

plotres

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1 Introduction

Residual plots are important in linear model analysis, but they also useful for non-parametric models. They can be used for an overview of the model's performance, to check for outliers, and to check if the response should be transformed.

The `plotres` function [6] can plot residuals for a wide variety of R models. Figure 1 shows an example. It was produced with the following code:

```
library(rpart); library(plotmo)           # plotres is in the plotmo package
library(earth); data(ozone1)              # get the ozone data
rpart.mod <- rpart(O3 ~ ., data = ozone1)  # generate an rpart model
plotres(rpart.mod)                        # plot its residuals
```

Since this is an `rpart` model [11], `plotres` draws a tree at the top left. Different figures will be drawn in that position for other types of model (Section 5).

The bottom left plot is a canonical Residuals Vs Fitted plot. In this example we see the quantized fits characteristic of RPART models. Each vertical line of points corresponds to a leaf of the tree. (We could use `plotres`'s `jitter` argument to reduce the overplotting caused by quantization here.) The red line is a lowess smooth.

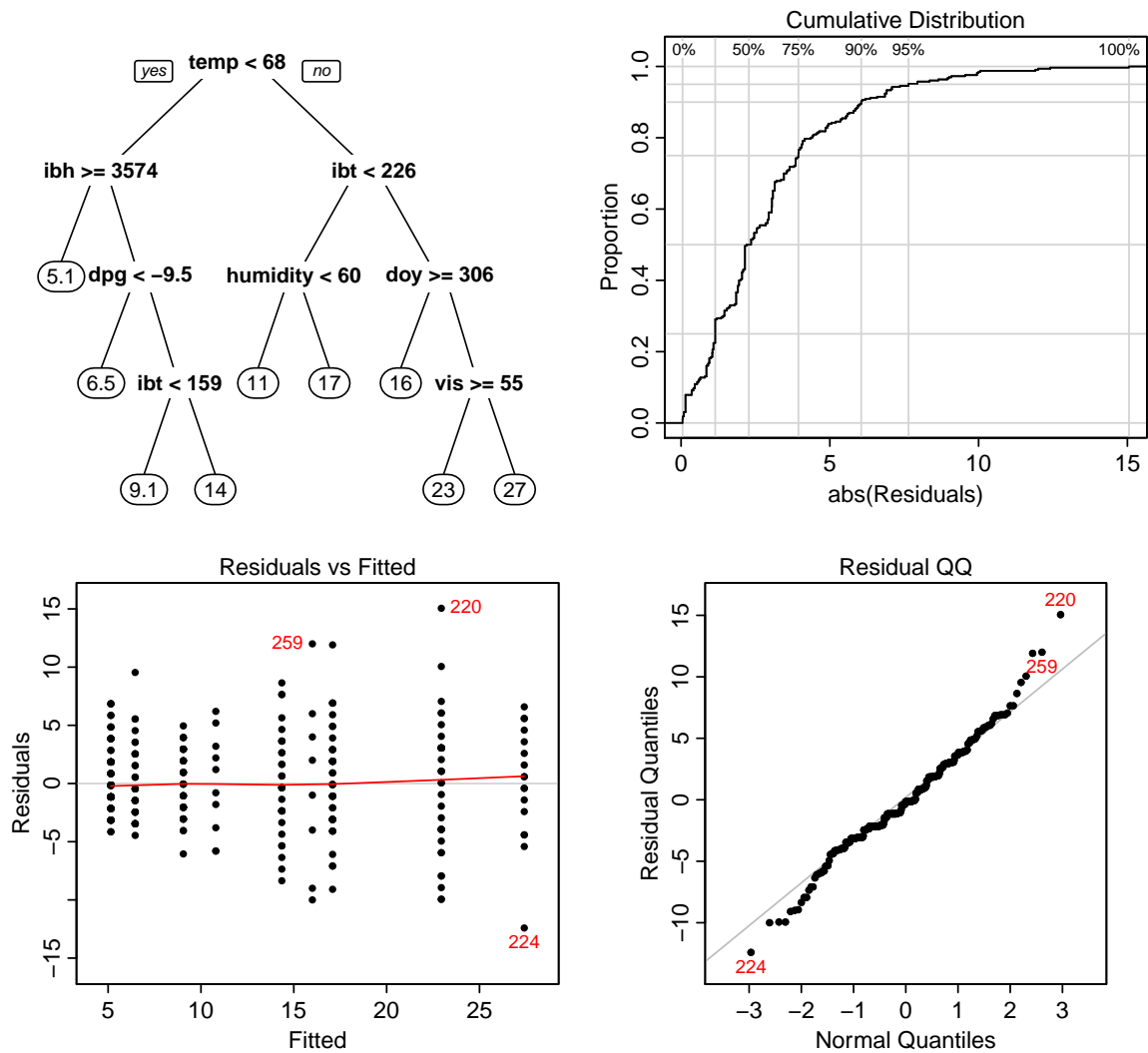


Figure 1: An example `plotres` plot. This example is for an `rpart` model.

In case 220, for example, we see that although the model estimates a high ozone level, the observed level was actually a lot higher. This residual determines the right bound of the Cumulative Distribution plot (top right). From the QQ plot (bottom right) we see that this residual is indeed abnormal.

A variety of other plots can be made, as described on the `plotres` help page.

2 Some examples

Here are some examples which illustrate `plotres` on various models. The models here are just for illustrating `plotres` and shouldn't be taken too seriously.

```
library(earth) # for ozone1 data
data(ozone1)

lm.mod <- lm(O3 ~ ., data=ozone1)           ## linear model
plotres(lm.mod)

