Package: lmSubsets (via r-universe)

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Title Exact Variable-Subset Selection in Linear Regression
Description Exact and approximation algorithms for variable-subset selection in ordinary linear regression models. Either compute all submodels with the lowest residual sum of squares, or determine the single-best submodel according to a pre-determined statistical criterion. Hofmann et al. (2020) <10.18637/jss.v093.i03>.
Depends R (>= 3.4.0)
SystemRequirements C++11
Imports stats, graphics, utils
License GPL (>= 3)
<pre>URL https://github.com/marc-hofmann/lmSubsets.R</pre>
Repository https://marc-hofmann.r-universe.dev
RemoteUrl https://github.com/marc-hofmann/lmsubsets.r
RemoteRef HEAD
RemoteSha 55787ffaaa66f1c76d97c688923b65501c7b71f7
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AirPollution

Air Pollution and Mortality

Description

Data relating air pollution and mortality, frequently used for illustrations in ridge regression and related tasks.

Usage

```
data("AirPollution")
```

Format

A data frame containing 60 observations on 16 variables.

precipitation Average annual precipitation in inches.

temperature1 Average January temperature in degrees Fahrenheit.

temperature7 Average July temperature in degrees Fahrenheit.

age Percentage of 1960 SMSA population aged 65 or older.

household Average household size.

education Median school years completed by those over 22.

housing Percentage of housing units which are sound and with all facilities.

population Population per square mile in urbanized areas, 1960.

noncauc Percentage of non-Caucasian population in urbanized areas, 1960.

whitecollar Percentage employed in white collar occupations.

income Percentage of families with income < USD 3000.

hydrocarbon Relative hydrocarbon pollution potential.

nox Relative nitric oxides potential.

so2 Relative sulphur dioxide potential.

humidity Annual average percentage of relative humidity at 13:00.

mortality Total age-adjusted mortality rate per 100,000.

Source

```
http://lib.stat.cmu.edu/datasets/pollution
```

References

McDonald GC, Schwing RC (1973). Instabilities of Regression Estimates Relating Air Pollution to Mortality. *Technometrics*, **15**, 463–482.

Miller AJ (2002). Subset Selection in Regression. New York: Chapman and Hall.

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Examples

```
## load data (with logs for relative potentials)
data("AirPollution", package = "lmSubsets")
for (i in 12:14) AirPollution[[i]] <- log(AirPollution[[i]])
## fit subsets
lm_all <- lmSubsets(mortality ~ ., data = AirPollution)
plot(lm_all)
## refit best model
lm6 <- refit(lm_all, size = 6)
summary(lm6)</pre>
```

IbkTemperature

Temperature Observations and Numerical Weather Predictions for Innsbruck

Description

00UTC temperature observations and corresponding 24-hour reforecast ensemble means from the Global Ensemble Forecast System (GEFS, Hamill et al. 2013) for SYNOP station Innsbruck Airport (11120; 47.260, 11.357) from 2011-01-01 to 2015-12-31.

Usage

```
data("IbkTemperature")
```

Format

A data frame containing 1824 daily observations/forecasts for 42 variables. The first column (temp) contains temperature observations at 00UTC (coordinated universal time), columns 2–37 are 24-hour lead time GEFS reforecast ensemble means for different variables (see below). Columns 38–42 are deterministic time trend/season patterns.

```
temp Observed temperature at Innsbruck Airport (deg C).

tp Total accumulated precipitation (kg\ m^{-2}).

t2m Temperature at 2 meters (K).

u10m U-component of wind at 10 meters (m\ s^{-1}).

v10m V-component of wind at 10 meters (m\ s^{-1}).

u80m U-component of wind at 80 meters (m\ s^{-1}).

v80m U-component of wind at 80 meters (m\ s^{-1}).

cape Convective available potential energy (J\ kg^{-1}).

ci Convective inhibition (J\ kg^{-1}).

sdlwrf Surface downward long-wave radiation flux (W\ m^{-2}).

sdswrf Surface downward short-wave radiation flux (W\ m^{-2}).
```

