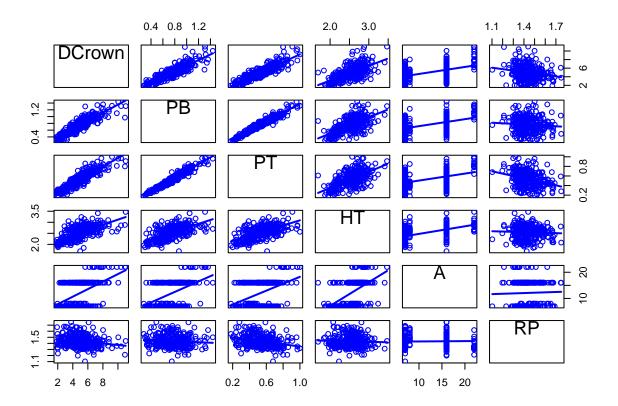
Linear Models, Problem 4

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```
data <- read.csv2("dcrown.csv")</pre>
data$RP <- data$PB/data$PT # Add the given ratio to the data.
head(data)
     DCrown
              PΒ
                   PT
                        HT A
                                    RP
#> 1
       5.19 0.92 0.62 2.70 16 1.483871
       7.03 1.00 0.72 2.54 7 1.388889
       3.51 0.52 0.36 2.09 7 1.444444
       5.25 0.72 0.54 2.88 16 1.333333
       5.33 0.93 0.67 2.90 16 1.388060
       5.46 0.90 0.65 2.72 16 1.384615
dim(data)
#> [1] 311
summary(data)
                           PΒ
                                           PT
                                                             HT
#>
        DCrown
#>
   Min.
          : 1.870
                     Min.
                            :0.280
                                            :0.1800
                                                      Min.
                                                              :1.690
#>
   1st Qu.: 3.905
                     1st Qu.:0.600
                                     1st Qu.:0.4050
                                                      1st Qu.:2.325
  Median : 4.920
                     Median :0.760
                                     Median :0.5400
                                                      Median :2.590
#> Mean
         : 4.991
                            :0.761
                                           :0.5345
                                                              :2.546
                     Mean
                                     Mean
                                                      Mean
   3rd Qu.: 5.930
                     3rd Qu.:0.905
                                     3rd Qu.:0.6400
                                                      3rd Qu.:2.770
#>
   Max.
          :11.030
                     Max.
                            :1.440
                                     Max.
                                          :1.0000
                                                      Max.
                                                              :3.470
#>
          Α
                         RP
#>
  Min.
          : 7.0
                   Min.
                          :1.103
   1st Qu.: 7.0
                   1st Qu.:1.367
#> Median :16.0
                   Median :1.418
  Mean
          :12.1
                   Mean :1.433
   3rd Qu.:16.0
                   3rd Qu.:1.487
#>
           :22.0
#> Max.
                   Max.
                         :1.750
```

scatterplotMatrix(data, smooth = F, diagonal = F)



Consider Two Different Linear Models

