

# Package ‘modmarg’

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**Title** Calculating Marginal Effects and Levels with Errors

**Version** 0.9.6

**Description** Calculate predicted levels and marginal effects, using the delta method to calculate standard errors. This is an R-based version of the 'margins' command from Stata.

**URL** <https://github.com/anniejw6/modmarg>

**BugReports** <https://github.com/anniejw6/modmarg/issues>

**Depends** R (>= 3.5.0)

**License** GPL-3

**Encoding** UTF-8

**LazyData** true

**Suggests** knitr, rmarkdown, testthat, sandwich, AER

**VignetteBuilder** knitr

**RoxygenNote** 7.1.1

**NeedsCompilation** no

**Author** Alex Gold [aut],  
Nat Olin [aut],  
Annie Wang [aut, cre]

**Maintainer** Annie Wang <anniejw6@gmail.com>

**Repository** CRAN

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`cvcov`*Clustered variance-covariance matrices and T statistic d.o.f.*

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**Description**

Variance-covariance matrices with robust clustered standard errors and degrees-of-freedom for T statistics, for tests and examples specifying `vcov` (d.o.f. defined as  $g - 1$ , where  $g$  is the number of clusters). Generated with `margex` data in this package.

**Usage**`cvcov`**Format**

A list of three lists, from an OLS model, logit model, and OLS with a polynomial interaction with missing data, each containing

**clust** 3-by-3 variance-covariance matrix

**dof** integer, degrees of freedom for the T statistic

**Details**

See `data-raw/make_cluster_vcov.R` for details.

**Source**

[http://cameron.econ.ucdavis.edu/research/Cameron\\_Miller\\_JHR\\_2015\\_February.pdf](http://cameron.econ.ucdavis.edu/research/Cameron_Miller_JHR_2015_February.pdf)

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`marg`*Estimating predictive margins on a model*

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**Description**

This function estimates the predictive effects and levels for variables within a model using the delta method.

**Usage**

```
marg(  
  mod,  
  var_interest,  
  data = NULL,  
  weights = NULL,  
  vcov_mat = NULL,  
  dof = NULL,
```

```

    type = "levels",
    base_rn = 1,
    at_var_interest = NULL,
    at = NULL,
    cofint = 0.95,
    ...
  )

```

## Arguments

mod	model object, currently only support those of class <code>glm</code> or <code>ivreg</code>
var_interest	name of the variable of interest, must correspond to a covariate in the model
data	data.frame that margins should run over, defaults changes based on class-specific method
weights	numeric, vector of weights used to generate predicted levels, defaults changes based on class-specific method. Must be equal to the number of rows in data.
vcov_mat	the variance-covariance matrix, defaults changes based on class-specific method
dof	integer, the degrees of freedom used for the T statistic in an OLS model, defaults changes based on class-specific method
type	either 'levels' (predicted outcomes) or 'effects' <i>dydx</i> , defaults to 'levels'
base_rn	numeric, if type == 'effects', the base level (taken as the index of one of the ordered unique values in var_interest). if type == 'levels', this parameter is ignored. Defaults to 1.
at_var_interest	vector, if type == 'levels', the values for the variable of interest at which levels should be calculated. If NULL, indicates all levels for a factor variable, defaults to NULL
at	list, should be in the format of <code>list('var_name' = c(values))</code> , defaults to NULL. This calculates the margins of the variable at these particular variables. If all values are needed, suggested syntax is <code>at = list('var' = unique(df\$var))</code> .
cofint	numeric, confidence interval (must be less than 1), defaults to 0.95
...	additional parameters passed to class-specific methods

## Details

The variable for the predictive margin is specified by `var_interest`. If margins are only needed at particular values of `var_interest`, `at_var_interest` should be used. If margins of `var_interest` are needed at across the levels of a *different* variable in the model, `at` should be used.

If higher-order polynomial terms (e.g.  $y = x + x^2$ ) are added using the R function `poly`, the `raw = TRUE` argument should be used to include the basic polynomial terms instead of orthogonal polynomial terms. If orthogonal polynomials are used, `marg` will fail when the user specifies `at` for a small set of values for the variable in question (e.g. `at = list(x = 10)`), since `poly` needs more data to calculate orthogonal polynomials (e.g. `poly(10, 2)` fails, but `poly(c(10, 8, 3), 2)` will run).

P values are calculated with T tests for gaussian families, and Z tests otherwise. If a new variance-covariance matrix is provided (e.g. for clustering standard errors), the degrees of freedom for the T test / p-value calculation may need to be specified using `dof`.

This function currently only supports `glm` and `ivreg` objects. If you would like to use `lm` objects, consider running a `glm` with family `gaussian`.

When calculating predicted levels and effects for models built using weights, `marg` returns weighted averages for levels and effects by default. Users can remove this option by setting `weights = NULL`.

## Value

list of dataframes with predicted margins/effects, standard errors, p-values, and confidence interval bounds

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marg.glm	<i>Predicted Margins for 'glm' objects</i>
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## Description

Obtains predicted margins and standard errors of those predictions from a fitted generalized linear model object.

