# Regression

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# **Exercises 6**

## **Exercise 1**

Consider normal random variables  $Y_1, \dots, Y_n$  which are iid from the  $N(\mu, \sigma^2)$  distribution. We assume that  $\sigma^2$  is known.

Use the maximum likelihood principle to find an estimate for the expectation parameter  $\mu$ .

### **Exercise 2**

Generate artificial data (say a sample of size n = 100) from the N(1,4) distribution.

- (a) Use the log-likelihood function from the previous Exercise. Plot it as a curve using the sample data and the known value  $\sigma^2=4$ .
- (b) Optimize the log-likelihood numerically using R (check for example <code>?optimize</code>). Add the estimated value  $\widehat{\mu}$  to your plot to check if it is really at the maximum. (You may also add the true value  $\mu=1$ .)

#### Exercise 3

Load the dataset Affairs from the R package AER. (Check the data documentation.)

To estimate a logit model, we need a dependent variable Y with only to values 1 and 0. A useful approach is to generate Y from the variable <code>affairs</code> (say Y=1 if the number of of affairs is positive and Y=0 otherwise).

Do the following analyses:

- (a) Explore graphically the effect of single explanatory variables on Y (you may use for example: spineplot, barplot, or mosaicplot).
- (b) Fit at least three different logit models (some of them should be nested) and do interpret the estimated coefficients.
- (c) Compare your estimated models. Instead of the F test for linear models we do now use  $\chi^2$  tests. The syntax is similar: anova (glm1, glm2, test="Chisq") Alternatively, you may also compare AIC values or apply stepAIC.

