Package ‘causact’

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Type Package
Title Accelerated Bayesian Analytics with DAGs
Version 0.4.1
Description Accelerate Bayesian analytics workflows in 'R' through interactive modelling, visualization, and inference. Define probabilistic graphical models using directed acyclic graphs (DAGs) as a unifying language for business stakeholders, statisticians, and programmers. This package relies on the sleek and elegant 'greta' package for Bayesian inference. 'greta', in turn, is an interface into 'TensorFlow' from 'R'. Install 'greta' using instructions available here: <https://www.causact.com/install-tensorflow-greta-and-causact.html>. See <https://github.com/flyaflya/causact> or <https://www.causact.com/> for more documentation.

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BugReports https://github.com/flyaflya/causact/issues

SystemRequirements Python and TensorFlow are needed for Bayesian inference computations; Python (>= 2.7.0) with header files and shared library; TensorFlow (= v1.14; https://www.tensorflow.org/); TensorFlow Probability (= v0.7.0; https://www.tensorflow.org/probability/)

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R topics documented:

- addPriorGroups ........................................ 3
- baseballData ........................................... 3
- beachLocDF ............................................ 4
- carModelDF ............................................ 5
- chimpanzeesDF ....................................... 5
- corruptDF ............................................. 6
- dagg_plot ............................................. 7
- dag_create ............................................ 9
- dag_diagrammer ...................................... 9
- dag_dim ............................................... 10
- dag_edge .............................................. 11
- dag_greta ............................................ 11
- dag_merge ............................................ 13
- dag_node ............................................. 14
- dag_plate ............................................ 16
- dag_render .......................................... 18
- delivDF .............................................. 19
- generate_dot2 ....................................... 20
- grViz2 ................................................. 20
- gymDF ................................................. 21
- houseDF ............................................... 22
- houseDFDescr ....................................... 23
- meaningfulLabels .................................... 24
- prodLineDF .......................................... 25
- rbern .................................................. 25
- replaceLabels ....................................... 26
- replace_in_spec2 .................................... 27
- schoolsDF ........................................... 27
- setDirectedGraphTheme ................................ 28
- ticketsDF ............................................ 28
- totalBeachgoersRepSample ........................... 29
- %>% ................................................... 29

Index 30
**addPriorGroups**

Add a column to tidy dataframe of draws that groups parameters by their prior distribution. All parameters with the same prior distribution receive the same index.

### Description

Add a column to tidy dataframe of draws that groups parameters by their prior distribution. All parameters with the same prior distribution receive the same index.

### Usage

```r
addPriorGroups(drawsDF)
```

### Arguments

- `drawsDF` : the dataframe created by `dag_greta()` where each row represents one draw of MCMC output. Two columns are expected, `param` - the parameter name, `value` - the realized value, and a third column, `priorGroup`, is appended as an integer grouping parameters by their prior distributions. The data for this third column is stored in an environment called `cacheEnv` when the `dag_greta()` function is called. Any parameters with the same prior end up in the same prior group. Used by `dagp_plot()` to group parameters when plotted.

### Value

a tidy dataframe of posterior draws. Useful for passing to `dagp_plot()` or for creating plots using `ggplot()`.

---

**baseballData**

Dataframe of 12,145 observations of baseball games in 2010

### Description

Dataframe of 12,145 observations of baseball games in 2010

### Usage

```r
baseballData
```
Format

A data frame with 12145 rows and 5 variables:

- **Date**  date game was played
- **Home**  abbreviation for home team (i.e. stadium where game played)
- **Visitor**  abbreviation for visiting team
- **HomeScore**  Runs scored by the home team
- **VisitorScore**  Runs scored by the visiting team

beachLocDF

Dataframe where each row represents data about one of the 26 mile markers (fake) from mile 0 to mile 2.5 along the Ocean City, MD beach/boardwalk.

Description

Dataframe where each row represents data about one of the 26 mile markers (fake) from mile 0 to mile 2.5 along the Ocean City, MD beach/boardwalk.

Usage

beachLocDF

Format

A data frame with 26 rows and 3 variables:

- **mileMarker**  a number representing a location on the Ocean City beach/boardwalk.
- **beachgoerProb**  The probability of any Ocean City, MD beachgoer (during the hot swimming days) exiting the beach at that mile marker.
- **expenseEst**  The estimated annual expenses of running a business at that location on the beach. It is assumed a large portion of the expense is based on commercial rental rates at that location. More populated locations tend to have higher expenses.
**carModelDF**

Dataframe of 1000 (fake) observations of whether certain car buyers were willing to get information on a credit card specializing in rewards for adventure travellers.

### Description

Dataframe of 1000 (fake) observations of whether certain car buyers were willing to get information on a credit card specializing in rewards for adventure travellers.