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```
sulwrf Surface upward long-wave radiation flux (W m^{-2}).
suswrf Surface upward short-wave radiation flux (W m^{-2}).
ghf Ground heat flux (W m^{-2}).
slhnf Surface latent heat net flux (W m^{-2}).
sshnf Surface sensible heat net flux (W m^{-2}).
mslp Mean sea level pressure (Pa).
psfc Surface pressure (Pa).
pw Precipitable water (kg m^{-2}).
vsmc Volumetric soil moisture content (fraction).
sh2m Specific humidity at 2 meters (kq kq^{-1}).
tcc Total cloud cover (percent).
tcic Total column-integrated condensate (kq m^{-2}).
tsfc Skin temperature (K).
tmax2m Maximum temperature (K).
tmin2m Minimum temperature (K).
st Soil temperature (0-10 cm below surface) (K).
ulwrf Upward long-wave radiation flux (W m^{-2}).
wr Water runoff (kq m^{-2}).
we Water equivalent of accumulated snow depth (kg m^{-2}).
wp Wind mixing energy (J).
w850 Vertical velocity at 850 hPa surface (Pa s^{-1}).
t2pvu Temperature on 2 PVU surface (K).
p2pvu Pressure on 2 PVU surface (Pa).
u2pvu U-component of wind on 2 PVU surface (m s^{-1}).
v2pvu U-component of wind on 2 PVU surface (m s^{-1}).
pv Potential vorticity on 320 K isentrope (K m^2 kg^{-1} s^{-1}).
time Time in years.
sin, cos Sine and cosine component of annual harmonic pattern.
sin2, cos2 Sine and cosine component of bi-annual harmonic pattern.
```

Source

```
Observations: http://www.ogimet.com/synops.phtml.en
Reforecasts: http://www.esrl.noaa.gov/psd/forecasts/reforecast2/
```

References

Hamill TM, Bates GT, Whitaker JS, Murray DR, Fiorino M, Galarneau Jr. TJ, Zhu Y, Lapenta W (2013). NOAA's Second-Generation Global Medium-Range Ensemble Reforecast Data Set. *Bulletin of the American Meteorological Society*, **94**(10), 1553–1565. doi:10.1175/BAMSD12-00014.1

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```
## load data and omit missing values
data("IbkTemperature", package = "lmSubsets")
IbkTemperature <- na.omit(IbkTemperature)</pre>
## fit a simple climatological model for the temperature
## with a linear trend and annual/bi-annual harmonic seasonal pattern
CLIM <- lm(temp ~ time + sin + cos + sin2 + cos2,
  data = IbkTemperature)
## fit a simple MOS with 2-meter temperature forecast in addition
## to the climatological model
MOSO \leftarrow lm(temp \sim t2m + time + sin + cos + sin2 + cos2,
  data = IbkTemperature)
## graphical comparison and MOS summary
plot(temp ~ time, data = IbkTemperature, type = "1", col = "darkgray")
lines(fitted(MOS0) ~ time, data = IbkTemperature, col = "darkred")
lines(fitted(CLIM) ~ time, data = IbkTemperature, lwd = 2)
MOS0
## best subset selection of remaining variables for the MOS
## (i.e., forcing the regressors of m1 into the model)
MOS1_all <- lmSubsets(temp ~ ., data = IbkTemperature,
  include = c("t2m", "time", "sin", "cos", "sin2", "cos2"))
plot(MOS1_all)
image(MOS1_all, size = 8:20)
## -> Note that soil temperature and maximum temperature are selected
## in addition to the 2-meter temperature
## best subset selection of all variables
MOS2_all <- lmSubsets(temp ~ ., data = IbkTemperature)
plot(MOS2_all)
image(MOS2_all, size = 2:20)
## -> Note that 2-meter temperature is not selected into the best
## BIC model but soil-temperature (and maximum temperature) are used instead
## refit the best BIC subset selections
MOS1 <- refit(lmSelect(MOS1_all))</pre>
MOS2 <- refit(lmSelect(MOS2_all))
## compare BIC
BIC(CLIM, MOS0, MOS1, MOS2)
## compare RMSE
sqrt(sapply(list(CLIM, MOS0, MOS1, MOS2), deviance)/
  nrow(IbkTemperature))
## compare coefficients
cf0 <- coef(CLIM)
cf1 <- coef(MOS0)
cf2 <- coef(MOS1)
```

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```
cf3 <- coef(MOS2)
names(cf2) <- gsub("^x", "", names(coef(MOS1)))
names(cf3) <- gsub("^x", "", names(coef(MOS2)))
nam <- unique(c(names(cf0), names(cf1), names(cf2), names(cf3)))
cf <- matrix(NA, nrow = length(nam), ncol = 4,
    dimnames = list(nam, c("CLIM", "MOS0", "MOS1", "MOS2")))
cf[names(cf0), 1] <- cf0
cf[names(cf1), 2] <- cf1
cf[names(cf2), 3] <- cf2
cf[names(cf3), 4] <- cf3
print(round(cf, digits = 3), na.print = "")</pre>
```

image

Variable Selection Heatmaps

Description

Visualization of variable subsets.

Usage

