# Package 'dotwhisker'

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Author Frederick Solt [aut],
     Yue Hu [aut, cre],
     Os Keyes [ctb],
     Ben Bolker [ctb],
     Stefan Müller [ctb],
     Thomas Leeper [ctb],
     Chris Wallace [ctb],
     Christopher Warshaw [ctb]
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# Description

add\_brackets draws brackets along the y-axis beyond the plotting area of a dot-and-whisker plot generated by dwplot, useful for labelling groups of predictors

# Usage

```
add_brackets(p, brackets, fontSize = 0.7, face = "italic", ...)
```

# Arguments

p	A plot generated by dwplot. Any 'ggplot' customization should be done before passing the plot to add_brackets. To pass the finalized plot to add_brackets without creating an intermediate object, simply wrap the code that generates it in braces (`{`and`}`).
brackets	A list of brackets; each element of the list should be a character vector consisting of (1) a label for the bracket, (2) the name of the topmost variable to be enclosed by the bracket, and (3) the name of the bottom most variable to be enclosed by the bracket.
fontSize	A number defining the size of the bracket label. The default value is .7.
face	A typeface for the bracket labels; options are "plain", "bold", "italic", "oblique", and "bold.italic".
	Extra arguments to pass to gpar.

## **Details**

The brackets are drawn by 'grid' functions. Apart from font size and typeface, users can customize the appearance of the bracket labels by setting 'gpar' arguments in 'add\_brackets'.

# Value

The function returns a ggplot object.

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#### **Examples**

by\_2sd

Rescale regression results by multiplying by 2 standard deviations

#### Description

by\_2sd rescales regression results to facilitate making dot-and-whisker plots using dwplot.

#### Usage

```
by_2sd(df, dataset)
```

#### Arguments

df

A data frame including the variables term (names of independent variables), estimate (corresponding coefficient estimates), std.error (corresponding standard errors), and optionally model (when multiple models are desired on a single plot) such as generated those by tidy.

dataset

The data analyzed in the models whose results are recorded in df, or (preferably) the *model matrix* used by the models in df; the information required for complex models can more easily be generated from the model matrix than from the original data set. In many cases the model matrix can be extracted from the original model via model.matrix.

#### **Details**

by\_2sd multiplies the results from regression models saved as tidy data frames for predictors that are not binary by twice the standard deviation of these variables in the dataset analyzed. Standardizing in this way yields coefficients that are directly comparable to each other and to those for untransformed binary predictors (Gelman 2008) and so facilitates plotting using dwplot. Note that the current version of by\_2sd does not subtract the mean (in contrast to Gelman's (2008) formula). However, all estimates and standard errors of the independent variables are the same as if the mean

was subtracted. The only difference from Gelman (2008) is that for all variables in the model the intercept is shifted by the coefficient times the mean of the variable.

An alternative available in some circumstances is to pass a model object to arm::standardize before passing the results to tidy and then on to dwplot. The advantages of by\_2sd are that (1) it takes a tidy data frame as its input and so is not restricted to only those model objects that standardize accepts and (2) it is much more efficient because it operates on the parameters rather than refitting the original model with scaled data.

#### Value

A tidy data frame

#### References

Gelman, Andrew. 2008. "Scaling Regression Inputs by Dividing by Two Standard Deviations." Statistics in Medicine, 27:2865-2873.

#### **Examples**

```
library(broom)
library(dplyr)

data(mtcars)
m1 <- lm(mpg ~ wt + cyl + disp, data = mtcars)
m1_df <- tidy(m1) %>% by_2sd(mtcars) # create data frame of rescaled regression results
```

dwplot

Dot-and-Whisker Plots of Regression Results

# **Description**

dwplot is a function for quickly and easily generating dot-and-whisker plots of regression models saved in tidy data frames.

