# Intro to R and RStudio

**FNCE5352** 

1/21/2025

#### Matt McDonald - CV

#### Work:













#### **Education:**







#### **Career Path**

Actuarial

IT/Software
Development

Credit
Modeling /
Model Risk
Management

Class Syllabus

## Why Learn To Code

- In Financial Institutions, Excel currently is the common medium for analysis, BUT...
- Coding languages like R and Python have distinct advantages, AND...
- Depending on where you work, coding ability will either be an expectation, OR...
- These skills could set you apart from your peers and make you a SUPER USER

## Advantages of a Code-based Analysis

- **Flexible:** No black box constraints. Access and combine data. Analyze and present it exactly as needed.
- Iterative: Quickly make changes and updates in response to feedback. Share updates with stakeholders
- **Reusable and extensible:** Tackle similar problems in the future. Extend to novel problems as circumstances change.
- Inspectable: Track changes over time, discover errors and audit the approach
- **Reproducible:** Combine with environment and package management, ensure analyses are repeatable and verifiable.

### Python

#### **Strengths**

- **Versatility**: Python is a general-purpose language with a simple, readable syntax, making it suitable for a wide range of applications beyond data analysis, such as web development, automation, and software development.
- Rich Libraries: For data analysis and machine learning, Python offers robust libraries like Pandas, NumPy, Scikitlearn, and TensorFlow.
- Community and Support: Python has a vast and active community, ensuring abundant resources, tutorials, and support.
- **Integration**: Python integrates well with other technologies and can be easily embedded in applications.
- **Scalability**: It's scalable and faster for general-purpose programming.

#### Weaknesses

- Statistical Analysis: While Python has statistical capabilities, they are not as extensive and sophisticated as R's.
- **Memory Usage**: Python's memory consumption can be high for large datasets.
- **Learning Curve**: For those specifically focused on data analysis, Python might introduce unnecessary complexity due to its broad scope.

#### R

#### **Strengths**

- Statistical Analysis and Visualization: R excels in statistical analysis and visualization. It has comprehensive packages like ggplot2 for data visualization and Im for regression, making it superior for statistical modeling.
- **Data Analysis Environment**: R is specifically designed for data analysis, which makes its environment and syntax more suited for statistical analysis tasks.
- Community: R has a strong community in academia and research, offering a wealth of packages and support for statistical analysis.
- Integrated Development Environment (IDE): RStudio provides an excellent IDE for R programming, making data analysis more intuitive.

#### Weaknesses

- General-Purpose Programming: R is less versatile for general software development compared to Python.
- **Speed**: For large datasets, R can be slower than Python, especially if not properly optimized.
- Memory Intensive: Like Python, R can also be memory-intensive, and it traditionally keeps all data in memory.

## Key Differences Between R and Python

- **Purpose and Design Philosophy**: Python is a general-purpose language designed for readability and versatility, while R is specifically designed for statistical analysis and data visualization.
- Data Handling: Python's Pandas library is excellent for data manipulation, offering a more programming-oriented approach, whereas R's data manipulation is deeply rooted in its statistical capabilities.
- Machine Learning and Deep Learning: Python is the go-to language for machine learning and deep learning due to libraries like Scikit-learn and TensorFlow.
- Statistical Analysis: R has a slight edge in advanced statistical modeling and tests, with a vast array of statistical packages.
- Visualization: R's ggplot2 is widely regarded as one of the best data visualization tools.
- **Community and Resources**: Python has a broader user base and community support, given its wide range of applications, while R's community is highly specialized in statistics and data analysis.

## A Gentle Introduction to R

#### Install R & RStudio

• <a href="https://stat545.com/install.html">https://stat545.com/install.html</a>

## R vs Rstudio

R: Engine



#### RStudio: Interface



# R Packages

R: New phone

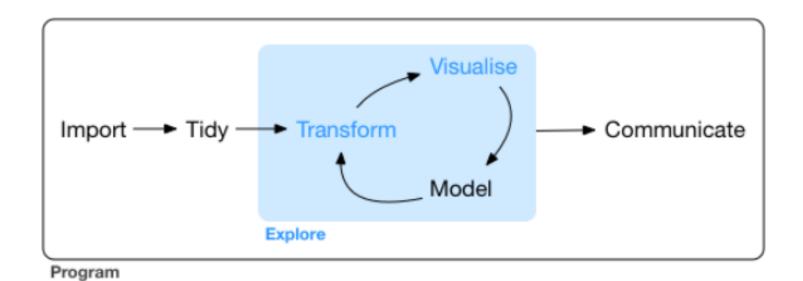


R Packages: Apps you can download





# **Modeling Workflow**



# Introduction to the Tidyverse

## **Project Based Workflow**

#### Why?

- work on more than 1 thing at a time
- collaborate, communicate, distribute
- start and stop

