

Data Visualization Assignment

Your Name

Getting Started

This assignment will test your ability to use the tidyverse to perform a complete data analysis workflow: from data wrangling to publication-quality visualization. You will be working with datasets from the psych package, which is commonly used in psychological research.

Your goal for each problem is to write a single, clean dplyr pipe (%>%) that creates the requested data, and then pipe that data directly into a ggplot() call to create a publication-quality plot.

This assignment uses datasets from the **psych** package. For each problem: - Briefly **describe your wrangling** (1–2 sentences). - Complete the r code in the # TODO sections. - Use tidy verbs (**filter**, **mutate**, **select**, **summarise**, **pivot_***). - Choose **appropriate plots** (when asked) that match variable types and questions. - Be sure to polish chart elements! For example, label numeric levels when necessary (e.g., 1 = Male, 2 = Female) - Be sure to comment your code!

Install/Load Required Packages

```
if (!require(psych)) install.packages("psych")
if (!require(tidyverse)) install.packages("tidyverse")
if (!require(corrplot)) install.packages("corrplot")
if (!require(ggplot2)) install.packages("ggplot2")

library(psych)
library(tidyverse)
library(corrplot)
library(ggplot2)
```

Part 1: Exploring Distributions (bfi Dataset)

The bfi dataset contains 25 personality items from the Big Five Inventory (BFI), along with gender, education, and age.

1) Distribution of Agreeableness

Data: 'psych::bfi'

Goal Create a density plot showing the distribution of the mean “Agreeableness” score.

Wrangling: Start with the bfi data. Use mutate() to create a new agreeableness column by calculating the row mean of the five agreeableness items (A1, A2, A3, A4, A5). Make sure to handle missing values (na.rm = TRUE).

Plotting: Pipe the wrangled data into ggplot(). * Use geom_density() to plot the agreeableness distribution.

- Add a fill to the geom (e.g., “skyblue”) and set alpha = 0.7.

- Add appropriate labels using `labs()` (title, subtitle, x-axis).
- Use `theme_minimal()`.

```
#YOUR CODE HERE
```

2) Neuroticism by Gender

Data: `psych::bfi`

Goal Create a boxplot comparing the distribution of mean “Neuroticism” scores for males and females.

Wrangling: Start with `bfi`. Create a neuroticism column (mean of N1...N5). Then, `filter()` out any NA values for gender. Finally, `mutate()` the gender column into a factor with descriptive labels (e.g., 1 = “Male”, 2 = “Female”).

Plotting: Use `geom_boxplot()` to map gender to the x-axis and neuroticism to the y-axis.

- Add `fill = gender` to the `aes()` mapping.
- Add `labs()` (e.g., title, subtitle, x=“Gender”, y=“Mean Neuroticism Score”).
- Use `theme_minimal()` and remove the legend (`theme(legend.position = “none”)`).

```
#YOUR CODE HERE
```

3) Conscientiousness by Education Level

Data: `psych::bfi`

Goal: Create a column chart showing the average “Conscientiousness” score by education level and showing the 95% confidence interval.

Wrangling: Create a conscientiousness column (mean of C1...C5). `filter()` out NA education levels. `group_by()` education. `summarise()` to find the `mean_c_score`.

Plotting: Use `geom_col()` to map education to x and `mean_c_score` to y. (Note: education should be a factor to prevent ggplot from treating it as a continuous number. Use `factor(education)` in your `aes()` mapping).

- Add `fill = factor(education)` to `aes()`.
- Add `labs()` (title, subtitle, x=“Education Level”, y=“Mean Conscientiousness”).
- Use `geom_errorbar()` to display the 95% CI for each mean.
- Use `scale_x_discrete()` to provide meaningful labels for the education levels (1=HS, 2=Finished HS, 3=Some College, 4=College Grad, 5=Grad School).

```
# Example use of scale_x_discrete()
#scale_x_discrete(
#  limits = c("3", "2", "1"), # limits controls order of factors
#  labels = c("Subcompact", "Compact", "Midsize") # create labels
# )
#YOUR CODE HERE
```

Part 2: Visualizing Relationships (sat.act Dataset)

4) SAT Verbal vs. SAT Quantitative

Data: `psych::sat.act`

Goal: Create a scatterplot to visualize the relationship between SAT Verbal (SATV) and SAT Quantitative (SATQ) scores.