earth.mod <- earth(O3 ~ ., data=ozone1, degree=2) ## earth
plotres(earth.mod) # equivalent to plot.earth

library(rpart)                               ## rpart
rpart.mod <- rpart(O3 ~ ., data=ozone1)
plotres(rpart.mod)

library(tree)                                ## tree
tree.mod <- tree(O3~., data=ozone1)
plotres(tree.mod)

library(randomForest)                        ## randomForest
set.seed(2015)
rf.mod <- randomForest(O3~., data=ozone1)
plotres(rf.mod)

library(gbm)                                 ## gbm
set.seed(2015)
gbm.mod <- gbm(O3~., data=ozone1, dist="gaussian",
               interact=2, shrink=.01, n.trees=1000)
plotres(gbm.mod)

library(nnet)                                ## nnet
set.seed(2015)
nnet.mod <- nnet(O3~., data=scale(ozone1), size=2, decay=0.01, trace=FALSE)
plotres(nnet.mod, type="raw")

library(neuralnet)                           ## neuralnet
set.seed(2015)
nn.mod <- neuralnet(O3~humidity+temp, data=scale(ozone1), hidden=2)
plotres(nn.mod)

library(caret)                               ## caret
set.seed(2015)
caret.earth.mod <- train(O3~., data=ozone1, method="earth",
                        tuneGrid=data.frame(degree=2, nprune=10))
plotres(caret.earth.mod, type="raw")
```

This definitely isn't an exhaustive list of models supported by `plotres`. The packages used in the above code are [2–4, 7–9, 11, 12].

3 Limitations

Plotres is designed primarily for displaying standard **response - fitted** residuals for regression models with a single continuous response. For some model types it supports multiple responses and other kinds of residuals.

In general, it won't work on models that don't save the call or data with the model in a standard way. For further discussion please see *Accessing the model data* in the *plotmo* vignette.

4 Generating the residuals

Plotres first tries to get the residuals by calling the **residuals** method for the model. If the call fails (which it will for models that don't have a **residuals** method), **plotres** must figure out the residuals manually. It does that using **predict**.