### Usage

```r
carModelDF
```

### Format

A data frame with 1000 rows and 3 variables:

- **customerID**: a unique id of a potential credit card customer. They just bought a car and are asked if they want information on the credit card.
- **carModel**: The model of car purchased.
- **getCard**: Whether the customer expressed interest in hearing more about the card.

**chimpanzeesDF**


### Description


### Usage

```r
chimpanzeesDF
```
Format

A data frame with 504 rows and 9 variables:

- **actor** name of actor
- **recipient** name of recipient (NA for partner absent condition)
- **condition** partner absent (0), partner present (1)
- **block** block of trials (each actor x each recipient 1 time)
- **trial** trial number (by chimp = ordinal sequence of trials for each chimp, ranges from 1-72; partner present trials were interspersed with partner absent trials)
- **prosoc_left** prosocial_left : 1 if prosocial (1/1) option was on left
- **chose_prosoc** choice chimp made (0 = 1/0 option, 1 = 1/1 option)
- **pulled_left** which side did chimp pull (1 = left, 0 = right)
- **treatment** narrative description combining condition and prosoc_left that describes the side the prosical food option was on and whether a partner was present

Source


---

**corruptDF**

*Dataframe of 174 observations where information on the human development index (HDI) and the corruption perceptions index (CPI) both exist. Each observation is a country.*

Description

Dataframe of 174 observations where information on the human development index (HDI) and the corruption perceptions index (CPI) both exist. Each observation is a country.

Usage

`corruptDF`

Format

A data frame with 174 rows and 7 variables:

- **country** country name
- **region** region name as given with CPI rating
- **countryCode** three letter abbreviation for country
- **regionCode** four letter or less abbreviation for country
- **population** 2017 country population
The Corruption Perceptions Index score for 2017: A country/territory’s score indicates the perceived level of public sector corruption on a scale of 0-100, where 0 means that a country is perceived as highly corrupt and a 100 means that a country is perceived as very clean.

The human development index score for 2017: the Human Development Index (HDI) is a measure of achievement in the basic dimensions of human development across countries. It is an index made from a simple unweighted average of a nation’s longevity, education and income and is widely accepted in development discourse.

Source


**Description**

The graph object should be of class `causact_graph` and created using `dag_create()`.

**Usage**

```r
dagp_plot(drawsDF, densityPlot = FALSE)
```

**Arguments**

- `drawsDF`: the dataframe output of `dag_greta(mcmc=TRUE)` where each column is a parameter and each row a single draw from a representative sample.
- `densityPlot`: If TRUE, each parameter gets its own density plot. If FALSE (recommended usage), parameters are grouped into facets based on whether they share the same prior or not. 10 and 90 percent credible intervals are displayed for the posterior distributions.

**Value**

a credible interval plot of all latent posterior distribution parameters.
Examples

# A simple example
posteriorDF = data.frame(x = rnorm(100),
y = rexp(100),
z = runif(100))
posteriorDF %>%
dagp_plot(densityPlot = TRUE)

# More complicated example requiring 'greta'
## Not run:
## Create a 2 node graph
library(greta)
graph = dag_create() %>%
dag_node("Get Card","y",
rhs = bernoulli(theta),
data = carModelDF$getCard) %>%
dag_node(descr = "Card Probability by Car", label = "theta",
rhs = beta(2,2),
child = "y")
graph %>% dag_render()

# below requires Tensorflow installation
drawsDF = graph %>% dag_greta(mcmc=TRUE)
drawsDF %>% dagp_plot()

## End(Not run)

# A multiple plate example
library(dplyr)
poolTimeGymDF = gymDF %>%
mutate(stretchType = ifelse(yogaStretch == 1,
   "Yoga Stretch",
   "Traditional")) %>%
group_by(gymID, stretchType, yogaStretch) %>%
summarize(nTrialCustomers = sum(nTrialCustomers),
nSigned = sum(nSigned))
graph = dag_create() %>%
dag_node("Cust Signed","k",
rhs = binomial(n,p),
data = poolTimeGymDF$nSigned) %>%
dag_node("Probability of Signing","p",
rhs = beta(2,2),
child = "k") %>%
dag_node("Trial Size","n",
data = poolTimeGymDF$nTrialCustomers,
child = "k") %>%
dag_plate("Yoga Stretch","x",
nodeLabels = c("p"),
data = poolTimeGymDF$stretchType,
addDataNode = TRUE) %>%
dag_plate("Observation","i",
nodeLabels = c("x","k","n")) %>%
dag_create

Create a graph object focused on drawing a DAG.

**Description**

Generates a causact_graph graph object that is set-up for drawing DAG graphs.

