UNIVERSITÀ DI ROMA TOR VERGATA EEBL - Business Statistics

Revision - week 2

1.

2. For the br.csv dataset, we regress log-price on log-sqft and log-age (with an intercept). There are N = 1080 observations. Try with regr = lm(log(price) log(sqft)+log(Age). This enables to interpret the estimated coefficients as elasticities (logarithmic derivatives). Illustrate the main estimation results.

```
> br = read.table("br.csv", sep = ",", header=T) # reads the data from a csv file
> attach(br)
> summary(lm(log(price)~log(sqft)+log(Age)))
Call:
lm(formula = log(price) ~ log(sqft) + log(Age))
Residuals:
    Min
              1Q
                   Median
                                 ЗQ
                                        Max
-1.30120 -0.17267 -0.01204 0.18235 1.29271
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
                      0.185793 25.21
                                          <2e-16 ***
(Intercept) 4.684008
                                 40.21
log(sqft)
             0.952429
                       0.023686
                                           <2e-16 ***
log(Age)
            -0.082439
                       0.007066 -11.67
                                           <2e-16 ***
Signif. codes: 0 *** 0.001 ** 0.01 * 0.05 . 0.1
                                                   1
Residual standard error: 0.3101 on 1077 degrees of freedom
Multiple R-squared: 0.6512, Adjusted R-squared: 0.6506
```

The elasticity of price to sqft is estimated equal to 0.95. A 10% increase in the house dimension is expected to yield a 9.5% increase in house price. The estimated coefficients are significantly different from zero. Both variables seem to have a good explanatory power.

F-statistic: 1005 on 2 and 1077 DF, p-value: < 2.2e-16

3. Knowing that

$$(\mathbf{X}'\mathbf{X})^{-1} = \begin{bmatrix} 0.3591 & -0.0455 & -0.0038\\ -0.0455 & 0.0058 & 0.0003\\ -0.0038 & 0.0003 & 0.0005 \end{bmatrix}, \quad \hat{\boldsymbol{\beta}} = \begin{bmatrix} 4.6840\\ 0.9524\\ -0.0824 \end{bmatrix}, \sum_{i} e_i^2 = \mathbf{e}'\mathbf{e} = 103.5403,$$

and that N = 1080, compute the residual standard error $\hat{\sigma}$ (see the slides for formula). Is it possible to compute the *t*-value for the coefficient β_1 from the information provided above?

The *t*-value for $H_0: \beta_2 = 0$ is -11.7 and the corresponding *p*-value is virtually 0; are you willing to accept the null hypothesis?

Knowing further that the total sum of squares is TSS = 296.8724, what is the value of R^2 ? Do you think it is satisfactory?

Solution We first need to estimate σ^2 :

$$\hat{\sigma}^2 = \frac{\sum_i e_i^2}{N - p - 1} = \frac{103.5403}{1080 - 2 - 1} = 0.0961$$

Then, the t-value is

$$t_1 = \frac{0.9524}{\sqrt{0.0961 \cdot 0.0058}} = 40.33281$$

(this is an approximation to the value obtained from the regression output in exercise 2. The estimated standard error of $\hat{\beta}_1$ is obtained as the square root of $\hat{\sigma}^2$ times the element in position (2,2) of the matrix $\mathbf{X}'\mathbf{X}$.

It is reasonable to reject the null $H_0: \beta_2 = 0$.

Finally, we know that RSS = 103.5403 and thus $R^2 = 1 - \frac{RSS}{TSS} = 1 - \frac{103.5403}{296.8724} = 0.65$.

- 4. (This is not an excercise!) The hat matrix etc.
- 5. Let \mathbf{x}_1 denote a vector of N = 10 temperatures in degrees Celsius, generated as $x_1 \sim N(20,3)$ and let \mathbf{y} be the corresponding consumption of beer (in cans), that is linearly related to temperature. Here, $N(\mu, \sigma^2)$ denotes the normal distribution with mean μ and variance σ^2 .

We create a new variable, $\mathbf{x}_2 = 32 + 1.8\mathbf{x}_1$, which represents temperatures converted to degrees Fahrenheit.

We regress Y on X_1 and X_2 . However, as we can see from the code below, something went wrong. What, in particular?

```
x1 = rnorm(10, 20, 3)
y = round(20 * x1 + rnorm(10,0,15))
plot(x1,y)
x2 = 32+1.8 * x1
summary(lm(y ~ x1+x2))
cor(x1,x2)
```

(To execute the above code, open RStudio; from the menu File select New \rightarrow R Script. Copy and paste the code, select and Run. Make sure that \sim is copied correctly).

```
Call:
lm(formula = y ~ x1 + x2)
Residuals:
     Min
                                  ЗQ
               1Q
                    Median
                                          Max
-16.5526 -5.0158 -0.6911
                              8.0913
                                       9.5560
Coefficients: (1 not defined because of singularities)
            Estimate Std. Error t value Pr(>|t|)
(Intercept)
              30.118
                          20.851
                                   1.444
                                            0.187
              18.837
                           1.063
                                 17.727 1.05e-07 ***
x1
x2
                  NA
                              NA
                                      NA
                                               NA
Signif. codes: 0 *** 0.001 ** 0.01 * 0.05 . 0.1
                                                     1
```

```
Residual standard error: 8.975 on 8 degrees of freedom
Multiple R-squared: 0.9752,Adjusted R-squared: 0.9721
F-statistic: 314.2 on 1 and 8 DF, p-value: 1.049e-07
```

The coefficient for the second variable is not estimable as the two regressors are perfectly collinear.