Package ‘stevemisc’

July 27, 2021

<table>
<thead>
<tr>
<th>Type</th>
<th>Package</th>
</tr>
</thead>
<tbody>
<tr>
<td>Title</td>
<td>Steve's Miscellaneous Functions</td>
</tr>
<tr>
<td>Version</td>
<td>1.2.0</td>
</tr>
<tr>
<td>Depends</td>
<td>R (&gt;= 3.6.0), stats</td>
</tr>
<tr>
<td>Description</td>
<td>These are miscellaneous functions that I find useful for my research and teaching. The contents include themes for plots, functions for simulating quantities of interest from regression models, functions for simulating various forms of fake data for instructional/research purposes, and many more. All told, the functions provided here are broadly useful for data organization, data presentation, data recoding, and data simulation.</td>
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<tr>
<td>License</td>
<td>GPL (&gt;= 2)</td>
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<tr>
<td>BugReports</td>
<td><a href="https://github.com/svmiller/stevemisc/issues">https://github.com/svmiller/stevemisc/issues</a></td>
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<td>LazyData</td>
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<td>no</td>
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<tr>
<td>Author</td>
<td>Steve Miller [aut, cre], Ben Bolker [ctb], Dave Armstrong [ctb], John Fox [ctb], Winston Chang [ctb], Brian Ripley [ctb], Bill Venables [ctb], Pascal van Kooten [ctb], Gerko Vink [ctb], Paul Williamson [ctb]</td>
</tr>
<tr>
<td>Maintainer</td>
<td>Steve Miller <a href="mailto:steven.v.miller@gmail.com">steven.v.miller@gmail.com</a></td>
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</table>
carrec

Create multivariate data by permutation
Description

This recodes a numeric vector, character vector, or factor according to fairly simple recode specifications that former Stata users will appreciate. Yes, this is taken from John Fox’s `recode()` function in his `car` package. I’m going with `carrec()` (i.e. shorthand for `car::recode()`, phonetically here: “car-wreck”) for this package, with an additional shorthand of `carr` that does the same thing.

The goal here is to minimize the number of function clashes with multiple packages that I use in my workflow. For example: `car`, `dplyr`, and `Hmisc` all have `recode()` functions. I rely on the `car` package just for this function, but it conflicts with some other `tidyverse` functions that are vital to my workflow.

Usage

carrec(var, recodes, as_fac, as_num = TRUE, levels)

carr(...)

Arguments

- **var**: numeric vector, character vector, or factor
- **recodes**: character string of recode specifications: see below, but former Stata users will find this stuff familiar
- **as_fac**: return a factor; default is TRUE if var is a factor, FALSE otherwise
- **as_num**: if TRUE (which is the default) and as.factor is FALSE, the result will be coerced to a numeric if all values in the result are numeric. This should be what you want in 99% of applications for regression analysis.
- **levels**: an optional argument specifying the order of the levels in the returned factor; the default is to use the sort order of the level names.
- **...**: optional, only to make the shortcut (carr) work

Details

Recode specifications appear in a character string, separated by semicolons (see the examples below), of the form input=output. If an input value satisfies more than one specification, then the first (from left to right) applies. If no specification is satisfied, then the input value is carried over to the result. NA is allowed on input and output.

Value

`carrec()` returns a vector, recoded to the specifications of the user. `carr()` is a simple shortcut for `carrec()`.

Author(s)

John Fox

References

Examples

```r
x <- seq(1,10)
carrec(x,"0=0;1:2=1;3:5=2;6:10=3")
```

---

**cor2data**  
*Simulate Data from Correlation Matrix*

**Description**

A function to simulate data from a correlation matrix. This is useful for illustrating some theoretical properties of regressions when population parameters are known and set in advance.

**Usage**

```r
cor2data(cor, n, seed)
```

**Arguments**

- `cor`: A correlation matrix (of class `matrix`)
- `n`: A number of observations to simulate
- `seed`: An optional parameter to set a seed. Omitting this generates new simulations every time.

**Value**

`cor2data()` returns a data frame where all observations are simulated from a standard normal distribution, but with those pre-set correlations.

**Author(s)**

Steven V. Miller

**Examples**

```r
vars <- c("control", "treat", "instr", "e")
Correlations <- matrix(cbind(1, 0.001, 0.001, 0.001, 0.001, 1, 0.85, -0.5, 0.001, 0.85, 1, 0.001, 0.001, -0.5, 0.001, 1), nrow=4)
rownames(Correlations) <- colnames(Correlations) <- vars
cor2data(Correlations, 1000, 8675309)
```
Description

corvectors() is a function to obtain a multivariate dataset by specifying the relation between those specified variables.

Usage

corvectors(
  data,
  corm,
  tol = 0.005,
  conv = 10000,
  cores = 2,
  splitsize = 1000,
  verbose = FALSE,
  seed
)

Arguments

data a data matrix containing the data
corm A value containing the desired correlation or a vector or data matrix containing the desired correlations
tol A single value or a vector of tolerances with length ncol(data) - 1. The default is 0.005
conv The maximum iterations allowed. Defaults to 1000.
cores The number of cores to be used for parallel computing
splitsize The size to use for splitting the data
verbose Logical statement. Default is FALSE
seed An optional seed to set

Details

This is liberally copy-pasted from van Kooten and Vink’s wonderful-but-no-longer-supported correlate package. They call it correlate() in their package, but I opt for corvectors() here.

