Package ‘cjbart’

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Title Heterogeneous Effects Analysis of Conjoint Experiments

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Description A tool for analyzing conjoint experiments using Bayesian Additive Regression Trees ('BART'), a machine learning method developed by Chipman, George and McCulloch (2010) <doi:10.1214/09-AOAS285>. This tool focuses specifically on estimating, identifying, and visualizing the heterogeneity within marginal component effects, at the observation- and individual-level. It uses a variable importance measure ('VIMP') with delete-d jackknife variance estimation, following Ishwaran and Lu (2019) <doi:10.1002/sim.7803>, to obtain bias-corrected estimates of which variables drive heterogeneity in the predicted individual-level effects.

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R topics documented:

- cjbart ................................................................. 2
- het_vimp .............................................................. 3
- IMCE ................................................................. 4
- plot.cjbart .......................................................... 6
- plot.cjbart.vimp .................................................. 7
- rf_vimp .............................................................. 8
- RMCE ............................................................... 8
- summary.cjbart .................................................... 9

Index 11

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cjbart Generate Conjoint Model Using BART

Description

A wrapper for the BART::pbart() function.

Usage

cjbart(data, Y, id = NULL, round = NULL, use_round = TRUE, cores = 1, ...)

Arguments

data A data.frame, containing all attributes, controls, the outcome and id variables to analyze.
Y Character string – the outcome variable
id Character string – variable identifying individual respondents (optional)
round Character string – variable identifying rounds of the conjoint experiment
use_round Boolean – whether to include the round indicator column when training the BART model (default = TRUE)
cores Integer – number of CPU cores used in model training
... Other arguments passed to BART::pbart()

Details

Please note, cjbart currently only works for a binary outcome.

Value

A trained BART::pbart() model that can be passed to IMCE()

See Also

BART::pbart()
Examples

subjects <- 5
rounds <- 2
profiles <- 2
obs <- subjects*rounds*profiles

fake_data <- data.frame(A = sample(c("a1","a2"), obs, replace = TRUE),
                        B = sample(c("b1","b2"), obs, replace = TRUE),
                        id1 = rep(1:subjects, each=rounds),
                        stringsAsFactors = TRUE)

fake_data$Y <- sample(c(0,1), obs, replace = TRUE)

cj_model <- cjbart(data = fake_data,
                    Y = "Y",
                    id = "id1")

het_vimp
Estimate Variable Importance Metrics for cjbart Object

Description

Estimates random forest variable importance scores for multiple attribute-levels of a conjoint experiment.

Usage

het_vimp(model, outcomes = NULL, covars = NULL)

Arguments

model Object of class cjbart, the result of running IMCE()
outcomes An optional vector of attribute levels to generate importance metrics for. By
default, all attribute-levels are analyzed.
covars An optional vector of covariates to include in the importance metric check. By
default, all covariates are included in each importance model.

Details

Having generated a schedule of individual-level marginal component effect estimates, this function fits a random forest model for each attribute-level using the supplied covariates as predictors. It then calculates a variable importance measure (VIMP) for each covariate. The VIMP method assesses how important each covariate is in terms of partitioning the predicted individual-level effects distribution, and can thus be used as an indicator of which variables drive heterogeneity in the IMCEs.
To recover a VIMP measure, we used permutation-based importance metrics recovered from random forest models estimated using `randomForestSRC::rfsrc()`. To permute the data, this function uses random node assignment, whereby cases are randomly assigned to a daughter node whenever a tree splits on the target variable (see Ishwaran et al. 2008). Importance is defined in terms of how random node assignment degrades the performance of the forest. Higher degradation indicates a variable is more important to prediction.

Variance estimates of each variable’s importance are subsequently recovered using the delete-d jackknife estimator developed by Ishwaran and Lu (2019). The jackknife method has inherent bias correction properties, making it particularly effective for variable selection exercises such as identifying drivers of heterogeneity.

