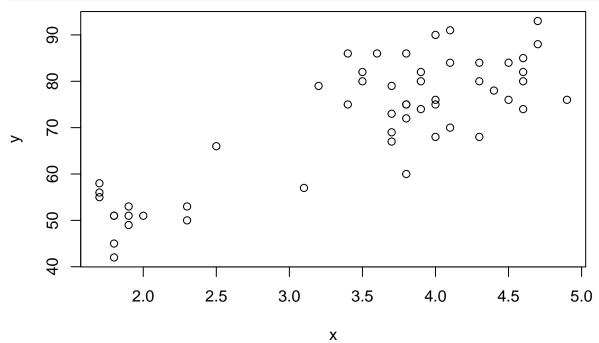
# Statistical inference: part 2, practice session 2

2023-11-30

# Exercicse 6.14

Data importation:

geyser = read.csv(file = "geyser.csv", sep = ";")
plot(y ~ x, data = geyser)



- (a) Implement the formulas given  $\hat{\beta}_1$  and  $\hat{\beta}_0$  (you can also use 1m to fit the model 1m(y ~ x, data = geyser))
- (b) Under  $H_0: \beta_1 = 0$ , we have:

$$t = \frac{\hat{\beta}_1}{s/\sqrt{\sum_{i=1}^n (x_i - \bar{x})^2}} \sim t(n-2)$$

where t(n-2) stands for the t distribution with n-2 degrees of freedom, and  $s^2$  is the unbiased estimator of the variance of the noise:

$$s^{2} = \frac{1}{n-2} \sum_{i=1}^{n} (y_{i} - \hat{y}_{i})^{2} = \frac{1}{n-2} \sum_{i=1}^{n} (y_{i} - \hat{\beta}_{0} - \hat{\beta}_{1} x_{i})^{2}$$

Thus, the course (p.132) says to reject  $H_0$  if  $|t| \ge t_{\alpha/2,n-2}$  where  $t_{\alpha/2,n-2}$  is the upper  $\alpha/2$  percentage point of the central t distribution with n-2 degrees of freedom. This is motivated by the fact that:

$$P_{H_0}(|t| \geq t_{\alpha/2,n-2}) = P_{H_0}(t \geq t_{\alpha/2,n-2}) + P_{H_0}(t \leq -t_{\alpha/2,n-2}) = 2P_{H_0}(t \geq t_{\alpha/2,n-2}) = 2 \times (\alpha/2) = \alpha$$

where the second equality comes from the symmetry of the student distribution. On R  $t_{\alpha/2,n-2}$  can be obtained by qt(alpha/2, n-2, lower.tail = F). Answer question by taking  $\alpha = 0.05$ .

(you can check your results by making summary(lm(y ~ x, data = geyser)) and looking at the p-value in the summary, you reject the test at risk  $\alpha$  if p-value  $\leq \alpha$ )

(c) Now come back to

$$t = \frac{\hat{\beta}_1 - \beta_1}{s/\sqrt{\sum_{i=1}^n (x_i - \bar{x})^2}} \sim t(n-2)$$

(here we do not assume  $H_0: \beta_1 = 0$  any more)

Thus the formula can be obtained page 133 of the book: the following  $100(1-\alpha)$  confidence interval for  $\beta_1$  is

$$\hat{\beta}_1 \pm t_{\alpha/2, n-2} \frac{s}{\sum_{i=1}^n (x_i - \bar{x})^2}$$

Compute the confidence interval with  $\alpha = 0.05$ .

(you can check you results with confint(lm(y~x, data = geyser)))

(d)  $r^{2} = \frac{SSR}{SST} = \frac{\sum_{i=1}^{n} (\hat{y}_{i} - \bar{y})^{2}}{\sum_{i=1}^{n} (y_{i} - \bar{y})^{2}}$ 

(you can also check that  $r^2 = \left(\frac{\text{COV}(x,y)}{\sigma_x \sigma_y}\right)^2$ ) where  $\frac{\text{COV}(x,y)}{\sigma_x \sigma_y}$  is the linear correlation coefficient (you can use cor(geyser\$x, geyser\$y) to get this coefficient)

(you can also look at Multiple R-squared value when making summary(lm(y ~ x, data = geyser)))