Value

corvectors() returns a matrix given the specified multivariate relation.

Author(s)

Pascal van Kooten and Gerko Vink
Examples

```r
set.seed(8675309)
library(tibble)
# bivariate example, start with zero correlation
as_tibble(data.frame(corvectors(replicate(2, rnorm(100)), .5)))

# multivariate example
as_tibble(data.frame(corvectors(replicate(4, rnorm(100)), c(.5, .6, .7))))
```

db_lselect  

*Lazily select variables from multiple tables in a relational database*

Description

`db_lselect()` allows you to select variables from multiple tables in an SQL database. It returns a lazy query that combines all the variables together into one data frame (as a tibble). The user can choose to run `collect()` after this query if they see fit.

Usage

```r
db_lselect(.data, connection, vars)
```

Arguments

- `.data`      a character vector of the tables in a relational database
- `connection` the name of the connection object
- `vars`       the variables (entered as class "character") to select from the tables in the database

Details

This is a wrapper function in which `purrr` and `dplyr` are doing the heavy lifting. The tables in the database are declared as a character (or character vector). The variables to select are also declared as a character (or character vector), which are then wrapped in a `one_of()` function within `select()` in `dplyr`.

Value

Assuming a particular structure to the database, the function returns a combined table including all the requested variables from all the tables listed in the data character vector. The returned table will have other attributes inherited from how `dplyr` interfaces with SQL, allowing the user to extract some information about the query (e.g. through `show_query()`).
References

Miller, Steven V. 2020. "Clever Uses of Relational (SQL) Databases to Store Your Wider Data (with Some Assistance from dplyr and purrr)" http://svmiller.com/blog/2020/11/smarter-ways-to-store-your-wide-

Examples

library(DBI)
library(RSQLite)
library(dplyr)
library(dbplyr)
set.seed(8675309)
A <- data.frame(uid = c(1:10),
    a = rnorm(10),
    b = sample(letters, 10),
    c = rbinom(10, 1, .5))
B <- data.frame(uid = c(11:20),
    a = rnorm(10),
    b = sample(letters, 10),
    c = rbinom(10, 1, .5))
C <- data.frame(uid = c(21:30), a = rnorm(10),
    b = sample(letters, 10),
    c = rbinom(10, 1, .5),
    d = rnorm(10))
con <- dbConnect(SQLite(), "::memory:"

copy_to(con, A, "A",
temporary=FALSE)
copy_to(con, B, "B",
temporary=FALSE)
copy_to(con, C, "C",
temporary=FALSE)

# This returns no warning because columns "a" and "b" are in all tables
c("A", "B", "C") %>% db_lselect(con, c("uid", "a", "b"))

# This returns two warnings because column "d" is not in 2 of 3 tables.
# ^ this is by design. It'll inform the user about data availability.
c("A", "B", "C") %>% db_lselect(con, c("uid", "a", "b", "d"))
dbDisconnect(con)
Some Labeled Data in the European Social Survey (Round 9)

Description
These are data to illustrate labeled data and how to process them with `get_var_info()` in this package.

Usage
ess9_labelled

Format
A data frame with 109 observations on the following 4 variables.

- **essround**: a numeric constant
- **edition**: another numeric constant
- **cntry**: a character vector (with label) for the country in the data
- **netusoft**: a numeric vector (with label) for self-reported internet consumption of a respondent

Details
Data are condensed summaries from the raw data. They amount to every unique combination of country and self-reported internet consumption. The data are here to illustrate the `get_var_info()` function in this package.

---

Reorganize a factor after "re-leveling" it

Description
`fct_reorg()` is a **forcats** hack that reorganizes a factor after re-leveling it. It has been situationally useful in my coefficient plots over the years.

Usage
`fct_reorg(fac, ...)`

Arguments

- **fac**: a character or factor vector
- **...**: optional parameters to be supplied to **forcats** functions.
Details

Solution comes by way of this issue on Github: https://github.com/tidyverse/forcats/issues/45

Value

This function takes a character or factor vector and first re-levels it before re-coding certain values. The end result is a factor.

Examples

```r
x<-factor(c("a","b","c"))
fct_reorg(x, B="b", C="c")
```

get_sims

*Get Simulations from a Model Object (with New Data)*

Description

get_sims() is a function to simulate quantities of interest from a multivariate normal distribution for "new data" from a regression model.

Usage

```r
get_sims(model, newdata, nsim, seed)
```

Arguments

- `model`: a model object
- `newdata`: A data frame on some quantities of interest to be simulated
- `nsim`: Number of simulations to be run
- `seed`: An optional seed to set

Details

This (should) be a flexible function that takes a `merMod` object (estimated from `lme4`, `blme`, etc.) or a `lm` or `glm` object and generates some quantities of interest when paired with new data of observations of interest. Of note: I’ve really only tested this function with linear models, generalized linear models, and their mixed model equivalents. For mixed models, this approach does not offer support for the incorporation of the random effects or the random slopes. It’s just for the fixed effects, which is typically what most people want anyway. Users who want to better incorporate the random intercepts or slope could find that support in the `merTools` package.
get_var_info

Get a small data frame of the variable label and values.

