## Homework 2

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## 2/15/2021

Calculate the residuals. Use the summary() function applied to your numeric vector of residuals to verify that this matches the Residuals: quantiles reported from the summary of your fitted model. Plot a histogram of your residuals. Do they appear reasonably Gaussian?

```
setwd("F:/Quant Eco")
getwd()
## [1] "F:/Quant Eco"
beach <- read.csv(file='beach.csv')</pre>
fit_i <- lm(OpenBeach ~ BeachID * Year, data = beach)</pre>
fit_i
##
## Call:
  lm(formula = OpenBeach ~ BeachID * Year, data = beach)
##
##
  Coefficients:
     (Intercept)
##
                         BeachIDB
                                         BeachIDC
                                                                    BeachIDB: Year
                                                              Year
##
        36945.73
                        -34211.98
                                        -28776.11
                                                            -18.36
                                                                             17.04
## BeachIDC:Year
##
           14.37
   <- residuals.lm(fit_i)
res
                                                                                      6
##
                             2
                                           3
                                                                        5
               1
    74.98497092
                  79.14988575
                                 7.01480058
                                             -20.32028459
                                                              6.24463025 -50.82554009
##
##
                                           9
                                                                       11
                                                                                     12
                                                         10
                 -35.53079560
                               -20.76588076
                                             -71.30096593
                                                             -5.27113627 -18.00622144
##
   -10.86062526
##
              13
                            14
                                          15
                                                         16
                                                                       17
                                                                                     18
   -66.84130661
                 -29.17639177
                               -12.11147694
                                                             22.41835272
                                                                           31.48326755
##
                                                0.15343789
##
              19
                            20
                                                        22
                                                                       23
                                                                                     24
                                          21
##
    57.24818238
                  62.31309722
                                -8.23403540
                                               23.18810936
                                                             15.91025413
                                                                          -38.76760111
              25
                            26
                                                         28
                                                                       29
##
                                          27
                                                                                     30
##
    -9.94545634
                   4.39883319
                                -3.47902205
                                               12.46526748
                                                              3.78741224
                                                                          -29.69044299
##
              31
                            32
                                          33
                                                        34
                                                                       35
                                                                                     36
##
    -0.06829823
                   0.75384653
                                17.47599130
                                               11.09813606
                                                             11.82028083
                                                                           12.34242559
##
              37
                            38
                                          39
                                                         40
                                                                       41
                                                                                     42
##
     6.16457035
                  17.78671512
                               -21.69114012
                                                1.63100465
                                                            -26.94685059
                                                                           -8.57438940
##
              43
                            44
                                                        46
                                                                                     48
    70.81920301
                  32.51279541 -33.59361218
                                               -2.80001978 -49.91283496
##
                                                                           -7.31924256
              49
                            50
                                                                                     54
##
                                          51
                                                                       53
   -32.73205775 -18.23846534 -40.74487294 -53.45128053 -11.95768812
                                                                           76.13590428
##
```