## Usage

```
dwplot(
    x,
    ci = 0.95,
    dodge_size = 0.4,
    vars_order = NULL,
    show_intercept = FALSE,
    margins = FALSE,
    model_name = "model",
    model_order = NULL,
    style = c("dotwhisker", "distribution"),
    by_2sd = FALSE,
```

```
vline = NULL,
  dot_args = list(size = 1.2),
 whisker_args = list(size = 0.5),
  dist_args = list(alpha = 0.5),
  line_args = list(alpha = 0.75, size = 1),
)
dw_plot(
  ci = 0.95,
  dodge_size = 0.4,
  vars_order = NULL,
  show_intercept = FALSE,
 margins = FALSE,
 model_name = "model",
 model_order = NULL,
  style = c("dotwhisker", "distribution"),
 by_2sd = FALSE,
  vline = NULL,
  dot_args = list(size = 1.2),
 whisker_args = list(size = 0.5),
  dist_args = list(alpha = 0.5),
  line_args = list(alpha = 0.75, size = 1),
)
```

# Arguments

vars\_order

margins

model\_name

x Either a model object to be tidied with tidy, or a list of such model objects, or a tidy data frame of regression results (see 'Details').

ci A number indicating the level of confidence intervals; the default is .95.

A number indicating how much vertical separation should be between different models' coefficients when multiple models are graphed in a single plot. Lower values tend to look better when the number of independent variables is small, while a higher value may be helpful when many models appear on the same plot; the default is 0.4.

A vector of variable names that specifies the order in which the variables are to appear along the y-axis of the plot. Note that the order will be overwritten by relabel\_predictors, if the function is following called.

show\_intercept A logical constant indicating whether the coefficient of the intercept term should be plotted.

A logical value indicating whether presenting the average marginal effects of the estimates. See the Details for more information.

The name of a variable that distinguishes separate models within a tidy data frame.

model_order	A character vector defining the order of the models when multiple models are involved.
style	Either "dotwhisker" or "distribution". "dotwhisker", the default, shows the regression coefficients' point estimates as dots with confidence interval whiskers. "distribution" shows the normal distribution with mean equal to the point estimate and standard deviation equal to the standard error, underscored with a confidence interval whisker.
by_2sd	When x is model object or list of model objects, should the coefficients for predictors that are not binary be rescaled by twice the standard deviation of these variables in the dataset analyzed, per Gelman (2008)? Defaults to FALSE. Note that when x is a tidy data frame, one can use by 2sd to rescale similarly.
vline	A geom_vline() object, typically with xintercept = 0, to be drawn behind the coefficients.
dot_args	When style is "dotwhisker", a list of arguments specifying the appearance of the dots representing mean estimates. For supported arguments, see <pre>geom_point</pre> .
whisker_args	When style is "dotwhisker", a list of arguments specifying the appearance of the whiskers representing the confidence intervals. For supported arguments, see geom_linerangeh.
dist_args	When style is "distribution", a list of arguments specifying the appearance of normally distributed regression estimates. For supported arguments, see <a href="mailto:geom_polygon">geom_polygon</a> .
line_args	When style is "distribution", a list of arguments specifying the appearance of the line marking the confidence interval beneath the normal distribution. For supported arguments, see <pre>geom_linerangeh</pre> .
	Extra arguments to pass to parameters.

#### **Details**

dwplot visualizes regression model objects or regression results saved in tidy data frames as dotand-whisker plots generated by ggplot.

Tidy data frames to be plotted should include the variables term (names of predictors), estimate (corresponding estimates of coefficients or other quantities of interest), std.error (corresponding standard errors), and optionally model (when multiple models are desired on a single plot; a different name for this last variable may be specified using the model\_name argument). In place of std.error one may substitute conf.low (the lower bounds of the confidence intervals of each estimate) and conf.high (the corresponding upper bounds).

For convenience, dwplot also accepts as input those model objects that can be tidied by tidy (or tidy\_parameters, parameters (with proper formatting)), or a list of such model objects.