**Description**

get_var_info() allows you to peek at your labelled data, extracting a given column's variable labels. The intended use here is mostly “peeking” for the purpose of recoding column's in the absence of a codebook or other form of documentation. gvi() is a shortcut for this function.

**Value**

get_sims() returns a data frame (as a tibble) with the quantities of interest and identifying information about the particular simulation number.

**Author(s)**

Steven V. Miller

**Examples**

# Note: these models are dumb, but they illustrate how it works.

```r
M1 <- lm(mpg ~ hp, mtcars)  
get_sims(M1, newdat, 100, 8675309)
```

# Note: this function requires the DV to appear somewhere, anywhere in the "new data"

```r
ewdat <- data.frame(mpg = 0,  
hp = c(mean(mtcars$hp) - sd(mtcars$hp),  
      mean(mtcars$hp),  
      mean(mtcars$hp) + sd(mtcars$hp)))
```

# Note: this is likely a dumb model, but illustrates how it works.

```r
mtcars$mpgd <- ifelse(mtcars$mpg > 25, 1, 0)
M2 <- glm(mpgd ~ hp, mtcars, family=binomial(link="logit"))
```

# Again: this function requires the DV to be somewhere, anywhere in the "new data"

```r
ewdat$mpgd <- 0
```

# Note: the simulations are returned on their original "link". Here, that's a "logit"  
# You can adjust that accordingly. plogis(y) will convert those to probabilities.

```r
get_sims(M2, newdat, 100, 8675309)
library(lme4)
M3 <- lmer(mpg ~ hp + (1 | cyl), mtcars)
```

# Random effects are not required here since we're passing over them.

```r
get_sims(M3, newdat, 100, 8675309)
```
**jenny**

Usage

```r
get_var_info(.data, x)
gvi(...)```

Arguments

<table>
<thead>
<tr>
<th>Argument</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>.data</td>
<td>a data frame</td>
</tr>
<tr>
<td>x</td>
<td>a column within the data frame</td>
</tr>
<tr>
<td>...</td>
<td>optional, only to make the shortcut (gvi) work</td>
</tr>
</tbody>
</table>

Details

This function leans on `var_label()` and `val_label()` in the `labelled` package, which is a dependency for this package. The function is designed to be used in a "pipe."

Value

If the column in the data frame is not labelled, the function returns a message communicating the absence of labels. If the column in the data frame is labelled, the function returns a small data frame communicating the `var_label()` output (var), the (often but not always) numeric "code" coinciding with `val_label()` (code), and the "label" attached to it (label).

Examples

```r
library(tibble)
library(dplyr)
library(magrittr)

ess9_labelled %>% get_var_info(netusoft) # works, as intended
ess9_labelled %>% get_var_info(cntry) # works, as intended
ess9_labelled %>% get_var_info(ess9round) # barks at you; data are not labelled```

---

**jenny**

*Set the Only Reproducible Seed That Matters*

Description

`jenny()` sets a reproducible seed of 8675309. It is the only reproducible seed you should use.

Usage

```r
jenny(x = 8675309)```
Arguments

x  a vector

Details

jenny() comes with some additional perks if you have the emo package installed. The package is optional.

Value

When ‘x’ is not specified or is 8675309, the function sets a reproducible seed of 8675309 and returns a nice message congratulating you for it. If ‘x’ is not 8675309, the function sets no reproducible seed and gently admonishes you for wasting its time.

Examples

jenny() # will work and reward you for it
ejenny(12345) # will not work and will result in a stern message

make_perclab

Make Percentage Label for Proportion and Add Percentage Sign

Description

make_perclab() takes a proportion, multiplies it by 100, optionally rounds it, and pastes a percentage sign next to it.

Usage

make_perclab(x, d = 2)

Arguments

x  a numeric vector

d  digits to round. Defaults to 2.

Details

This function is useful if you’re modeling proportions in something like a bar chart (for which proportions are more flexible) but want to label each bar as a percentage. The function here is mostly cosmetic.

Value

The function takes a proportion, multiplies it by 100, (optionally) rounds it to a set decimal point, and pastes a percentage sign next to it.
**Description**

`make_scale()` will rescale any vector to have a user-defined minimum and maximum.

**Usage**

```r
make_scale(x, minim, maxim)
```

**Arguments**

- `x`: a numeric vector
- `minim`: a desired numeric minimum
- `maxim`: a desired numeric maximum

**Details**

This function is useful if you wanted to do some kind of minimum-maximum rescaling of a variable on some given scale, prominently rescaling to a minimum of 0 and a maximum of 1 (thinking ahead to a regression). The function is flexible enough for any minimum or maximum.

**Value**

The function takes a numeric vector and returns a rescaled version of it with the observed (desired) minimum, the observed (desired) maximum, and rescaled values between both extremes.

**Examples**

```r
x <- runif(100, 1, 100)
make_scale(x, 2, 5) # works
make_scale(x, 5, 2) # results in message
make_scale(x, 0, 1) # probably why you're using this.
```
mround

Multiply a Number by 100 and Round It (By Default: 2)

Description

mround() is a convenience function I wrote for my annotating bar charts that I make. Assuming a proportion variable, mround() will multiply each value by 100 and round it for presentation. By default, it rounds to two. The user can adjust this.

Usage

mround(x, d = 2)

Arguments

x a numeric vector
d the number of decimal points to which the user wants to round. If this is not set, it rounds to two decimal points.

Details

This is a sister function of make_perclab() in the same package. This, however, won’t add a percentage sign.

Value

The function takes a numeric vector, multiplies it by 100, rounds it (to two digits by default), and returns it to the user.

Examples

x <- runif(100)
mround(x)
mround(x, 2) # same as above
mround(x, 3)

normal_dist

Make and annotate a normal distribution with ggplot2

Description

normal_dist() is a convenience function for making a plot of a normal distribution with annotated areas underneath the normal curve.
Usage