**Usage**

dag_create()

**Value**

a list object of class ‘causact_graph’ consisting of 6 dataframes. Each data frame is responsible for storing information about nodes, edges, plates, and the relationships among them.

**Examples**

# With `dag_create()` we can create an empty graph and
# add in nodes (`dag_node()`), add edges (`dag_edge`), and
# view the graph with `dag_render()`.

dag_create()

dag_diagrammer

The graph object should be of class causact_graph and created using dag_create().

**Description**

The graph object should be of class causact_graph and created using dag_create().

**Usage**

dag_diagrammer(graph, wrapWidth = 24, shortLabel = FALSE)
Arguments

- **graph**: A graph object of class `causact_graph` created using `dag_create()`.
- **wrapWidth**: A required character label that describes the node.
- **shortLabel**: A longer more descriptive character label for the node.

Value

A graph object of class `dgr_graph`. Useful for further customizing graph displays using the `DiagrammeR` package.

Examples

```r
library("DiagrammeR")
dag_create() %>%
dag_node("Get Card","y",
  rhs = bernoulli(theta),
  data = carModelDF$getCard) %>%
dag_diagrammer() %>%
render_graph(title = "DiagrammeR Version of causact_graph")
```

---

**dag_dim**

*Add dimension information to causact_graph*

Description

Internal function that is used as part of rendering graph or running greta.

Usage

```r
dag_dim(graph)
```

Arguments

- **graph**: A graph object of class `causact_graph` created using `dag_create()`.

Value

A graph object of class `causact_graph` with populated dimension information.
**dag_edge**

_Add an edge (or edges) between nodes in a graph object._

---

**Description**

With a graph object of class `causact_graph` created from `dag_create`, add an edge between nodes in the graph. Vector recycling is used for all arguments.

**Usage**

```
dag_edge(graph, from, to, type = as.character(NA))
```

**Arguments**

- `graph` a graph object of class `causact_graph`.
- `from` a character vector representing the parent nodes label or description from which the edge is connected.
- `to` the child node label or description from which the edge is connected.
- `type` character string used to represent the DiagrammeR line type (e.g. "solid").

Use `type = "extract"` to encourage causact to only pass indexed elements of the parent node to each instance of the child node. Specify `type = "solid"` to override any automated extract behavior.

**Value**

a graph object of class `dgr_graph` with additional edges created by this function.

**Examples**

```r
# Create a graph with 2 connected nodes
dag_create() %>%
  dag_node("X") %>%
  dag_node("Y") %>%
  dag_edge(from = "X", to = "Y") %>%
  dag_render(shortLabel = TRUE)
```

---

**dag_greta**

_Generate a representative sample of the posterior distribution_

---

**Description**

The input graph object should be of class `causact_graph` and created using `dag_create()`. The specification of a completely consistent joint distribution is left to the user. Helpful error messages are scheduled for future versions of the causact package.
Usage

dag_greta(graph, mcmc = TRUE, meaningfulLabels = TRUE, ...)

Arguments

graph a graph object of class causact_graph representing a complete and consistent specification of a joint distribution.

mcmc a logical value indicating whether to sample from the posterior distribution. When mcmc=FALSE, the greta code is printed to the console, but not executed. The user can cut and paste the code to another script for running line-by-line. This option is most useful for debugging purposes. When mcmc=TRUE, the code is executed and outputs a dataframe of posterior draws.

meaningfullLabels a logical value indicating whether to replace the indexed variable names in draws with abbreviated names representing the factor value corresponding to the index. This argument is treated as TRUE regardless of user input. The ability to retain numerical indexing will be in a subsequent release.

... additional arguments to be passed onto greta::mcmc().

Value

If mcmc=TRUE, returns a dataframe of posterior distribution samples corresponding to the input 'causact_graph'. Each column is a parameter and each row a draw from the posterior sample output. If mcmc=FALSE, running dag_greta returns a character string of code that would help the user create three objects representing the posterior distribution:

1. draws: An mcmc.list object containing raw output from the HMCMC sampler used by greta.
2. drawsDF: A wide data frame with all latent variables as columns and all draws as rows. This data frame is useful for calculations based on the posterior.
3. tidyDrawsDF: A long data frame with each draw represented on one line. This data frame is useful for plotting posterior distributions.

Examples

library(greta)
graph = dag_create() %>%
dag_node("Get Card","y",
  rhs = bernoulli(theta),
  data = carModelDF$getCard) %>%
dag_node(descr = "Card Probability by Car",label = "theta",
  rhs = beta(2,2),
  child = "y") %>%
dag_node("Car Model","x",
  data = carModelDF$carModel, child = "y") %>%
dag_plate("Car Model","x",
  data = carModelDF$carModel, nodeLabels = "theta")
dag_merge

Merge two non-intersect causact_graph objects

Description
Generates a single causact_graph graph object that combines the multiple provided graphs.