**Value**

A "long" data.frame of variable importance scores for each combination of covariates and attribute-levels, as well as the estimated 95% confidence intervals for each metric.

**References**


**See Also**

`randomForestSRC::rfsrc()` and `randomForestSRC::subsample()`

**IMCE**

Heterogeneous Effects Analysis of Conjoint Results

**Description**

IMCE calculates the individual-level marginal component effects from a BART-estimated conjoint model.

**Usage**

```r
IMCE(
  data,
  model,
  attribs,
  ref_levels,
  method = "bayes",
  alpha = 0.05,
  keep_omce = FALSE,
  cores = 1,
  skip_checks = FALSE
)
```
Arguments

data A data.frame, containing all attributes, covariates, the outcome and id variables to analyze.

model A model object, the result of running cjbart()

attrs Vector of attribute names

ref_levels Vector of reference levels, used to calculate marginal effects

method Character string, setting the variance estimation method to use. When method is "parametric", a typical combined variance estimate is employed; when method = "bayes", the 95% posterior interval is calculated; and when method = "rubin", combination rules are used to combine the variance analogous to in multiple imputation analysis.

alpha Number between 0 and 1 – the significance level used to compute confidence/posterior intervals. When method = "bayes", the posterior interval is calculated by taking the alpha/2 and (1-alpha/2) quantiles of the posterior draws. When method = "rubin", the confidence interval equals the IMCE +/- qnorm(alpha/2). By default, alpha is 0.05 i.e. generating a 95% confidence/posterior interval.

keep_omce Boolean, indicating whether to keep the OMCE-level results (default = FALSE)

cores Number of CPU cores used during prediction phase

skip_checks Boolean, indicating whether to check the structure of the data (default = FALSE). Only set this to TRUE if you are confident that the data is structured appropriately

Details

The OMCE estimates are the result of subtracting the predicted value of each observation under the reference-level category from the predicted value of each observation under the given attribute level. If an attribute has k levels, then this will yield k-1 estimates per observation. The IMCE is the average of the OMCEs for each individual within the data.

Value

IMCE returns an object of type "cjbart", a list object.

omce A data.frame containing the observation-level marginal effects
imce A data.frame containing the individual-level marginal effects
imce_upper A data.frame containing the upper bound of the IMCE confidence/credible interval
imce_lower A data.frame containing the lower bound of the IMCE confidence/credible interval
att_levels A vector containing the attribute levels

See Also

cjbart()
Examples

```r
subjects <- 5
rounds <- 2
profiles <- 2
obs <- subjects*rounds*profiles
fake_data <- data.frame(A = sample(c("a1","a2"), obs, replace = TRUE),
                        B = sample(c("b1","b2"), obs, replace = TRUE),
                        id1 = rep(1:subjects, each=rounds),
                        stringsAsFactors = TRUE)
fake_data$Y <- sample(c(0,1), obs, replace = TRUE)
cj_model <- cjbart(data = fake_data, 
                    Y = "Y", 
                    id = "id1")
## Skip if not Unix due to longer CPU time
if (.Platform$OS.type="unix") {
het_effects <- IMCE(data = fake_data, 
                    model = cj_model, 
                    attribs = c("A","B"), 
                    ref_levels = c("a1","b1"), 
                    cores = 1)
summary(het_effects)
}
```

plot.cjbart

Plot Marginal Component Effects of a \texttt{cjbart} Object

Description

Plots observation-level or individual-level marginal component effects (OMCE and IMCE respectively). By default, all attribute-levels in the model are plotted.

Usage

```r
## S3 method for class 'cjbart'
plot(x, covar = NULL, plot_levels = NULL, se = TRUE, ...)
```

Arguments

- **x** Object of class \texttt{cjbart}, the result of running \texttt{IMCE()}
- **covar** Character string detailing the covariate over which to analyze heterogeneous effects


**plot.cjbart.vimp**

Optional vector of conjoint attribute names to plot. If not supplied, all attributes within the conjoint model will be plotted.

