

Second Edition (Chapman and Hall/CRC, 2019)

## Exercises of Chapter 6:

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## Exercises

### Exercise 6.1

Use the `womensrole` data (see below) and modify the R code given in the **Examples of Chapter 6** to analyse and visualize the data as instructed.

[illegible]

##	education	sex	agree	disagree
## 1	0	Male	4	2
## 2	1	Male	2	0
## 3	2	Male	4	0
## 4	3	Male	6	3
## 5	4	Male	5	5
## 6	5	Male	13	7

### Exercise 6.3

Use the accidents data (see below) and modify the R code given in the **Examples of Chapter 6** to analyse the data as instructed.

```
accidents <- structure(list(
  weight = structure(c(1L, 1L, 1L, 1L, 2L, 2L, 2L, 2L),
    .Label = c("Small", "Standard"), class = "factor"),
  eject = structure(c(1L, 1L, 2L, 2L, 1L, 1L, 2L, 2L),
    .Label = c("No", "Yes"), class = "factor"),
  type = structure(c(1L, 2L, 1L, 2L, 1L, 2L, 1L, 2L),
    .Label = c("Collision", "Rollover"), class = "factor"),
  severely = c(150, 112, 23, 80, 1022, 404, 161, 265),
  nseverely = c(350, 60, 26, 19, 1878, 148, 111, 22)),
  .Names = c("weight", "eject", "type", "severely", "nseverely"),
  row.names = c(NA, -8L), class = "data.frame")

head(accidents)
```

##	weight	eject	type	severely	nseverely
## 1	Small	No	Collision	150	350
## 2	Small	No	Rollover	112	60
## 3	Small	Yes	Collision	23	26
## 4	Small	Yes	Rollover	80	19
## 5	Standard	No	Collision	1022	1878
## 6	Standard	No	Rollover	404	148

```
tail(accidents)
```

##	weight	eject	type	severely	nseverely
## 3	Small	Yes	Collision	23	26
## 4	Small	Yes	Rollover	80	19
## 5	Standard	No	Collision	1022	1878
## 6	Standard	No	Rollover	404	148
## 7	Standard	Yes	Collision	161	111
## 8	Standard	Yes	Rollover	265	22

## Exercise 6.4

Use the statistics data (see below) and modify the R code given in the **Examples of Chapter 6** to analyse the data as instructed.

```
statistics <- structure(list(

  result = structure(c(1L, 1L, 2L, 1L, 2L, 2L, 2L, 2L, 2L,
                      2L, 2L, 2L, 2L, 2L, 1L, 2L, 1L, 1L, 2L),
                    .Label = c("Fail", "Pass"), class = "factor"),

  test = c(525, 533, 545, 582, 581, 576, 572, 609, 559, 543,
           576, 525, 574, 582, 574, 471, 595, 557, 557, 584),

  grade = structure(c(2L, 3L, 2L, 1L, 3L, 4L, 2L, 1L, 3L, 4L,
                    2L, 1L, 4L, 3L, 2L, 2L, 3L, 1L, 1L, 1L),
                  .Label = c("A", "B", "C", "D"), class = "factor")),
.Names = c("result", "test", "grade"), row.names = c(NA, -20L), class = "data.frame")

head(statistics)
```

```
##   result test grade
## 1   Fail  525     B
## 2   Fail  533     C
## 3   Pass  545     B
## 4   Fail  582     A
## 5   Pass  581     C
## 6   Pass  576     D
```

```
tail(statistics)
```

```
##   result test grade
## 15   Pass  574     B
## 16   Fail  471     B
## 17   Pass  595     C
## 18   Fail  557     A
## 19   Fail  557     A
## 20   Pass  584     A
```

