To cite Zelig as a whole, please reference these two sources:

Kosuke Imai, Gary King, and Olivia Lau. 2007. "Zelig: Everyone's Statistical Software," magentahttp://GKing.harvard.edu/zelig.

Imai, Kosuke, Gary King, and Olivia Lau. (2008). "Toward A Common Framework for Statistical Analysis and Development." Journal of Computational and Graphical Statistics, Vol. 17, No. 4 (December), pp. 892-913.

1 gamma: Gamma Regression for Continuous, Positive Dependent Variables

Use the gamma regression model if you have a positive-valued dependent variable such as the number of years a parliamentary cabinet endures, or the seconds you can stay airborne while jumping. The gamma distribution assumes that all waiting times are complete by the end of the study (censoring is not allowed).

1.0.1 Syntax

```
> z.out <- zelig(Y ~ X1 + X2, model = "gamma", data = mydata)
> x.out <- setx(z.out)
> s.out <- sim(z.out, x = x.out, x1 = NULL)</pre>
```

1.0.2 Additional Inputs

In addition to the standard inputs, zelig() takes the following additional options for gamma regression:

• robust: defaults to FALSE. If TRUE is selected, zelig() computes robust standard errors via the sandwich package (see ?). The default type of robust standard error is heteroskedastic and autocorrelation consistent (HAC), and assumes that observations are ordered by time index.

In addition, robust may be a list with the following options:

- method: Choose from
 - * "vcovHAC": (default if robust = TRUE) HAC standard errors.
 - * "kernHAC": HAC standard errors using the weights given in ?.
 - * "weave": HAC standard errors using the weights given in ?.
- order.by: defaults to NULL (the observations are chronologically ordered as in the original data). Optionally, you may specify a vector of weights (either as order.by = z, where z exists outside the data frame; or as order.by = ~z, where z is a variable in the data frame). The observations are chronologically ordered by the size of z.
- ...: additional options passed to the functions specified in method. See the sandwich library and? for more options.

1.0.3 Example

Attach the sample data:

```
> data(coalition)
```

Estimate the model:

```
> z.out <- zelig(duration ~ fract + numst2, model = "gamma", data = coalition)
```

View the regression output:

```
> summary(z.out)
```

Set the baseline values (with the ruling coalition in the minority) and the alternative values (with the ruling coalition in the majority) for X:

 $> x.low \leftarrow setx(z.out, numst2 = 0)$

> x.high <- setx(z.out, numst2 = 1)

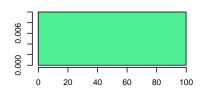
Simulate expected values (qi\$ev) and first differences (qi\$fd):

> s.out <- sim(z.out, x = x.low, x1 = x.high)

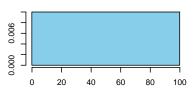
> summary(s.out)

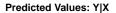
> plot(s.out)

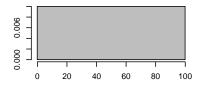
Expected Values: E(Y|X)



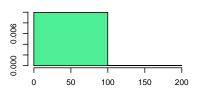
Expected Values (for X1): E(Y|X1)

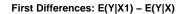


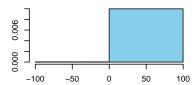




Predicted Values: Y|X1







1.0.4 Model

• The Gamma distribution with scale parameter α has a stochastic component:

$$Y \sim \operatorname{Gamma}(y_i \mid \lambda_i, \alpha)$$

$$f(y) = \frac{1}{\alpha^{\lambda_i} \Gamma \lambda_i} y_i^{\lambda_i - 1} \exp -\left\{\frac{y_i}{\alpha}\right\}$$

for $\alpha, \lambda_i, y_i > 0$.

• The systematic component is given by

$$\lambda_i = \frac{1}{x_i \beta}$$

1.0.5 Quantities of Interest

• The expected values (qi\$ev) are simulations of the mean of the stochastic component given draws of α and β from their posteriors:

$$E(Y) = \alpha \lambda_i$$
.

- The predicted values (qi\$pr) are draws from the gamma distribution for each given set of parameters (α, λ_i) .
- If x1 is specified, sim() also returns the differences in the expected values (qi\$fd),

$$E(Y \mid x_1) - E(Y \mid x)$$

.

• In conditional prediction models, the average expected treatment effect (att.ev) for the treatment group is

$$\frac{1}{\sum_{i=1}^{n} t_i} \sum_{i:t_i=1}^{n} \left\{ Y_i(t_i=1) - E[Y_i(t_i=0)] \right\},\,$$

where t_i is a binary explanatory variable defining the treatment $(t_i = 1)$ and control $(t_i = 0)$ groups. Variation in the simulations are due to uncertainty in simulating $E[Y_i(t_i = 0)]$, the counterfactual expected value of Y_i for observations in the treatment group, under the assumption that everything stays the same except that the treatment indicator is switched to $t_i = 0$.

• In conditional prediction models, the average predicted treatment effect (att.pr) for the treatment group is

$$\frac{1}{\sum_{i=1}^{n} t_i} \sum_{i:t_i=1}^{n} \left\{ Y_i(t_i=1) - \widehat{Y_i(t_i=0)} \right\},\,$$

where t_i is a binary explanatory variable defining the treatment $(t_i = 1)$ and control $(t_i = 0)$ groups. Variation in the simulations are due to uncertainty in simulating $\widehat{Y_i(t_i = 0)}$, the counterfactual predicted value of Y_i for observations in the treatment group, under the assumption that everything stays the same except that the treatment indicator is switched to $t_i = 0$.

1.0.6 Output Values

The output of each Zelig command contains useful information which you may view. For example, if you run z.out <- zelig(y ~ x, model = "gamma", data), then you may examine the available information in z.out by using names(z.out), see the coefficients by using z.out\$coefficients, and a default summary of information through summary(z.out). Other elements available through the \$ operator are listed below.

• From the zelig() output object z.out, you may extract:

- coefficients: parameter estimates for the explanatory variables.
- residuals: the working residuals in the final iteration of the IWLS fit.
- fitted.values: the vector of fitted values.
- linear.predictors: the vector of $x_i\beta$.
- aic: Akaike's Information Criterion (minus twice the maximized log-likelihood plus twice the number of coefficients).
- df.residual: the residual degrees of freedom.
- df.null: the residual degrees of freedom for the null model.
- zelig.data: the input data frame if save.data = TRUE.
- From summary(z.out), you may extract:
 - coefficients: the parameter estimates with their associated standard errors, p-values, and t-statistics.
 - cov.scaled: a $k \times k$ matrix of scaled covariances.
 - cov.unscaled: a $k \times k$ matrix of unscaled covariances.
- From the sim() output object s.out, you may extract quantities of interest arranged as matrices indexed by simulation × x-observation (for more than one x-observation). Available quantities are:
 - qi\$ev: the simulated expected values for the specified values of x.
 - qi\$pr: the simulated predicted values drawn from a distribution defined by (α, λ_i) .
 - qi\$fd: the simulated first difference in the expected values for the specified values in x and x1.
 - qi\$att.ev: the simulated average expected treatment effect for the treated from conditional prediction models.
 - qi\$att.pr: the simulated average predicted treatment effect for the treated from conditional prediction models.

How to Cite

See also

The gamma model is part of the stats package by ?. Advanced users may wish to refer to help(glm) and help(family), as well as ?. Robust standard errors are implemented via the sandwich package by ?. Sample data are from ?.