```
## S3 method for class 'lmSubsets'
image(x, size = NULL, best = 1, which = NULL,
    hilite, hilite_penalty, main, sub, xlab = NULL, ylab,
    ann = par("ann"), axes = TRUE, col = c("gray40", "gray90"),
    lab = "lab", col_hilite = cbind("red", "pink"),
    lab_hilite = "lab", pad_size = 3, pad_best = 1,
    pad_which = 3, axis_pos = -4, axis_tck = -4,
    axis_lab = -10, ...)

## S3 method for class 'lmSelect'
image(x, best = NULL, which = NULL, hilite,
    hilite_penalty, main, sub = NULL, xlab = NULL, ylab,
    ann = par("ann"), axes = TRUE, col = c("gray40", "gray90"),
    lab = "lab", col_hilite = cbind("red", "pink"),
    lab_hilite = "lab", pad_best = 2, pad_which = 2,
    axis_pos = -4, axis_tck = -4, axis_lab = -10, ...)
```

Arguments

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```
hilite, hilite_penalty
Submodels to be highlighted.

col_hilite, lab_hilite
Highlighting style.

pad_size, pad_best, pad_which
Padding.

axis_pos, axis_tck, axis_lab
Position of axes, tick length, and position of labels
... Ignored.

axes
Plot axes.
ann
Annotate plot.
```

See Also

lmSubsets, lmSelect.

```
## data
data("AirPollution", package = "lmSubsets")
###################
## 1mSubsets ##
##################
lm_all \leftarrow lmSubsets(mortality \sim ., data = AirPollution, nbest = 20)
## heatmap
image(lm_all, best = 1:3)
## highlight 5 best (BIC)
image(lm_all, best = 1:3, hilite = 1:5, hilite_penalty = "BIC")
###################
## lmSelect ##
#################
## default criterion: BIC
lm_best <- lmSelect(lm_all)</pre>
## highlight 5 best (AIC)
image(lm_best, hilite = 1:5, hilite_penalty = "AIC")
## axis labels
image(lm_best, lab = c("bold(lab)", "lab"), hilite = 1,
      lab_hilite = "underline(lab)")
```

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1mSelect

Best-Subset Regression

Description

Best-subset regression for ordinary linear models.

Usage

Arguments

formula, data, subset, weights, na.action, model, x, y, contrasts, offset Standard formula interface.

intercept Include intercept.

include, exclude

Force regressors in or out.

penalty Penalty per parameter.

tolerance Approximation tolerance.

nbest Number of best subsets.

... Forwarded to lmSelect_fit.

pradius Preordering radius.

Details

The lmSelect generic provides a convenient interface for best variable-subset selection in linear regression: The nbest best – according to an information criterion of the AIC family – subset models are returned.

The information criterion is specified with the penalty parameter. Accepted values are "AIC", "BIC", or a numeric value representing the penalty per model parameter (see AIC).

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A custom selection criterion may be specified by passing an R function as the penalty argument. The expected signature is function(size, rss), where size is the number of predictors (including intercept, if any), and rss the residual sum of squares. The function must be non-decreasing in both parameters.

A low-level matrix interface is provided by lmSelect_fit.

See lmSubsets for further information.

Value

An object of class "lmSelect", i.e., a list with the following components:

nobs, nvar Number of observations, of variables.

intercept TRUE if model has intercept term; FALSE otherwise.

include, exclude

Included, excluded variables.

size Subset sizes.

tolerance Approximation tolerance.

nbest Number of best subsets.

submodel Submodel information.

subset Selected variables.

x, and y. See 1m for more information.

Further components include call, na.action, weights, offset, contrasts, xlevels, terms, mf,

References

Hofmann M, Gatu C, Kontoghiorghes EJ, Colubi A, Zeileis A (2020). lmSubsets: Exact Variable-Subset Selection in Linear Regression for R. *Journal of Statistical Software*. **93**, 1–21. doi:10.18637/jss.v093.i03.

See Also

1mSubsets, summary, methods.

