Practical Exercises for Exercise Collection

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Exercise 1: Statistical terminologies

Group the following terminology items into the three categories:

- (1) sample & variables
- (2) hypothesis testing & statistical modelling
- (3) descriptive statistics

 alternative hypothesis 	degree of freedom	interceptIQR	paired sam- ples	single-sided test
• anova	• dependent	linear model	• poisson	 skewed data
barplot	variable	• linear regres-	population	• slope
binary	effect size	sion	predictor	 standard devi-
binomial	• error	• logistic re-	proportion	ation
• Bonferroni	explanatory	gression	• p-value	• standard er-
boxplot	variable	• mean	QQ-plot	ror
 categorical 	• factor	median		• student t -
• Chisquare test	• Fisher's exact	• multiple com-	quantile	distribution
confounding	test	parison	• range	• treatment ef-
contingency	histogram	nominal	• regression co-	fect
table	hypothesis	normal	efficient	• t-test
• continuous	testing	• null hypothe-	residuals	
 correlation 	hypothesis	sis	response	 two-sided test
coefficient	tests	• numeric	• sample	unpaired
• count	 independent 	observation	sampling vari-	samples
 data format 	variable	 odds ratio 	ation	variable
• data point	• integer	ordinal	• scatter plot	variance
• data type	interaction	• outcome	 significance 	vector

Exercise 2: Getting to know R and chickwts

- (a) Open R Studio.
- (b) Open a new R-Script.
- (c) Load data set chickwts.

```
# ?chickwts
data(chickwts)
head(chickwts)
```

Exercise 3: Summary statistics for the chickwts data set

- (a) Do summary statistics (numerically and graphically).
- (b) For advanced R users: Try an anova (are the assumptions fulfilled?) and a Tukey-Anscombe plot. Try a histogram with a density line on top. ...

Exercise 4: Data import to R and summary statistics perulung_ems.csv

- (a) Import the data set perulung_ems.csv (taken from Kirkwood and Sterne, 2nd edition) into R.

 Data from a study of lung function among children living in a deprived suburb of Lima, Peru.

 Variables:
 - fev1: in liter, "Forced Expiratory Volume in 1 second" measured by a spirometer. This is the maximum volume of air which the children could breath out in 1 second
 - age: in years
 - height: in cm
 - sex: 0 = girl, 1 = boy
 - respsymp: respiratory symptoms experienced by the child over the previous 12 months
- (b) What delimiter do you need to choose?
- (c) Do summary statistics (numerically and graphically).

```
summary(lung)
lung$sex <- factor(lung$sex, levels = c("0", "1"))
levels(lung$sex) <- c("female", "male")
lung$respsymptoms <- factor(lung$respsymptoms, levels = c("0", "1"))
# Continuous and factor
tapply(lung$height, lung$sex, mean)
tapply(lung$height, lung$respsymptoms, mean)
# Factor and factor
table(lung$respsymptoms, lung$sex)</pre>
```

```
prop.table(table(lung$respsymptoms, lung$sex),2)
# Continuous and factor
tapply(lung$age, lung$sex, mean)
tapply(lung$age, lung$respsymptoms, mean)
# Continuous and factor
tapply(lung$fev1, lung$sex, mean)
tapply(lung$fev1, lung$respsymptoms, mean)
```

(d) Plot a boxplot.

Exercise 5: Defining a new data frame

(a) Create a data frame with 3 columns.

Exercise 6: Different bracket types within R

What is conceptionally the difference between the bracket types [...] and (...)?

```
chickwts[, 2]
summary(aov(weight ~ feed, data = chickwts))
```

Exercise 7: Data type of perulung_ems data set

(a) Do all variables have the correct data type (numeric, integer, factor)? If not, do correct and / or define them.

Exercise 8: Get to know bacteria data set

- (a) Install package MASS. Load data set bacteria.
- (b) Describe in your own words what the data set bacteria contains.
- (c) Do summary statistic (numerically and graphically).
- (d) Select only observations collected during the second week.
- (e) How many levels has the factor variable trt from bacteria?
- (f) Define a new variable trt.new in which you combine the levels drug and drug+ into one single level and label it as treated. The new variable trt.new should in the end have two levels: placebo and treated.
- (g) Do summary statistics for placebo and treated group.

Exercise 9: Get to know ToothGrowth data set

- (a) Load data set ToothGrowth.
- (b) Do summary statistic (numerically and graphically).
- (c) Define additional column dose.factor by converting the numeric variable dose into a factor variable.

(d) Are the tooth length measurements normally distributed within the treatment (supp: VC or OJ) and within in the different doses (dose: 0.5, 1, 2)?

Exercise 10: Data plausibility checks

- (a) What can go wrong?
- (b) Identify different strategies for spotting these potential errors.
 - · Logical errors
 - · Spelling mistakes
- (c) Import the data set bacteria_plausibility_check.csv to R.
- (d) Detect the six errors in the imported data set bacteria_plausibility_check.csv in R.
- (e) Find possible solutions in R how to handle these challenges.
- (f) Do all variables have the correct data type (numeric, integer, factor)? If not, do correct / define them.

