Module 3: R Professional Skills

Amy Farrow

Data Sciences Institute, University of Toronto

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Overview

- Presenting Data Science
- Presenting Your Work in R
- Managing Data Science Projects

Presenting Data Science

Challenges

- A target audience that will likely not have an equivalent technical understanding to your own
- Communicating the limitations while promoting your work
- Keeping it interesting
- Including the appropriate level of persuasion



Strategies

- Don't just think in terms of performance and technical ability of your analysis and models. Consider: -- Are your models believable? -- What evidence have you offered? How have you built trust?
- Don't overcomplicate it -- What details are necessary to understanding the core goals, abilities, and limitations of your project? Which are not?
- Tailor your message -- Who are you presenting to? What level of knowledge can you assume?
- Always have takeaway messages -- In a presentation where some viewers might get confused along the way, a strong structure and final message can keep everyone on track

Practical Considerations: Presenting Your Work in R

Presenting Data Tables

There are many libraries to format tables for readable and professional-looking outputs. Some only work for certain kinds of outputs (html, pdf, or docx), so you may have to pick a library that is the best fit for you.

kableExtra is a library for formatting table output in html and pdf.

A table without formatting looks like this:

With formatting:

My Table		
Categories		
Α	В	Other
Run 1		
1	one	14.3
2	two	182.5
Run 2		
3	three	54.0
4	four	33.1

Formatting Reports

The package **bookdown** can be used to cross-reference figures and tables, add citations, create a table of contents, and more.

For the output to work correctly, you want to have blank lines in your markdown in between parts: i.e. between a paragraph and the next paragraph, a paragraph and a code chunk, a code chunk and a paragraph, or a code chunk and the next code chunk.

A typical header for such a report would look like:

Breaking down the header

```
title: "TITLE"
subtitle: "SUBTITLE"
author: "YOUR NAME"
date: "`r format(Sys.time(), '%d %B %Y')`"
output:
  bookdown::pdf_document2:
    toc: yes
abstract: "ABSTRACT"
bibliography: references.bib
---
```

- The title, subtitle, author, and abstract can be inserted in the quotes.
- The date is automatically generated based on the system time and date.
- The output for the report is pdf_document2, a **bookdown** format.
- A table of contents will be included in the output.
- The bibliography will be created based on a bib file called references.

Breaking down the setup chunk

```
knitr::opts_chunk$set(
  echo = FALSE,  # hide source code in output
  message = FALSE, # hide messages from code in output
  warning = FALSE  # hide warnings from code in output
  )

library() # load libraries here, including bookdown
```

- We set all the chuns to hide code, warnings, and messages by default, while still showing the code output.
- All necessary libraries are loaded in the chunk.

A typical end of a report would look like this:

```
196
197 - # Conclusion
198
199
     CONCLUSION
200
     \newpage
201
202
203
204 - # (APPENDIX) Appendix {-}
205
206 - # Appendix A
207
208
    APPENDICES
209
210
     \newpage
211
212
213 - # References
214
```

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Breaking down the end of the report

```
# Conclusion
CONCLUSTON
\newpage
# (APPENDIX) Appendix {-}
# Appendix A
APPENDICES
\newpage
# References
```

- Our last report section is the conclusion. We want a page break in between the conclusion and the appendices.
- We also want a page break before our references section. References will be automatically attached to the end of the report by **bookdown**.

Referencing

Each work that you wish to cite will need to be present in the .bib file, where it will be given a unique nickname:

```
## CITED WORKS

dbook{mycitedwork,
    author={Hastie, Trevor and Tibshirani, Robert and Friedman, Jerome},
    title={The Elements of Statistical Learning: Data Mining, Inference, and Prediction},
    publisher={Springer},
    edition={2nd edition},
    year={2009}
}
```

To reference it in your text, you will refer to that nickname again, in the form <code>@mycitedwork</code> or <code>[@mycitedwork]</code>:

```
28
29 Include some citations [@mycitedwork].
30
```

Cross-referencing figures

```
121 Refer to your Figure \@ref(fig:myfigure) in your text.
122
123 ~ ```{r myfigure, fig.cap = "A CAPTION"}
```

- Each chunk with a figure (ex. ggplot graph) will require a name (alphanumeric, excluding spaces and underscores) and a caption ("A CAPTION"). These must be unique to each figure.
- To reference the figure, you can use the code \aref(fig:), with the figure name included in the brackets.

Cross-referencing tables

```
133
134 Refer to your Table \@ref(tab:mytable1) in your text.
135
136 - \``\{r mytable1}
```

- Each chunk with a table will also need a unique name.
- In the body of your code chunk, you also need to give the table a caption. Exactly how this is accomplished will depend on what table formatting package you are using.
- To reference the table, you can use the code \aref(tab:), with the table name included in the brackets

Managing Data Science Projects

Objectives

Before embarking on a project, the team should be able to provide answers to each of the following:

- 1. Regulatory requirements
- 2. Frequency of model updating
- 3. Consequences of being wrong
- 4. Volume of data
- 5. Method for users to access results
- 6. Level of access/connection between the data science team and end users

Skills

- 1. Do the skills currently exist?
- 2. Are the people with the skills available?
- 3. What are the consequences of failing to complete?
- 4. How urgent is the project?
- 5. Would it be difficult to hire a temp?

Data

- 1. Has the team worked with this data before?
- 2. What is the data provenance?
- 3. Would you benefit from more data, and would acquiring it be feasible?
- 4. Do you have permissions required to use the data?
- 5. Will the data be refreshed frequently enough for your model?

Discussion questions

- 1. What do you think is the biggest challenge in communicating data science to non-data scientists?
- 2. How can you balance giving a thorough presentation and making the work accessible?
- 3. What questions would you add to the data science project checklist?