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```
## Not run:
lm_all \leftarrow lmSubsets(mortality \sim ., data = AirPollution, nbest = 20)
lm_best <- lmSelect(lm_all)</pre>
## End(Not run)
## summary statistics
summary(lm_best)
## visualize
plot(lm_best)
##########################
## custom criterion ##
## the same as above, but with a custom criterion:
M <- nrow(AirPollution)</pre>
11 <- function (rss) {</pre>
  -M/2 * (log(2 * pi) - log(M) + log(rss) + 1)
aic <- function (size, rss, k = 2) {
  -2 * ll(rss) + k * (size + 1)
bic <- function (size, rss) {</pre>
  aic(size, rss, k = log(M))
lm_cust <- lmSelect(mortality ~ ., data = AirPollution,</pre>
                    penalty = bic, nbest = 20)
lm_cust
```

1mSubsets

All-Subsets Regression

Description

All-subsets regression for linear models estimated by ordinary least squares (OLS).

Usage

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Arguments

formula, data, subset, weights, na.action, model, x, y, contrasts, offset

Standard formula interface.

intercept Include intercept.

include, exclude

Force regressors in or out.

nmin, nmax Minimum and maximum number of regressors.

tolerance Approximation tolerance (vector).

nbest Number of best subsets.

... Forwarded to lmSubsets.default and lmSubsets_fit.

pradius Preordering radius.

Details

The lmSubsets generic provides various methods to conveniently specify the regressor and response variables. The standard formula interface (see lm) can be used, or the information can be extracted from an already fitted "lm" object. The regressor matrix and response variable can also be passed in directly (see Examples).

The call is forwarded to lmSubsets_fit, which provides a low-level matrix interface.

The nbest best subset models for every subset size are computed, where the "best" models are the models with the lowest residual sum of squares (RSS). The scope of the search can be limited to a range of subset sizes by setting nmin and nmax. A tolerance vector (expanded if necessary) may be specified to speed up the search, where tolerance[j] is the tolerance applied to subset models of size j.

By way of include and exclude, variables may be forced in to or out of the regression, respectively.

The extent to which variables are preordered is controlled with the pradius parameter.

A set of standard extractor functions for fitted model objects is available for objects of class "1mSubsets". See methods for more details.

The summary method can be called to obtain summary statistics.

Value

An object of class "lmSubsets", i.e., a list with the following components:

nobs, nvar Number of observations, of variables.

intercept TRUE if model has intercept term; FALSE otherwise.

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include, exclude

Included, excluded regressors.

size Subset sizes.

tolerance Approximation tolerance (vector).

nbest Number of best subsets.

submodel Submodel information.

subset Selected variables.

Further components include call, na.action, weights, offset, contrasts, xlevels, terms, mf, x, and y. See lm for more information.

References

Hofmann M, Gatu C, Kontoghiorghes EJ, Colubi A, Zeileis A (2020). lmSubsets: Exact Variable-Subset Selection in Linear Regression for R. *Journal of Statistical Software*. **93**, 1–21. doi:10.18637/jss.v093.i03.

Hofmann M, Gatu C, Kontoghiorghes EJ (2007). Efficient Algorithms for Computing the Best Subset Regression Models for Large-Scale Problems. *Computational Statistics* & *Data Analysis*, **52**, 16–29. doi:10.1016/j.csda.2007.03.017.

Gatu C, Kontoghiorghes EJ (2006). Branch-and-Bound Algorithms for Computing the Best Subset Regression Models. *Journal of Computational and Graphical Statistics*, **15**, 139–156. doi:10.1198/106186006x100290.

See Also

lmSelect, summary, methods.

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methods

Methods for 'lmSubsets' and 'lmSelect' Objects

Description

Extractor methods for "lmSubsets" and "lmSelect" objects.

Usage

