Data Manipulation with dplyr and tidyr

1 Introduction

The dplyr and tidyr packages are essential tools in R for efficient data manipulation and tidying. This document introduces key functionalities of these packages, including filtering, summarizing, grouping, and reshaping data. Each section includes explanations, examples, best practices, and exercises.

2 Filtering Data with dplyr

2.1 Overview

Filtering involves selecting rows from a dataset based on specific conditions. The filter() function in dplyr allows you to extract subsets of data.

2.2 Examples

2.3 Best Practices

- Use logical operators (&, |, !) for complex conditions.
- Avoid filtering directly in your original dataset; store the filtered result in a new variable.

3 Summarizing Data with dplyr

3.1 Overview

The summarize() function computes summary statistics (mean, median, sum, etc.) for datasets.

3.2 Examples

3.3 Best Practices

- Use na.rm = TRUE to handle missing values in summary functions.
- Combine group_by() with summarize() for grouped summaries.

4 Grouping Data with dplyr

4.1 Overview

The group_by() function divides a dataset into groups, allowing for grouped operations.

4.2 Examples

```
# Group by Species and compute mean Sepal.Length
grouped_stats <- iris %>%
  group_by(Species) %>%
  summarize(mean_sepal = mean(Sepal.Length))
```

4.3 Best Practices

- Always use ungroup() after grouped operations to avoid unintended behavior.
- Combine group_by() with other dplyr functions like mutate() or summarize().

5 Reshaping Data with tidyr

5.1 Overview

The tidyr package provides functions to transform data between wide and long formats.

5.2 Examples

```
# Example dataset: pivot_wider
library(tidyr)
data <- data.frame(
  id = c(1, 1, 2, 2),
   variable = c("A", "B", "A", "B"),
  value = c(10, 20, 15, 25)
)

# Wide format
wide_data <- pivot_wider(data, names_from = variable,
    values_from = value)

# Long format
long_data <- pivot_longer(wide_data, cols = A:B, names_
    to = "variable", values_to = "value")</pre>
```

5.3 Best Practices

- Ensure the dataset is in the desired structure (wide or long) before analysis.
- Use gather() and spread() (deprecated) for backward compatibility.

6 Best Practices for Data Manipulation

- Always inspect your data before and after applying transformations.
- Use the pipe operator (%>%) for readable and efficient workflows.
- Document your data manipulation steps for reproducibility.

7 Practice Exercises

7.1 Datasets

Use datasets from the datasets package in R (iris, mtcars, airquality, etc.) for these exercises.

7.2 Exercises

- 1. Filter rows in the mtcars dataset where mpg is greater than 20.
- 2. Compute the mean of Sepal.Length for each species in the iris dataset.
- 3. Group the mtcars dataset by the number of cylinders (cyl) and calculate the average horsepower (hp) for each group.
- 4. Reshape the airquality dataset to a long format with month as a key column.
- 5. Use the group_by() and summarize() functions to compute the maximum wind speed for each month in the airquality dataset.
- 6. Filter the iris dataset for rows where Sepal. Width is less than 3 and Species is not setosa.
- 7. Reshape the mtcars dataset into a long format and back to wide format using pivot_longer() and pivot_wider().
- 8. Create a new column in the iris dataset that contains the ratio of Sepal.Length to Sepal.Width.
- 9. Use summarize() to compute the total number of observations for each species in the iris dataset.
- 10. Combine filtering and summarizing to calculate the average mileage (mpg) for cars with more than 100 horsepower in the mtcars dataset.

8 Conclusion

The dplyr and tidyr packages provide powerful tools for efficient data manipulation in R. Mastering these functions will significantly enhance your data analysis capabilities.