## Exercise 7.53

Import the data:

```
gas = read.csv("gas.csv",sep=";")
head(gas)
      y x1 x2
                xЗ
## 1 29 33 53 3.32 3.42
## 2 24 31 36 3.10 3.26
## 3 26 33 51 3.18 3.18
## 4 22 37 51 3.39 3.08
## 5 27 36 54 3.20 3.41
## 6 21 35 35 3.03 3.03
X = cbind(1, as.matrix(gas[,-1])) # add a first column of 1 for the intercept
head(X)
          x1 x2
                  хЗ
## [1,] 1 33 53 3.32 3.42
## [2,] 1 31 36 3.10 3.26
## [3,] 1 33 51 3.18 3.18
## [4,] 1 37 51 3.39 3.08
## [5,] 1 36 54 3.20 3.41
## [6,] 1 35 35 3.03 3.03
y = gas[,1]
```

Use the formula

$$\hat{\beta} = (X'X)^{-1}X'y$$

With R, the function t stands for transposition, solve for matrix inversion, and the operator %\*% stand for matrix multiplication.

And  $s^2$  is given by

$$s^{2} = \frac{1}{n-k-1} \sum_{i=1}^{n} (y_{i} - \hat{y}_{i})^{2}$$

You can compare you results with the results of 1m:

#### ## [1] 2.729995

(b) We know that  $cov(\hat{\beta}) = \sigma^2(X'X)^{-1}$  and  $\sigma^2$  can be estimated by  $s^2$  thus

$$\widehat{\operatorname{cov}(\hat{\beta})} = s^2 (X'X)^{-1}$$

Thus do the computation of  $\widehat{\operatorname{cov}(\hat{\beta})}$  using previous results:

You can check the value of  $(X'X)^{-1}$ 

### reg\_summary\$cov.unscaled

```
x2
                                                                      xЗ
##
                 (Intercept)
                                          x1
                 0.464850012 \quad 0.0019446557 \quad -0.0085652766 \quad -0.15591023
## (Intercept)
                               0.0011014057 -0.0002584188 -0.02187407
## x1
                 0.001944656
## x2
                -0.008565277 -0.0002584188
                                              0.0006152966
                                                             0.01394149 -0.01677282
## x3
                -0.155910227 -0.0218740653 0.0139414927 1.09058795 -0.96667144
                 0.143881820 \quad 0.0105129832 \ -0.0167728207 \ -0.96667144
```

or directly obtain  $s^2(X'X)^{-1}$  by making:

#### vcov(reg)

```
##
               (Intercept)
                                                  x2
                                                              xЗ
                                                                          x4
                                     x1
## (Intercept)
               3.46446861
                            0.014493274 -0.063835928 -1.1619793
## x1
                            0.008208638 -0.001925963 -0.1630247
                                                                  0.07835194
               -0.06383593 -0.001925963 0.004585728
                                                      0.1039042 -0.12500572
## x2
               -1.16197929 -0.163024654 0.103904189 8.1280146 -7.20448052
## x3
                           0.078351940 -0.125005720 -7.2044805
```

The diagonal elements gives the estimated variance of each parameter.

(c)

Find  $R^2$  and  $R_a^2$ , you can use the formulas of exercise 7.29:

$$R^2 = 1 - SSE / \sum_i (y_i - \bar{y})^2$$

and

$$R_a^2 = 1 - \frac{SSE/(n-k-1)}{\sum_i (y_i - \bar{y})^2/(n-1)}$$

You can check you results:

reg\_summary

```
##
## Call:
## lm(formula = y ~ ., data = gas)
##
## Residuals:
     Min
             1Q Median
                           ЗQ
                                 Max
## -5.586 -1.221 -0.118 1.320 5.106
##
## Coefficients:
##
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 1.01502
                        1.86131
                                   0.545 0.59001
                          0.09060 -0.316 0.75461
## x1
              -0.02861
## x2
              0.21582
                          0.06772
                                   3.187 0.00362 **
              -4.32005
                          2.85097
                                   -1.515 0.14132
## x3
## x4
              8.97489
                          2.77263
                                    3.237 0.00319 **
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.73 on 27 degrees of freedom
## Multiple R-squared: 0.9261, Adjusted R-squared: 0.9151
## F-statistic: 84.54 on 4 and 27 DF, p-value: 7.249e-15
where Multiple R-squared gives the R^2 and Adjusted R-squared gives R_a^2
```