#### How?

- dedicated directory
- RStudio <u>P</u>roject
- Git repo, probably syncing to a remote

## **Project Based Workflow**



```
smell-test.R
wrangle.R
model.R >>> everything.R
make-figs.R
report.Rmd
```

## Good Enough Practices for Data Science

#### Box 1. Summary of practices

- 1. Data management
  - a Save the raw data.
  - b Ensure that raw data are backed up in more than one location.
  - c Create the data you wish to see in the world.
  - d Create analysis-friendly data.
  - e Record all the steps used to process data.
  - f Anticipate the need to use multiple tables, and use a unique identifier for every record.
  - g Submit data to a reputable DOI-issuing repository so that others can access and cite it.

#### 2. Software

- a Place a brief explanatory comment at the start of every program.
- b Decompose programs into functions.
- c Be ruthless about eliminating duplication.
- d Always search for well-maintained software libraries that do what you need.
- e Test libraries before relying on them.
- f Give functions and variables meaningful names.
- g Make dependencies and requirements explicit.
- h Do not comment and uncomment sections of code to control a program's
- i Provide a simple example or test data set.
- j Submit code to a reputable DOI-issuing repository.
- Collaboration
- a Create an overview of your project.
- b Create a shared "to-do" list for the project.
- c Decide on communication strategies.
- d Make the license explicit.
- e Make the project citable.

#### 4. Project organization

- a Put each project in its own directory, which is named after the project.
- b Put text documents associated with the project in the doc directory.
- c Put raw data and metadata in a data directory and files generated during cleanup and analysis in a results directory.
- d Put project source code in the src directory.
- e Put external scripts or compiled programs in the bin directory.
- f Name all files to reflect their content or function.
- 5. Keeping track of changes
  - a Back up (almost) everything created by a human being as soon as it is created.
  - b Keep changes small.
  - c Share changes frequently.
  - d Create, maintain, and use a checklist for saving and sharing changes to the project.
  - e Store each project in a folder that is mirrored off the researcher's working machine.
  - $f\,$  Add a file called CHANGELOG . txt to the project's docs subfolder.
  - g Copy the entire project whenever a significant change has been made.
  - h Use a version control system.
- 6. Manuscripts
  - a Write manuscripts using online tools with rich formatting, change tracking, and reference management.
  - b Write the manuscript in a plain text format that permits version control.

## Getting to Know GIT

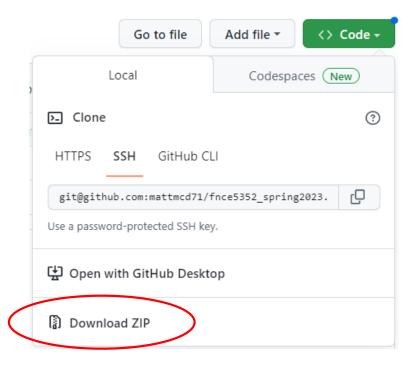
https://happygitwithr.com/

### Class Repo:

https://github.com/mattmcd71/ fnce5352\_spring2023

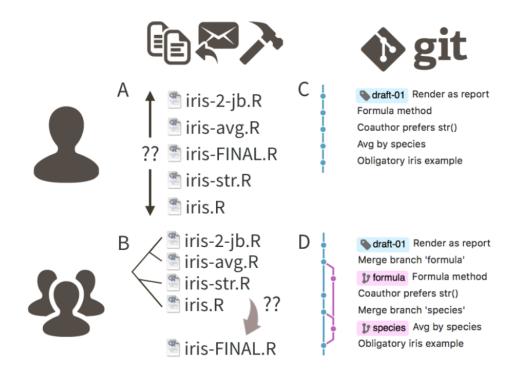
### How to Access the Class GIT Repo

Download ZIP



- Clone
  - HTTPS or SSH
  - SSH will require a setup of your SSH keys
    - See happygitwithr page for complete instructions on that

# Why GIT?



#### GIT vs GitHub

