Package ‘intRegGOF’

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Description

Integrated Regression Goodness of Fit to test the adequacy of different model to represent the regression function for a given data.
Usage

```r
anovarIntReg(objh0, ..., covars = NULL, B = 499,
               LINMOD = FALSE, INCREMENTAL = FALSE)
print.anovarIntReg(x, ...)
```

Arguments

- **objh0**: An object of class `lm`, `glm` or `nls` which will be considered as null hypotheses model or the base reference mode when `INCREMENTAL` is set to `TRUE`.
- ...: One or more objects of class `lm`, `glm` or `nls`
- **covars**: Names of continuous (numerical) variates used to compute Integrated Regression. They should be variables contained in the data frame used to compute the regression fit. When `NULL` it is obtained as the max. number of different covariates in all tested models. It also can be a `formula` like `~ x1 + x2 + ...`.
- **B**: Bootstrap resampling size.
- **LINMOD**: When `TRUE` and if `obj` is an object of class `print.intRegGOFlm` Linear Model matrix fitting equations are used.
- **INCREMENTAL**: When is `FALSE` all models in ... are tested against `objh0`, while when `TRUE` each of the models are checked against the next one starting in `objh0`.
- **x**: An object of class `anovarIntReg`.

Details

This function implements the test

\[ H_0 : m \in M_0 \text{ vs } H_1 : m \in M_1 \]

for two different models \( M_0, M_1 \) using the Integrated Regression Goodness of Fit as os done in `intRegGOF`, but instead of the accumulation of the residual of a given model, in this case, the accumulation of the difference in the fits is considered:

\[ R_n^w(x) = n^{-1/2} \sum_{i=1}^{n} (\hat{y}_{0i} - \hat{y}_{1i}) I(x_i \leq x). \]

The test statistics considered are $K_n$ and $W_n^2$.

If `objh0` and `objh1` are `lm`, `glm` or `nls` fits for the models in classes \( M_0 \) and \( M_1 \) respectively, then `anovarIntReg(objh0, objh1)` computes test \( H_0 : m \in M_0 \text{ vs } H_1 : m \notin M_1 \). When `anovarIntReg(objh0, objh1, ..., objhk)` is executed (notice that by default `INCREMENTAL=FALSE`) we obtain a table with the statistics \( K_n \) and \( W_n^2 \) and its associated \( p \)-values for each of the tests \( H_0 : m \in M_0 \text{ vs } H_i : m \notin M_i \) being \( i = 1, \ldots, k \). On the other hand, if the parameter `INCREMENTAL` is set to `TRUE`, the command returns the results for the tests \( H_i : m \in M_i \text{ vs } H_{i+1} : m \notin M_{i+1} \) being \( i = 1, \ldots, k - 1 \).

Value

This function returns an object of class `anovarIntReg`, a matrix like structure whose rows refers to models and columns to statistics and its \( p \)-values. It also has an attribute `heading` to support printing the object.
Core Functions

Note

This method requires more testing, and careful study of the effect of factors (discrete random variables) when fitting the model.

Author(s)

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See Also

lm, glm, nls, and intRegGOF.

Examples

```r
n <- 50
d <- data.frame(X1=runif(n),X2=runif(n))
d$Y <- 1 - 2*d$X1 - 5*d$X2 + rnorm(n, sd=1.25)
a0 <- lm(Y~1,d)
a1 <- lm(Y~X1,d)
a2 <- lm(Y~X1+X2,d)
anovarIntReg(a0,a1,a2,B=50)
anovarIntReg(a0,a1,a2,B=50,INCREMENTAL=TRUE)
```

---

**Core Functions**

Utility functions for Integrated Regression Goodness of Fit

Description

Core functions for the computation of the Integrated Regression Goodness of Fit

Usage

```r
compIntRegProc(y, xord, weig = rep(1, length(y)))
compBootSamp(obj, datLT, B = 999, LINMOD = FALSE)
plotIntRegProc(y, x, weig = rep(1, length(y)), ADD = FALSE, ...)
getModelFrame(obj)
getResiduals(obj,type)
```

Arguments

- **y**: vector, values to add to compute the Integrated Regression.
- **xord**: list of list with the index of covariate points that are less than covariate data. This tells how to cumulate according to covariates.
- **weig**: vector of weights, specifically used to fit and compute test statistics when data is selection biased.
- **obj**: An object of class lm, glm or nls.
- **datLT**: structure as xord telling how to cumulate according to covariates.
intRegGOF

B  Bootstrap resampling size.
LINMOD  When TRUE and if obj is an object of class lm Linear Model matrix fitting equations are used.

x  vector with covarates to plot
ADD  If TRUE the plot is added to existing plot.
type  Type of residual.
...  Further parameters to plot.

Details

... TODO: Each of them computes what in which way

Note

Surely they can better implemented.

Author(s)

Jorge Luis Ojeda Cabrera (<jojeda@unizar.es>).

---

intRegGOF  Integrated Regression Goodness of Fit

Description

Integrated Regression Goodness of Fit to test if a given model is suitable to represent the regression function for a given data.

Usage

intRegGOF(obj, covars = NULL, B = 499, LINMOD = FALSE)
print.intRegGOF(x,...)

Arguments

obj  An object of class lm, glm or nls.
covars  Names of continuous (numerical) variates used to compute Integrated Regression. They should be variables contained in the data frame used to compute the regression fit.
B  Bootstrap resampling size.
LINMOD  When TRUE and if obj is an object of class lm Linear Model matrix fitting equations are used.
x  An object of class intRegGOF.
...  Further parameters for print command.
Details

The Integrated Regression Goodness of Fit technique is introduce in Stute(1997). The main idea is to study the process that results from the cumulation of the residuals up to a given value of the covariates. Once this process is built, different functional over it can be considered to measure the discrepancy between the true regression function and its estimation.