Usage
dag_merge(graph1, ...)

Arguments
graph1 A causact_graph objects to be merged with
...
As many causact_graph’s as wish to be merged

Value
a merged graph object of class causact_graph. Useful for creating simple graphs and then merging them into a more complex structure.

Examples
# With `dag_merge()` we
# reset the node ID’s and all other item ID’s,
# bind together the rows of all given graphs, and
# add in nodes and edges later
# with other functions
# to connect the graph.
#
# THE GRAPHS TO BE MERGED MUST BE DISJOINT
# THERE CAN BE NO IDENTICAL NODES OR PLATES
# IN EACH GRAPH TO BE MERGED, AT THIS TIME

g1 = dag_create() %>%
dag_node("Demand for A","dA",
  rhs = normal(15,4)) %>%
dag_node("Supply for A","sA",
  
```r
graph %>% dag_render()
gretaCode = graph %>% dag_greta(mcmc=FALSE)
## Not run:
## default functionality returns a data frame
# below requires Tensorflow installation
drawsDF = graph %>% dag_greta()
drawsDF %>% dagp_plot()
## End(Not run)
```
dag_node

rhs = uniform(0,100)) %>%
dag_node("Profit for A","pA",
  rhs = min(sA,dA)) %>%
dag_edge(from = c("dA","sA"),to = c("pA"))

g2 <- dag_create() %>%
dag_node("Demand for B","dB",
  rhs = normal(20,8)) %>%
dag_node("Supply for B","sB",
  rhs = uniform(0,100)) %>%
dag_node("Profit for B","pB",
  rhs = min(sB,dB)) %>%
dag_edge(from = c("dB","sB"),to = c("pB"))

g1 %>% dag_merge(g2) %>%
dag_node("Total Profit","TP",
  rhs = sum(pA,pB)) %>%
dag_edge(from = c("pA","pB"),to = c("TP")) %>%
dag_render()

---

dag_node  Add a node to an existing causact_graph object

Description

The graph object should be of class causact_graph and created using dag_create().

Usage

dag_node(
  graph,
  descr = as.character(NA),
  label = as.character(NA),
  rhs = NA,
  child = as.character(NA),
  data = NULL,
  obs = FALSE,
  keepAsDF = FALSE,
  extract = as.logical(NA),
  dec = FALSE,
  det = FALSE
)

Arguments

graph a graph object of class causact_graph. An initial object gets created using dag_create().

descr a longer more descriptive character label for the node.
**dag_node**

- **label**: a shorter character label for referencing the node (e.g. "X","beta").
- **rhs**: either a greta distribution such as `uniform`, `normal`, `lognormal`, `bernoulli`, etc. or an R expression. Greta distribution arguments are optional. Valid values include `normal(mu, sigma)`, `greta::normal`, `normal`, and `normal(6,2)`. R computation/expression examples include `alpha+beta*x` or `ilogit(alpha + gamma + beta)`. If a distribution is given, this is a random/stochastic node, if a formula is given it is a deterministic node once given the values of its parents. Quotes should not be used as all function/computations should consist of R objects, functions, and constants.
- **child**: an optional character vector of existing node labels. Directed edges from the newly created node to the supplied nodes will be created.
- **data**: a vector or data frame (with observations in rows and variables in columns).
- **obs**: a logical value indicating whether the node is observed. Assumed to be **TRUE** when data argument is given.
- **keepAsDF**: a logical value indicating whether the data argument should be split into one random variable node per column or kept together as a random matrix for matrix computation. Defaults to creating one node per column of the data frame.
- **extract**: a logical value. When **TRUE**, child nodes will try to extract an indexed value from this node. When **FALSE**, the entire random object (e.g. scalar, vector, matrix) is passed to children nodes. Only use this argument when overriding default behavior seen using `dag_render()`.
- **dec**: a logical value indicating whether the node is a decision node. Used to show nodes as rectangles instead of ovals when using `dag_render()`.
- **det**: a logical value indicating whether the node is a deterministic function of its parents. Used to draw a double-line (i.e. peripheries = 2) around a shape when using `dag_render()`. Assumed to be **TRUE** when rhs is a formula.

**Value**

a graph object of class `causact_graph` with an additional node(s).