Use **trace = 1** to see the arguments passed to **predict** and friends. Plotres tries to use sensible default arguments for **predict**, but they won't always be correct (**plotres** doesn't know about every kind of model). Change the defaults if necessary using arguments with a **predict.** prefix. Plotres passes any argument prefixed with **predict.** to **predict**, after removing the prefix.

For example, **predict.gbm** has a **n.trees** argument, which **plotres** defaults to the total number of trees. But that can be changed, for example:

```
library(gbm); library(plotmo)

example(gbm)                # create gbm1, a gbm model

plotres(gbm1)               # call predict.gbm with total n.trees

nbest <- gbm.perf(gbm1, method = "OOB") # get "best" number of trees

plotres(gbm1, predict.n.trees = nbest)  # pass n.trees = nbest to predict.gbm
```

5 The which=1 plot

The top left plot is the model-specific “**which=1**” plot.¹ What gets plotted here depends on the model class. For example, for **earth** models this is a model selection plot. For **glmnet** models it's a coefficient profile plot.

Nothing will be displayed for some models. This isn't really an issue. You will see three instead of four plots when you call **plotres**.

For some models, the **which=1** plot is called with default arguments programmed into **plotres**. Use **trace = 1** to see those arguments. Change the arguments with **plotres**

¹This plot is included by default, since by default **which = 1:4**.

arguments with a `w1.` prefix. `Plotres` passes any argument prefixed with `w1.` to the `which=1` plot, after removing the prefix. See the example below.

It may happen that the `which=1` plot is partially plotted or messes up the page for further plots. In that case, call `plotres` with a `which` argument that excludes 1 (don't use the default `which`). We would be interested in hearing about this kind of bad behaviour.

An example

Here's a `which=1` plot showing `glmnet` [1] using the lasso to tame the highly collinear `longley` data (Figure 2).

```
library(glmnet); library(plotmo); data(longley)

mod <- glmnet(data.matrix(longley[,1:6]), longley[,7])

plotres(mod)                                     # left side of the figure
```

Change the parameters passed to the plot using the `w1.` prefix:

```
plotres(mod,                                     # right side of the figure
  w1.xvar="norm",                                # pass xvar="norm" to the plot
  w1.col=c("black", "red", "sienna"),           # change the color scheme
  grid.col="gray")                              # add a grid to all plots
```

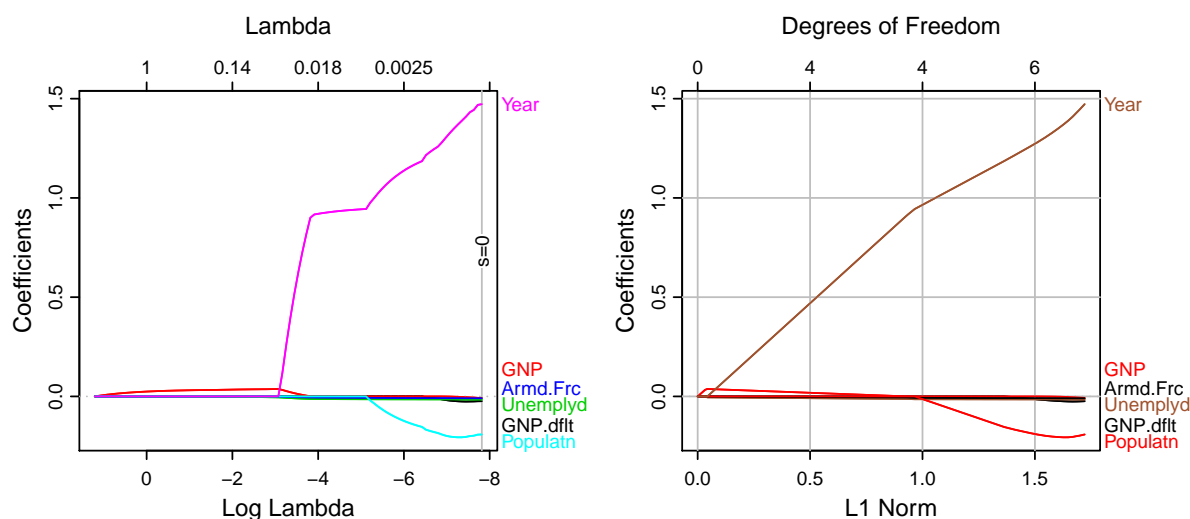


Figure 2: `glmnet` on the `longley` data [5]. Just the `which=1` plot is shown.