By default, the plot will display 95-percent confidence intervals. To display a different interval when passing a model object or objects, specify a ci argument. When passing a data frame of results, include the variables conf.low and conf.high describing the bounds of the desired interval.

Because the function can take a data frame as input, it is easily employed for a wide range of models, including those not supported by broom, broomExtra, or parameters. And because the output is a ggplot object, it can easily be further customized with any additional arguments and layers supported by ggplot2. Together, these two features make dwplot extremely flexible.

dwplot provides an option to present the average marginal effect directly based on margins. Users can alter the confidence intervals of the margins through the ci argument. See the full list of supported functions in the document of the package margins. The 'margins' argument also works for small\_multiple and secret\_weapon.

#### Value

The function returns a ggplot object.

#### References

Kastellec, Jonathan P. and Leoni, Eduardo L. 2007. "Using Graphs Instead of Tables in Political Science." \*Perspectives on Politics\*, 5(4):755-771.

Gelman, Andrew. 2008. "Scaling Regression Inputs by Dividing by Two Standard Deviations." \*Statistics in Medicine\*, 27:2865-2873.

# **Examples**

```
library(dplyr)
# Plot regression coefficients from a single model object
data(mtcars)
m1 <- lm(mpg ~ wt + cyl + disp, data = mtcars)
dwplot(m1, vline = geom_vline(xintercept = 0, colour = "grey50", linetype = 2)) +
    xlab("Coefficient")
# using 99% confidence interval
dwplot(m1, ci = .99)
# Plot regression coefficients from multiple models
m2 \leftarrow update(m1, . \sim . - disp)
dwplot(list(full = m1, nodisp = m2))
# Change the appearance of dots and whiskers
dwplot(m1, dot_args = list(size = 3, pch = 21, fill = "white"))
# Plot regression coefficients from multiple models on the fly
mtcars %>%
    split(.$am) %>%
   purrr::map(\sim lm(mpg \sim wt + cyl + disp, data = .x)) %>%
   dwplot() %>%
    relabel_predictors(c(wt = "Weight", cyl = "Cylinders", disp = "Displacement")) +
    theme_bw() + xlab("Coefficient") + ylab("") +
   geom_vline(xintercept = 0, colour = "grey60", linetype = 2) +
   ggtitle("Predicting Gas Mileage, OLS Estimates") +
    theme(plot.title = element_text(face = "bold"),
          legend.position = c(.995, .99),
          legend.justification = c(1, 1),
          legend.background = element_rect(colour="grey80"),
          legend.title.align = .5) +
    scale_colour_grey(start = .4, end = .8,
                      name = "Transmission",
                      breaks = c("Model 0", "Model 1"),
                      labels = c("Automatic", "Manual"))
```

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relabel\_predictors

Relabel the Predictors in a Tidy Data Frame of Regression Results

# **Description**

relabel\_predictors is a convenience function for relabeling the predictors in a tidy data frame to be passed to dwplot or a plot generated by dwplot

#### Usage

```
relabel_predictors(x, ...)
```

#### **Arguments**

x

Either a tidy data frame to be passed to dwplot or a plot generated by dwplot.

. . .

Named replacements, as in recode. The argument names should be the current values to be replaced, and the argument values should be the new (replacement) values. For backwards compatibility, a named character vector, with new values as values, and old values as names may also be used. The order of the named replacements will be preserved, so this function also serves the purpose of reordering variables.

#### Value

The function returns an object of the same type as it is passed: a tidy data frame or a plot generated by dwplot.

#### **Examples**

relabel\_y\_axis 9

relabel_y_axis	Relabel the Y-Axis of a Dot-Whisker Plot

## **Description**

relabel\_y\_axis DEPRECATED. A convenience function for relabeling the predictors on the y-axis of a dot-whisker plot created by dwplot. It is deprecated; use relabel\_predictors instead.

