

# Lab3

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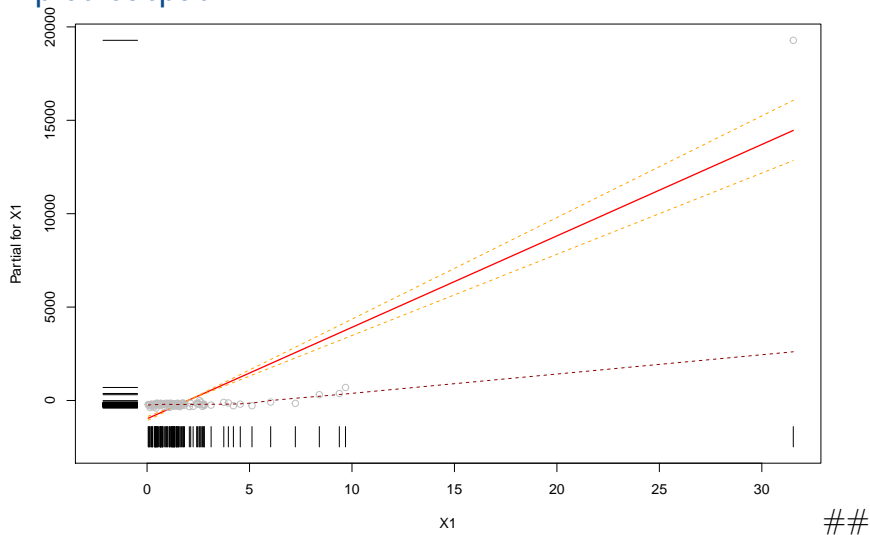
# Topics

- ▶ wercker
- ▶ termplot
- ▶ residuals in GLMS
- ▶ more on interactions

- ▶ Check your builds

## Termplot Example

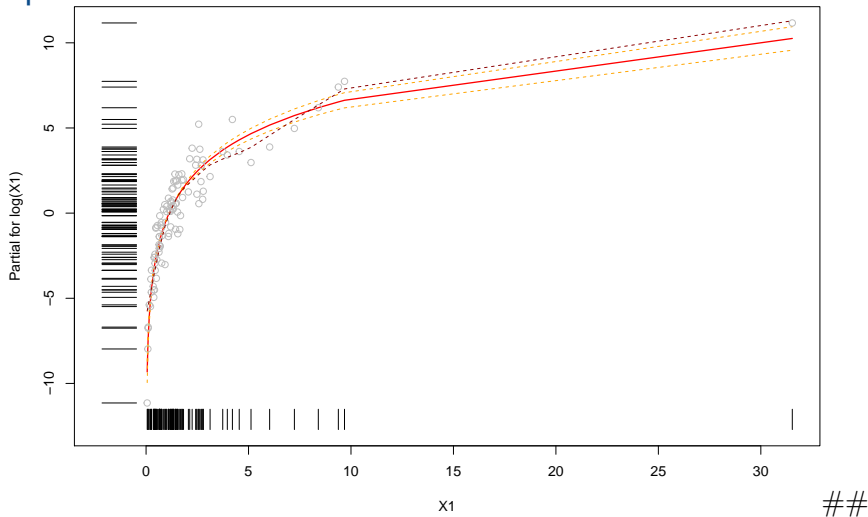
## Termplot output



## Termplot output



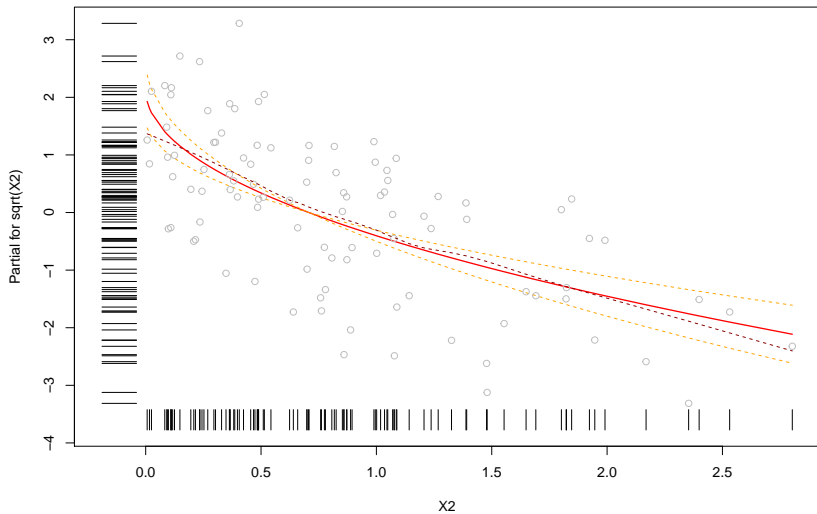
## termplot with transformation of Y and X1



## termplot with transformation of Y and X1



## termplot with transformation of Y, X1, and X2



## What is in a term plot?

- ▶ x-axis is the (untransformed) variable in your dataframe ( $X_1, X_2$ )
- ▶ line is the “term” of that variable’s contribution to  $f(x)$
- ▶ y-axis is partial residuals for term
- ▶ `partial.resid = T` adds the partial residuals to the plot
- ▶ `rug = T` shows location of data on axes
- ▶ `se = T` adds the SE of the term’s contribution to  $f(x)$
- ▶ `smooth = panel.smooth` adds “smoothed” means to plot



## Terms

$$Y = \hat{\beta}_0 + \hat{\beta}_1 X_1 + \hat{\beta}_2 X_2 + e$$

Equivalent to centered model

$$Y = \bar{Y} + \hat{\beta}_1 (X_1 - \bar{X}_1) + \hat{\beta}_2 (X_2 - \bar{X}_2) + e$$

Terms are coefficient estimates times centered predictors

$$\hat{\beta}_1 (X_1 - \bar{X}_1)$$

$$\hat{\beta}_2 (X_2 - \bar{X}_2)$$

## Terms with transformations

$$\log(Y) = \hat{\beta}_0 + \hat{\beta}_1 \log(X_1) + \hat{\beta}_2 X_2 + e$$

Equivalent to centered model

$$\log(Y) = \log(\bar{Y}) + \hat{\beta}_1(\log(X_1) - \log(\bar{X}_1)) + \hat{\beta}_2(X_2 - \bar{X}_2) + e$$

Terms are coefficient estimates times centered “predictors”

$$\hat{\beta}_1(\log(X_1) - \log(\bar{X}_1))$$

## partial residuals for a term

$$\log(Y) = \log(\bar{Y}) + \hat{\beta}_1(\log(X1) - \log(\bar{X1})) + \hat{\beta}_2(X2 - \bar{X2}) + e$$

$$\log(Y) - (\log(\bar{Y}) + \hat{\beta}_1(\log(X1) - \log(\bar{X1}))) = \hat{\beta}_2(X2 - \bar{X2}) + e$$

- ▶ Lefthand side takes response and removes the part of the response that is explained by  $X1$
- ▶ Equal to the term for  $X2$  plus the residual  $e$
- ▶ part of residual variation that is not explained by the other terms that potentially can be explained by  $X2 =$  partial residual for  $X2$
- ▶ partial residual for  $X1$

$$\hat{\beta}_1(\log(X1) - \log(\bar{X1})) + e$$