Left: `plotres(glmnet.mod)`

Right: `plotres(glmnet.mod, w1.xvar="norm", w1.col=newcolors, grid.col="gray")`

Notes on glmnet

Plotres uses an internal version of `plot.glmnet`. This is like the `plot.glmnet` described in the `glmnet` help pages, but it has an extra plot type "rlambda", which `plotres` uses by default (left side of Figure 2).

The plot is annotated with the `s` parameter, the penalty `lambda` passed to `predict` when generating the residuals in the other plots (but those plots aren't shown here).

By default `plotres` passes `s = 0` to `predict`, no penalty. Change that using something like `predict.s = 0.002` to see how the residuals change for models at different points on the coefficient plot.

For multiple response models, use `plotres`'s `nresponse` argument to select which response is plotted (`plot.glmnet`'s `type.coef` argument isn't supported).

By default, the names of the twenty variables with the largest final coefficients are printed on the right (using `spread.labs` from the `TeachingDemos` package [10]). Use `w1.label=TRUE` to print all names; use `w1.label=FALSE` to remove the names.

The usual plot arguments like `col` and `lty` can be passed using a prefix, for example `w1.lty=c(1,2,3)`. Use `w1.s.col=0` to remove the `s` vertical line.

Notes on gbm

Plotres uses an internal function for plotting `gbm` models. Figure 3 shows an example, generated with the following code:

```
library(earth); data(ozone1) # get the ozone data
gbm.mod <- gbm(O3~., data=ozone1, dist="gaussian", interact=2,
               shrink=.01, train.frac=.8, cv.folds=10, n.trees=1000)
plotres(gbm.mod)
```

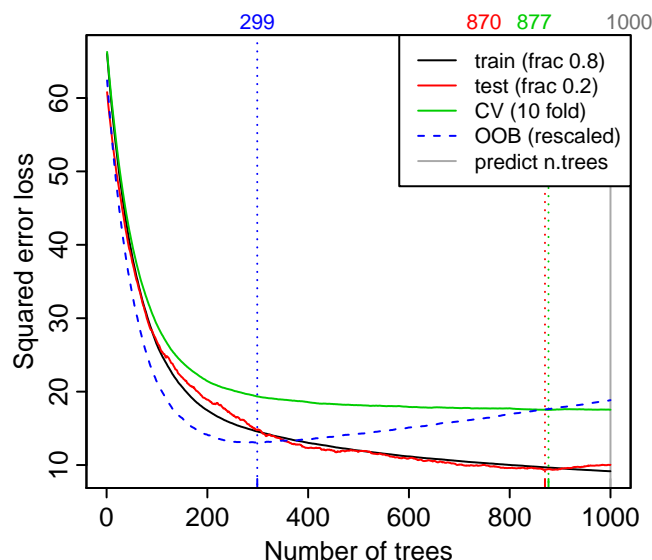


Figure 3: `gbm` on the `ozone1` data. Just the `which=1` plot is shown.

The gray vertical line on the right shows that all 1000 trees were used when generating the residuals (Section 4), although the residual plots aren't shown here.

The vertical dotted lines and the corresponding colored numbers along the top of the plot show the number of trees selected by various criteria.