**Examples**

```r
library(greta)
# Create an empty graph and add 2 nodes by using
# the `dag_node()` function twice
graph2 = dag_create() %>%
dag_node("Get Card","y",
    rhs = bernoulli(theta),
    data = carModelDF$getCard) %>%
dag_node(descr = "Card Probability by Car",label = "theta",
    rhs = beta(2,2),
    child = "y")
graph2 %>% dag_render()
```

# The Eight Schools Example from Gelman et al.:
schools_dat <- data.frame(y = c(28, 8, -3, 7, -1, 1, 18, 12),
sigma = c(15, 10, 16, 11, 9, 11, 10, 18), schoolName = paste0("School",1:8))

graph = dag_create() %>%
dag_node("Treatment Effect","y",
  rhs = normal(theta, sigma),
  data = schools_dat$y) %>%
dag_node("Std Error of Effect Estimates","sigma",
  data = schools_dat$sigma,
  child = "y") %>%
dag_node("Exp. Treatment Effect","theta",
  child = "y",
  rhs = avgEffect + schoolEffect) %>%
dag_node("Pop Treatment Effect","avgEffect",
  child = "theta",
  rhs = normal(0,30)) %>%
dag_node("School Level Effects","schoolEffect",
  rhs = normal(0,30),
  child = "theta") %>%
dag_plate("Observation","i",nodeLabels = c("sigma","y","theta")) %>%
dag_plate("School Name","school",
  nodeLabels = "schoolEffect",
  data = schools_dat$schoolName,
  addDataNode = TRUE)

graph %>% dag_render()
## Not run:
# below requires Tensorflow installation
graph %>% dag_greta(mcmc=TRUE)
tidyDrawsDF %>% dagp_plot()
## End(Not run)

---

**dag_plate**

*Create a plate representation for repeated nodes.*

**Description**

Given a graph object of class `causact_graph`, create collections of nodes that should be repeated i.e. represent multiple instances of a random variable, random vector, or random matrix. When nodes are on more than one plate, graph rendering will treat each unique combination of plates as separate plates.

**Usage**

```r
dag_plate(
  graph,
  descr,  
  label,  
  nodeLabels,  
)```

`dag_plate`

data = as.character(NA),
addErrorNode = FALSE,
rhs = NA
)

Arguments

- **graph**
  a graph object of class `dgr_graph` created using `dag_create()`.

- **descr**
  a longer more descriptive label for the cluster/plate.

- **label**
  a short character string to use as an index.

- **nodeLabels**
  a character vector of node labels or descriptions to include in the list of nodes.

- **data**
  a vector representing the categorical data whose unique values become the plate index. To use with `addErrorNode = TRUE`, this vector should represent observations of a variable that can be coerced to a factor.

- **addErrorNode**
  a logical value. When `addErrorNode = TRUE`, the code attempts to add a node of observed data that is used as an index for extracting the correct parameter from parent nodes that are on the newly created plate. Verify the graphical model using `dag_render()` to ensure correct behavior.

- **rhs**
  Optional rhs expression for when `addErrorNode = TRUE`. This can be either a greta distribution such as `uniform`, `normal`, `lognormal`, `bernoulli`, etc. or an R expression. Greta distribution arguments are optional. Valid values include `normal(mu,sigma), greta::normal, normal, and normal(6,2)`. R computation/expression examples include `alpha+beta*x` or `ilogit(alpha + gamma + beta)`. If a distribution is given, this is a random/stochastic node, if a formula is given it is a deterministic node once given the values of its parents. Quotes should not be used as all function/computations should consist of R objects, functions, and constants.

Value

an expansion of the input causact_graph object with an added plate representing the repetition of `nodeLabels` for each unique value of `data`.

Examples

# single plate example
graph = dag_create() %>%
dag_node("Get Card","y",
  rhs = bernoulli(theta),
  data = carModelDF$getCard) %>%
dag_node(dscr = "Card Probability by Car",label = "theta",
  rhs = beta(2,2),
  child = "y") %>%
dag_node("Car Model","x",
  data = carModelDF$carModel,
  child = "y") %>%
dag_plate("Car Model","x",
  data = carModelDF$carModel,

nodeLabels = "theta"

# multiple plate example
library(dplyr)
poolTimeGymDF = gymDF %>%
mutate/stretchType = ifelse(yogaStretch == 1,
"Yoga Stretch",
"Traditional") %>%
group_by(gymID,stretchType,yogaStretch) %>%
summarize(nTrialCustomers = sum(nTrialCustomers),
nSigned = sum(nSigned))

graph = dag_create() %>%
dag_node("Cust Signed","k",
rhs = binomial(n,p),
data = poolTimeGymDF$nSigned) %>%
dag_node("Probability of Signing","p",
rhs = beta(2,2),
child = "k") %>%
dag_node("Trial Size","n",
data = poolTimeGymDF$nTrialCustomers,
child = "k") %>%
dag_plate("Yoga Stretch","x",
nodeLabels = c("p"),
data = poolTimeGymDF$stretchType,
addDataNode = TRUE) %>%
dag_plate("Observation","i",
nodeLabels = c("x","k","n")) %>%
dag_plate("Gym","j",
nodeLabels = "p",
data = poolTimeGymDF$gymID,
addDataNode = TRUE)

dag_render()
Arguments

- **graph**
  a graph object of class `dgr_graph`.

- **shortLabel**
  a logical value. If set to `TRUE`, distribution and formula information is suppressed. Meant for communication with non-statistical stakeholders.

- **wrapWidth**
  a numeric value. Used to restrict width of nodes. Default is wrap text after 24 characters.

- **width**
  a numeric value. an optional parameter for specifying the width of the resulting graphic in pixels.

- **height**
  a numeric value. an optional parameter for specifying the height of the resulting graphic in pixels.

Value

Returns an object of class `grViz` and `htmlwidget` that is also rendered in the RStudio viewer for interactive building of graphical models.

Examples