```r
normal_dist(curvecolor, fillcolor, fontfamily)
```

Arguments

- **curvecolor**: What color should the curve itself be. Any `ggplot2`-recognized format should do here.
- **fillcolor**: What color should the area underneath the curve be. Any `ggplot2`-recognized format should do here.
- **fontfamily**: Font family for labeling areas underneath the curve. OPTIONAL. You can omit this if you’d like.

Details

The normal distribution is a standard normal distribution with a mean of 0 and a standard deviation of 1.

Value

The function returns a fancy plot of a normal distribution annotated with areas underneath the hood. Note that whatever color is supplied in `fillcolor` is automatically lightened for areas further from the center of the curve.

Examples

```r
library(stevemisc)
normal_dist("blue","red")
normal_dist("purple","orange")
```

---

**prepare.refs**

Prepare `bib2df` Data Frame for Formatting to Various Outputs

Description

`prepare.refs` does some last-minute formatting of a data frame created by `bib2df` so that it can be formatted nicely to various outputs.

Usage

```r
prepare.refs(bib2df.refs, toformat = "plain")
```

Arguments

- **bib2df.refs**: a data frame created by `bib2df`
- **toformat**: what type of output you are ultimately going to want from `print.refs()`. Default is "plain".
Details

The function is designed to work more generally in the absence of various fields. Assume, for example, that your data frame has no BOOK field. The function uses the one_of() wrapper to work around this. The "warning" returned by the function is more of a message. This function may be expanded as I think of more use cases.

Value

print.refs() does some last-minute formatting to a data frame created by bib2df so that rendering in R Markdown is a little easier and less code-heavy.

See Also

print.refs() for formatting a .bib references to various outputs.

Examples

prepare.refs(stevepubs)

Description

print.refs() is a convenience function I found and edited that will allow a user to print and format .bib entries as if they were references. This function is useful if you want to load a .bib entry or set of entries and print them in the middle of a document in R Markdown.

Usage

print.refs(
  bib,
  csl = "american-political-science-association.csl",
  toformat = "latex",
  cslrepo = "https://raw.githubusercontent.com/citation-style-language/styles/master",
  spit_out = TRUE,
  delete_after = TRUE
)

Arguments

bib a valid .bib entry

csl a CSL file, matching one available on the Github repository, that the user wants to format the references. Default is "american-political-science-association.csl".

toformat the output wanted by the user. Default is "latex".
ps_btscs

cslrepo
spit_out
delete_after

da directory of CSL files. Defaults to the one on Github.
logical, defaults to TRUE. If TRUE, wraps ("spits out") formatted citations in a writeLines() output for the console. If 'FALSE', returns a character vector.
logical, defaults to TRUE. If TRUE, deletes CSL file when it’s done. If FALSE, retains CSL for (potential) future use.

Details

print.refs() assumes an active internet connection in the absence of the appropriate CSL file in the working directory. The citation style language (CSL) file supplied by the user must match a file in the massive Github repository of CSL files. Users interested in potential outputs should read more about Pandoc (https://pandoc.org/MANUAL.html). The Github repository of CSL files is available here: https://github.com/citation-style-language/styles.

Value

print.refs() takes a .bib entry and returns a formatted reference from it.

Examples

```r
example <- "@Book{vasquez2009twp, Title = {The War Puzzle Revisited},
Author = {Vasquez, John A}, Publisher = {New York, NY: Cambridge University Press},
Year = {2009}}"

print.refs(example)
```

---

ps_btscs

Create "peace years" or "spells" by cross-sectional unit, more generally

Description

ps_btscs() allows you to create spells ("peace years" in the international conflict context) between observations of some event. This will allow the researcher to better model temporal dependence in binary time-series cross-section ("BTSCS") models. It is an improvement on sbtscs() (included in this package) by its ability to more flexibly work with data that have lots of NAs that bracket the observed event data. It is used in the peacescience package.

Usage