**se**

Boolean determining whether to show an estimated 95% confidence interval

**...**

Additional arguments for plotting the marginal component effects (see below).

---

**Value**

Plot of marginal component effects.

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**Description**

Plots a heatmap of variable importance, across predicted IMCEs. By default, all attribute-levels and covariates in the model are plotted.

**Usage**

```r
## S3 method for class 'cjbart.vimp'
plot(x, covars = NULL, att_levels = NULL, ...)
```

**Arguments**

- **x** Object of class `cjbart`, the result of running `IMCE()`
- **covars** Optional vector of covariate names to plot. By default, all included covariates are shown.
- **att_levels** Optional vector of attribute-levels to plot. By default, all attribute-levels are shown.
- **...** Additional arguments (not currently used)

**Value**

Plot of covariate importance scores
rf_vimp

Estimate a Single Variable Importance Metric for cjbart Object

Description
Estimates random forest variable importance scores for a single attribute-level of a conjoint experiment. This function is for advanced use. Users should typically use the \texttt{het_vimp()} function.

Usage
\begin{verbatim}
rf_vimp(model, outcome, covars = NULL)
\end{verbatim}

Arguments
\begin{itemize}
\item \texttt{model} Object of class \texttt{cjbart}, the result of running \texttt{IMCE()}
\item \texttt{outcome} Character string detailing the covariate over which to analyze heterogeneous effects
\item \texttt{covars} An optional vector of covariates to include in the importance metric check. When \texttt{covars = NULL} (the default), all covariates are included in the importance model.
\end{itemize}

Value
Data.frame of variable importance scores for each covariate in the model, as well as values for the estimated 95\% confidence interval for each importance score.

RMCE

Inspect Round-Level Marginal Component Effect (RMCE)

Description
RMCE calculates the round-level marginal component effects from a cjbart model.

Usage
\begin{verbatim}
RMCE(imces)
\end{verbatim}

Arguments
\begin{itemize}
\item \texttt{imces} An object of class "cjbart", the result of calling the \texttt{IMCE} function
\end{itemize}

Details
The RMCE estimates are the result of averaging the OMCEs within each round, for each subject in the experiment. The RMCE is the intermediate causal quantity between OMCEs and IMCEs, and can be useful for inspecting whether there are any carryover or stability issues across rounds.
Value

IMCE returns a data frame of RMCEs.

See Also

cjbart() and IMCE()

Description

summary method for class "cjbart"

Usage

## S3 method for class 'cjbart'
summary(object, ...)

Arguments

object Object of class cjbart, the result of running IMCE()
...
Further arguments (not currently used)

Value

Data frame summarizing the average marginal component effect, the minimum and maximum values, and standard deviations for each attribute-level.

Examples

subjects <- 5
rounds <- 2
profiles <- 2
obs <- subjects*rounds*profiles

fake_data <- data.frame(A = sample(c("a1","a2"), obs, replace = TRUE),
                    B = sample(c("b1","b2"), obs, replace = TRUE),
                    id1 = rep(1:subjects, each=rounds),
                    stringsAsFactors = TRUE)

fake_data$Y <- sample(c(0,1), obs, replace = TRUE)

cj_model <- cjbart(data = fake_data,
                    Y = "Y",
                    id = "id1")

## Skip if not Unix due to longer CPU time
if (.Platform$OS.type=='unix') {

het_effects <- IMCE(data = fake_data,
                     model = cj_model,
                     attrs = c("A","B"),
                     ref_levels = c("a1","b1"),
                     cores = 1)

summary(het_effects)
}

Index

BART::pbart(), 2

cjbart, 2
cjbart(). 5, 9

het_vimp, 3
het_vimp(). 8

IMCE, 4
IMCE(). 2, 3, 6–9

plot.cjbart, 6
plot.cjbart.vimp, 7

randomForestSRC::rfsrc(), 4
randomForestSRC::subsamp(), 4
rf_vimp, 8
RMCE, 8

summary.cjbart, 9