```r
# Render a simple graph
dag_create() %>%
  dag_node("Demand","X") %>%
  dag_node("Price","Y", child = "X") %>%
  dag_render()

# Hide the mathematical details of a graph
dag_create() %>%
  dag_node("Demand","X") %>%
  dag_node("Price","Y", child = "X") %>%
  dag_render(shortLabel = TRUE)
```

---

**delivDF**

117,790 line items associated with 23,339 shipments.

Description

A dataset containing the line items, mostly parts, associated with 23,339 shipments from a US-based warehouse.

Usage

- `delivDF`
Format

A data frame (tibble) with 117,790 rows and 5 variables:

- **shipID** unique ID for each shipment
- **plannedShipDate** shipment date promised to customer
- **actualShipDate** date the shipment was actually shipped
- **partID** unique part identifier
- **quantity** quantity of partID in shipment

Source

Adam Fleischhacker

---

`generate_dot2`  
*Generate DOT code using a graph object*

Description

Generates Graphviz DOT code as an R character object using DiagrammeR graph object.

Usage

`generate_dot2(graph)`

Arguments

- `graph` A graph object of class dgr_graph.

Value

A character vector of length 1 containing Graphviz DOT code.

---

`grViz2`  
*R + viz.js*

Description

Make diagrams in R using viz.js with infrastructure provided by htmlwidgets.
gymDF

Usage

```r
grViz2(
  diagram = "",
  engine = "dot",
  allow_subst = TRUE,
  options = NULL,
  width = NULL,
  height = NULL
)
```

Arguments

- `diagram`: spec for a diagram as either text, filename string, or file connection.
- `engine`: string for the Graphviz layout engine; can be `dot` (default), `neato`, `circo`, or `twopi`. For more information see `viz.js usage`.
- `allow_subst`: a boolean that enables/disables substitution functionality.
- `options`: parameters supplied to the htmlwidgets framework.
- `width`: an optional parameter for specifying the width of the resulting graphic in pixels.
- `height`: an optional parameter for specifying the height of the resulting graphic in pixels.

Value

An object of class `htmlwidget` that will intelligently print itself into HTML in a variety of contexts including the R console, within R Markdown documents, and within Shiny output bindings.

gymDF

Dataframe of 44 observations of free crossfit classes data Each observation indicates how many students that participated in the free month of crossfit signed up for the monthly membership afterwards

Description

Dataframe of 44 observations of free crossfit classes data Each observation indicates how many students that participated in the free month of crossfit signed up for the monthly membership afterwards

Usage

