

# PSYC 259: Principles of Data Science

Prof. John M. Franchak

Winter 2021

E-mail: [franchak@ucr.edu](mailto:franchak@ucr.edu)

<https://github.com/PSYC-259-Data-Science>

Office Hours: By Appointment

Class Hours: W 1:30-4:20 PM

<https://ucr.zoom.us/my/johnfranchak>

<https://ucr.zoom.us/j/97985633776>

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## Course Description

Most quantitative courses (importantly) focus on the final steps of data analysis—conducting and understanding statistical tests. However, much of the work in data science is taking raw data, often from multiple, incompatible sources, and processing those data into a usable form. This course will emphasize the importance of robust, documented, and automated workflows for processing data to save time, reduce errors, improve reproducibility, and facilitate collaboration among multiple researchers. We will also spend time on data visualization and communication—an important part of creating, checking, and collaborating on data workflows. We will use the R programming language, Github, and Rmarkdown to work through examples, but the focus is on concepts/best practices that can be applied to any software or programming language. The course is open to students who have little programming experience or experience with R. The goal is for students at all levels of programming experience to set goals to improve their data science skills.

## Course Objectives

The goals of this course are for you to critically analyze and improve your data analysis workflows. Implementing robust, automated procedures for handling data will allow you to:

- Foster open science through increased transparency, reproducibility, and easier data sharing
- Increase the fidelity of your data and analyses by detecting and preventing errors
- Better understand and communicate about your data
- Save time by preventing errors, automating tasks, and reusing code
- Facilitate collaboration with organized and documented workflows

## Course Materials

- The course [Github page](#) has links to project files, readings, and the syllabus
- We will work through examples and complete assignments through a course workspace on [RStudio Online](#). To sign up, follow this [link](#). You will need to create a free user account. We will go over how to access materials on the first day of class.
- Readings from *R for Data Science* (Wickham & Golemund, 2017) are available [online](#).
- PDFs of other course readings referenced below are available on Github.

## Other helpful resources

- [R Markdown: The Definitive Guide](#)
- [papaja: Reproducible APA manuscripts with R Markdown](#)
- RStudio [cheatsheets](#) for base R, data import, ggplot, R Markdown, and other packages.

## Course Policies

TBD

## Assignments and Grading

Grading for the class is S/NC. Your grade will be based on the following:

Component	Weight
Participation	20%
Weekly assignments	20%
Workflow critique	20%
Final project	40%

### *Weekly Assignments*

ALLOW GROUP WORK FOR THE WEEKLY ASSIGNMENTS (MAX OF 3) (SOME PART SHOULD BE CREATIVE/OPEN ENDED)

### *Workflow Critique*

### *Final Project*

## Remote Instruction

## Class Schedule

Readings should be completed prior to each class.

### Week 01, 01/06: Course Goals

- Goals of the class
- Logistics (RStudio/Github)
- Readings and assignments

*R4DS*: [Introduction](#) and [Workflow Basics](#)

*OPTIONAL*: Intro R lessons using `swirl()`

### Week 02, 01/13: Data Workflow

- File organization
- Version control
- Workflow examples
- *SKILLS*: Basic I/O functions, getting help, using Github

*R4DS*: [Data import](#), [Tibbles](#), and [Projects](#)

*OTHER READINGS*: [Github beginner guide](#) and Minimizing mistakes (Rouder, Haaf, & Snyder, 2019)

### Week 03, 01/20: Data Structure

- Data types
- Data organization (variables/observations)
- *SKILLS*: Factors/strings, reshaping/aggregating data

*R4DS*: [Data transformation](#), [Tidy data](#), and [Factors](#)

*OTHER READINGS*: Tidy Data (Wickham, 2014)

### Week 04, 01/27: Automation

- Copy/paste is evil (automating data pipeline)
- Drop-down menus are evil (automating your analyses)
- Machine-readable formats
- *SKILLS*: Writing functions, iteration, vectors

*R4DS*: [Functions](#), [Vectors](#) and [Iteration](#)

### Week 05, 02/03: Error Checking

- Data validation
- Automating visualizations
- Exploratory data analysis
- *SKILLS*: Basic visualizations, handling missing data

*R4DS*: [Data visualization](#) and [Exploratory data analysis](#)

**Week 06, 02/10: Reuse Part 1: Encapsulation and Writing Functions**

- Best option: Use existing APIs and packages
- Next best option: Writing your own general-purpose functions
- Not great option: Writing overly-specific functions that you can never use again
- *SKILLS*: Sourcing functions, parameterization

*OTHER READING*: Technical debt (Suryanarayana, Samarthayam, & Sharma, 2014)

**Week 07, 02/17: Reuse Part 2: Data Sharing, Archiving, and Documentation**

- Code as documentation
- Reuse-minded project management

*OTHER READING*: Care and feeding of data (Goodman et al., 2014)

**Week 08, 02/24: Communication Part 1: Reproducible Reports**

- Reproducible reports
- Preserving programming environment and analyses
- *SKILLS*: R Markdown, papaja, package control

*R4DS*: [R Markdown](#)

*OTHER READING*: Transparency in psychological science (Klein et al., 2018)

**Week 09, 03/03: Communication Part 2: Visualization**

- Communicating through graphical styles
- Interactive/animated plots for data exploration
- *SKILLS*: ggplot and extensions

*R4DS*: [Graphics for communication](#)

*OTHER READING*: Designing graphs for decision-makers (Zacks & Franconeri, 2020), [Chartjunk](#) from (Tufte, 1990, 2001, 2006), and (optional) Graph construction (Witt, 2019).

**Week 10, 03/10: Project Presentations**

- How have you changed your workflow?
- What have you learned about your data?
- What problems are still unresolved?

## References

- Goodman, A., Pepe, A., Blocker, A. W., Borgman, C. L., Cranmer, K., Crosas, M., ... Slavkovic, A. (2014). Ten simple rules for the care and feeding of scientific data. *PLoS Computational Biology*, 10(4), e1003542.
- Klein, O., Hardwicke, T. E., Aust, F., Breuer, J., Danielsson, H., Hofelich Mohr, A., ... Frank, M. C. (2018). A practical guide for transparency in psychological science. *Collabra: Psychology*, 4(1), 1–15.
- Rouder, J. N., Haaf, J. M., & Snyder, H. K. (2019). Minimizing mistakes in psychological science. *Advances in Methods and Practices in Psychological Science*, 2(1), 3–11.
- Suryanarayana, G., Samarthiyam, G., & Sharma, T. (2014). *Refactoring for software design smells: Managing technical debt*. Elsevier Science.
- Tufte, E. R. (1990). *Envisioning information*.
- Tufte, E. R. (2001). *The visual display of quantitative information* (2nd ed.).
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- Wickham, H. (2014). Tidy data. *Journal of Statistical Software*, 59(10), 1–23.
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- Witt, J. K. (2019). Graph construction. *Meta-Psychology*, 3.
- Zacks, J. M., & Franconeri, S. L. (2020). Designing graphs for decision-makers. *Policy Insights from the Behavioral and Brain Sciences*, 7(1), 52–63.