

Introduction to programming in R and Python by Natalia Kazakova and Andrey Yakovenko

Code ▾

Project 1

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```
install.packages("stringi", dependencies=TRUE, INSTALL_opts = c('--no-lock'))
```

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```
install.packages("stringr", dependencies=TRUE, INSTALL_opts = c('--no-lock'))
```

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```
install.packages("tidyverse")
```

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```
library(tidyverse)
```

1-2-3. Find a dataset, download it and load the data into R

We will study the species repartition and weight of animals caught in plots in our study area. The dataset is stored as a comma separated value (CSV) file. Each row holds information for a single animal, and the columns represent:

record_id - Unique id for the observation
month - month of observation
day - day of observation
year - year of observation
plot_id - ID of a particular plot
species_id - 2-letter code
sex - sex of animal ("M", "F")
hindfoot_length - length of the hindfoot in mm
weight - weight of the animal in grams
genus - genus of animal
species - species of animal
taxa - some taxon, e.g. Rodent, Reptile, Bird, Rabbit
plot_type - type of plot

Before the work, we need to check and change (if necessary) our working directory:

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```
getwd( )
```

```
[1] "/Users/19820585/Desktop"
```

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```
setwd('/Users/19820585/Documents/hse/Introduction to programming in R and Python')
```

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```
surveys <- read.csv("/Users/19820585/Documents/hse/Introduction to programming in  
R and Python/portal_data_joined.csv")
```

Let's check the top of this data frame:

	record_id	mo...	...	year	plot_id	species_id	...	hindfoot_length	weight	
	<int>	<int>	<int>	<int>	<int>	<chr>	<chr>	<int>	<int>	
1	1	7	16	1977	2	NL	M	32	NA	
2	72	8	19	1977	2	NL	M	31	NA	
3	224	9	13	1977	2	NL		NA	NA	
4	266	10	16	1977	2	NL		NA	NA	
5	349	11	12	1977	2	NL		NA	NA	
6	363	11	12	1977	2	NL		NA	NA	

6 rows | 1-10 of 13 columns

... and size:

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```
dim(surveys)
```

```
[1] 34786    13
```

Let's inspect the structure of a data frame with the function str():

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```
str(surveys)
```

```
'data.frame': 34786 obs. of 13 variables:
 $ record_id      : int  1 72 224 266 349 363 435 506 588 661 ...
 $ month          : int  7 8 9 10 11 11 12 1 2 3 ...
 $ day            : int  16 19 13 16 12 12 10 8 18 11 ...
 $ year           : int  1977 1977 1977 1977 1977 1977 1977 1977 1978 1978 1978 ...
 $ plot_id        : int  2 2 2 2 2 2 2 2 2 2 ...
 $ species_id     : chr   "NL" "NL" "NL" "NL" ...
 $ sex            : chr   "M" "M" "" "" ...
 $ hindfoot_length: int  32 31 NA NA NA NA NA NA NA NA ...
 $ weight         : int  NA NA NA NA NA NA NA NA 218 NA ...
 $ genus          : chr   "Neotoma" "Neotoma" "Neotoma" "Neotoma" ...
 $ species        : chr   "albigula" "albigula" "albigula" "albigula" ...
 $ taxa           : chr   "Rodent" "Rodent" "Rodent" "Rodent" ...
 $ plot_type      : chr   "Control" "Control" "Control" "Control" ...
```

Moreover, it can be useful to check summary statistics for each column:

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```
summary(surveys)
```