The tests that implements this function is

\[ H_0 : m \in M \text{ vs } H_1 : m \notin M \]

being \( m \) the regression function, and \( M \) a given class of functions. The statistics considered are

\[ K_n = \sup_{x \in R^d} |R_n^w(x)| \]
\[ W_n^2 = \int_{R^d} R_n^w(z)^2 \, dF(z). \]

where \( R_n^w(z) \) is the cumulated residual process:

\[ R_n^w(x) = n^{-1/2} \sum_{i=1}^{n} (y_i - \hat{y}_i)I(x_i \leq x). \]

As the stochastic behaviour of this cumulated residual process is quite complex, the implementation of the technique is based on resampling techniques. In particular the chosen implementation is based on Wild Bootstrap methods.

The method also handles selection biased data by means of compensation, by means of the weights used to fit the regression function when computing the cumulated residual process.

At the moment only 'response' type of residuals are considered, jointly with wild bootstrap re-sampling technique and the result for discrete responses might no be proper.

Value

This function returns an object of class \texttt{intRegGOF}, a list which contains following objects:

- \texttt{call} \hspace{1cm} The call to the function
- \texttt{regObj} \hspace{1cm} String with the \texttt{lm}, \texttt{glm} or \texttt{nls} object whose fit is checked
- \texttt{regModel} \hspace{1cm} \texttt{lm}, \texttt{glm} or \texttt{nls} object call.
- \texttt{p.value} \hspace{1cm} \( p \)-values for \( K_n \) and \( W_n^2 \) statistics.
- \texttt{datStat} \hspace{1cm} value of \( K_n \) and \( W_n^2 \) statistics.
- \texttt{covars} \hspace{1cm} continuous (numerical) variates used to compute Integrated Regression.
- \texttt{intErr} \hspace{1cm} cumulated residual process at the values of \texttt{covars} in data.
- \texttt{xLT} \hspace{1cm} structure with the order of \texttt{covars} summation.
- \texttt{bootSamp} \hspace{1cm} Bootstrap samples for \( K_n \) and \( W_n^2 \).

Note

This method requires more testing, and careful study of the effect of factors (discrete random variables) when fitting the model.
Author(s)

Jorge Luis Ojeda Cabrera (<jojeda@unizar.es>).

References


See Also

`lm`, `glm`, `nls` and its methods `summary`, `print`, `plot`, etc...

Examples

```r
n <- 50
d <- data.frame(X1=runif(n),X2=runif(n))
d$Y <- 1 + 2*d$X1 + rnorm(n,sd=.125)
plot( d )
intRegGOF(lm(Y~X1+X2,d),B=99)
intRegGOF(a <- lm(Y~X1-1,d),B=99)
intRegGOF(a,c("X1","X2"),B=99)
intRegGOF(a,-X2+X1,B=99)
```

Description

Methods to develop model validation and visualization of Integrated Regression Goodness of Fit technique.

Usage

```r
plotAsIntRegGOF(obj, covar = 1, ADD = FALSE, ...)
pointsAsIntRegGOF(obj, covar = 1, ...)
linesAsIntRegGOF(obj, covar = 1, ...)
```
Arguments

obj  An object of class \texttt{lm, glm or nls}.

covar  Variable name, number or vector for which Int. Reg. is computed. If it is a number, it reference a covariate in the model frame, while if it is a name refer to data in data frame using in the fitting process.

ADD  If TRUE the plot is added to existing plot.

Further parameters to for plotobj command.

Details

Currently, the implementation computes the accumulated residual process against a single covariate (covar). When the value of covar is set to 0, the response is used as the variable whose residual are accumulated against.

Notice that if covar is a vector its lenght should be equal to the number of residuals.

Note

\texttt{lm} objects that does not have a data parameter set when the call is executed does not work presently when the covar parameter is different than 0.

Author(s)

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See Also

\texttt{lm, glm, nls} its associated \texttt{plot} method and \texttt{intRegGOF}.

Examples

```r
n <- 50
d <- data.frame( X1=runif(n),X2=runif(n))
d$Y <- 1 + 2*d$X1 + rnorm(n,sd=.125)
par(ask=TRUE)
plot( d )
plotAsIntRegGOF(lm(Y~X1+X2,d),covar="X1")
plotAsIntRegGOF(a <- lm(Y~X1-1,d))
plotAsIntRegGOF(a,c("X1"))
plotAsIntRegGOF(a,0)
plotAsIntRegGOF(a,fitted(a))
par(ask=FALSE)
```
Utility Functions

Utility functions for Integrated Regression Goodness of Fit

Description
Functions that are basic or/and useful for the computation of the Integrated Regression Goodness of Fit.

Usage

getLessThan(x, d)
mvCumSum(x, ord)
mvPartOrd(x1, x2)
getContVar(df, vars = NULL)
getModelCovars(obj)
getModelWeights(obj)
rWildBoot(n)

Arguments

x, d          matrix like structure.

x1, x2        vectors with the same length.

df            a data frame.

ord           list of list structure with the ordering to add data points according to a given covariates.

obj            An object of class lm, glm or nls.

vars          vector with variable names in observations data frame.

n              integer, sample size.

Details

...TODO: Each of them computes what in which way

Note

gtLessThan can be certainly better implemented.

Author(s)

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