```r
gymDF
```
houseDF

Format

A data frame with 44 rows and 5 variables:

- **gymID**  unique gym identifier
- **nTrialCustomers**  number of unique customers taking free trial classes
- **nSigned**  number of customers from trial that sign up for membership
- **yogaStretch**  whether trial classes included a yoga type stretch
- **timePeriod**  month number, since inception of company, for which trial period was offered

---

Dataframe of 1,460 observations of home sales in Ames, Iowa. Known as The Ames Housing dataset, it was compiled by Dean De Cock for use in data science education. Each observation is a home sale. See houseDFDescr for more info.

Description

Dataframe of 1,460 observations of home sales in Ames, Iowa. Known as The Ames Housing dataset, it was compiled by Dean De Cock for use in data science education. Each observation is a home sale. See houseDFDescr for more info.

Usage

houseDF

Format

A data frame with 1,460 rows and 37 variables:

- **SalePrice**  the property’s sale price in dollars. This is the target variable
- **MSSubClass**  The building class
- **MSZoning**  The general zoning classification
- **LotFrontage**  Linear feet of street connected to property
- **LotArea**  Lot size in square feet
- **Street**  Type of road access
- **LotShape**  General shape of property
- **Utilities**  Type of utilities available
- **LotConfig**  Lot configuration
- **Neighborhood**  Physical locations within Ames city limits
- **BldgType**  Type of dwelling
- **HouseStyle**  Style of dwelling
- **OverallQual**  Overall material and finish quality
OverallCond  Overall condition rating
YearBuilt  Original construction date
YearRemodAdd  Remodel date
ExterQual  Exterior material quality
ExterCond  Present condition of the material on the exterior
BsmtQual  Height of the basement
BsmtCond  General condition of the basement
BsmtExposure  Walkout or garden level basement walls
BsmtUnfSF  Unfinished square feet of basement area
TotalBsmtSF  Total square feet of basement area
1stFlrSF  First Floor square feet
2ndFlrSF  Second floor square feet
LowQualFinSF  Low quality finished square feet (all floors)
GrLivArea  Above grade (ground) living area square feet
FullBath  Full bathrooms above grade
HalfBath  Half baths above grade
BedroomAbvGr  Number of bedrooms above basement level
TotRmsAbvGrd  Total rooms above grade (does not include bathrooms)
Functional  Home functionality rating
GarageCars  Size of garage in car capacity
MoSold  Month Sold
YrSold  Year Sold
SaleType  Type of sale
SaleCondition  Condition of sale

Source

Dataframe of 523 descriptions of data values from "The Ames Housing dataset", compiled by Dean De Cock for use in data science education. Each observation is a possible value from a variable in the houseDF dataset.
meaningfulLabels

Usage
	houseDFDescr

Format

A data frame with 260 rows and 2 variables:

- **varName** the name and description of a variable stored in the houseDF dataset
- **varValueDescr** The value and accompanying interpretation for values in the houseDF dataset

Source


meaningfulLabels

Store meaningful parameter labels prior to running dag_greta() of greta::mcmc(). When greta creates posterior distributions for multi-dimensional parameters, it creates an often meaningless number system for the parameter (e.g. beta[1,1], beta[2,1], etc.). Since parameter dimensionality is often determined by a factor, this function creates labels from the factors unique values. replaceLabels() applies the text labels stored using this function to the greta output. The meaningful parameter names are stored in an environment, cacheEnv.

Description

Store meaningful parameter labels prior to running dag_greta() of greta::mcmc(). When greta creates posterior distributions for multi-dimensional parameters, it creates an often meaningless number system for the parameter (e.g. beta[1,1], beta[2,1], etc.). Since parameter dimensionality is often determined by a factor, this function creates labels from the factors unique values. replaceLabels() applies the text labels stored using this function to the greta output. The meaningful parameter names are stored in an environment, cacheEnv.

Usage

meaningfulLabels(graph)

Arguments

- **graph** a causact_graph object.

Value

A data frame meaningfulLabels stored in an environment named cacheEnv that contains a lookup table between greta labels and meaningful labels.
prodLineDF

**Description**

A dataset containing partID attributes.

**Usage**

prodLineDF

**Format**

A data frame (tibble) with 117,790 rows and 5 variables:

- **partID** unique part identifier
- **productLine** a product line associated with the partID
- **prodCategory** a product category associated with the partID

**Source**

Adam Fleischhacker

---

rbern

Make dbern, pbern, qbern, rbern available for implementation of the Bernoulli distribution functions.

**Description**

Density, distribution function, quantile function and random generation for the benoulli distribution with parameter prob.

**Usage**

rbern(n, prob)

**Arguments**

- **n** number of observations. If `length(n) > 1`, the length is taken to be the number required.
- **prob** probability of success of each trial

**Value**

A vector of 0’s and 1’s representing failure and success.
Examples

\#Return a random result of a Bernoulli trial given \code{prob}.
\rbern(n =1, prob = 0.5)

---

replaceLabels  
Replace parameter labels in a \code{mcmc.list} with more meaningful labels after they are created by running \code{dag_greta()}. When \code{greta} creates posterior distributions for multi-dimensional parameters, it creates an often meaningless number system for the parameter (e.g. beta[1,1], beta[2,1], etc.). Since parameter dimensionality is often determined by a factor, this functionality restores the text labels associated with the underlying factor whose coefficients are being estimated (e.g. beta_varValue1, beta_varValue2). The meaningful parameter names are stored in an environment, cacheEnv, created by a call to \code{dag_greta()}.  

Description

Replace parameter labels in a \code{mcmc.list} with more meaningful labels after they are created by running \code{dag_greta()}. When \code{greta} creates posterior distributions for multi-dimensional parameters, it creates an often meaningless number system for the parameter (e.g. beta[1,1], beta[2,1], etc.). Since parameter dimensionality is often determined by a factor, this functionality restores the text labels associated with the underlying factor whose coefficients are being estimated (e.g. beta_varValue1, beta_varValue2). The meaningful parameter names are stored in an environment, cacheEnv, created by a call to \code{dag_greta()}.  

Usage

\funs{replaceLabels}{draws}

Arguments

draws  
an \code{mcmc.list} object created by \code{dag_greta()}.  

Value

an \code{mcmc.list} with more meaningful names that get created during a \code{dag_greta} function call.
**replace_in_spec2**  
*Razor-like template for diagram specification*

**Description**

Use Razor-like syntax to define a template for use in a `grViz` diagram.

**Usage**