## Exercise 6.5

Use the menstruation data (see below) and modify the R code given in the **Examples of Chapter 6** to analyse and visualize the data as instructed.

```
menstruation <- structure(list(
  age = c(11.08, 11.33, 11.58, 11.83, 12.08, 12.33, 12.58, 12.83, 13.08,
          13.33, 13.58, 14.08, 14.33, 14.58, 15.08, 15.33, 15.58, 17.58),
  bmens = c(2, 5, 10, 17, 16, 29, 39, 51, 47, 67, 81, 79, 90, 93, 117, 107, 92, 1049),
  n = c(120, 88, 105, 111, 100, 93, 100, 108, 99, 106, 117,
        98, 97, 100, 122, 111, 94, 1049)),
  .Names = c("age", "bmens", "n"), row.names = c(NA, -18L), class = "data.frame")

menstruation
```

##	age	bmens	n
## 1	11.08	2	120
## 2	11.33	5	88
## 3	11.58	10	105
## 4	11.83	17	111
## 5	12.08	16	100
## 6	12.33	29	93
## 7	12.58	39	100
## 8	12.83	51	108
## 9	13.08	47	99
## 10	13.33	67	106
## 11	13.58	81	117
## 12	14.08	79	98
## 13	14.33	90	97
## 14	14.58	93	100
## 15	15.08	117	122
## 16	15.33	107	111
## 17	15.58	92	94
## 18	17.58	1049	1049

## Exercise 6.6

Use the gambling data (see below) and modify the R code given in the **Examples of Chapter 6** to analyse and visualize the data as instructed. Below you will find some information on the variables of the data. For more details, see the original article:

Donati, M. A., Chiesi, F. and Primi, C. (2013). A model to explain at-risk/problem gambling among male and female adolescents: Gender similarities and differences. *Journal of Adolescence*, 36, 129-137.

```
gambling <- read.table("data/gambling.txt", sep = '\t', header = TRUE)

head(gambling)
```

##	Id	Gender	Age	G1	G2	G3	G4	G5	G6	G7	G8	G9	G10	G11	GamblF	GFalla
## 1	1	2	13.41667	0	1	0	0	0	0	1	0	0	0	0	2	5
## 2	2	1	13.83333	0	0	0	0	1	0	0	0	2	2	0	5	2
## 3	3	2	13.83333	1	0	0	0	0	0	0	0	3	1	0	5	3
## 4	4	2	13.83333	0	0	0	0	0	0	0	0	1	0	0	1	6
## 5	5	2	13.83333	1	0	0	0	0	3	0	0	0	0	0	4	6
## 6	6	2	13.83333	3	1	2	0	0	1	1	2	4	0	0	14	2

  

##	SensaS	SuperS	EconG	PareG	PeerG	Nsymp	P3prob	P2prob
## 1	21	18	30	0	0	0	0	0
## 2	36	11	35	1	NA	0	0	0
## 3	17	11	39	1	0	0	0	0
## 4	24	26	38	0	0	0	0	0
## 5	25	24	36	1	0	1	1	1
## 6	34	18	26	1	1	3	2	1

- Gender: 1 = male, 2 = female
- Age: Age in years
- G1: Cards (0 = never, 1 = less than monthly, 2 = monthly, 3 = weekly, 4 = daily)
- G2: Coin toss
- G3: Bets on games of skill
- G4: Bets on sport games
- G5: Bets on horse races
- G6: Bingo
- G7: Dice games
- G8: Slot machines
- G9: Instant scratch-cards
- G10: Lotteries
- G11: Online games
- GamblF: Gambling frequency (sum of G1 to G11)
- GFalla: Probabilistic reasoning (Gambler's fallacy)
- SensaS: Sensation seeking
- SuperS: Superstitious thinking
- EconG: Economic perception of gambling (high score = more negative attitude)
- PareG: Parental gambling behavior
- PeerG: Peer gambling behavior
- Nsymp: Problem gambling (total of symptoms)
- P3prob: Problem gambling (0 = non-problem, 1 = at-risk, 2 = problem)
- P2prob: Problem gambling (0 = non-problem, 1 = at-risk/problem)