Wrangling: None needed, just pipe `sat.act` directly into `ggplot()`.

Plotting: Use `geom_point()` to map SATV to x and SATQ to y. Set `alpha = 0.5` to handle overplotting.

- Add a linear regression line using `geom_smooth`.
- Add `labs()` (e.g., title, subtitle, x="SAT Verbal", y="SAT Quantitative").
- Use `theme_bw()`.

#YOUR CODE HERE

5) Faceting the SAT Relationship by Gender

Data: `psych::sat.act`

Goal: Expand on Problem 4. Does the relationship between SATV and SATQ differ by gender?

Wrangling: `mutate()` the gender column to a factor with labels (1="Male", 2="Female").

Plotting: Create the same plot as Problem 4 (`geom_point` + `geom_smooth`).

- Add `color = gender` to the `aes()` in the main `ggplot()` call.
- Add `facet_wrap(~ gender)` to create two separate plots.
- Use `labs()` and `theme_bw()`.

#YOUR CODE HERE

6) ACT Scores by Education

Data: `psych::sat.act`

Goal: Create a bar chart showing the average ACT score by education level, ordered from highest to lowest.

Wrangling: `group_by(education)`, `summarise(mean_act = mean(ACT, na.rm = TRUE))`. Use `mutate()` to `reorder()` the education factor by `mean_act` (e.g., `education = reorder(factor(education), mean_act)`).

Plotting: Use `geom_col()` to map education to x and `mean_act` to y.

- Add `fill = education` to the `aes()` and remove the legend.
- Add `labs()` and `theme_minimal()`.

#YOUR CODE HERE

7) BFI: Trait Tradeoffs (E vs N)

Data: `psych::bfi`

Goal: Compute mean **Extraversion** and **Openness** and visualize their **bivariate relationship**. Pick one: **2D density (filled)**, or **scatter + ggMarginal**. Briefly interpret the pattern.

Wrangling: `summarize()` (e.g., mean of E1...E5).

Plotting: Use `geom_density_2d_filled()` or `ggMarginal()`.

#YOUR CODE HERE

8) Visualizing Intercorrelations Among Cognitive Ability Tests

Data: `ability.cov`

Goal: Convert the `ability.cov` covariance matrix in `$cov` to a correlation matrix using `cov2cor()` and create a correlogram using `corrplot`.

- The top half of the matrix should represent the correlations using shapes or colors and the bottom half of the matrix should contain the values.
- Change the default colors from the default palette

#YOUR CODE HERE

9) SAT & ACT: Relationships

Data: `psych::sat.act` (columns typically SATV, SATQ, ACT)

Goal: Tidy column names, then visualize relationships:

Plotting: Set up the `aes()` to plot SATV vs SATQ with ACT as size.

- Add a smooth trend. Comment on any nonlinearity.
- Add labels
- Add a theme

#YOUR CODE HERE

10) Harman74: Correlogram with Numbers

Data: `psych::Harman74.cor` (correlation matrix + N)

Goal: Plot a **mixed correlogram**: symbols/colors on one triangle and numbers on the other.

`data(Harman74.cor)`

#YOUR CODE HERE

Part 3: Reshaping and Comparing (msq and bfi Datasets)

The `msq` (Motivation State Questionnaire) dataset is in a “wide” format, with different mood scales as columns.

11) Tidying the msq Data

Data: `psych::msq`

Goal The msq dataset is wide. Use `pivot_longer()` to make it tidy.

Wrangling: Use `pivot_longer()` to gather all the mood scale columns (from Active to Sleepy) into two new columns: `mood_scale` and `score`.

Plotting: No plot needed. Just show the `head()` of your new tidy `msq_tidy` data frame.

#YOUR CODE HERE

12) Faceted Density of Moods

Data: Your `msq_tidy` data from #9

Goal: Using your `msq_tidy` data, create faceted density plots for the “Calm”, “Energetic”, and “Tense” mood scales.