58

59

60

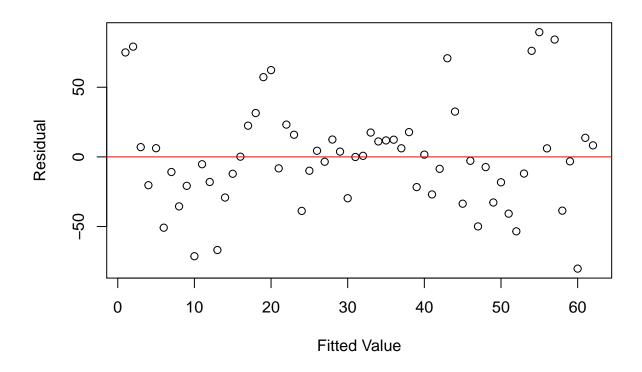
57

##

55

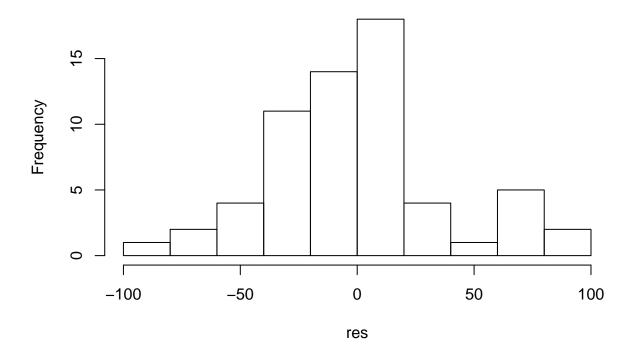
56

```
## 89.52949669
                 6.12308909 84.21668150 -38.58972610 -3.19613369 -80.20254128
##
            61
                         62
## 13.69105112
                 8.28464353
summary(res)
     Min. 1st Qu. Median
                            Mean 3rd Qu.
                                            Max.
## -80.203 -21.460 -1.434 0.000 13.385 89.529
summary(fit_i)
##
## Call:
## lm(formula = OpenBeach ~ BeachID * Year, data = beach)
## Residuals:
      Min
               1Q Median
                              ЗQ
                                     Max
## -80.203 -21.460 -1.434 13.385 89.529
##
## Coefficients:
                 Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                 36945.725 2547.493 14.503 < 2e-16 ***
                            3602.404 -9.497 2.87e-13 ***
## BeachIDB
                -34211.979
## BeachIDC
                -28776.115
                            3602.404 -7.988 8.09e-11 ***
## Year
                   -18.365
                               1.272 -14.443 < 2e-16 ***
## BeachIDB:Year
                   17.043
                               1.798
                                      9.478 3.07e-13 ***
## BeachIDC:Year
                   14.371
                               1.798
                                     7.992 7.96e-11 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 39.46 on 56 degrees of freedom
## Multiple R-squared: 0.8298, Adjusted R-squared: 0.8146
## F-statistic: 54.62 on 5 and 56 DF, p-value: < 2.2e-16
plot(res, xlab= 'Fitted Value', ylab= 'Residual')
abline(h=0, col='red')
```



hist(res)

## Histogram of res



The numeric vector of residuals matche the Residuals: quantiles reported from the summary of the fitted model. The residuals appear reasonably Gaussian.

Calculate test statistics for your regression coefficients. Verify by comparing to test statistics reported from model output.

```
summary(fit_i)$coefficients
                      Estimate Std. Error
##
                                                t value
                                                             Pr(>|t|)
## (Intercept)
                   36945.72537 2547.492995
                                              14.502778 1.296196e-20
## BeachIDB
                  -34211.97897 3602.403682
                                              -9.496986 2.868075e-13
## BeachIDC
                  -28776.11491 3602.403682
                                              -7.988032 8.090204e-11
## Year
                     -18.36491
                                   1.271577 -14.442627 1.558806e-20
## BeachIDB:Year
                      17.04277
                                   1.798122
                                               9.478096 3.074370e-13
## BeachIDC:Year
                      14.37132
                                   1.798122
                                               7.992408 7.957165e-11
b_1 <- coef(fit_i)[2]
s_b <- summary(fit_i)[['coefficients']]['BeachIDB', 'Std. Error']</pre>
B_1 <- 0
t1 \leftarrow (b_1 - B_1) / s_b
t1
##
  BeachIDB
## -9.496986
b_2 <- coef(fit_i)[3]</pre>
s_b1 <- summary(fit_i)[['coefficients']]['BeachIDC', 'Std. Error']</pre>
B 2 <- 0
t2 \leftarrow (b_2 - B_2) / s_b1
```

```
t2
## BeachIDC
## -7.988032
b_3 <- coef(fit_i)[4]</pre>
s_b2 <- summary(fit_i)[['coefficients']]['Year', 'Std. Error']</pre>
B_3 <- 0
t3 \leftarrow (b_3 - B_3) / s_b2
##
        Year
## -14.44263
b_4 <- coef(fit_i)[5]
s_b3 <- summary(fit_i)[['coefficients']]['BeachIDB:Year', 'Std. Error']</pre>
B_4 < -0
t4 \leftarrow (b_4 - B_4) / s_b3
t4
## BeachIDB:Year
##
        9.478096
b_5 <- coef(fit_i)[6]
s_b4 <- summary(fit_i)[['coefficients']]['BeachIDC:Year', 'Std. Error']</pre>
B_5 <- 0
t5 \leftarrow (b_5 - B_5) / s_b4
## BeachIDC:Year
##
        7.992408
Calculate p-values for your regression coefficients. Verify by comparing to p-values reported from modeloutput.
What are the associated null hypotheses? Do you reject or fail to reject these null hypotheses?
p1 < -pt(-1*abs(t1), df=56) + (1 - pt(abs(t1), df = 56))
p1
##
       BeachIDB
## 2.868446e-13
p2 < -pt(-1*abs(t2), df=56) + (1 - pt(abs(t2), df = 56))
p2
##
       BeachIDC
## 8.090199e-11
p3 < -2*pt(-1*abs(t3), df=56) + (1 - pt(abs(t3), df = 56))
рЗ
##
            Year
## 1.558806e-20
p4 < -pt(-1*abs(t4), df=56) + (1 - pt(abs(t4), df = 56))
р4
## BeachIDB:Year
## 3.073734e-13
```

```
p5<-pt(-1*abs(t5), df=56) + (1 - pt(abs(t5), df = 56))
p5
```

```
## BeachIDC:Year
## 7.957167e-11
```

