# Exam questions for: Advanced Data Analysis with R (700540) $${\rm Master}$ Ecosystem Analysis and Modelling

# Johannes Signer & Kai Husmann

## Summer term 2025

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# 1 Vector data

Use the following data set

- 1. Use df1 and create a geometry column.
- 2. Buffer each tree with its canopy radius.
- 3. Calculate the crown area of each tree and save it in a new column (hint, you my want to use the function st\_area()).
- 4. Find the tree with the largest canopy area.
- 5. Find the tree with the largest canopy area for each species.

## 2 Wokring with raster data

The aim of this exercise is to get you start working with raster data.

- 1. Load the Digital Elevation Model (DEM) of Germany saved in data/raster/dem 3035.tif.
- 2. What is the spatial resolution and the CRS of the raster?
- 3. Cut the DEM to the state of Lower Saxony (use the data set on German states for this; data/ger/ger\_states\_3035.shp).
- 4. What is the mean elevation of Lower Saxony?
- 5. Find all pixels in Lower Saxony that have an elevation of 100 m or more. What is the percentage of Lower Saxony with an elevation of 100 m or more?

## 3 Calculating probabilities

Draw six numbers with replacement from the numbers 1,2,3. Write code for the experiment above and repeat it 10000 times and then calculate the probabilities for

- 1. each number (1, 2, 3) being drawn exactly twice.
- 2. there is one number (e.g., 1, 2, 3) that is never drawn.
- 3. one number being drawn 4 times and one number being drawn twice.
- 4. one number being drawn three times, one twice and one only once.

## 4 Random Variables

Consider the very simple random trial of rolling four dices. Use R to answer the following questions:

- 1. What is the probability that the sum of the four dices is more than 10?
- 2. What is the probability that the sum of the four dices is less or equal to 5?

#### 5 PMFs

Use R to verify that  $f(x) = \frac{2x}{k(k+1)}$  can serve as a PMF for a random variable X, with  $x = 1, 2, 3, \dots, k$ .

## 6 Distributions

Many bird species migrate between winter and summer ranges. Assume we have a sample of 20 independent birds and we know that the probability of arriving at the winter range is 0.86.

- What is the probability that at exactly 10 birds arrive at their winter range?
- What is the probability that at least 10 birds arrive at their winter range?
- What is the probability that more than 5 and less than 16 birds arrive at their winter range?

# 7 p-value

A test statistics from a two-tailed t-test with 9 degrees of freedom is -1.8452. What is the corresponding p-value?

## 8 Estimating parameters of a distribution

A researcher measured the distance a bird dispersed (the distance between the nest where it hatched and the position where it established its nest). The measured distances in km are:

```
## c(1.3, 3, 0.5, 1.9, 3.7, 2.2, 0.8, 0.5, 3.1, 2.1, 2.1, 1.9)
```

- Which distribution do you think is suitable to model this data?
- Which parameters does this distribution have?
- Use optim() to estimate the parameters of this distribution.

## 9 Fitting a linear model

Use the trees data set (located in data/trees.rds) and select only spruce trees for the year 2020. First, investigate graphically, if you think that bio18 (precipitation in the warmest month) could have a linear effect on mean\_damage. Then fit a linear model for this relationship and give an interpretation of the estimated intercept, slope and variance parameter.

#### 10 Visualize a fitted model

Use the trees data set (located in data/trees.rds) and select only spruce trees for the year 2020. Use bio18 and elevation as covariates. Give a brief interpretation of both covariates. Then visualize model by choosing three typical values for elevation (colored lines) and show the values for bio18 on the x-axis and the predicted mean loss on the y-axis. Can you create three variants of the plot? Once without uncertainty, once with a prediction interval and once with a confidence interval?

## 11 Linear model 2

Use the iris data set that is already included in R. You can load it with data(iris). Fit the model:

```
sepal.length = \beta_0 + \beta_1 \times \text{sepal.width} + \beta_2 \times \text{species} + \beta_3 \times \text{sepal.width} \times \text{species}
```

- 1. Before looking at the model summary, think about how many coefficients need to be estimated.
- 2. Write the first 5 entries of the model design or model matrix X. Verify with model.matrix(m1)[1:5, ] that you did this correct.
- 3. Give a brief interpretation for each coefficient.
- 4. Manually predict the sepal.length for *I. setosa* with a sepal width of 3.1 cm. Use the function predict() to verify your prediction.
- 5. Create a plot of the expected sepal lengths for a reasonable range of sepal widths for the different species.