## Usage

```
relabel_y_axis(x)
```

## **Arguments**

x A vector of labels for predictors, listed from top to bottom

#### See Also

relabel\_predictors to relabel predictors on the y-axis of a dot-whisker plot or in a tidy data.frame

secret_weapon	Generate a 'Secret Weapon' Plot of Regression Results from Multiple Models

# Description

secret\_weapon is a function for plotting regression results of multiple models as a 'secret weapon' plot

# Usage

```
secret_weapon(x, var = NULL, ci = 0.95, margins = FALSE, by_2sd = FALSE, ...)
```

# Arguments

X	Either a model object to be tidied with tidy, or a list of such model objects, or a tidy data frame of regression results (see 'Details').
var	The predictor whose results are to be shown in the 'secret weapon' plot
ci	A number indicating the level of confidence intervals; the default is .95.
margins	A logical value indicating whether presenting the average marginal effects of the estimates. See the Details for more information.
by_2sd	When x is a list of model objects, should the coefficients for predictors that are not binary be rescaled by twice the standard deviation of these variables in the dataset analyzed, per Gelman (2008)? Defaults to TRUE. Note that when x is a tidy data frame, one can use by_2sd to rescale similarly.
	Arguments to pass to dwplot.

#### **Details**

Andrew Gelman has coined the term "the secret weapon" for dot-and-whisker plots that compare the estimated coefficients for a single predictor across many models or datasets. secret\_weapon takes a tidy data frame of regression results or a list of model objects and generates a dot-and-whisker plot of the results of a single variable across the multiple models.

Tidy data frames to be plotted should include the variables term (names of predictors), estimate (corresponding estimates of coefficients or other quantities of interest), std.error (corresponding standard errors), and model (identifying the corresponding model). In place of std.error one may substitute 1b (the lower bounds of the confidence intervals of each estimate) and ub (the corresponding upper bounds).

Alternately, secret\_weapon accepts as input a list of model objects that can be tidied by tidy (or tidy\_parameters, parameters (with proper formatting)), or a list of such model objects.

#### Value

The function returns a ggplot object.

# **Examples**

```
library(dplyr)
library(broom)

# Estimate models across many samples, put results in a tidy data frame
by_clarity <- diamonds %>% group_by(clarity) %>%
    do(tidy(lm(price ~ carat + cut + color, data = .))) %>%
    ungroup %>% rename(model = clarity)

# Generate a 'secret weapon' plot of the results of diamond size
secret_weapon(by_clarity, "carat")
```

small\_multiple

Generate a 'Small Multiple' Plot of Regression Results

# Description

small\_multiple is a function for plotting regression results of multiple models as a 'small multiple' plot

#### Usage

```
small_multiple(
   x,
   ci = 0.95,
   margins = FALSE,
```

```
dodge_size = 0.4,
show_intercept = FALSE,
model_order = NULL,
submodel_order = NULL,
axis_switch = FALSE,
by_2sd = FALSE,
dot_args = list(size = 0.3),
...
)
```

#### **Arguments**

X	Either a model object to be tidied with tidy, or a list of such model objects, or a tidy data frame of regression results (see 'Details').
ci	A number indicating the level of confidence intervals; the default is .95.
margins	A logical value indicating whether presenting the average marginal effects of the estimates. See the Details for more information.
dodge_size	A number (typically between 0 and 0.3; the default is .06) indicating how much horizontal separation should appear between different submodels' coefficients when multiple submodels are graphed in a single plot. Lower values tend to look better when the number of models is small, while a higher value may be helpful when many submodels appear on the same plot.
show_intercept	A logical constant indicating whether the coefficient of the intercept term should be plotted
model_order	A character vector defining the order of the models when multiple models are involved.
submodel_order	A character vector defining the order of the submodels when multiple submodels are involved.
axis_switch	A logical constant indicating the position of variable labels and y axis ticks. Default is FALSE, when the variable label is on the right side, and y axis ticks is on the left size.
by_2sd	When x is model object or list of model objects, should the coefficients for predictors that are not binary be rescaled by twice the standard deviation of these variables in the dataset analyzed, per Gelman (2008)? Defaults to TRUE. Note that when x is a tidy data frame, one can use by_2sd to rescale similarly.
dot_args	A list of arguments specifying the appearance of the dots representing mean estimates. For supported arguments, see <a href="mailto:geom_pointrange">geom_pointrange</a> h.
	Arguments to pass to dwplot.