record_id	month	day	year	plot_id	s
species_id	sex				
Min. : 1	Min. : 1.000	Min. : 1.0	Min. : 1977	Min. : 1.00	Length: 34786
1st Qu.: 8964	1st Qu.: 4.000	1st Qu.: 9.0	1st Qu.: 1984	1st Qu.: 5.00	Class : character
Median : 17762	Median : 6.000	Median : 16.0	Median : 1990	Median : 11.00	Mode : character
Mean : 17804	Mean : 6.474	Mean : 16.1	Mean : 1990	Mean : 11.34	
3rd Qu.: 26655	3rd Qu.: 10.000	3rd Qu.: 23.0	3rd Qu.: 1997	3rd Qu.: 17.00	
Max. : 35548	Max. : 12.000	Max. : 31.0	Max. : 2002	Max. : 24.00	
hindfoot_length	weight	genus	species	taxa	
plot_type					
Min. : 2.00	Min. : 4.00	Length: 34786	Length: 34786	Length: 34786	
1st Qu.: 21.00	1st Qu.: 20.00	Class : character	Class : character	Class : character	
Median : 32.00	Median : 37.00	Mode : character	Mode : character	Mode : character	
Mean : 29.29	Mean : 42.67				
3rd Qu.: 36.00	3rd Qu.: 48.00				
Max. : 70.00	Max. : 280.00				
NA's : 3348	NA's : 2503				

4. Pick the variables that you think can be relevant for your analysis. Copy these variables into a separate dataset

We decided to drop two variables from our dataset - taxon and type of plot:

5. Check that all R data types actually match the data (e.g. no numbers are stored as characters). If there are, convert the data to the relevant types

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```
str(surveys_new)
```

```
'data.frame':  34786 obs. of  11 variables:
 $ record_id      : int  1 72 224 266 349 363 435 506 588 661 ...
 $ month          : int  7 8 9 10 11 11 12 1 2 3 ...
 $ day           : int  16 19 13 16 12 12 10 8 18 11 ...
 $ year          : int  1977 1977 1977 1977 1977 1977 1977 1977 1978 1978 1978 ...
 $ plot_id       : int  2 2 2 2 2 2 2 2 2 2 ...
 $ species_id    : chr   "NL" "NL" "NL" "NL" ...
 $ sex           : chr   "M" "M" "" "" ...
 $ hindfoot_length: int  32 31 NA NA NA NA NA NA NA NA ...
 $ weight        : int  NA NA NA NA NA NA NA NA 218 NA ...
 $ genus         : chr   "Neotoma" "Neotoma" "Neotoma" "Neotoma" ...
 $ species       : chr   "albigula" "albigula" "albigula" "albigula" ...
```

It looks fine

6. In this dataset, check if there are any missing data. If there are, clean your data by removing the observations where data are missing

Let's start by removing observations of animals for which weight and hindfoot_length are missing, or the sex has not been determined:

Because we are interested in plotting how species abundances have changed through time, we are also going to remove observations for rare species (i.e., that have been observed less than 50 times). We will do this in two steps: first, we are going to create a data set that counts how often each species has been observed, and filter out the rare species. Second, we will extract only the observations for these more common species:

Let's check our new data frame:

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```
dim(surveys_complete)
```

```
[1] 30521  11
```

7. After cleaning the dataset, save your result for further analysis to the RDS format

Also, we will save it to the CSV format

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```
write_csv(surveys_complete, path = "/Users/19820585/Documents/hse/Introduction to programming in R and Python//surveys_complete.csv")
```

Project 3

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```
install.packages("ggfortify")
```