Exercise 11: Missing values

(a) Check out the difference between the different missing values.

```
y1 <- c(2, 4, 3, NA, 6, 1)
y2 <- c("diseased", "healthy", NA, "NA")
y3 <- c(1, "NA", 0, 1, NaN)
#
is.na(y1)
which(is.na(y1))
is.na(y2)
which(is.na(y2))
is.na(y3)
which(is.na(y3))
is.na(y3)</pre>
```

- (b) Create a vector with missing values and determine the mean and median.
- (c) If x < -c(22,3,7,NA,NA,67) what will be the output for the R statement length(x)?
- (d) If x = c(NA, 3, 14, NA, 33, 17, NA, 41) which line of R code removes all occurrences of NA in x.
- (e) If y = c(1, 3, 12, NA, 33, 7, NA, 21) what R statement will replace all occurrences of NA with 11?
- (f) If x = c(34, 33, 65, 37, 89, NA, 43, NA, 11, NA, 23, NA) then what will count the number of occurrences of NA in x?

(g) Create the vector x1. Then, find again the number of missing values and their position.

```
x1 \leftarrow c(rnorm(10,5,2), NA, 5:12, NA, 6, 7.5, NA)
```

(h) Now, create the vector x2 and assess the difference to x1.

```
x2 <- c(rnorm(10,5,2), NA, 5:12, NA, 6, 7.5, NA, log(-2))
```

- (i) What is the meaning of "NA" versus "NaN"?
- (j) Replace the missing values in x1 with a 0. Check then that the NAs are no longer present. Try two different commands to coerce the NAs into 0.

Exercise 12: t-test in R

(a) Load the data set ToothGrowth within R and apply the two-sided two sample t-test to suitable variables of the data set.

```
data(ToothGrowth)
```

- (b) Interpret the results.
- (c) Read in the data set perulung_ems and apply the two-sided t-test to suitable variables of the perulung_ems data set and interpret the results.

Exercise 13: Chi-square test in R

- (a) Apply the Chi-square test and the fisher exact test to the whole bacteria data set.
- (b) Apply the Chi-square test and the fisher exact test to the subset of bacteria containing only the observations taken in week 2 (cf. Exercise 3). Are there any issues?
- (c) Repeat this exercise by using the (previously defined) combined trt.new variable (cf. Exercise 5) with the two levels treated and drug.
- (d) Could you also obtain the odds ratios?
- (e) Try also a logistic regression in R. Ask Google for help!

Exercise 14: Anova in R

- (a) Open the .R file ANOVA_with_chickwts.R from your RCourse folder and have another look on how we applied the anova to the chickwts data set. Check line for line.
- (b) Load the ToothGrowth data set into R and encode the numeric variable dose as a factor variable. Define the new factor variable as dose.factor with the three levels low, med and high and add it to the data frame of ToothGrowth.

```
data(ToothGrowth)
```

- (c) Visualize the variable len per dose.factor level in a boxplot.
- (d) With the help of the R-commands written in the ANOVA_with_chickwts.R file, apply a analysis of variance (ANOVA) to the data set ToothGrowth

Exercise 15: Linear Regression Model I

- (a) Reuse the commands from the lecture slides to fit a simple as well as a multiple linear regression model to the data set of perulung_ems. Use fev1 as your response variable y.
- (b) Check the model assumptions.
- (c) Which model is best?

Exercise 16: Linear Regression Model II

(a) Load the ToothGrowth data set and run the following four linear regression models.

- (b) Have a look at the summary of these models.
- (c) How do you interpret the model coefficients?
- (d) Which model is best?

Exercise 17: Linear Regression Model III

- (a) Load the water data set and fit a multiple linear regression model. Use mortality as your response variable and add hardness and location as an explanatory variable.
- (b) Check the underlying model assumptions.
- (c) Add an interaction term between hardness and location to the above estimated multiple linear regression model.
- (d) Interpret the interaction coefficient hardness:locationSouth.
- (e) Check the underlying model assumptions.
- (f) Which one is the better model? With or without the interaction term?
- (g) How to derive confidence intervals for the regression coefficient of hardness and location?

Exercise 18: Combining everthing we have learnt

Hypothetical example from Kirkwood and Sterne, Medical Statistics, 2nd ed., p. 177

- (a) Read in the data set lepto. This study presents a serology survey of leptospira sero-prevalence in rural and urban areas of the west indies.
- (b) Encode the numeric variable antibodies as a factor with levels 0 and 1.
- (c) Make a crosstable with the risk factor exposure and antibodies.
- (d) Run a Chi-squared test, a Fisher's exact test and a logistic regression (glm) to assess if the exposure (living in rural vs. urban areas) is a risk factor.
- (e) Create a subset for male and female based on the variable gender.
- (f) Repeat the crosstable (2-by-2 table), Chi-squared test, Fisher's exact test and a logistic regression (glm) for the subsets **separately**.
- (g) Does the conclusion of your research question change with the analysis of the subsets? (Research question: Is the exposure (rural and urban areas) a risk factor?)
- (h) Fit a logistic regression model (glm) with exposure and gender as explanatory variables.