```r
ps_btscs(data, event, tvar, csunit, pad_ts = FALSE)
```
Arguments

data the data set with which you are working

event some event (0, 1) for which you want spells or peace years

tvar the time variable (e.g. a year)

csunit the cross-sectional unit (likely a dyad if you’re doing boilerplate international conflict stuff)

pad_ts should time-series be filled when panels are unbalanced/have gaps? Defaults to FALSE.

Details

This function is derived from sbtscs(). See documentation there for more information.

Value

ps_btscs() takes a data frame and returns the data frame with a new variable named spell.

Author(s)

David A. Armstrong, Steven V. Miller

References


Examples

library(dplyr)
library(stevemisc)
data(usa_mids)

# notice: no quotes
ps_btscs(usa_mids, midongoing, year, dyad)
Convert the p-value you want to the z-value it actually is

Description
I loathe how statistical instruction privileges obtaining a magical p-value by reference to an area underneath the standard normal curve, only to botch what the actual z-value is corresponding to the magical p-value. This simple function converts the p-value you want (typically .05, thanks to R.A. Fisher) to the z-value it actually is for the kind of claims we typically make in inferential statistics. If we’re going to do inference the wrong way, let’s at least get the z-value right.

Usage
\[ p_z(x, \text{ts} = \text{TRUE}) \]

Arguments
- \text{x} \quad \text{a numeric vector (one or multiple) between 0 or 1}
- \text{ts} \quad \text{a logical, defaults to TRUE. If TRUE, returns two-sided critical z-value. If FALSE, the function returns a one-sided critical z-value.}

Details
\( p_z() \) takes a p-value of interest and converts it, with precision, to the z-value it actually is. The function takes a vector and returns a vector. The function assumes you’re doing something akin to calculating a confidence interval or testing a regression coefficient against a null hypothesis of zero. This means the default output is a two-sided critical z-value. We’re taught to use two-sided z-values when we’re agnostic about the direction of the effect or statistic of interest, which is, to be frank, hilarious given how most research is typically done.

Value
This function takes a numeric vector, corresponding to the p-value you want, and returns a numeric vector coinciding with the z-value you want under the standard normal distribution. For example, the z-value corresponding with the magic number of .05 (the conventional cutoff for assessing statistical significance) is not 1.96, it’s something like 1.959964 (rounding to the default six decimal points).

Examples

```r
library(stevemisc)

p_z(.05)
p_z(c(.001, .01, .05, .1))
p_z(.05, \text{ts=FALSE})
p_z(c(.001, .01, .05, .1), \text{ts=FALSE})
```
r1sd  \hspace{1cm} \textit{Scale a vector by one standard deviation}

Description

\texttt{r1sd} allows you to rescale a numeric vector such that the ensuing output has a mean of 0 and a standard deviation of 1.

Usage

\texttt{r1sd(x, na = TRUE)}

Arguments

- \texttt{x} \hspace{1cm} a numeric vector
- \texttt{na} \hspace{1cm} what to do with NAs in the vector. Defaults to TRUE (i.e. passes over the missing observations)

Details

This is a convenience function since the default ‘\texttt{rescale()}’ function has some additional weirdness that is not welcome for my use cases. By default, ‘\texttt{na.rm}’ is set to TRUE.

Value

The function returns a numeric vector rescaled with a mean of 0 and a standard deviation of 1.

Examples

\begin{verbatim}
x <- rnorm(100)
r1sd(x)
\end{verbatim}

r2sd  \hspace{1cm} \textit{Scale a vector (or vectors) by two standard deviations}

Description

\texttt{r2sd} allows you to rescale a numeric vector such that the ensuing output has a mean of 0 and a standard deviation of .5. \texttt{r2sd_at} is a wrapper for \texttt{mutate_at} and \texttt{rename_at} from \texttt{dplyr}. It both rescales the supplied vectors to new vectors and renames the vectors to each have a prefix of \texttt{z}_.

Usage

\texttt{r2sd(x, na = TRUE)}

\texttt{r2sd_at(data, x)}
**Arguments**

- `x`  a vector, likely in your data frame
- `na` what to do with NAs in the vector. Defaults to TRUE (i.e. passes over the missing observations)
- `data` a data frame

**Details**

By default, `na.rm` is set to TRUE. If you have missing data, the function will just pass over them. Gelman (2008) argues that rescaling by two standard deviations puts regression inputs on roughly the same scale no matter their original scale. This allows for some honest, if preliminary, assessment of relative effect sizes from the regression output. This does that, but without requiring the `rescale` function from `arm`. I’m trying to reduce the packages on which my workflow relies. Importantly, I tend to rescale only the ordinal and interval inputs and leave the binary inputs as 0/1. So, my `r2sd` function doesn’t have any of the fancier if-else statements that Gelman’s `rescale` function has.

**Value**

The function returns a numeric vector rescaled with a mean of 0 and a standard deviation of .5.

**References**


**Examples**

```r
x <- rnorm(100)
r2sd(x)
r2sd_at(mtcars, c("mpg", "hp", "disp"))
```

---

**rbnorm**  
*Bounded Normal (Really: Scaled Beta) Distribution*

**Description**

`rbnorm()` is a function to randomly generate values from a bounded normal (really: a scaled beta) distribution with specified mean, standard deviation, and upper/lower bounds. I use this function to randomly generate data that we treat as interval for sake of getting means and standard deviations, but have discernible bounds (and even skew) to teach students about things like random sampling and central limit theorem.
Usage

```
rbnorm(n, mean, sd, lowerbound, upperbound, round = FALSE, seed)
```

Arguments

- **n**: the number of observations to simulate
- **mean**: a mean to approximate
- **sd**: a standard deviation to approximate
- **lowerbound**: a lower bound for the data to be generated
- **upperbound**: an upper bound for the data to be generated
- **round**: whether to round the values to whole integers. Defaults to FALSE
- **seed**: set an optional seed

Details

I call it "bounded normal" when it’s really a beta distribution. I’m aware of this. I took much of this code from somewhere. I forget where.

Value

The function returns a vector of simulated data approximating the user-specified conditions.

Examples

