

Introduction to R for Data Management and Analysis

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Session 4

Announcements

- Additional topics to cover
 - Reshaping your data
- Piping operator `magrittr::%>%` or `|>` (new; R > 4.2)
 - Takes the LHS as input to the RHS
 - Readable
 - Allows easy command chaining

Outline for today

- Review exercises
- Using dplyr to combine data manipulations
- Reshaping data
- Plotting in base R
- Exploratory Data Analysis
- Intro to ggplot2
- Saving graphics

But first, a quote...

The data may not contain the answer. The combination of some data and an aching desire for an answer does not ensure that a reasonable answer can be extracted from a given body of data.
-John Tukey

Review

- Exercises 1 - 3

Using the *nycflights13* dataset

```
library(nycflights13); library(dplyr)
flights |> group_by(carrier) |>
  summarise(avg_depdelay = mean(dep_delay, na.rm = TRUE),
            count = n()) |> left_join(airlines) |>
  arrange(avg_depdelay) |> head()
```

```
## # A tibble: 6 x 4
##   carrier avg_depdelay count name
##   <chr>      <dbl> <int> <chr>
## 1 US          3.78 20536 US Airways Inc.
## 2 HA          4.90   342 Hawaiian Airlines Inc.
## 3 AS          5.80   714 Alaska Airlines Inc.
## 4 AA          8.59 32729 American Airlines Inc.
## 5 DL          9.26 48110 Delta Air Lines Inc.
## 6 MQ         10.6 26397 Envoy Air
```

Reshaping data

- Useful to prepare data for visualizations
- long vs wide
- long format - multiple observations per row (survival data)
- wide format - a single observation per row

Reshaping using *pivot_wider*

```
library(tidyr); library(tidycensus)
```

```
us_rent_income
```

```
## # A tibble: 104 x 5
```

```
##   GEOID NAME      variable estimate   moe
##   <chr> <chr>      <chr>      <dbl> <dbl>
## 1 01     Alabama income    24476  136
## 2 01     Alabama rent       747    3
## 3 02     Alaska income    32940  508
## 4 02     Alaska rent      1200   13
## 5 04     Arizona income    27517  148
## 6 04     Arizona rent       972    4
## 7 05     Arkansas income    23789  165
## 8 05     Arkansas rent       709    5
## 9 06     California income    29454  109
## 10 06     California rent      1258    3
```


Reshaping using `pivot_longer`

```
relig_income
```

```
## # A tibble: 18 x 11
```

```
##   religion `<$10k` ` $10-20k` ` $20-30k` ` $30-40k` ` $40-50k`
```

```
##   <chr>      <dbl>      <dbl>      <dbl>      <dbl>      <dbl>
```

```
## 1 Agnostic      27        34        60        81        76
```

```
## 2 Atheist       12        27        37        52        35
```

```
## 3 Buddhist      27        21        30        34        33
```

```
## 4 Catholic     418       617       732       670       638
```

```
## 5 Don't know    15        14        15        11        10
```

```
## 6 Evangelical  575       869      1064       982       881
```

```
## 7 Hindu         1         9         7         9         11
```

```
## 8 Historical    228       244       236       238       197
```

```
## 9 Jehovah's    20        27        24        24        21
```

```
## 10 Jewish       19        19        25        25        30
```

```
## 11 Mainline     289       495       619       655       651
```

```
## 12 Mormon       20        40        48        51        54
```

Long dataset

```
relig_income |> pivot_longer(-religion, names_to = "income",  
  values_to = "count") |> head()
```

```
## # A tibble: 6 x 3  
##   religion income    count  
##   <chr>    <chr>    <dbl>  
## 1 Agnostic <$10k      27  
## 2 Agnostic $10-20k    34  
## 3 Agnostic $20-30k    60  
## 4 Agnostic $30-40k    81  
## 5 Agnostic $40-50k    76  
## 6 Agnostic $50-75k   137
```

- `-religion` - don't include religion when reshaping
- `names_to` - create an income variable out of the columns
- `values_to` - cell values are counts

group_by operations

- Allow users to group different levels of categories of 1 or more variables
- Efficient summarization

Using group_by (1)

```
relig_income |> pivot_longer(-religion,  
  names_to = "income", values_to = "count") |>  
  group_by(income) |> summarise(totals = sum(count))
```

```
## # A tibble: 10 x 2
```

	income	totals
	<chr>	<dbl>
## 1	<\$10k	1930
## 2	>150k	2608
## 3	\$10-20k	2781
## 4	\$100-150k	3197
## 5	\$20-30k	3357
## 6	\$30-40k	3302
## 7	\$40-50k	3085
## 8	\$50-75k	5185
## 9	\$75-100k	3990
## 10	Don't know/refuse	6181

Using group_by (2)

```
relig_income |> pivot_longer(-religion,  
  names_to = "income", values_to = "count") |>  
  group_by(religion) |> summarise(totals = sum(count))
```

```
## # A tibble: 18 x 2
```

##	religion	totals
##	<chr>	<dbl>
##	1 Agnostic	826
##	2 Atheist	515
##	3 Buddhist	411
##	4 Catholic	8054
##	5 Don't know/refused	272
##	6 Evangelical Prot	9472
##	7 Hindu	257
##	8 Historically Black Prot	1995
##	9 Jehovah's Witness	215
##	10 Jewish	680

Plotting and Graphing

- Exploratory Data Analysis
- Base R graphics
- Intro ggplot2
- Saving graphics

Plotting systems in R

- 'Base' graphics
- lattice
- ggplot2

Exploratory Data Analysis

- Informal representation data
- Looking for patterns, outliers, etc.
- Get familiar with your data!

Types of graphs

- Histogram
- Scatterplot
 - Scatterplot matrix
- Boxplots / dotplots (ggplot2)
- Violin plots (ggplot2)
- Q-Q plots
- Mosaic plots
- and many more!

ggplot2 - Grammar of Graphics

- Different syntax
 - Slight learning curve
- Plots are built in layers
- Operations add layers to the plot

Saving outputs

- Common formats for saving plots:
 - PDF
 - SVG
 - PNG/TIFF
- but there are more
- ggsave

Output sandwich

- Start with a function `pdf`, `png`, `jpeg`, etc.



- End in `dev.off()` for closing the graphics window

Saving plots in ggplot2

- ggplot2 graphics require a `print` (or a call) before it gets rendered in the file.
- `ggsave` - added to make it easier to save plotting objects

Recommended resources

- Fundamentals of Data Visualization
 - Claus O. Wilke
- R Graphics Cookbook
 - Winston Chang