The OOB line is the out-of-bag improvement, similar to the `gbm.perf` plot. It is on a different scale to the other lines, so shouldn't really be shown on the same plot. We force it in by rescaling and shifting it. Although helpful for comparing model-selection criteria, this can also lead to confusion: the scale on the left doesn't apply to the OOB line. A dashed line is used as a reminder.

Pass arguments to the plot using the `w1.` prefix. For more resolution on the vertical axis use something like `w1.ylim=c(10,30)`. Use `w1.n.trees=NA` to remove the vertical gray line.

6 Comparison to `plot.lm`

The function `plot.lm` automatically standardizes residuals for some of the plots. In contrast, with `plotres` we must explicitly specify when the residuals should be standardized. This is because standardization isn't possible or appropriate for many of the non-parametric models that `plotres` is designed for.

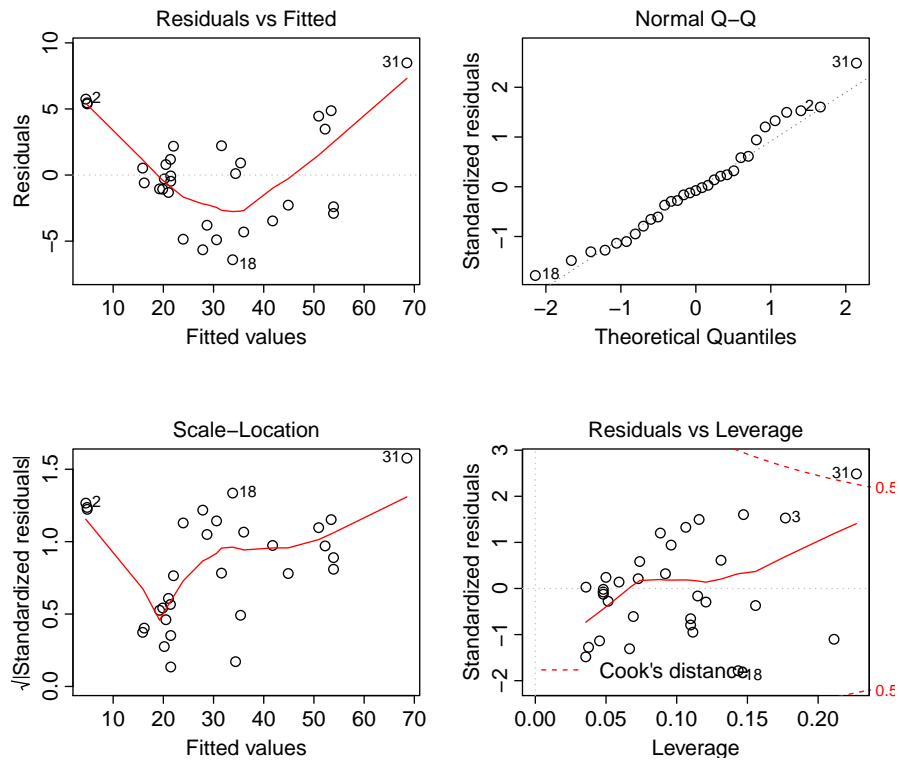
The different `which` numbering scheme used by `plotres` is mostly an historical legacy.

As a somewhat academic exercise, here's how `plot.lm` can be emulated with `plotres` (Figure 6):

```
plotlm <- function(object) # similar to plot.lm
{
  # residuals vs fitted
  plotres(object, which=3, center=FALSE,
    caption=paste(deparse(object$call), collapse=" "))
  # QQ plot
  plotres(object, which=4, standardize=TRUE)
  # scale-location plot
  plotres(object, which=6, standardize=TRUE, main="Scale-Location")
  # leverage plot
  plotres(object, which=3, versus=4, standardize=TRUE)
}

fit <- lm(Volume ~ ., data = trees) # simple linear model
par(mfrow = c(2,2), oma = c(0,0,3,0)) # four plots on page, space for caption
plot(fit) # call plot.lm
plotlm(fit) # call our version of plot.lm
```