## 12 Interactions

Use the trees data set (located in data/trees.rds) and select only the year 2020. Select the following species: spruce, beeches and oaks. Fit two models: 1) a linear model where you try to explain variation in the mean leaf loss as a function of bio18 and species, and 2) second where you allow an interaction between bio18 and species. Give a brief interpretation of the results. Visualize model both model, what do you notice?

# 13 Fitting a GLM

Linden and Rohloff (2015) studied white-headed woodpeckers in California (USA) using point counts. The file woodp\_Occ.csv contains data from three visits at 66 sites. For each site and visit we know the if a woodpecker

was present or not (columns y.1, y.2, y.3), the day of the year (columns date.1, date.2, date.3) and the snag density (the number of tree snags per ha; column snags). Tree snags are standing but dead or dying trees. Only use the first session of data collection for this exercise (i.e., y.1).

- 1. What would be a suitable model for this kind of data (i.e., think of the distribution of y)?
- 2. Fit the model that you decided on in 1).
- 3. Give an interpretation of the model results.
- 4. Create a suitable plot to communicate the results.
- 5. Validate your model.

## 14 Splines

The code, below, reads in data containing estimates of the number of Moose in Minnesota between 2005 and 2020:

Fit different linear regression (polynomials up to order 5 and natural splines with 3 and 5 knots) model to the data and evaluate whether the assumptions are reasonable.

Plot the data and the model.

## 15 Linear Mixed Models

The file pines/Data1.txt contains estimates of DBH and age for each tree at different ages. Each tree has an individual id (core.code). Perform the following tasks:

- 1. Plot a growth curve for each tree (in one plot). Where you plot the age on the x-axis and the DBH on the y-axis.
- 2. Fit the following models:
  - a. A model  $DBH = \beta_0 + \beta_1 age$  for each individual tree.
  - b. A global model  $DBH = \beta_0 + \beta_1 age$ .
  - c. A global model  $DBH = \beta_0 + \beta_1 age + \beta_2 id + \beta_3 age \cdot id$ .
  - d. The same model as in b), but with a random intercept.
  - e. The same model as in b), but with a random intercept and random slope.
- 3. Create a plot with the model type (2b, 2d, 2e) on the x-axis and the estimate (with a confidence interval) on the y-axis. Distinguish between the two terms (intercept and slope) by using different panels. Hint, you may find the function(s) broom::tidy() with the argument conf.int = TRUE, and bind\_rows useful.

#### 16 Time series 1

Run the following code to generate an annual time series from year 1 to 70.

```
set.seed(1)
c(arima.sim(n = 30, list(ar = c(.9, -.5, .3))) + .05 * c(1 : 30) + runif(30),
    arima.sim(n = 10, list(ar = c(.9))) + .05 * c(31 : 40) + runif(10),
    arima.sim(n = 30, list(ar = c(.9, -.5, .3))) + .05 * c(41 : 70) + runif(30)) |> ts()
```

- 1. Create an appealing plot.
- 2. Visualise all relevant information that are of interest for further analyses of the time series.
- 3. Create an autocorrelation function.

#### 17 Time series 2

• Load the number of births per month in New York city, from January 1946 to December 1959 as births <- scan("http://robjhyndman.com/tsdldata/data/nybirths.dat").

- Transform and save it as a time series object.
- 1. Plot the time Series
- 2. Estimate parameters for an intercept, a linear trend and seasonality.
- 3. Which of these 3 components are significant?

## 18 Time series 3

- Load the number of births per month in New York city, from January 1946 to December 1959 as births <- scan("http://robjhyndman.com/tsdldata/data/nybirths.dat").
- Transform and save it as a time series object.
- 1. Describe the series using descriptive statistics.
- 2. Which model(s) do you use?
- 3. How do you interpret the results of that model(s)?

## 19 Time series 4

- Load the number of births per month in New York city, from January 1946 to December 1959 as births <- scan("http://robjhyndman.com/tsdldata/data/nybirths.dat").
- Transform and save it as a time series object.
- 1. Create an appropriate ar model.
- 2. Do a one year forecast.