Package 'allestimates'

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Type Package

Title Effect Estimates from All Models

Version 0.2.3

Description Estimates and plots effect estimates from models with all possible combinations of a list of variables. It can be used for assessing treatment effects in clinical trials or risk factors in bio-medical and epidemiological research. Like Stata command 'confall' (Wang Z (2007) <doi:10.1177/1536867X0700700203>), 'allestimates' calculates and stores all effect estimates, and plots them against p values or Akaike information criterion (AIC) values. It currently has functions for linear regression: all_lm(), logistic and Poisson regression: all_glm(), and Cox proportional hazards regression: all_cox().

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Encoding UTF-8

LazyData true

Imports broom, ggplot2, survival, tidyr, utils, stringr, dplyr

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Contents

`allestimates` *Effect estimates from models with all possible combinations of variables*

Description

To assess treatment effects in clinical trials and risk factors in bio-medical and epidemiological research, we use *regression coefficients*, *odds ratios* or *hazard ratios* as *effect estimates*. `allestimates` allows users to quickly obtain effect estimates from models with all possible combinations of a list of variables specified by users. `all_lm` for linear regression, `all_glm` for logistic regression, `all_speedglm` using `speedlm` as a faster alternative of `all_glm`, and `all_cox` for Cox Proportional Hazards Models. Users can further use those values in a returned list of results. `all_plot` draws scatter plots with all effect estimate values against p values, as `Stata confall` command (Wang Z (2007) <doi:10.1177/1536867X0700700203>). Those plots divide estimates into four categories:

Details

- positive and significant: left-top quarter
- negative and significant: left-bottom quarter
- positive and non-significant: right-top quarter
- negative and non-significant: right-bottom quarter

`all_plot2` draws multiple plots. Each of those plots indicates whether a specific variable is included or not included in models. Those effect estimates help users better understand confounding effects, uncertainty of their estimates, as well as inappropriately including variables in the models. This is a tool for calculating and exploring effect estimates from all possible models. Interpretation of the results should be in the context of other analyses and biological knowledge.

Examples

```
? all_speedglm
? all_glm
? all_cox
? all_lm
? all_plot
? all_plot2
```

`all_cox` *Estimates all possible effect estimates using Cox Proportional Hazards regression models*

Description

Estimates hazard ratios using Proportional Hazards Regression models ("coxph" from **survival** package) from models with all possible combinations of a list of variables.

Usage

```
all_cox(crude, xlist, data, na_omit = TRUE, ...)
```

Arguments

crude	An object of *formula* for initial model, generally crude model. However, any other variables can also be included here as the initial model. The left-hand side of ~ is the outcome of interest, and the variable on the right-hand side of ~ is the exposure of the interest (either a treatment or a risk factor)
xlist	A *vector* of a list of variable names.
data	*Data frame*.
na_omit	Remove all missing values. Default is "na_omit = TRUE".
...	Further optional arguments.

Value

A list of all effect estimates.

See Also

surival

Examples

```
vlist <- c("Age", "Sex", "Married", "BMI", "Education", "Income")
results <- all_cox(crude = "Surv(t0, t1, Endpoint) ~ Diabetes", xlist = vlist, data = diab_df)
results
```

all_glm *Estimates all possible effect estimates using* glm

Description

all_glm estimates odds ratios or rate ratios using generalized linear models (glm) with all possible combinations of a list of variables (potential confounding factors).

Usage

```
all_glm(crude, xlist, data, family = "binomial", na_omit = TRUE, ...)
```

Arguments

crude	An object of *formula* for initial model, generally crude model. However, any other variables can also be included here as the initial model.
xlist	A *vector* of a list of variable names (potential confounding factors).
data	*Data frame*.
family	*family* Description of the error distribution. Default is "binomial".
na_omit	Remove all missing values. Default is "na_omit = TRUE".
...	Further optional arguments.

Value

A list of all effect estimates.

See Also

stats

Examples

```
diab_df$Overweight <- as.numeric(diab_df$BMI >= 25)
vlist <- c("Age", "Sex", "Income")
all_glm(crude = "Diabetes ~ Overweight", xlist = vlist, data = diab_df)
```

all_lm *Estimates all possible effect estimates using* lm

Description

all_lm estimates coefficients of a specific variable using linear models (lm) with all possible combinations of other variables (potential confounding factors).

Usage

```
all_lm(crude, xlist, data, na_omit = TRUE, ...)
```

Arguments

crude	An object of *formula* for initial model, generally crude model. However, additional variables can also be included here as the initial model.
xlist	A *vector* of a list of variable names (potential confounding factors).
data	*Data frame*.
na_omit	Remove all missing values. Default is "na_omit = TRUE".
...	Further optional arguments.

Value

A list of all effect estimates.

See Also

lm

Examples

```
vlist <- c("Age", "Sex", "Cancer", "CVD", "Education", "Income")
all_lm(crude = "BMI ~ Married", xlist = vlist, data = diab_df)
```

all_plot *Plot all effect estimates against p values*

Description

all_plot() generates a scatter plot with effect estimates of all possible models again p values.