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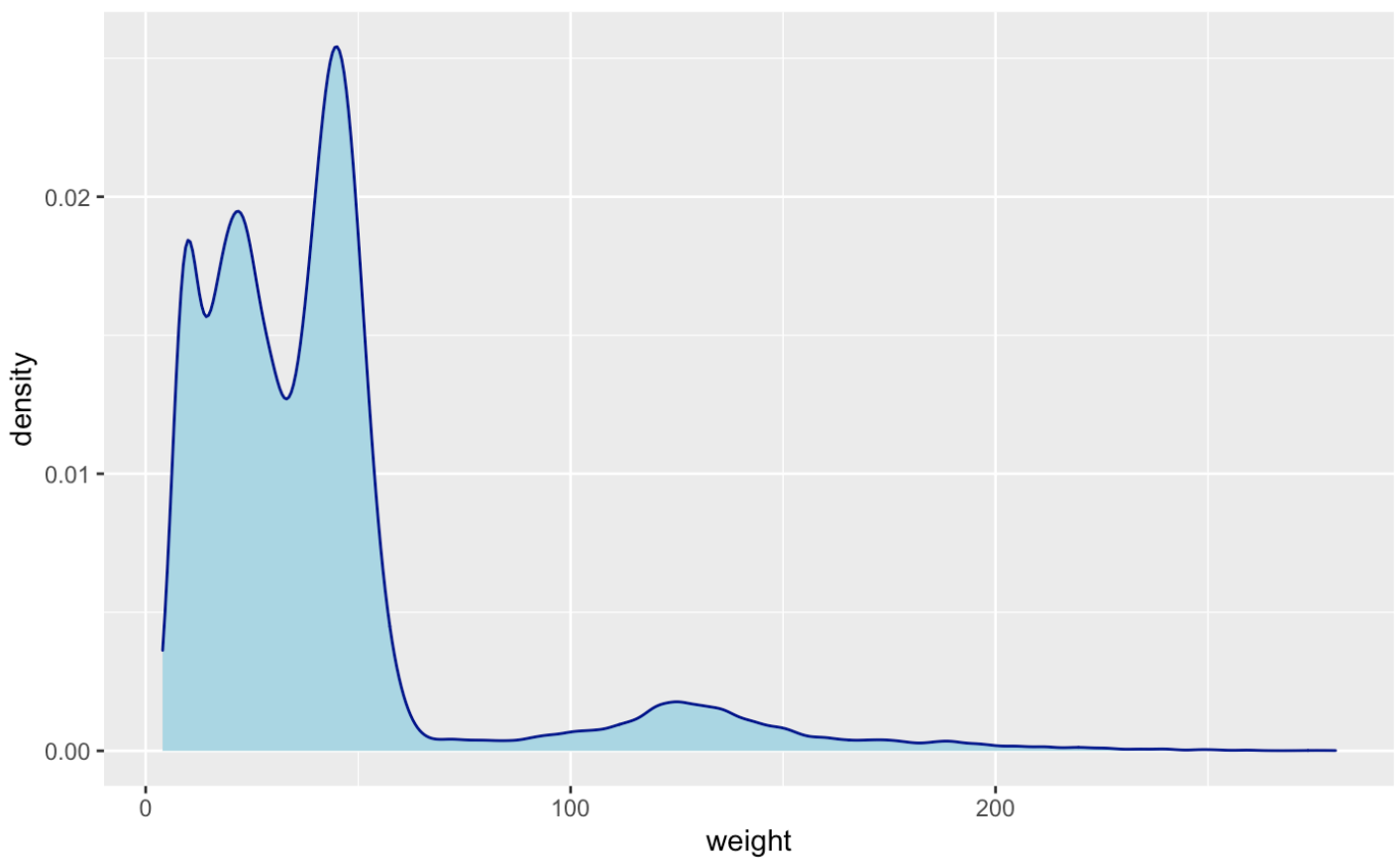