#### **Details**

small\_multiple, following Kastellec and Leoni (2007), provides a compact means of representing numerous regression models in a single plot.

Tidy data frames to be plotted should include the variables term (names of predictors), estimate (corresponding estimates of coefficients or other quantities of interest), std.error (corresponding standard errors), and model (identifying the corresponding model). In place of std.error one may

substitute conf. low (the lower bounds of the confidence intervals of each estimate) and conf. high (the corresponding upper bounds).

Alternately, small\_multiple accepts as input a list of model objects that can be tidied by tidy (or tidy\_parameters, parameters (with proper formatting)), or a list of such model objects.

Optionally, more than one set of results can be clustered to facilitate comparison within each model; one example of when this may be desirable is to compare results across samples. In that case, the data frame should also include a variable submodel identifying the submodel of the results.

#### Value

The function returns a ggplot object.

#### References

Kastellec, Jonathan P. and Leoni, Eduardo L. 2007. "Using Graphs Instead of Tables in Political Science." Perspectives on Politics, 5(4):755-771.

#### **Examples**

```
library(broom)
library(dplyr)
# Generate a tidy data frame of regression results from six models
m <- list()
ordered_vars <- c("wt", "cyl", "disp", "hp", "gear", "am")</pre>
m[[1]] \leftarrow lm(mpg \sim wt, data = mtcars)
m123456_df <- m[[1]] %>% tidy %>% by_2sd(mtcars) %>%
  mutate(model = "Model 1")
for (i in 2:6) {
 m[[i]] <- update(m[[i-1]], paste(". ~ . +", ordered_vars[i]))</pre>
 m123456_df <- rbind(m123456_df, m[[i]] %>% tidy %>% by_2sd(mtcars) %>%
   mutate(model = paste("Model", i)))
}
# Generate a 'small multiple' plot
small_multiple(m123456_df)
## Using submodels to compare results across different samples
# Generate a tidy data frame of regression results from five models on
# the mtcars data subset by transmission type (am)
ordered_vars <- c("wt", "cyl", "disp", "hp", "gear")</pre>
mod <- "mpg ~ wt"
by_trans <- mtcars %>% group_by(am) %>% # group data by transmission
  do(tidy(lm(mod, data = .))) %>%  # run model on each group
  rename(submodel = am) %>%
                                         # make submodel variable
  mutate(model = "Model 1") %>%
                                        # make model variable
 ungroup()
```

```
for (i in 2:5) {
  mod <- paste(mod, "+", ordered_vars[i])</pre>
   by_trans <- rbind(by_trans, mtcars %>% group_by(am) %>%
                         do(tidy(lm(mod, data = .))) %>%
                         rename(submodel = am) %>%
                         mutate(model = paste("Model", i)) %>%
                         ungroup())
}
small_multiple(by_trans) +
theme_bw() + ylab("Coefficient Estimate") +
    geom_hline(yintercept = 0, colour = "grey60", linetype = 2) +
    theme(axis.text.x = element_text(angle = 45, hjust = 1),
          legend.position=c(0, 0), legend.justification=c(0, 0),
          legend.title = element_text(size=9),
          legend.background = element_rect(color="gray90"),
          legend.spacing = unit(-3, "pt"),
          legend.key.size = unit(10, "pt")) +
    scale_colour_hue(name = "Transmission",
    breaks = c(0, 1),
    labels = c("Automatic", "Manual"))
```

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