```
library(tibble)

tibble(x = rbnorm(10000, 57, 14, 0, 100))
tibble(x = rbnorm(10000, 57, 14, 0, 100, round = TRUE))
tibble(x = rbnorm(10000, 57, 14, 0, 100, seed = 8675309))
```

---

**revcode**  
Reverse code a numeric variable

Description

`revcode` allows you to reverse code a numeric variable. If, say, you have a Likert item that has values of 1, 2, 3, 4, and 5, the function inverts the scale so that 1 = 5, 2 = 4, 3 = 3, 4 = 2, and 5 = 1.

Usage

```
revcode(x)
```

Arguments

- **x**: a numeric vector
Details
This function passes over NAs you may have in your variable. It does assume, reasonably might I add, that the observed values include both the minimum and the maximum. This is usually the case in a discrete ordered-categorical variable (like a Likert item). It also assumes that the numeric vector supplied to it contains all possible values and that the minimum observed value is 1. This is usually a safe assumption in survey data where the variable of interest is ordinal (either on a 1:4 scale, or 1:5 scale, or 1:10 scale). No matter, use the function with that in mind.

Value
The function returns a numeric vector that reverse codes the numeric vector that was supplied to it.

Examples

data.frame(x1 = rep(c(1:7, NA), 2),
             x2 = c(1:10, 1:4, NA, NA),
             x3 = rep(c(1:4), 4)) -> example_data

library(dplyr)
library(magrittr)

example_data %>% mutate_at(vars("x1", "x2", "x3"), ~revcode(.))

sbtscs
Create "peace years" or "spells" by cross-sectional unit

Description
sbtscs() allows you to create spells ("peace years" in the international conflict context) between observations of some event. This will allow the researcher to better model temporal dependence in binary time-series cross-section ("BTSCS") models.

Usage
sbtscs(data, event, tvar, csunit, pad_ts = FALSE)

Arguments
data the data set with which you are working
event some event (0, 1) for which you want spells or peace years
tvar the time variable (e.g. a year)
csunit the cross-sectional unit (likely a dyad if you’re doing boilerplate international conflict stuff)
pad_ts should time-series be filled when panels are unbalanced/have gaps? Defaults to FALSE.
Details

I should confess outright, and it should be obvious to anyone who looks at the code, that I liberally copy from Dave Armstrong’s btscs() function in the DAMisc package. I offer two such improvements. One, the btscs() function chokes when a large number of cross-sectional units have no recorded “event.” I don’t know why this happens but it does. Further, “tidying” up the code by leaning on dplyr substantially speeds up computation. Incidentally, this concerns the same cross-sectional units with no recorded events that can choke the btscs() function in large numbers.

Value

sbtscs() takes a data frame and returns the data frame with a new variable named spell.

Author(s)

David A. Armstrong, Steven V. Miller

References


Miller, Steven V. 2017. “Quickly Create Peace Years for BTSCS Models with sbtscs in stevemisc.”
http://svmiller.com/blog/2017/06/quickly-create-peace-years-for-btscs-models-with-stevemisc/

Examples

```r
# Not run:
library(dplyr)
library(stevemisc)
data(usa_mids)