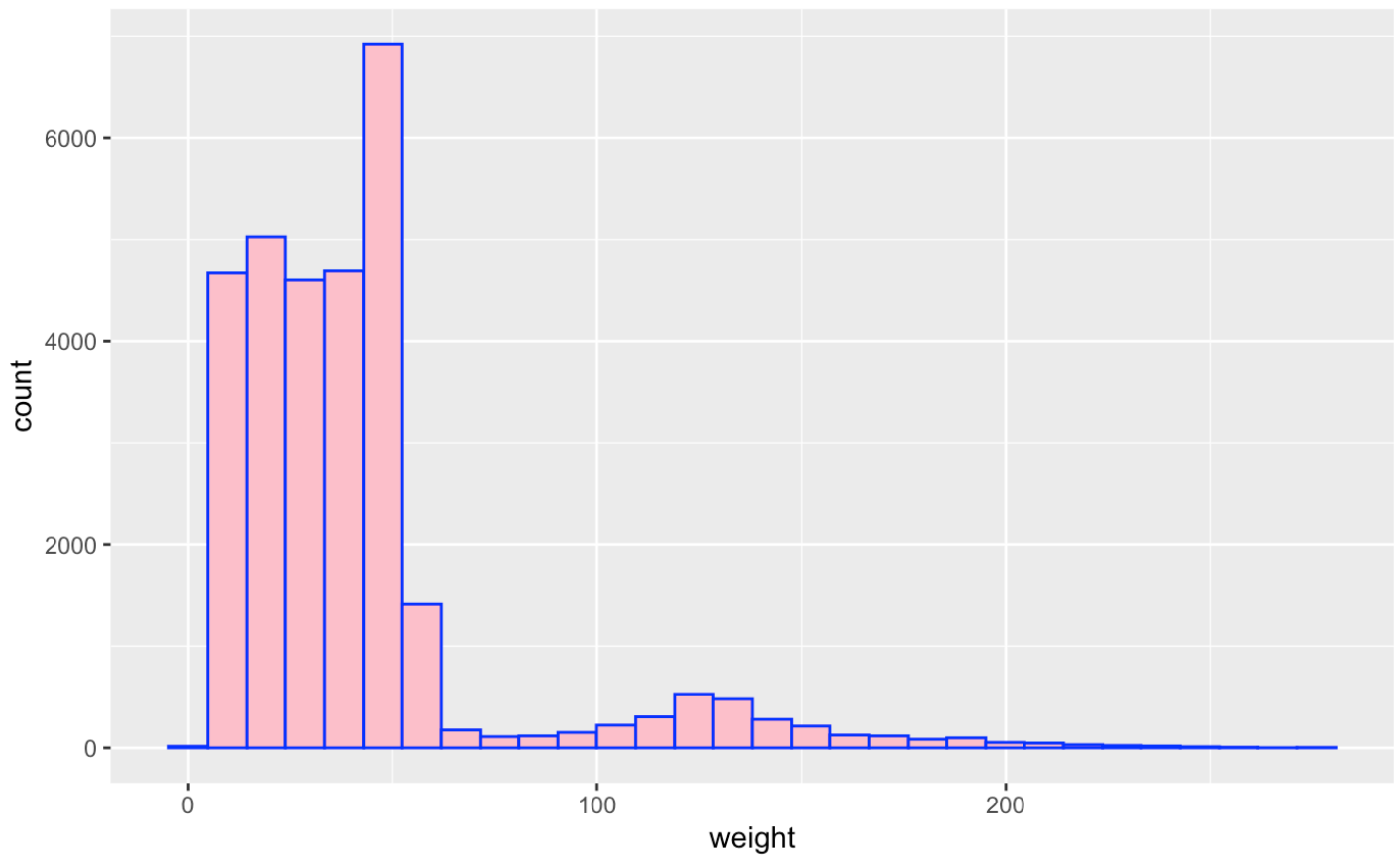
```
library(ggfortify)
```

