Chapter 5 - OLS Asymptotics

Thalles Quinaglia Liduares

24/03/2022

Exercise 5.1

Upload packages

```
library(wooldridge)
library(lmreg)
library(lmtest)
```

Upload database

```
data<-wooldridge::wage1
attach(data)</pre>
```

Use the data in WAGE1.RAW for this exercise.

(i) Estimate the equation

```
wage = \beta_0 + \beta_1 educ + \beta_2 exper + \beta_3 tenure + u
```

Save the residuals and plot a histogram.

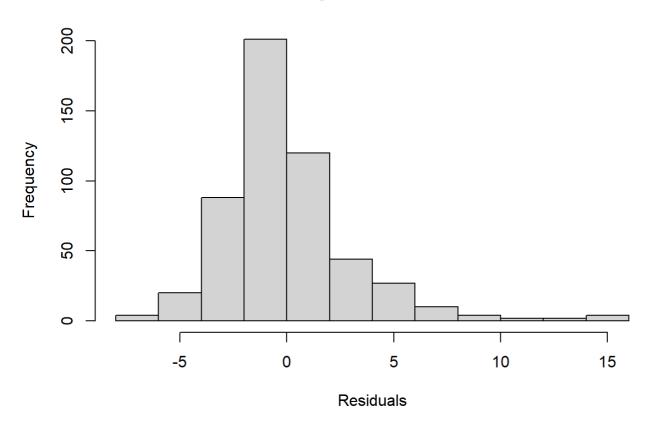
```
lm1<-lm(wage~educ+exper+tenure)
summary(lm1)</pre>
```

```
##
## lm(formula = wage ~ educ + exper + tenure)
##
## Residuals:
##
     Min
             1Q Median
                          3Q
                               Max
## -7.6068 -1.7747 -0.6279 1.1969 14.6536
##
## Coefficients:
            Estimate Std. Error t value Pr(>|t|)
##
## educ
           0.02234 0.01206 1.853 0.0645 .
## exper
           0.16927
                     0.02164 7.820 2.93e-14 ***
## tenure
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.084 on 522 degrees of freedom
## Multiple R-squared: 0.3064, Adjusted R-squared: 0.3024
## F-statistic: 76.87 on 3 and 522 DF, p-value: < 2.2e-16
```

Histogram

```
hist(lm1$residuals,
main="Histogram of residuals",
xlab="Residuals")
```

Histogram of residuals



Normality of residuals test

```
Box.test(lm1$residuals, type="Ljung-Box") #Lm1
```

```
##
## Box-Ljung test
##
## data: lm1$residuals
## X-squared = 5.6728, df = 1, p-value = 0.01723
```

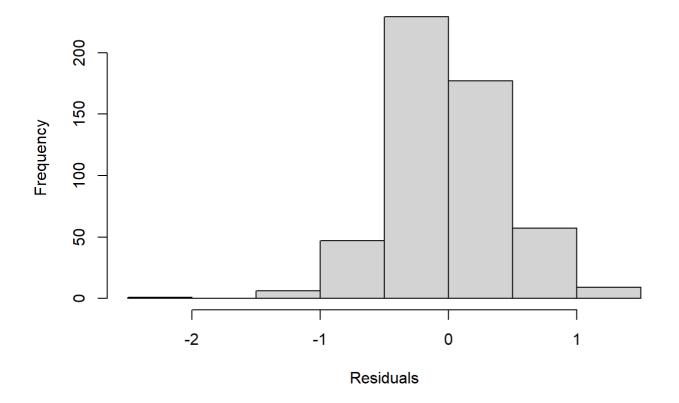
(ii) Repeat part (i), but with log(wage) as the dependent variable.

```
lm2<-lm(lwage~educ+exper+tenure)
summary(lm2)</pre>
```

```
##
## Call:
## lm(formula = lwage ~ educ + exper + tenure)
##
## Residuals:
##
       Min
                 1Q
                      Median
                                  3Q
                                          Max
## -2.05802 -0.29645 -0.03265 0.28788 1.42809
##
## Coefficients:
##
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 0.284360 0.104190
                                  2.729 0.00656 **
              0.092029
## educ
                        0.007330 12.555 < 2e-16 ***
## exper
              0.004121 0.001723 2.391 0.01714 *
## tenure
              0.022067 0.003094
                                  7.133 3.29e-12 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.4409 on 522 degrees of freedom
## Multiple R-squared: 0.316, Adjusted R-squared: 0.3121
## F-statistic: 80.39 on 3 and 522 DF, p-value: < 2.2e-16
```

```
hist(lm2$residuals,
main="Histogram of residuals",
xlab="Residuals")
```

Histogram of residuals



Box.test(lm2\$residuals, type="Ljung-Box") #Lm2

```
##
## Box-Ljung test
##
## data: lm2$residuals
## X-squared = 6.9406, df = 1, p-value = 0.008426
```

(iii) Would you say that Assumption MLR.6 is closer to being satisfied for the level-level model or the log-level model?

The first case is more close of a normal distribution of residuals.