

# Assignment 6

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2. If you're feeling more adventurous, try the code from Chapter 8 in Kabacoff's *R In Action*, which offers further diagnostics and more exploration related to linear regression, such as doing cross-validation or trying to evaluate the relative importance of predictors. Not all of it will work; in particular, I think some of the functions in the "car" package are deprecated. Try FOUR from the following:

## Listing 8.6

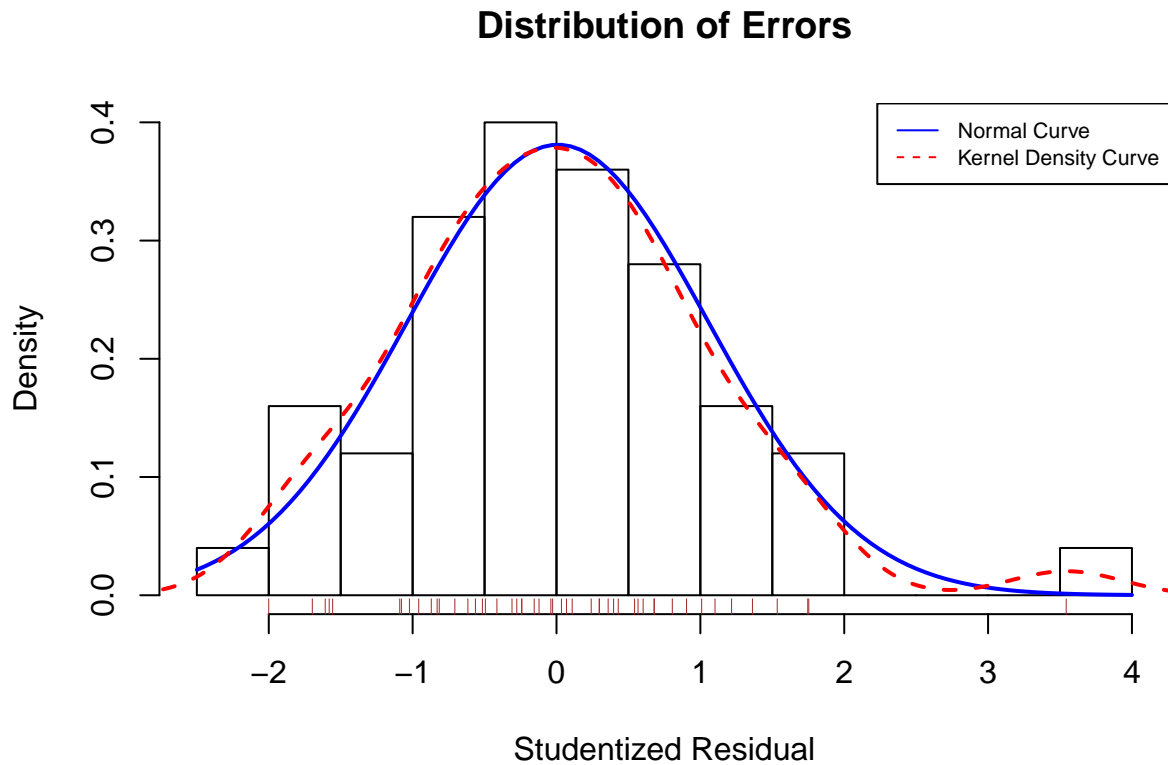
Function for plotting studentized residuals

```
states <- as.data.frame(state.x77[,c("Murder","Population","Illiteracy","Income",
                                     "Frost")])
fit <- lm(Murder ~ Population + Illiteracy + Income + Frost, data = states )

residplot <- function(fit,nbreaks =10) {

  z <-rstudent(fit)
  hist(z,breaks=nbreaks,freq= FALSE, xlab ="Studentized Residual",
  main = "Distribution of Errors")
  rug(jitter(z),col ="brown")
  curve(dnorm(x,mean =mean(z),sd =sd(z)),
  add = TRUE, col ="blue" ,lwd = 2)
  lines(density(z)$x,density(z)$y,
  col ="red",lwd = 2, lty = 2)
  legend("topright",
  legend = c("Normal Curve", "Kernel Density Curve"),
  lty = 1:2, col =c("blue","red"), cex =.7)
}

residplot(fit)
```



