

Lesson 6 Chapter 12 & 13

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3/9/2022

0. Load libraries

```
knitr::opts_chunk$set(echo = TRUE)

library(tidyverse)
library(rlang)

# Load NYC flight dataset
library(nycflights13)
```

\$12.3 Pivoting

- `pivot_longer(data, cols, names_to = "name", values_to = "value")`

```
relig_income
```

```
## # A tibble: 18 x 11
##   religion ' <$10k' '$10-20k' '$20-30k' '$30-40k' '$40-50k' '$50-75k' '$75-100k'
##   <chr>      <dbl>      <dbl>      <dbl>      <dbl>      <dbl>      <dbl>      <dbl>
## 1 Agnostic    27         34         60         81         76        137        122
## 2 Atheist     12         27         37         52         35         70         73
## 3 Buddhist    27         21         30         34         33         58         62
## 4 Catholic   418        617        732        670        638       1116       949
## 5 Don't k~    15         14         15         11         10         35         21
## 6 Evangel~   575        869       1064       982        881       1486       949
## 7 Hindu        1          9          7          9         11         34         47
## 8 Histori~   228        244        236        238        197        223        131
## 9 Jehovah~    20         27         24         24         21         30         15
## 10 Jewish     19         19         25         25         30         95         69
## 11 Mainlin~   289        495        619        655        651       1107       939
## 12 Mormon     29         40         48         51         56        112         85
## 13 Muslim      6          7          9         10          9         23         16
## 14 Orthodox   13         17         23         32         32         47         38
## 15 Other C~    9          7         11         13         13         14         18
## 16 Other F~   20         33         40         46         49         63         46
## 17 Other W~    5          2          3          4          2          7          3
## 18 Unaffil~  217        299        374        365        341        528       407
```

```
## # ... with 3 more variables: $100-150k <dbl>, >150k <dbl>,
## #   Don't know/refused <dbl>
```

```
relig_income %>%
  pivot_longer(cols = !religion, names_to = "income", values_to = "count")
```

```
## # A tibble: 180 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
## 7 Agnostic $75-100k   122
## 8 Agnostic $100-150k  109
## 9 Agnostic >150k     84
## 10 Agnostic Don't know/refused 96
## # ... with 170 more rows
```

- pivot_wider(data, names_from, values_from)

```
fish_encounters
```

```
## # A tibble: 114 x 3
##   fish station seen
##   <fct> <fct>   <int>
## 1 4842 Release     1
## 2 4842 I80_1      1
## 3 4842 Lisbon     1
## 4 4842 Rstr       1
## 5 4842 Base_TD    1
## 6 4842 BCE       1
## 7 4842 BCW       1
## 8 4842 BCE2      1
## 9 4842 BCW2      1
## 10 4842 MAE       1
## # ... with 104 more rows
```

```
fish_encounters %>%
  pivot_wider(names_from = station, values_from = seen)
```

```
## # A tibble: 19 x 12
##   fish Release I80_1 Lisbon Rstr Base_TD BCE BCW BCE2 BCW2 MAE MAW
##   <fct>   <int> <int>   <int> <int>   <int> <int> <int> <int> <int> <int> <int>
## 1 4842      1     1     1     1     1     1     1     1     1     1     1
## 2 4843      1     1     1     1     1     1     1     1     1     1     1
## 3 4844      1     1     1     1     1     1     1     1     1     1     1
## 4 4845      1     1     1     1     1     NA     NA     NA     NA     NA     NA
## 5 4847      1     1     1     NA     NA     NA     NA     NA     NA     NA     NA
```

## 6 4848	1	1	1	1	NA	NA	NA	NA	NA	NA	NA
## 7 4849	1	1	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 8 4850	1	1	NA	1	1	1	1	NA	NA	NA	NA
## 9 4851	1	1	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 10 4854	1	1	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 11 4855	1	1	1	1	1	NA	NA	NA	NA	NA	NA
## 12 4857	1	1	1	1	1	1	1	1	1	NA	NA
## 13 4858	1	1	1	1	1	1	1	1	1	1	1
## 14 4859	1	1	1	1	1	NA	NA	NA	NA	NA	NA
## 15 4861	1	1	1	1	1	1	1	1	1	1	1
## 16 4862	1	1	1	1	1	1	1	1	1	NA	NA
## 17 4863	1	1	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 18 4864	1	1	NA	NA	NA	NA	NA	NA	NA	NA	NA
## 19 4865	1	1	1	NA	NA	NA	NA	NA	NA	NA	NA

- `separate(data, col, into, sep)`

```
df <- data.frame(x = c(NA, "x.y", "x.z", "y.z"))
df
```

```
##      x
## 1 <NA>
## 2  x.y
## 3  x.z
## 4  y.z
```

```
df %>% separate(x, c("A", "B"))
```

```
##      A      B
## 1 <NA> <NA>
## 2    x    y
## 3    x    z
## 4    y    z
```

```
# use regular expression in separator.
df %>% separate(x, c("A", "B"), sep = '\\.')
```

```
##      A      B
## 1 <NA> <NA>
## 2    x    y
## 3    x    z
## 4    y    z
```

- `unite()`

```
df_sep <- df %>% separate(x, c("A", "B"))
```

```
# Different separators.
df_sep %>% unite(x, A, B)
```

```
##          x
## 1 NA_NA
## 2  x_y
## 3  x_z
## 4  y_z
```

```
# don't need to use regular express for the separator
df_sep %>% unite(x, A, B, sep = ".")
```

```
##          x
## 1 NA.NA
## 2  x.y
## 3  x.z
## 4  y.z
```

\$13.4 Mutating joins

Compare the joins in dplyr and SQL.

dplyr	SQL
inner_join(x, y, by = "z")	SELECT * FROM x INNER JOIN y USING (z)
left_join(x, y, by = "z")	SELECT * FROM x LEFT OUTER JOIN y USING (z)
right_join(x, y, by = "z")	SELECT * FROM x RIGHT OUTER JOIN y USING (z)
full_join(x, y, by = "z")	SELECT * FROM x FULL OUTER JOIN y USING (z)

\$13.7 Set operations

- intersect(x, y): return only obs. in both x and y.
- union(x, y): return unique obs. in x and y.
- setdiff(x, y): return obs. in x, but not in y.

Exercise:

Compare the sheet **PBI New** and **SAS New** of Excel file: **Test PBI and SAS new and endorsement 20220214.xlsx**

- Policy number is the primary key for each sheet. Find out the difference records between the two sheets.
- use join function and setdiff function to see if you can get the same results.