df.residual the residual degrees of freedom.

call the matched call.

terms the <u>terms</u> object used.

contrasts (only where relevant) the contrasts used.

xlevels (only where relevant) a record of the levels of the factors used in fitting.

offset the offset used (missing if none were used).

y if requested, the response used.
x if requested, the model matrix used.

model if requested (the default), the model frame used.

na.action (where relevant) information returned by model.frame on the special handling of NAs.

In addition, non-null fits will have components assign, effects and (unless not requested) qr relating to the linear fit, for use by extractor functions such as summary and effects.

Using time series

Considerable care is needed when using lm with time series.

Unless na.action = NULL, the time series attributes are stripped from the variables before the regression is done. (This is necessary as omitting NAs would invalidate the time series attributes, and if NAs are omitted in the middle of the series the result would no longer be a regular time series.)

Even if the time series attributes are retained, they are not used to line up series, so that the time shift of a lagged or differenced regressor would be ignored. It is good practice to prepare a data argument by ts.intersect(..., dframe = TRUE), then apply a suitable na.action to that data frame and call lm with na.action = NULL so that residuals and fitted values are time series.

Note

Offsets specified by offset will not be included in predictions by predict.lm, whereas those specified by an offset term in the formula will be.

Author(s)

The design was inspired by the S function of the same name described in Chambers (1992). The implementation of model formula by Ross Ihaka was based on Wilkinson & Rogers (1973).

References

Chambers, J. M. (1992) Linear models. Chapter 4 of Statistical Models in S eds J. M. Chambers and T. J. Hastie, Wadsworth & Brooks/Cole.

Wilkinson, G. N. and Rogers, C. E. (1973) Symbolic descriptions of factorial models for analysis of variance. Applied Statistics, 22, 392-9.

See Also

summary.lm for summaries and anova.lm for the ANOVA table; aov for a different interface.

The generic functions <u>coef</u>, <u>effects</u>, <u>residuals</u>, <u>fitted</u>, <u>vcov</u>.

predict.lm (via predict) for prediction, including confidence and prediction intervals; confidence intervals of parameters.

<u>lm.influence</u> for regression diagnostics, and <u>glm</u> for **generalized** linear models.

The underlying low level functions, lm. fit for plain, and lm. wfit for weighted regression fitting.

More lm() examples are available e.g., in ans.combe, attitude, freeny, LifeCycleSavings, longley, stackloss, <a href="swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-swiss-

biglm in package biglm for an alternative way to fit linear models to large datasets (especially those with many cases).

Examples

```
require(graphics)
```

```
## Annette Dobson (1990) "An Introduction to Generalized Linear Models". ## Page 9: Plant Weight Data.  \text{ctl} <- \text{c}(4.17,5.58,5.18,6.11,4.50,4.61,5.17,4.53,5.33,5.14) }    \text{trt} <- \text{c}(4.81,4.17,4.41,3.59,5.87,3.83,6.03,4.89,4.32,4.69) }    \text{group} <- \text{gl}(2, 10, 20, \text{labels} = \text{c}(\text{"Ctl","Trt"}))    \text{weight} <- \text{c}(\text{ctl, trt})    \text{lm.D9} <- \text{lm}(\text{weight} \sim \text{group})    \text{lm.D90} <- \text{lm}(\text{weight} \sim \text{group} - 1)   # omitting intercept  \text{anova}(\text{lm.D9})    \text{summary}(\text{lm.D90})    \text{opar} <- \text{par}(\text{mfrow} = \text{c}(2,2), \text{oma} = \text{c}(0, 0, 1.1, 0))    \text{plot}(\text{lm.D9}, \text{las} = 1)   # Residuals, Fitted, ...  \text{par}(\text{opar})   ### less simple examples in "See Also" above
```