```
## S3 method for class 'lmSubsets'
variable.names(object, size, best = 1, ..., na.rm = TRUE, drop = TRUE)
## S3 method for class 'lmSelect'
variable.names(object, best = 1, ..., na.rm = TRUE, drop = TRUE)
## S3 method for class 'lmSubsets'
formula(x, size, best = 1, ...)
## S3 method for class 'lmSelect'
formula(x, best, ...)
## S3 method for class 'lmSubsets'
model.frame(formula, ...)
## S3 method for class 'lmSelect'
model.frame(formula, ...)
## S3 method for class 'lmSubsets'
model.matrix(object, size, best = 1, ...)
## S3 method for class 'lmSelect'
model.matrix(object, best, ...)
## S3 method for class 'lmSubsets'
model_response(data, ...)
## S3 method for class 'lmSelect'
```

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```
model_response(data, ...)
## S3 method for class 'lmSubsets'
refit(object, size, best = 1, ...)
## S3 method for class 'lmSelect'
refit(object, best = 1, ...)
## S3 method for class 'lmSubsets'
deviance(object, size, best = 1, ..., na.rm = TRUE, drop = TRUE)
## S3 method for class 'lmSelect'
deviance(object, best = 1, ..., na.rm = TRUE, drop = TRUE)
## S3 method for class 'lmSubsets'
sigma(object, size, best = 1, ..., na.rm = TRUE, drop = TRUE)
## S3 method for class 'lmSelect'
sigma(object, best = 1, ..., na.rm = TRUE, drop = TRUE)
## S3 method for class 'lmSubsets'
logLik(object, size, best = 1, ..., na.rm = TRUE, drop = TRUE)
## S3 method for class 'lmSelect'
logLik(object, best = 1, ..., na.rm = TRUE, drop = TRUE)
## S3 method for class 'lmSubsets'
AIC(object, size, best = 1, ..., k = 2, na.rm = TRUE, drop = TRUE)
## S3 method for class 'lmSelect'
AIC(object, best = 1, ..., k = 2, na.rm = TRUE, drop = TRUE)
## S3 method for class 'lmSubsets'
BIC(object, size, best = 1, ..., na.rm = TRUE, drop = TRUE)
## S3 method for class 'lmSelect'
BIC(object, best = 1, ..., na.rm = TRUE, drop = TRUE)
## S3 method for class 'lmSubsets'
coef(object, size, best = 1, ..., na.rm = TRUE, drop = TRUE)
## S3 method for class 'lmSelect'
coef(object, best = 1, ..., na.rm = TRUE, drop = TRUE)
## S3 method for class 'lmSubsets'
vcov(object, size, best = 1, ...)
## S3 method for class 'lmSelect'
vcov(object, best = 1, ...)
## S3 method for class 'lmSubsets'
fitted(object, size, best = 1, ...)
## S3 method for class 'lmSelect'
fitted(object, best = 1, ...)
## S3 method for class 'lmSubsets'
```

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```
residuals(object, size, best = 1, ...)
## S3 method for class 'lmSelect'
residuals(object, best = 1, ...)
```

Arguments

object, formula, data, x
An object of class "lmSubsets" or "lmSelect".

size The submodel size.

best The submodel ranking.
... Forwarded arguments.

k AIC penalty.

drop Control shape of return value.

na.rm Remove missing submodels.

Details

The extractor methods work for "lmSubsets" and "lmSelect" objects, i.e., objects that have been generated using the formula interface.

If drop == FALSE, the extractor methods variable.names, deviance, sigma, logLik, AIC, BIC and coef return a data.frame object. If drop == TRUE, the return value is a logical matrix (variable.names), a numeric matrix (coef), or a numeric vector. If the drop parameter is not set explicitly when calling variable.names or coef, one-dimensional values are represented in a compact form.

If a desired extractor function is not available, refit can be called explicitly to obtain the corresponding "lm" object.

See Also

```
lmSubsets, lmSelect, refit.
```

```
## load data
data("AirPollution", package = "lmSubsets")

## fit subsets (5 best subsets per size)
lm_all <- lmSubsets(mortality ~ ., data = AirPollution, nbest = 5)

## extract information (for best submodel of size 3)
coef(lm_all, size = 3)
vcov(lm_all, size = 3)
residuals(lm_all, size = 3)
fitted(lm_all, size = 3)
model.matrix(lm_all, size = 3)

## select best (BIC) submodels
lm_best <- lmSelect(lm_all)</pre>
```

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```
## extract information
deviance(lm_best)
logLik(lm_best)
AIC(lm_best)
BIC(lm_best, best = 1:5)

## refit model
lm5 <- refit(lm_all, size = 5)
summary(lm5)
## (Note that the p-values are not valid due to model selection.)</pre>
```

refit

Refitting Models

Description

Generic function for refitting a model on various subsets or reweighted data sets.

Usage

```
refit(object, ...)
```

Arguments

object An object.
... Forwarded arguments.

Details

The refit generic is a new function for refitting a certain model object on multiple versions of a data set (and is hence different from update). Applications refit models after some kind of model selection, e.g., variable subset selection, partitioning, reweighting etc.

The generic is similar to the one provided in **modeltools** and **fxregime** (and should fulfill the same purpose). To avoid dependencies, it is also provided here.

See Also

methods

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