# notice: no quotes
sbtscs(usa_mids, midongoing, year, dyad)

# End(Not run)
```

show_ranef

Get a caterpillar plot of random effects from a mixed model

Description

show_ranef() allows a user estimating a mixed model to quickly plot the random intercepts (with conditional variances) of a given random effect in a mixed model. In cases where there is a random slope over the intercept, the function plots the random slope as another caterpillar plot (as another facet)
Usage

show_ranef(data, grp, reorder = TRUE)

Arguments

data a fitted mixed model with random intercepts
grp What random intercept/slopes do you want to see as a caterpillar plot? Declare it as a character
reorder optional argument. DEFAULT is TRUE, which “re-orders” the intercepts by the original value in the data. If FALSE, the ensuing caterpillar plot defaults to a default method of ordering the levels of the random effect by their estimated conditional mode.

Details

This function is a simple wrapper in which broom.mixed and, obviously ggplot2 are doing the heavy lifting.

Value

show_ranef() returns a caterpillar plot of the random intercepts from a given mixed model. If broom.mixed::augment() can process it, this function should work just fine.

Author(s)

Steven V. Miller

Examples

library(lme4)
library(stevemisc)
data(sleepstudy)

M1 <- lmer(Reaction ~ Days + (Days | Subject), data=sleepstudy)
show_ranef(M1, "Subject")
show_ranef(M1, "Subject", reorder=FALSE)

smvrnorm

Simulate from a Multivariate Normal Distribution

Description

smvrnorm() simulates data from a multivariate normal distribution.
Usage

smvrnorm(
  n = 1,
  mu,
  sigma,
  tol = 1e-06,
  empirical = FALSE,
  eispack = FALSE,
  seed
)

Arguments

n the number of observations to simulate
mu a vector of means
sigma a positive-definite symmetric matrix specifying the covariance matrix of the variables.
tol tolerance (relative to largest variance) for numerical lack of positive-definiteness in sigma.
empirical logical. If true, mu and sigma specify the empirical not population mean and covariance matrix.
eispack logical. values other than FALSE result in an error
seed set an optional seed

Details

This is a simple port and rename of mvrnorm() from the MASS package. I elect to plagiarize/port it because the MASS package conflicts with a lot of things in my workflow, especially select(). This is useful for "informal Bayes" approaches to generating quantities of interest from a regression model.

Value

The function returns simulated data from a multivariate normal distribution.

References


Examples

M1 <- lm(mpg ~ disp + cyl, mtcars)
smvrnorm(100, coef(M1), vcov(M1))
Description

These are data on my publications, barring a few things like book reviews and some forthcoming pieces. I use these data to illustrate the \texttt{print.refs()} function. You should cite my publications more.

Usage

\texttt{stevepubs}

Format

A data frame the following 14 variables.

\begin{itemize}
  \item \texttt{CATEGORY} the entry type
  \item \texttt{BIBTEXKEY} the unique entry key
  \item \texttt{AUTHOR} a list of authors for this entry
  \item \texttt{BOOKTITLE} the book title, if appropriate
  \item \texttt{JOURNAL} the journal title, if appropriate
  \item \texttt{NUMBER} the journal volume number, if appropriate
  \item \texttt{PAGES} the range of page numbers, if appropriate
  \item \texttt{PUBLISHER} the book publisher, if appropriate
  \item \texttt{TITLE} the title of the publication
  \item \texttt{VOLUME} the journal volume number, if appropriate
  \item \texttt{YEAR} the year of publication, as a character. Publications with no year are assumed to be forthcoming
  \item \texttt{DOI} a DOI, if I entered one
\end{itemize}

Details

Cite my publications more, you goons. *Extremely Smokey Bear voice* Only YOU can jack my h-index to infinity.
**strategic_rivalries**  
*Strategic Rivalries, 1494-2010*

**Description**
A simple summary of all strategic (inter-state) rivalries from Thompson and Dreyer (2012).

**Usage**
```r
data("strategic_rivalries")
```

**Format**
A data frame with 197 observations on the following 10 variables.

- `rivalryno`  a numeric vector for the rivalry number
- `rivalryname`  a character vector for the rivalry name
- `sidea`  a character vector for the first country in the rivalry
- `sideb`  a character vector for the second country in the rivalry
- `styear`  a numeric vector for the start year of the rivalry
- `endyear`  a numeric vector for the end year of the rivalry
- `region`  a character vector for the region of the rivalry, per Thompson and Dreyer (2012)
- `type1`  a character vector for the primary type of the rivalry (spatial, positional, ideological, or interventionary)
- `type2`  a character vector for the secondary type of the rivalry, if applicable (spatial, positional, ideological, or interventionary)
- `type3`  a character vector for the tertiary type of the rivalry, if applicable (spatial, positional, ideological, or interventionary)

**Details**
Information gathered from the appendix of Thompson and Dreyer (2012). Ongoing rivalries are right-bound at 2010, the date of publication for Thompson and Dreyer’s handbook. Users are free to change this if they like.

**References**

**Examples**
```r
data(strategic_rivalries)
```
The Student-t Distribution (Location-Scale)

Description
These are density, distribution function, quantile function and random generation for the Student-t distribution with location $\mu$, scale $\sigma$, and degrees of freedom $df$. Base R gives you the so-called "standard" Student-t distribution, with just the varying degrees of freedom. This generalizes that standard Student-t to the three-parameter version.

Usage

dst(x, df, mu, sigma)
pst(q, df, mu, sigma)
qst(p, df, mu, sigma)
rst(n, df, mu, sigma)

Arguments

x, q          a vector of quantiles
df            a vector of degrees of freedom
mu            a vector for the location value
sigma         a vector of scale values
p              Vector of probabilities.
n              Number of samples to draw from the distribution.

Details
This is a simple hack taken from Wikipedia. It's an itch I've been wanting to scratch for a while. I can probably generalize this outward to allow the tail and log stuff, but I wrote this mostly for the random number generation. Right now, I haven't written this to account for the fact that sigma should be non-negative, but that's on the user to know that (for now).

Value

dst() returns the density. pst() returns the distribution function. qst() returns the quantile function. rst() returns random numbers.

See Also

TDist
**tbl_df**

Convert data frame to an object of class "tibble"

**Description**

`tbl_df()` ensures legacy compatibility with some of my scripts since the function is deprecated in `dplyr`. `to_tbl()` also added for fun.

**Usage**

```r
tbl_df(...)  
```

```

```r
to_tbl(...)  
```

**Arguments**

... optional parameters, but don’t put anything here. It’s just there to quell CRAN checks.

**Value**

This function takes a data frame and turns it into a tibble.