The null hypotheses associated with these p-values is that value of the slope coefficient (B\_1,B\_2,B\_3,B\_4,B\_5) is equal to 0. Each p-value is less than 0.05 so we reject the null hypotheses.

Select a single regression coefficient (your choice) and devise a null hypothesis that is different from the default in lm(). Report the test statistics, your p-value, and whether you reject or fail to reject your null hypothesis.

```
H0:B_1=5 HA:B_1 > 5

ts_new <- (coef(fit_i)[2] - 5) /
summary(fit_i)[['coefficients']]['BeachIDB', 'Std. Error']
ts_new

## BeachIDB
## -9.498374

pnew<-pt(-1*abs(ts_new), df=56) + (1 - pt(abs(ts_new), df = 56))
pnew

## BeachIDB</pre>
```

The test statistic is -9.498374. The p-value is 2.853376e-13. We reject the null hypothesis.

Interpret output of your fitted model. Tell me how beach area does (or does not change) through time at each of the 3 beaches.

```
summary(fit_i)
```

## 2.853376e-13

```
##
## Call:
## lm(formula = OpenBeach ~ BeachID * Year, data = beach)
##
## Residuals:
##
      Min
                1Q Median
                                3Q
                                       Max
  -80.203 -21.460
                   -1.434
                           13.385
                                    89.529
##
## Coefficients:
##
                   Estimate Std. Error t value Pr(>|t|)
                              2547.493 14.503 < 2e-16 ***
## (Intercept)
                  36945.725
## BeachIDB
                 -34211.979
                              3602.404
                                       -9.497 2.87e-13 ***
## BeachIDC
                 -28776.115
                              3602.404
                                       -7.988 8.09e-11 ***
## Year
                    -18.365
                                 1.272 -14.443 < 2e-16 ***
## BeachIDB:Year
                     17.043
                                 1.798
                                         9.478 3.07e-13 ***
## BeachIDC:Year
                     14.371
                                 1.798
                                         7.992 7.96e-11 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 39.46 on 56 degrees of freedom
## Multiple R-squared: 0.8298, Adjusted R-squared: 0.8146
## F-statistic: 54.62 on 5 and 56 DF, p-value: < 2.2e-16
```

The F-statistic p-value is < 2.2e-16, which is highly significant. This means that, at least, one of the predictor variables is significantly related to the outcome variable. The predictor variables that are significant, according to the coefficients table, have t-statitic p-values less than 0.05 or have a \*\*\*. The beach area at beach A decreases by 18.36491 hectares per year. The beach area at beach B decreases by 1.32214 hectares per year. The beach area at beach C decreases by 3.99359 hectares per year.