```r
replace_in_spec2(spec)
```

**Arguments**

<table>
<thead>
<tr>
<th>Name</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>spec</td>
<td>string spec to be parsed and evaluated.</td>
</tr>
</tbody>
</table>

**schoolsDF**  
*This example, often referred to as 8-schools, was popularized by its inclusion in Bayesian Data Analysis (Gelman, Carlin, & Rubin 1997).*

**Description**

This example, often referred to as 8-schools, was popularized by its inclusion in Bayesian Data Analysis (Gelman, Carlin, & Rubin 1997).

**Usage**

```r
schoolsDF
```

**Format**

A data frame with 8 rows and 3 variables:

- `y` estimated treatment effect at a particular school
- `sigma` standard error of the treatment effect estimate
- `schoolName` an identifier for the school represented by this row
**setDirectedGraphTheme**  
*Set DiagrammeR defaults for graphical models*

**Description**

`setDirectedGraph` returns a graph with good defaults.

**Usage**

```r
setDirectedGraphTheme(dgrGraph)
```

**Arguments**

- `dgrGraph`  
  A DiagrammeR graph

**Value**

An updated version of `dgrGraph` with good defaults for graphical models.  
return a `dgrGraph` object with the color and shape defaults used by the causact package.

**Examples**

```r
ticketsDF
```

---

**ticketsDF**

*Dataframe of 55,167 observations of the number of tickets written by NYC precincts each day Data modified from https://github.com/stan-dev/stancon_talks/tree/master/2018/Contributed-Talks/01_auerbach which originally sourced data from https://opendata.cityofnewyork.us/*

**Description**

Dataframe of 55,167 observations of the number of tickets written by NYC precincts each day Data modified from https://github.com/stan-dev/stancon_talks/tree/master/2018/Contributed-Talks/01_auerbach which originally sourced data from https://opendata.cityofnewyork.us/

**Usage**

```r
ticketsDF
```
Format

A data frame with 55167 rows and 4 variables:

- **precinct**: unique precinct identifier representing precinct of issuing officer
- **date**: the date on which ticket violations occurred
- **month_year**: the month_year extracted from date column
- **daily_tickets**: Number of tickets issued out of precinct on this day

**totalBeachgoersRepSample**

A representative sample from a random variable that represents the annual number of beach goers to Ocean City, MD beaches on hot days. Think of this representative sample as coming from either a prior or posterior distribution. An example using this sample is can be found in The Business Analyst’s Guide To Business Analytics at [https://www.causact.com/](https://www.causact.com/).

Description

A representative sample from a random variable that represents the annual number of beach goers to Ocean City, MD beaches on hot days. Think of this representative sample as coming from either a prior or posterior distribution. An example using this sample is can be found in The Business Analyst’s Guide To Business Analytics at [https://www.causact.com/](https://www.causact.com/).

Usage

```r
totalBeachgoersRepSample
```

Format

A 4,000 element vector.

**totalBeachgoersRepSample** a draw from a representative sample of total beachgoers to Ocean City, MD.

```r
%>%
```

The magrittr pipe

Description

causact uses the pipe function, `%>%` to turn function composition into a series of imperative statements.

Value

Pipe a value forward into a function- or call expression and return the function on the ‘rhs’ with the ‘lhs’ used as the first argument.
Index

* datasets
  baseballData, 3
  beachLocDF, 4
  carModelDF, 5
  chimpanzeesDF, 5
  corruptDF, 6
  delivDF, 19
  gymDF, 21
  houseDF, 22
  houseDFDescr, 23
  prodLineDF, 25
  schoolsDF, 27
  ticketsDF, 28
  totalBeachgoersRepSample, 29
  %>%, 29

  addPriorGroups, 3
  delivDF, 19
  generate_dot2, 20
  grViz2, 20
  gymDF, 21
  houseDF, 22
  houseDFDescr, 23
  meaningfulLabels, 24
  prodLineDF, 25
  rbern, 25
  replace_in_spec2, 27
  replaceLabels, 26
  schoolsDF, 27
  setDirectedGraphTheme, 28
  ticketsDF, 28
  totalBeachgoersRepSample, 29