- a. A histogram and a density plot of some numeric (ratio or interval scale 😊) variables. These plots show us the distribution of weight within our data set.

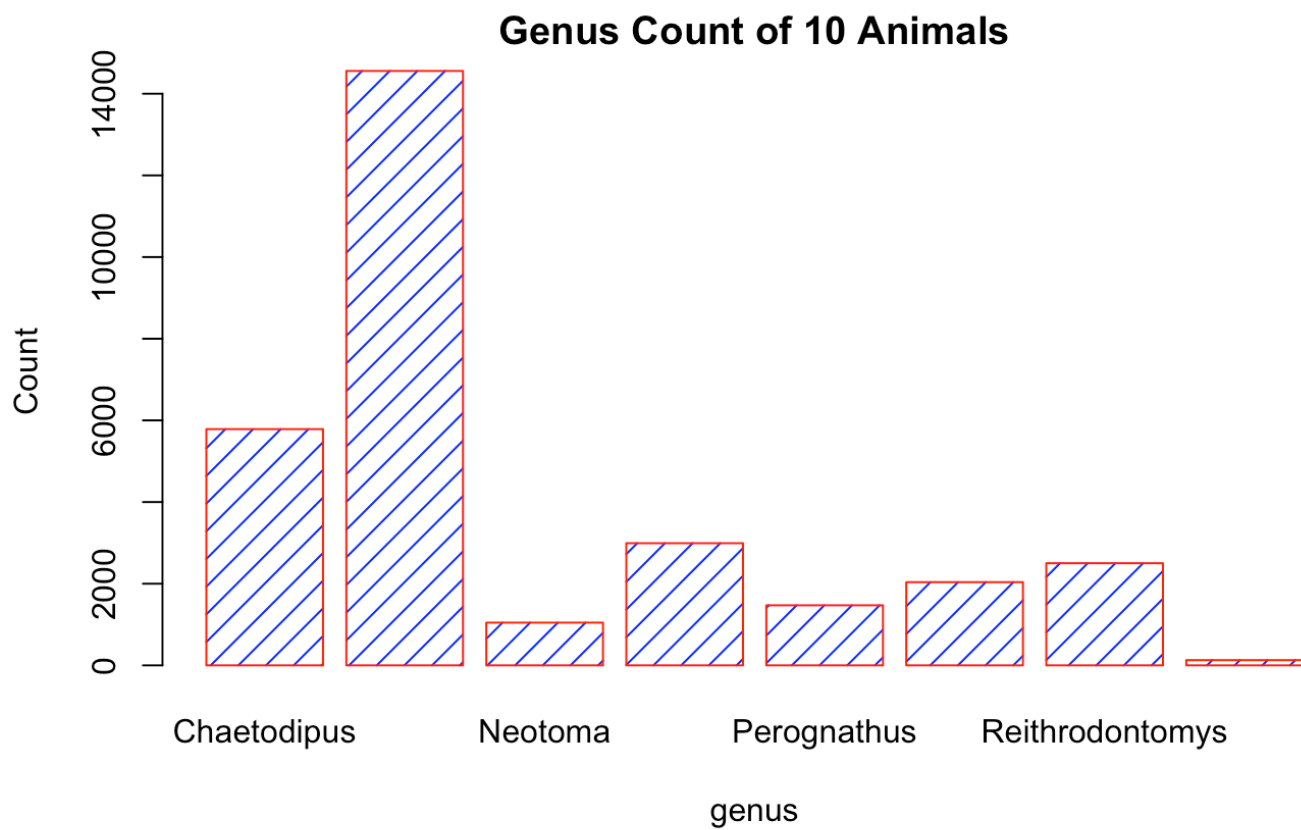
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```
ggplot(data=surveys_complete, aes(x=weight))+geom_histogram(color = "blue", fill = "pink")
```

```
`stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```

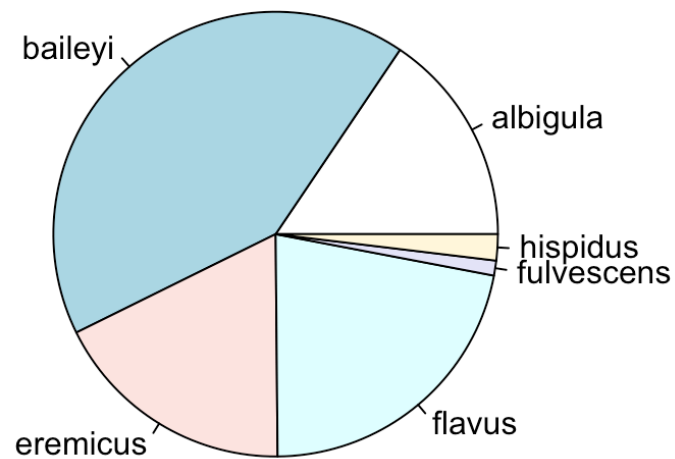


- b. A bar and a pie chart for some factor variables (if you do not have factor variables in your dataset, take some numeric variables and convert them to factors)

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```
head(df_pie2)
```

df_pie						
	albigula	baileyi	eremicus	flavus	fulvescens	hispidus
	1046	2808	1200	1471	73	128

Pie Chart of Species

c. Box (or violin) plots for some numeric variables Here we can see box plots, that reflect animals' weight within their species.