Wrangling: Start with `msq_tidy`. `filter()` so that `mood_scale` is one of “Calm”, “Energetic”, or “Tense”.

Plotting: Use `geom_density(aes(x = score, ...))`.

- Add `facet_wrap(~ mood_scale)`.
- Use `labs()` and `theme_minimal()`. Remove the legend.

#YOUR SCORE HERE

13) All BFI Traits by Gender

Data: `psych::bfi`

Goal: Create a grouped bar chart comparing the mean scores for all five BFI traits (A, C, E, N, O) grouped by gender.

Wrangling: This will be a multi-step process.

- Start with `bfi`.
- `mutate()` to create 5 new mean-score columns (`agree`, `consc`, `extra`, `neuro`, `open`).
- `select()` only the gender column and your 5 new score columns.
- `pivot_longer()` to gather the 5 score columns into `trait` and `score`.
- `group_by(gender, trait)` and `summarise(mean_score = mean(score, na.rm = TRUE))`.
- `filter()` out NA genders and `mutate(gender = factor(gender, labels = c(“Male”, “Female”)))`.

Plotting:

- `ggplot(aes(x = trait, y = mean_score, fill = gender))`.
- Use `geom_col(position = “dodge”)`.
- Use `geom_errorbar()` to display your choice of measure of uncertainty.
- Use `labs()` (`title`, `subtitle`, `x=“Personality Trait”`, `y=“Mean Score”`).
- Use `scale_fill_brewer(palette = “Set1”)` for a nice color palette.
- Use `theme_minimal()`.

```
#YOUR CODE HERE
```

Part 4: Open Ended

14) BFI: Age Trends in Traits

Data: `psych::bfi`

Goal: For any **two traits** of your choice, examine how scores change with **age**.

```
#YOUR CODE HERE
```

15) SAT & ACT: Choose the View

Data: `psych::sat.act`

Goal: Decide on an **appropriate plot** to communicate how **SATQ** relates to **ACT**, conditioned on **SATV** (e.g., faceting SATV bins or using color gradients). Justify your choice.

```
#YOUR CODE HERE
```

16) BFI: Trait Interactions by Gender

Data: `psych::bfi`

Goal: Investigate whether the relationship between two traits (you choose) **differs by gender**.

```
#YOUR CODE HERE
```

17) “Tell a Data Story”

Use your own data. Ask a meaningful question, **tidy** the data, and produce **1–2 plots** that answer it. Write a short caption (3–5 sentences) explaining your findings and why the charts fit the question.

If you do not have any of your own data, you can pick a dataset from `psych` (`bfi`, `msq`, `msqR`, `sat.act`, `ability.cov`, `Harman74.cor`), but your analysis must be sufficiently different from those covered in this assignment!.

Your “story” should include:

- At least two continuous DVs and two categorical IVs with 2 or more levels
- Publication quality plots of the data distributions (use `aes()` and/or faceting)
- Publication quality plots of the data summaries with estimates of uncertainty (e.g., `geom_col`)
- All appropriate labels

```
# TODO: choose data, tidy, visualize, and caption
```

About the datasets in the `psych` package

Here’s a summary of some prominent datasets within the `psych` package:

bfi (Big Five Inventory):

- Content: Contains item responses from 2800 participants on 25 personality self-report items designed to measure the Big Five personality traits (Agreeableness, Conscientiousness, Extraversion, Neuroticism, Openness). Also includes demographic variables like gender, education, and age.

epi.bfi:

- Content: Data from 231 participants who completed both the Eysenck Personality Inventory (EPI - measuring Extraversion and Neuroticism) and the Big Five Inventory (bfi items).

sat.act:

- Content: Self-reported scores on the SAT Verbal (SATV), SAT Quantitative (SATQ), and ACT for 700 students. Also includes gender and education level.

How to Explore All Datasets in psych: You can see the full list of datasets available in the currently loaded version of the psych package and access their documentation directly within R:

```
# Make sure the psych package is loaded
library(psych)

# List all datasets in the psych package
data(package = "psych")

# Get detailed help documentation for a specific dataset (e.g., bfi)
?bfi
help(bfi)
```