Usage

```
all_plot(
  data,
  xlabels = c(0, 0.001, 0.01, 0.05, 0.2, 0.5, 1),
  xlim = c(0, 1),
  xlab = "P value",
  ylim = NULL,
  ylab = NULL,
  yscale_log = FALSE,
  title = NULL
)
```

Arguments

data	*Object* from all_cox, all_glm, all_speedglm, or all_glm, including all effect estimate values.
xlabels	*Numeric vector* x-axis tick labels. Default is "c(0, 0.001, 0.01, 0.05, 0.2, 0.5, 1)".
xlim	*Vector* of 2 numeric values for x-axis limits. Default is "c(0, 1)".
xlab	*Character* string for x-axis name. Default is "P value".
ylim	*Vector* of 2 numeric values for y-axis limits.
ylab	*Character* string for y-axis name. Default depends on original model types.
yscale_log	*TRUE or FALSE* to re-scale y-axis to "log10". Default is "FALSE".
title	*Character* for plot title. Default is "NULL".

Value

A **ggplot2** object: scatter plot

Examples

```
vlist <- c("Age", "Sex", "Married", "BMI", "Education", "Income")
results <- all_cox(crude = "Surv(t0, t1, Endpoint) ~ Diabetes", xlist = vlist, data = diab_df)
all_plot(results)
```

---

all_plot2 *Plots all effect estimates against p values with each specific variable in the models*

Description

all_plot2() generates a panel of scatter plots with effect estimates of all possible models again p values. Each plot includes effect estimates from all models including a specific variable.

Usage

```
all_plot2(
  data,
  xlabels = c(0, 0.001, 0.01, 0.05, 0.2, 0.5, 1),
  xlim = c(0, 1),
  xlab = "P value",
  ylim = NULL,
  ylab = NULL,
  yscale_log = FALSE,
  title = NULL
)
```

Arguments

data	*Object* from all_cox, all_glm, all_speedglm, or all_glm, including all effect estimate values.
xlabels	*numeric vector* x-axis tick labels. Default is "c(0, 0.001, 0.01, 0.05, 0.2, 0.5, 1)"
xlim	*vector* of 2 numeric values for x-axis limits. Default is "c(0, 1)".
xlab	*Character* string for x-axis name. Default is "P value".
ylim	*vector* of 2 numeric values for y-axis limits.
ylab	*Character* string for y-axis name. Default depends on original model types.
yscale_log	*TRUE or FALSE* re-scale y-axis to "log10". Default is "FALSE".
title	*Character* title. Default is "NULL".

Value

A **ggplot2** object: scatter plot

Examples

```
vlist <- c("Age", "Sex", "Married", "BMI", "Income")
results <- all_cox(crude = "Surv(t0, t1, Endpoint) ~ Diabetes", xlist = vlist, data = diab_df)
all_plot2(results)
```

---

all_plot_aic *Draws scatter plot with all effect estimates against AIC*

Description

all_plot_aic() generates a scatter plot with all effect estimates against AIC.

Usage

```
all_plot_aic(data, xlab = "AIC", ylab = NULL, title = NULL)
```

Arguments

data	*Object* from all_cox, all_glm, all_speedglm, or all_glm, including all effect estimate values.
xlab	*Character* string for x-axis name. Default is "AIC"
ylab	*Character* string for y-axis name. Default depends on original model types.
title	*Character* for plot title. Default is "NULL".

Value

A **ggplot2** object: scatter plot

Examples

```
vlist <- c("Age", "Sex", "Married", "BMI", "Education", "Income")
results <- all_cox(crude = "Surv(t0, t1, Endpoint) ~ Diabetes", xlist = vlist, data = diab_df)
all_plot_aic(results)
```

---

all_plot_aic2 *Draws multiple scatter plots of all effect estimates against AIC*

Description

all_plot_aic2() draws multiple scatter plots of all effect estimates against AIC. Each plot indicates if a specific variable is included in the models.

Usage

```
all_plot_aic2(data, xlab = "AIC", ylab = NULL, title = NULL)
```

Arguments

data	*Object* from all_cox, all_glm, all_speedglm, or all_glm, including all effect estimate values.
xlab	*Character* string for x-axis name. Default is "AIC".
ylab	*Character* string for y-axis name. Default depends on original model types.
title	*Character* for plot title. Default is "NULL".

Value

A **ggplot2** object: scatter plot.

Examples

```
vlist <- c("Age", "Sex", "Married", "BMI", "Education", "Income")
results <- all_cox(crude = "Surv(t0, t1, Endpoint) ~ Diabetes", xlist = vlist, data = diab_df)
all_plot_aic(data = results)
```

diab_df *Example data: Health outcomes of 2372 adults with and without diabetes*

Description

A data frame with 2372 rows and 14 variables with diabetes status `diabetes` and mortality status `endpoint`. For the purpose of demonstrate, assume that we are interested in the association between `diabetes` and `endpoint`. Other variables are considered as possible confounding factors. The purposes of this dataset is to illustrate those functions in **chest** and **allestimates** packages only. Therefore, we assume it is a cohort design for Cox Proportional Hazard regression, and a case-control design for logistic regression.

Usage

```
diab_df
```

Format

A data frame with 2372 rows and 14 variables:

Diabetes diabetes status 1: with diabetes 0: without diabetes

Endpoint mortality status 1: reached end point, and 0: survived

Age Age, in years

Sex sex, 1: male, 2: Female

BMI Body mass index

Married marital status 1: married, 0: not

Smoke smoking status 1: smoker, 0: non-smoker

CVD cardiovascular disease 1: yes 0: no

Cancer cancer 1: yes, 0: no

Education education 1: high, 0: low

Income income 1: high, 0: low

t0 time (age) at the start of the follow-up

t1 time (age) at the end of the follow-up

mid matched set id, for conditional logistic regression

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