gives the test-control balance for each of the bins. All others are self-explanatory.
<code>n_bins</code>	integer, number of population bins; default 20
<code>show_theoretical_max</code>	boolean, optional Toggle theoretical maximal qini curve, if overfitting to treatment/control. Only works for Qini-style curves.
<code>show_practical_max</code>	boolean, optional Toggle theoretical maximal qini curve, if not overfitting to treatment/control. Only works for Qini-style curves.
<code>show_random_selection</code>	boolean, optional Toggle straight line indicating a random ordering. Only works for Qini-style curves.
<code>show_no_dogs</code>	boolean, optional Toggle theoretical maximal qini curve, if you believe there are no sleeping dogs. Only works for Qini-style curves.
<code>...</code>	additional arguments

Value

a pylift plot

Examples

```

set.seed(0)
rl <- function(x){
  round(1/(1+exp(-x)))
}
n <- 2000; p <- 3
beta <- -0.5
X <- matrix(rnorm(n*p), n, p)
W <- rbinom(n, 1, 0.5)
Y <- rl(pmax(beta+X[,1], 0) * W + X[,2])
p1 <- 1/(1+exp(-(beta+X[,1])))
pue <- plUpLiftEval(W, Y, p1)
pl_plot(pue,
         show_practical_max = TRUE,
         show_theoretical_max = TRUE,
         show_no_dogs = TRUE,
         n_bins=20)

library(grf)
set.seed(123)

rl <- function(x){
  round(1/(1+exp(-x)))
}
n <- 2000; p <- 10
X <- matrix(rnorm(n*p), n, p)
W <- rbinom(n, 1, 0.2)
Y <- rl(X[,1]) * W - rl(X[,3]) * W + rnorm(n)
tau.forest <- causal_forest(X, Y, W)
tau.hat <- predict(tau.forest, X)
pue <- plUpLiftEval(W, Y, tau.hat$predictions)
pue
pl_plot(pue,
         show_practical_max = TRUE,
         show_theoretical_max = TRUE,
         show_no_dogs = TRUE,
         n_bins=20)

```

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