**Examples**

```r
tbl_df(mtcars)  
tbl_df(iris)  
```

---

**theme_steve**

*Steve’s Preferred ggplot2 Themes and Assorted Stuff*

**Description**

`theme_steve()` was a preferred theme of mine a few years ago. It is basically `theme_bw()` from `ggplot2` theme, but with me tweaking a few things. I’ve since moved to `theme_steve_web()` for most things now, prominently on my website. It incorporates the "Open Sans" and "Titillium Web" fonts that I like so much. `post_bg()` is for changing the backgrounds on plots to better match my website for posts that I write. `theme_steve_ms()` is for LaTeX manuscripts that use the cochinel font package. `theme_steve_font()` is for any purpose, allowing you to supply your own font.
Usage

```r
theme_steve(...)  
theme_steve_web(...)  
post_bg(...)  
theme_steve_ms(axis_face = "italic", caption_face = "italic", ...)  
theme_steve_font(axis_face = "italic", caption_face = "italic", font, ...)```

Arguments

- `...`: optional stuff, but don’t put anything in here. You won’t need it.
- `axis_face`: font face ("plain", "italic", "bold", "bold.italic"). Optional, defaults to "italic". Applicable only to `theme_steve_ms()`.
- `caption_face`: font face ("plain", "italic", "bold", "bold.italic"). Optional, defaults to "italic". Applicable only to `theme_steve_ms()`.
- `font`: font family for the plot. Applicable only to `theme_steve_font()`.

Details

- `theme_steve_web()` depends on having the fonts installed on your end. It’s ultimately optional for you to have them.

Value

- `post_bg()` takes a `ggplot2` plot and changes the background to have a color of "#fdfdfd". `theme_steve()` takes a `ggplot2` plot and formats it to approximate `theme_bw()` from `ggplot2`, but with some other tweaks. `theme_steve_web()` extends `theme_steve()` to add custom fonts, notably "Open Sans" and "Titillium Web". In all cases, these functions take a `ggplot2` plot and return another `ggplot2` plot, but with some cosmetic changes. `theme_steve_ms()` takes a `ggplot2` plot and overlays "Crimson Text" fonts, which is the basis of the `cochineal` font package in LaTeX. `theme_steve_font()` takes a `ggplot2` plot and overlays a font of your choosing.

See Also

- `ggplot2::theme`

Examples

```r
## Not run:
library(ggplot2)

mtcars <- ggplot(mtcars, aes(x = mpg, y = hp)) +  
  geom_point() + theme_steve() +  
  labs(title = "A ggplot2 Plot from the Motor Trend Car Road Tests Data",  
       subtitle = "We’ve all seen this plot over a hundred times.",  
       caption = "Data: ?mtcars in {datasets} in base R.")
```
ggplot(mtcars, aes(x = mpg, y = hp)) +
  geom_point() + theme_steve_web() +
  labs(title = "A ggplot2 Plot from the Motor Trend Car Road Tests Data",
       subtitle = "Notice the prettier fonts, if you have them.",
       caption = "Data: ?mtcars in {datasets} in base R.")

ggplot(mtcars, aes(x = mpg, y = hp)) +
  geom_point() + theme_steve_web() +
  post_bg() +
  labs(title = "A ggplot2 Plot from the Motor Trend Car Road Tests Data",
       subtitle = "Notice the slight change in background color",
       caption = "Data: ?mtcars in {datasets} in base R.")

ggplot(mtcars, aes(x = mpg, y = hp)) +
  geom_point() + theme_steve_ms() +
  labs(title = "A ggplot2 Plot from the Motor Trend Car Road Tests Data",
       subtitle = "Notice the fonts will match the 'cochineal' font package in LaTeX.",
       caption = "Data: ?mtcars in {datasets} in base R.")

ggplot(mtcars, aes(x = mpg, y = hp)) +
  geom_point() + theme_steve_font(font = "Comic Sans MS") +
  labs(title = "A ggplot2 Plot from the Motor Trend Car Road Tests Data",
       subtitle = "Notice the fonts will match the 'cochineal' font package in LaTeX.",
       caption = "Data: ?mtcars in {datasets} in base R.")

## End(Not run)

---

**usa_mids**  
*United States Militarized Interstate Disputes (MIDs)*

**Description**

This is a non-directed dyad-year data set for militarized interstate disputes involving the United States. I use this to illustrate the `sbttscs()` function.

**Usage**

```r
data("usa_mids")
```

**Format**

A data frame with 14586 observations on the following 6 variables.

- **dyad** a numeric vector for a unique dyad-identifier
- **ccode1** a numeric vector for the lower Correlates of War country code in the dyad (i.e. `ccode1 == 2`)
- **ccode2** a numeric vector for the other Correlates of War country code in the dyad (i.e. not involving the United States)
year  a numeric vector for the year
midongoing  a numeric vector for whether there is a dispute ongoing in the dyad-year
midonset  a numeric vector for whether there is a unique dispute onset in the dyad-year

Details
Rare cases where there were multiple disputes ongoing in a given dyad-year were first whittled by isolating 1) unique dispute onsets. Thereafter, the data select the 2) highest fatality, then 3) the highest hostility level, and then 4) the longer dispute, until 5) just picking whichever one came first. There are no duplicate non-directed dyad-year observations.

Source

References

Examples
data(usa_mids)

%\texttt{\texttt{\textbackslash nin}}%  \hspace{1cm} \textit{Find Non-Matching Elements}

Description
%\texttt{\textbackslash nin}% finds non-matching elements in a given vector. It is the negation of %\texttt{\textbackslash in}%.

Usage
a \%\texttt{\textbackslash nin}% b

Arguments
a  \hspace{1cm} a vector (character, factor, or numeric)
b  \hspace{1cm} a vector (character, factor, or numeric)

Details
This is a simple negation of %\texttt{\textbackslash in}%. I use it mostly for columns in a data frame.
Value

%nin% finds non-matching elements and returns one of two things, depending on the use. For two simple vectors, it will report what matches and what does not. For comparing a vector within a data frame, it has the effect of reporting the rows in the data frame that do not match the supplied (second) vector.

Examples

```r
library(tibble)
library(dplyr)

# Watch this subset stuff

dat <- tibble(x = seq(1:10), d = rnorm(10))
filter(dat, x %nin% c(3, 6, 9))
```
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