plot.lm



plot.lm emulated with plotres

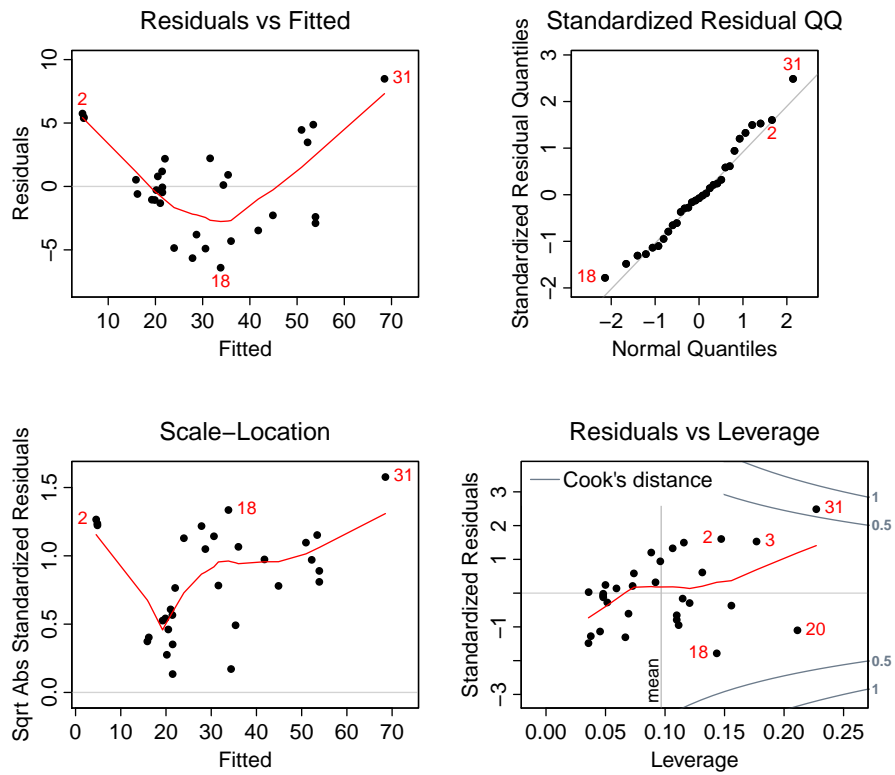


Figure 4: `plot.lm` and `plotres`.

7 FAQ

Please see the FAQ and *Common error messages* in the `plotmo` vignette.

Also see the *Notes on some packages* in that vignette.

Why is nothing displayed for `which=1` ?

See Section 5. For some model classes, nothing will displayed for `which=1`. Plotres will plot three instead of four plots.

References

- [1] Jerome Friedman, Trevor Hastie, and Robert Tibshirani. *Regularization Paths for Generalized Linear Models via Coordinate Descent*. JASS, 2010. Cited on page 5.
- [2] Stefan Fritsch and Frauke Guenther; following earlier work by Marc Suling. *neuralnet: Training of neural networks*, 2012. R package. Cited on page 3.
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- [4] Andy Liaw, Mathew Weiner; Fortran original by Leo Breiman, and Adele Cutler. *randomForest: Breiman and Cutler’s random forests for regression and classification*, 2014. R package. Cited on page 3.
- [5] J. W. Longley. *An appraisal of least-squares programs from the point of view of the user*. JASS, 1967. Cited on page 5.
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- [10] Greg Snow. *TeachingDemos: Demonstrations for teaching and learning*, 2013. R package version. Cited on page 6.
- [11] Terry Therneau and Beth Atkinson. *rpart: Recursive Partitioning and Regression Trees*, 2014. R package. Cited on pages 1 and 3.
- [12] W. N. Venables and B. D. Ripley. *nnet: Feed-forward Neural Networks and Multinomial Log-Linear Models*, 2014. R package. Cited on page 3.