

Introduction to \mathcal{R}

Session 5: Basic Statistics

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Introduction

So far, you have learned...

- to navigate the \mathcal{R} environment,
- to create, index, and modify objects,
- to take advantage of control flow statements,
- generate decent looking graphs.
- You are now ready to unleash \mathcal{R} on real data.

What are we going to do?

We are going to plunge head first into data analysis, taking advantage of a data set which includes all *named* characters from George R.R. Martin's "A Song of Ice and Fire".² In the end, you will be able to offer much insight on the mother of all cocktail party questions: **Is Jon Snow going to die?**

To get started:

1. Quit & reopen \mathcal{R} .
2. Load `"/05/dta/asoiaf.csv"` from the course material.
 - **Note:** Uncheck the option "Strings as factors".
3. Open a new script file.
4. Load `"ggplot2"`.

²O'Neill, M. 2016. Game of Thrones. <https://bit.ly/2qjUfQ2> (last access: 10/08/2018).

Outline

- 1 Introduction
- 2 Many Values
- 3 Few Values
- 4 Adventures in Association
- 5 Summary

Few or Many Values?

We are interested in the variables `allegiances`, `age_in_chapters`, `gender`, and `nobility`.

- Proper tools for description change with data type
- Indicator: How many different values does a variable take?
- **Exercise:** Create a `for()` loop that returns the number of unique values on said variables.

Few or Many Values?, contd.

A possible solution

```
str(asoiaf)
vars <- c("allegiances", "age_in_chapters", "gender",
         "nobility"
)
for (v in vars) { print(length(unique(asoiaf[, v]))) }
```

Many Values

Central Tendency and Spread

Enter each of the following commands. Explain the output.

```
mean(asoiaf[, "age_in_chapters"], na.rm = TRUE)
sd(asoiaf[, "age_in_chapters"], na.rm = TRUE)
quantile(asoiaf[, "age_in_chapters"],
  probs = c(0, .01, .05, .25, .5, .75, .9, .95, 1),
  na.rm = TRUE
)
summary(asoiaf[, "age_in_chapters"])
```

■ What have we learned?

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- What have we learned?
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- **What have we learned?**
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- Each requires instructions on how to process NAs.

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```

■ What have we learned?

- `mean()`, `sd()`, and `quantile()` return just that.
- Each requires instructions on how to process NAs.
- `summary()` returns the 5-point-summary plus mean and NAs.

Graphical EDA

ggplot2 offers numerous exploratory graphs.³ Create each of the graphs below. What do they return?

```
p <- ggplot(data = asoiaf, aes(x = age_in_chapters))  
p + geom_histogram()  
p + geom_density() + labs(y = "PDF")  
p + stat_ecdf() + labs(y = "CDF")  
p + geom_boxplot(aes(x = 0, y = age_in_chapters))  
ggplot(data = asoiaf, aes(sample = age_in_chapters)) +  
  geom_qq() + geom_qq_line()
```

³For an entire theory of graphical EDA using ggplot2 see Unwin, A. 2015. Graphical Data Analysis with R. Boca Raton: CRC Press.

Grouping Values

- **Goal:** Controlled loss of information for, e.g., tables

```
# Variant a. Aggregate data -----  
mu_age_by_allegiance <- aggregate(  
  x = asoiaf[, "age_in_chapters"],  
  by = list(allegiances = asoiaf[, "allegiances"]),  
  FUN = mean, na.rm = TRUE  
); mu_age_by_allegiance
```

Grouping Values, contd.

```
# Variant b. Recode the data -----
tmp <- cut(x = asoiaf[, "age_in_chapters"],
  breaks = 5
  # divides data into <breaks> pieces of equal length
); summary(tmp) # Note something weird?
tmp <- cut(x = asoiaf[, "age_in_chapters"],
  breaks = quantile(
    asoiaf[, "age_in_chapters"], na.rm = TRUE
  ), # vector of values at which to cut x.
  include.lowest = FALSE
); summary(tmp) # Note something weird?
typeof(tmp); class(tmp) # Note something weird?
```

Few Values

What are factors?

- Special instance (“class”) of atomic vectors
- Store nominal and ordinal data, e.g., eye color & letter grades
- Look like character strings, but behave like integers
- Create factors only when needed

```
grades <- c("A", "B", "B", "C")
grades <- factor(grades,
  levels = c("C", "B", "A"), # state ALL values
  labels = c("C", "B", "A"), # name EACH value
  ordered = TRUE # defaults to FALSE (nominal data)
)
typeof(grades); attributes(grades) # Try these.
```

What are factors?, contd.

- Components of a factor: numeric value & character label
- labels (BUT NOT VALUES) can be used for logical indexing

```
grades; as.numeric(grades)
```

```
## [1] A B B C
```

```
## Levels: C < B < A
```

```
## [1] 3 2 2 1
```

```
grades > "C" # will work fine
```

```
## [1] TRUE TRUE TRUE FALSE
```

```
grades > 1 # will generally not work
```

```
## [1] NA NA NA NA
```

Simple N-way Contingency Tables

- `table()` creates N-way contingency tables

```
table(asoiaf[, "gender"]) # single 1way table
```

```
##  
##    0    1  
## 157 760
```

Explain the output of these statements. Do you notice anything?

```
apply(  
  asoiaf[, c("gender", "nobility")], 2, table  
)  
table(asoiaf[, "book_of_death"], asoiaf[, "nobility"])
```

Add Information to Contingency Tables

- You must state explicitly what information you require.
- Examples: Proportions & Totals

```
mytable <- table(  
  "gender" = asoiaf[, 'gender'],  
  "nobility" = asoiaf[, 'nobility']  
)  
prop.table(mytable)  
# Add argument margin = {1; 2}. What happens?  
addmargins(mytable)  
# Add argument margin = {1; 2}. What happens?
```

Adventures in Association

Two-way Contingency Tables

- Numerous methods provided
- Most defined by individual functions
- See packages **vcd** & **vcdExtra** for more options

```
mytable <- table(  
  "gender" = asoiaf[, 'gender'],  
  "nobility" = asoiaf[, 'nobility']  
)  
fit <- chisq.test(mytable); fit # Chi-Square Test  
fit <- fisher.test(mytable); fit # Fisher's Exact Test
```

Correlation Analysis

- Scatter plots are the starting point for any correlation analysis
- `base::pairs()` & `car::scatterplotMatrix()` return plot matrices
- Quantify associations using `cor()` and `cor.test()` functions

```
# a. Explain the code and plot. -----
ggplot(data = asoiaf,
  aes(x = chapter_of_intro, y = chapter_of_death)
) + geom_point() +
  geom_smooth(aes(col = "loess"), method = "loess") +
  geom_smooth(aes(col = "ols"), method = "lm")
# b. Correlation Analysis -----
cor(x = asoiaf[, c(6, 8)],
  use = "complete.obs", # What does <use> do?
  method = "pearson" # {pearson; kendall; spearman}
) # Now try cor.test() on your own.
```

Mean Comparison Tests

- Question: Do two groups come from the same population?
- In ASOIAF: Do nobles survive longer than other social strata?

```
t.test( # Alternative:  $\mu_0 < \mu_1$   
  age_in_chapters ~ nobility, data = asoiaf,  
  alternative = "less" # "greater" for  $\mu_0 > \mu_1$   
)
```

- Defaults to unequal variances & returns Welch's t-test
- Set paired = TRUE for paired t-tests
- For detailed instructions see Field, A. et al. 2012. Discovering Statistics Using R. Los Angeles: SAGE, Ch. 9.

Summary