```
modA <- lm(DCrown~PT+RP+HT+A, data = data)</pre>
summary(modA)
#>
#> lm(formula = DCrown ~ PT + RP + HT + A, data = data)
#>
#> Residuals:
      Min
                1Q Median
                                30
                                       Max
#> -2.4770 -0.4398 -0.0358 0.3922 3.4693
#>
#> Coefficients:
#>
               Estimate Std. Error t value Pr(>|t|)
#> (Intercept) -1.84147
                           0.73373 -2.510 0.01260 *
                                    24.228 < 2e-16 ***
#> PT
                8.45921
                           0.34915
#> RP
                0.77674
                           0.44621
                                     1.741 0.08273 .
#> HT
                0.33440
                           0.19159
                                     1.745 0.08192 .
                0.02863
                           0.01026
                                     2.792 0.00558 **
#> A
#> ---
#> Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
#>
#> Residual standard error: 0.7213 on 306 degrees of freedom
#> Multiple R-squared: 0.8124, Adjusted R-squared: 0.81
#> F-statistic: 331.4 on 4 and 306 DF, p-value: < 2.2e-16
```

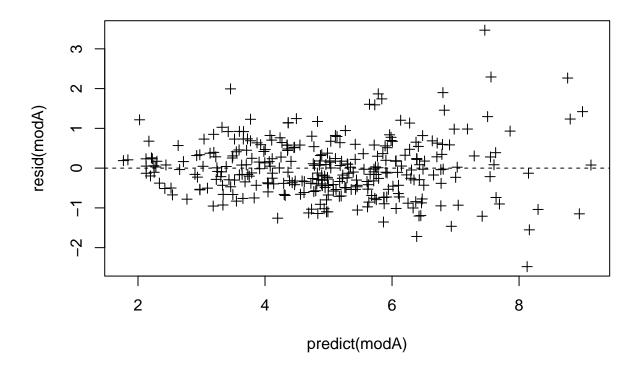
```
modB <- lm(log(DCrown)~log(PT)+log(RP)+log(HT)+log(A), data = data)</pre>
summary(modB)
#>
#> Call:
\# lm(formula = log(DCrown) ~ log(PT) + log(RP) + log(HT) + log(A),
#>
       data = data)
#>
#> Residuals:
#>
       Min
                  1Q
                       Median
                                    3Q
                                        0.45920
#> -0.35655 -0.09336 -0.00394 0.09086
#>
#> Coefficients:
#>
               Estimate Std. Error t value Pr(>|t|)
                           0.09994
#> (Intercept) 1.69533
                                    16.964
                                              <2e-16 ***
#> log(PT)
                0.88111
                           0.03435
                                    25.653
                                              <2e-16 ***
#> log(RP)
                0.28075
                           0.12723
                                     2.207
                                              0.0281 *
#> log(HT)
                0.23307
                           0.09806
                                     2.377
                                              0.0181 *
#> log(A)
                0.05636
                           0.02291
                                     2.461
                                              0.0144 *
#> ---
#> Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
#>
#> Residual standard error: 0.1422 on 306 degrees of freedom
#> Multiple R-squared: 0.8345, Adjusted R-squared: 0.8323
\#> F-statistic: 385.7 on 4 and 306 DF, p-value: < 2.2e-16
```

a) Questions about modA

- (1) All the coefficient estimates are not significantly different from zero, with a significance level of 5%. Only PT and A (plus the intercept) are significant to this level.
- (2) The estimation of the residual variance is (Residual standard error)² = 0.5202814.

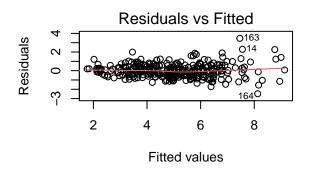
Residual plot is given below. Looks like it is hard to assume homoscedasticity in this case.

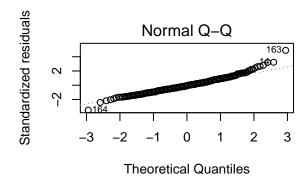
```
plot(predict(modA), resid(modA), pch = 3)
abline(h=0, lty = 2)
```

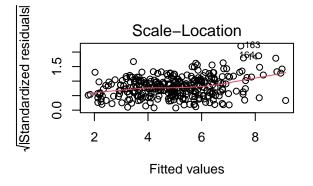


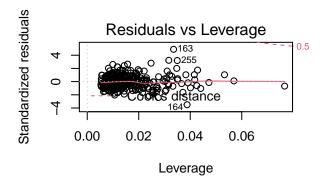
Model diagnostics are done below.

```
par(mfrow = c(2,2))
plot(modA)
```





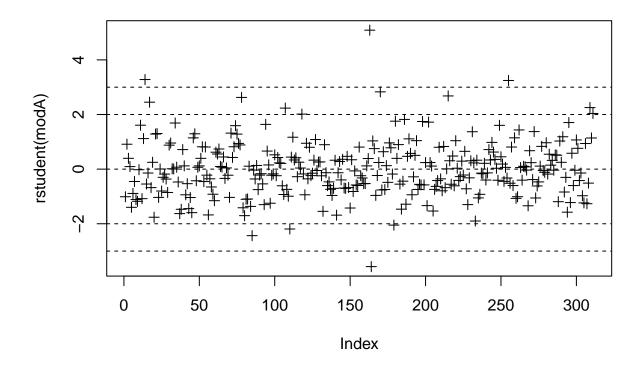




par(mfrow = c(1,1))

(3) The studentized residuals are plotted below.