## Usage

```
## S3 method for class 'glm'
marg(
  mod,
  var_interest,
  data = mod$data[names(mod$prior.weights), ],
  weights = mod$prior.weights,
  ...
)
```

## Arguments

<code>mod</code>	model object, currently only support those of class <code>glm</code> or <code>ivreg</code>
<code>var_interest</code>	name of the variable of interest, must correspond to a covariate in the model
<code>data</code>	data.frame that margins should run over, defaults changes based on class-specific method
<code>weights</code>	numeric, vector of weights used to generate predicted levels, defaults changes based on class-specific method. Must be equal to the number of rows in data.
<code>...</code>	additional parameters passed to <code>?marg</code> .

## Examples

```
data(mtcars)
mod <- glm(vs ~ as.factor(gear) + mpg, data = mtcars, family = 'binomial')

# Get the level of the outcome variable at different values of `gear`
marg(mod, var_interest = 'gear', type = 'levels')
# Get the effect of `gear` on the outcome value, holding values of `mpg`
```

```

# constant
marg(mod, var_interest = 'gear', type = 'effects',
      at = list(mpg = c(15, 21)))

data(margex)
mod <- glm(outcome ~ as.factor(treatment) + distance,
           data = margex, family = 'binomial')
# Get the level of the outcome variable at different values of `treatment`
marg(mod, var_interest = 'treatment', type = 'levels', at = NULL)
# Get the effect of `treatment` on the outcome variable
marg(mod, var_interest = 'treatment', type = 'effects', at = NULL)
# Get the level of the outcome variable at different values of `distance`
marg(mod, var_interest = 'distance', type = 'levels',
      at = NULL, at_var_interest = c(10, 20, 30))

# Using a custom variance-covariance matrix for clustered standard errors
# (also requires custom degrees of freedom for T statistic with OLS model),
# clustering on the "arm" variable

data(margex)
data(cvcov)
# ?cvcov
v <- cvcov$ols$clust
d <- cvcov$ols$stata_dof
mod <- glm(outcome ~ treatment + distance,
           data = margex, family = 'binomial')
marg(mod, var_interest = 'treatment', type = 'levels',
      vcov_mat = v, dof = d)

# Using weights
data(margex)
mm <- glm(y ~ as.factor(treatment) + age, data = margex, family = 'gaussian',
          weights = distance)
z1 <- marg(mod = mm, var_interest = 'treatment', type = 'levels')[[1]]
z2 <- marg(mod = mm, var_interest = 'treatment', type = 'effects')[[1]]

```

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marg.ivreg

*Predicted Margins for 'ivreg' objects from the AER package*


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## Description

Obtains predicted margins and standard errors of those predictions from a fitted `ivreg` model object.

## Usage

```

## S3 method for class 'ivreg'
marg(mod, var_interest, data, weights = NULL, ...)

```

**Arguments**

mod	model object, currently only support those of class <code>glm</code> or <code>ivreg</code>
var_interest	name of the variable of interest, must correspond to a covariate in the model
data	data.frame that margins should run over, defaults changes based on class-specific method
weights	numeric, vector of weights used to generate predicted levels, defaults changes based on class-specific method. Must be equal to the number of rows in data.
...	additional parameters passed to <code>?marg</code> .

**Examples**

```
# From ?AER::ivreg

# data
data("CigarettesSW", package = "AER")
CigarettesSW$rprice <- with(CigarettesSW, price/cpi)
CigarettesSW$rincome <- with(CigarettesSW, income/population/cpi)
CigarettesSW$tdiff <- with(CigarettesSW, (taxs - tax)/cpi)

# model
fm <- AER::ivreg(log(packs) ~ log(rprice) + log(rincome) |
                 log(rincome) + tdiff + I(tax/cpi),
                 data = CigarettesSW, subset = year == "1995")

# Get margins for different levels of price/cpi
rprice_levs <- round(quantile(CigarettesSW$rprice))

marg(fm, data = subset(CigarettesSW, year == "1995"),
     var_interest = 'rprice', at_var_interest = rprice_levs)
```

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margex

*Artificial data for margins*


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**Description**

A fictitious dataset outcome, treatment, and demographic variables for 3000 observations.

**Usage**

```
margex
```

**Format**

A data frame with 3000 rows and 11 variables:

**y** numeric

**outcome** integer, 0 or 1  
**sex** character: "female" or "male"  
**group** integer  
**age** integer  
**distance** numeric  
**ycn** numeric  
**yc** numeric, 0 or 1  
**treatment** integer  
**agegroup** character: "20-29", "30-39", or "40+"  
**arm** integer

### Source

<https://www.stata-press.com/data/r14/margex.dta>

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pred\_se

*Main wrapper function to calculate margins and standard errors*

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### Description

For one set of transformed covariates (not including the variable of interest), calculate the predicted level and standard error for the variable of interest.

### Usage

```
pred_se(  
  df_levels,  
  model,  
  type,  
  base_rn,  
  vcov_mat,  
  weights,  
  deriv_func,  
  link_func  
)
```

### Arguments

df_levels	data.frame, already transformed for variables not related to the variable of interest
model	model object
type	either effects or levels
base_rn	numeric, row number of the base level

<code>vcov_mat</code>	matrix, variance-covariance matrix
<code>weights</code>	vector of weights, or NULL
<code>deriv_func</code>	function for the derivative of the predicted outcomes
<code>link_func</code>	function to transform output of 'predict' method into response scale

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