Welcome to instats

The Session Will Begin Shortly

START

Statistics in R with Tidyverse

Session 1: Introduction to R: Basics and Advanced Techniques

Welcome and Introduction

Dr. Chester Ismay

- PhD in Statistics
- Worked in academia, online education, corporate training, tech bootcamps, and independent consulting
- Currently,
 - Faculty Member in Data Analytics,
 Portland State University
 - Vice President of Data and Automation,
 MATE Seminars
 - Freelance data scientist and educator
- Fun Fact: Slept a night or eaten a meal in all 50 US states



Course Learning Objectives

By the end of this course, you will be able to

- Perform data wrangling techniques in R via the tidyverse
- Develop skills in data visualization with ggplot2
- Apply fundamental concepts of statistical inference with infer
- Build and interpret regression models with moderndive
- Integrate Theory-Based and Simulation-Based Approaches



Agenda

Day 1: Working with Data in R - Explore, Visualize, Wrangle, Import

- Session 1: Introduction to R Basics and Advanced Techniques
- Session 2: Data Visualization using ggplot2
- Session 3: Data Wrangling and Tidy Data



Introduction to R and RStudio

- R: programming language mainly for statistical computing and data analysis
- RStudio: IDE
- R vs RStudio

R: Engine



RStudio: Dashboard





Installing R and RStudio

- R: https://cloud.r-project.org/
- RStudio: https://posit.co/download/rstudio-desktop/
- Download and install for your operating system



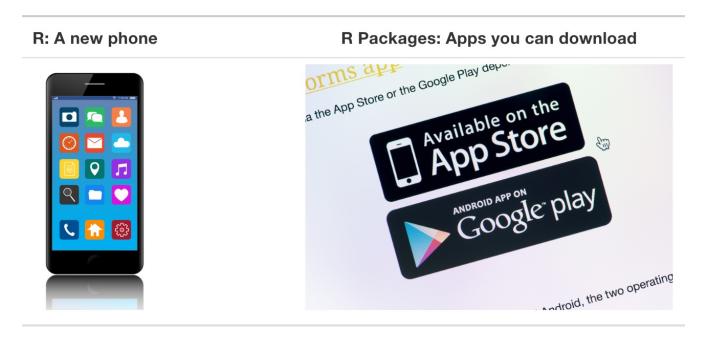
Coding in R

- Commands entered as code in the Console or via scripts.
- Key concepts include objects, vectors, and data types
- Conditional statements and functions help perform tasks
- Learning to code takes frequent practice, but it is one of the most rewarding things you can do!



Using R packages

- Extend R's capabilities with additional functions and/or datasets
- First install the package with install.packages()
- Load the package using library()





Exploring Data in R with RStudio

- Data frames are like tables with rows and columns
- Use View(), glimpse(), or kable() to inspect
- The \$ operator extracts columns from data frames
- Identification versus measurement variables/columns



Demo & Exercises

Q&A

STOP

Welcome to instats

The Session Will Begin Shortly

START

Statistics in R with Tidyverse

Session 2: Data Visualization using ggplot2

Introduction to Data Visualization

- Insights that raw data alone cannot provide
- ggplot2 package based on Grammar of Graphics by Leland Wilkinson
- Visualizations help to identify outliers, distributions, and relationships



Grammar of Graphics

- A statistical graphic maps data variables to aesthetic attributes
- Key components:
 - 1. data: The dataset
 - 2. geom: The geometric objects (points, lines, bars)
 - 3. aes: Aesthetic attributes like position, color, shape, size
- Create visualizations by layering these components in ggplot()

The Five Named Graphs

- Essential tools for data visualization
- Scatterplots, linegraphs, histograms, boxplots, and barplots
 - Each type works best for different data relationships and distributions
 - Goal is to uncover trends, patterns, and outliers in data



Scatterplots

- Display relationships between two numerical variables
- Using geom_point()
- Customizing points (color, shape, size)
- Tip: Handling overplotting
 - alpha transparency
 - jittering with geom_jitter()

Linegraphs

- Display trends over time or relationships between two sequential variables
- Use geom_line()
- Commonly used for time-based data (hours, days, weeks, etc.)
- **Tip**: Avoid using linegraphs when the x-axis variable has no inherent order

Histograms

- Display the distribution of a single numerical variable
- Use geom_histogram()
- Visualize data spread, center, and frequency of values
- **Tip**: Adjust bin width or number of bins for better data representation



Boxplots

- Summarize numerical data using quartiles and medians
- Use geom_boxplot()
- Effective for identifying data spread and detecting outliers
- **Tip**: Use boxplots for comparing distributions across groups



Barplots

- Display the distribution of a categorical variable's frequencies
- Use geom_bar()or geom_col()
- Barplots are ideal for comparing frequencies of categories or groups
- Tip: Use geom_bar() for raw (uncounted) data and geom_col() for pre-counted data



Demo & Exercises

Q&A

STOP

Welcome to instats

The Session Will Begin Shortly

START

Statistics in R with Tidyverse

Session 3: Data Wrangling and Tidy Data

Data Wrangling

- Overview of the tidyverse
- Importance of Data Wrangling in Research
- Key Packages: tidyr, dplyr



Filter Rows

- Use filter()to select rows based on conditions
- Focuses on rows
 - Similar to slice() which selects rows by position, not condition
- Combine conditions with & (AND) and | (OR)
- **Tip**: Use != to filter out specific values

Mutate Columns

- Use mutate()to create new columns based on existing ones
- Adds new columns; unlike transmute(), which drops all other columns
- Useful for transforming or calculating new values from existing data
- Tip: Can also be used to modify an existing column



Summarize Data

- Use summarize() to calculate summary statistics
- Reduces data to a single row or value; unlike mutate() which keeps
 original data format
- **Tip**: Can handle missing data with na.rm = TRUE

Group By and Summarize

- Use group_by() to split data into groups, then apply summarize()
- Organizes data into groups; unlike arrange(), which only sorts data
- Combine group_by() with summarize() to create grouped statistics
- **Tip**: ungroup() data after grouping if further processing is needed



Arrange Data

- Use arrange() to sort rows based on specific columns
- Sorts data; unlike filter() which selects rows without changing order
- **Tip**: Sort in ascending order by default; use desc() for descending



Select Columns

- Use select() to choose specific columns
- Different from mutate(), which adds new columns
- Can deselect columns using (e.g., select(-year))
- Tip: Use helpers like starts_with() to select columns by pattern

Tidy Data

- "Tidy" data means
 - each variable has its own column
 - each observation has its own row
 - each kind of thing you're observing is its own table
- Different from "wide" data in that it is often longer to be tidy
- Tip: Use pivot_longer() to convert wide data for easier analysis

Pipe Operator (|>)

- Use the pipe operator to chain multiple operations together
- Chains operations unlike using nested functions, which is harder to read
- Often improves workflows
- **Tip**: Think of |> as "then" to improve readability



Demo & Exercises

Q&A

STOP