```
plot(rstudent(modA), pch = 3) abline(h = c(-3, -2, 0, 2, 3), lty = 2)
```



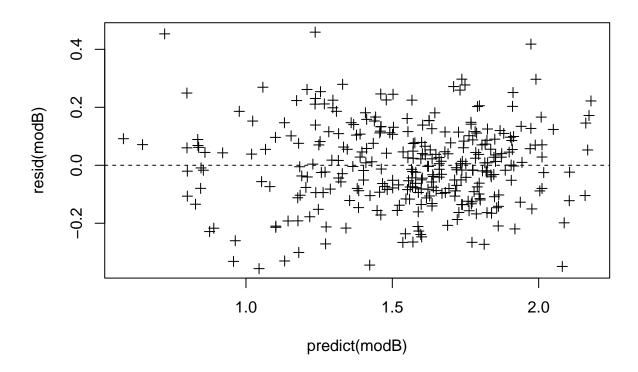
As is apparent, 15 points are outside the interval (-2,2). This represents a percentage of approximately 4.82 % of the points, which is reasonable, considering $1.96 \approx 2$, where 1.96 is the 0.975 quantile of the standard normal distribution. Hence, (-2,2) is an approximation of a 5% confidence interval.

b) Questions about modB

- (1) In the second model all the coefficient estimates are significantly different from zero, with a significance level of 5%.
- (2) The estimation of the residual variance is (Residual standard error) $^2 = 0.0202191$.

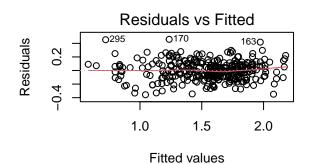
Residual plot is given below. Looks better.

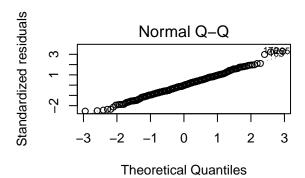
```
plot(predict(modB), resid(modB), pch = 3)
abline(h=0, lty = 2)
```

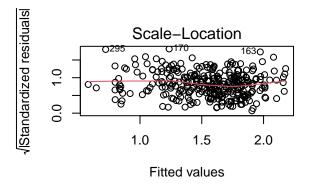


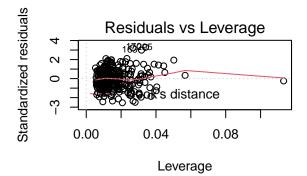
Model diagnostics are done below.

```
par(mfrow = c(2,2))
plot(modB)
```





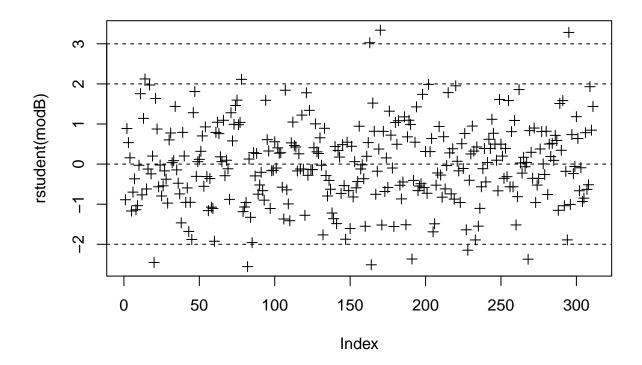




```
par(mfrow = c(1,1))
```

(3) The studentized residuals are plotted below.

```
plot(rstudent(modB), pch = 3) abline(h = c(-3, -2, 0, 2, 3), lty = 2)
```



As is apparent, 11 points are outside the interval (-2,2). This represents a percentage of approximately 3.54% of the points, which is reasonable, considering $1.96 \approx 2$, where 1.96 is the 0.975 quantile of the standard normal distribution. Hence, (-2,2) is an approximation of a 5% confidence interval.

c)

I would choose modB since, compared to modB, R^2 is larger, the F-statistic is larger (even though both have small p-values for the F-test) and all the coefficients are significant (in contrast to the first model, which only has a few estimates that are significantly non-zero, to a 5% level). Note that I have assumed that the assumptions of the linear model are verified for both models (as it says in the problem), which is why these assumptions are not discussed.

d)

#> 1 3.764868 2.843311 4.985115

```
# Do not need to do the following (see row below) since the variables are log-transformed in the model. #new.data <- data.frame(PT = log(0.4), PB = log(0.6), HT = log(2.3), A = log(10), RP = log(0.6/0.4))
new.data <- data.frame(PT = 0.4, PB = 0.6, HT = 2.3, A = 10, RP = 0.6/0.4)
exp(predict(modB, new.data, interval = "prediction", level = 0.95)) # Transform back to correct dimensi
#> fit lwr upr
```