#### Listing 8.8 ####Global Test of Linear Model Assumptions

```
library(gvlma)
gvmodel <- gvlma(fit)

summary(gvmodel)
```

```
##
## Call:
## lm(formula = Murder ~ Population + Illiteracy + Income + Frost,
##     data = states)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.7960 -1.6495 -0.0811  1.4815  7.6210
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  1.235e+00  3.866e+00   0.319   0.7510
## Population    2.237e-04  9.052e-05   2.471   0.0173 *
## Illiteracy    4.143e+00  8.744e-01   4.738  2.19e-05 ***
## Income        6.442e-05  6.837e-04   0.094   0.9253
## Frost         5.813e-04  1.005e-02   0.058   0.9541
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
```

```
## Residual standard error: 2.535 on 45 degrees of freedom
## Multiple R-squared:  0.567, Adjusted R-squared:  0.5285
## F-statistic: 14.73 on 4 and 45 DF,  p-value: 9.133e-08
##
##
## ASSESSMENT OF THE LINEAR MODEL ASSUMPTIONS
## USING THE GLOBAL TEST ON 4 DEGREES-OF-FREEDOM:
## Level of Significance =  0.05
##
## Call:
## gvlma(x = fit)
##
##              Value p-value              Decision
## Global Stat      2.7728  0.5965 Assumptions acceptable.
## Skewness         1.5374  0.2150 Assumptions acceptable.
## Kurtosis         0.6376  0.4246 Assumptions acceptable.
## Link Function    0.1154  0.7341 Assumptions acceptable.
## Heteroscedasticity 0.4824  0.4873 Assumptions acceptable.
```

`gvlma(x=fit)`

```
##
## Call:
## lm(formula = Murder ~ Population + Illiteracy + Income + Frost,
##     data = states)
##
## Coefficients:
## (Intercept)  Population  Illiteracy      Income      Frost
##  1.235e+00   2.237e-04   4.143e+00   6.442e-05   5.813e-04
##
##
## ASSESSMENT OF THE LINEAR MODEL ASSUMPTIONS
## USING THE GLOBAL TEST ON 4 DEGREES-OF-FREEDOM:
## Level of Significance =  0.05
##
## Call:
## gvlma(x = fit)
##
##              Value p-value              Decision
## Global Stat      2.7728  0.5965 Assumptions acceptable.
## Skewness         1.5374  0.2150 Assumptions acceptable.
## Kurtosis         0.6376  0.4246 Assumptions acceptable.
## Link Function    0.1154  0.7341 Assumptions acceptable.
## Heteroscedasticity 0.4824  0.4873 Assumptions acceptable.
```

## Listing 8.11

### Comparing nested models using anova()

```
states <- as.data.frame(state.x77[,c("Murder", "Population", "Illiteracy", "Income",
                                     "Frost")])
fit1 <- lm(Murder ~ Population + Illiteracy + Income + Frost, data = states )
```

```

fit2 <- lm(Murder ~ Population + Illiteracy, data = states )

anova(fit2,fit1)

## Analysis of Variance Table
##
## Model 1: Murder ~ Population + Illiteracy
## Model 2: Murder ~ Population + Illiteracy + Income + Frost
##   Res.Df    RSS Df Sum of Sq    F Pr(>F)
## 1      47 289.25
## 2      45 289.17  2  0.078505 0.0061 0.9939

```

## Listing 8.12

### Comparing models with AIC

```

states <- as.data.frame(state.x77[,c("Murder","Population","Illiteracy","Income",
                                     "Frost")])
fit1 <- lm(Murder ~ Population + Illiteracy + Income + Frost, data = states )
fit2 <- lm(Murder ~ Population + Illiteracy, data = states )

AIC(fit2,fit1)

##      df      AIC
## fit2  4 237.6565
## fit1  6 241.6429

```