What we'll be covering:

- Row-wise operation the column-wise way with pivot longer
- Row-wise data operation with transpose
- Row-wise operations with rowSums
- Row-wise operations with pmap and apply
- ID columns for doing row-wise operations the column-wise way

Get the #tidytuesday Data

```
library(tidyverse)
library(tidytuesdayR)
plastic <- tidytuesdayR::tt download(tt load gh("2021-01-26"))</pre>
plastic[[1]]
## # A tibble: 13,380 x 14
      country year parent company empty hdpe ldpe
                                                                    pet
                                                                            pp
     pvc
ps
##
                                                215
##
    1 Argent... 2019 Grand Total
                                           \cap
                                                        55
                                                              607
                                                                   1376
                                                                           281
116
       18
   2 Argent... 2019 Unbranded
                                                              532
##
                                                155
                                                        50
                                                                    848
                                                                           122
114
       17
##
    3 Argent... 2019 The Coca-Cola...
                                           0
                                                  0
                                                         0
                                                                0
                                                                    222
                                                                            35
\Omega
      0
    4 Argent... 2019 Secco
                                            0
                                                                     39
                                                                             4
0
##
    5 Argent... 2019 Doble Cola
                                           0
                                                  0
                                                         \cap
                                                                0
                                                                     38
                                                                             \cap
0
    6 Argent... 2019 Pritty
                                            0
                                                         0
                                                                     22
                                                                             7
0
    7 Argent... 2019 PepsiCo
##
                                            0
                                                  \cap
                                                         \cap
                                                                0
                                                                     21
                                                                             6
0
##
    8 Argent... 2019 Casoni
                                                                     26
                                                                             0
##
    9 Argent... 2019 Villa Del Sur...
                                            0
                                                  0
                                                         0
                                                                0
                                                                     19
                                                                             1
0
## 10 Argent... 2019 Manaos
                                            0
                                                  0
                                                         0
                                                                     14
                                                                             4
\#\# \# ... with 13,370 more rows, and 3 more variables: grand total ,
       num events , volunteers
```

We can see that there are certain names for different plastic types in the columns. The columns that count the different types of plastic are from empty until pvc. There is also a $grand_total$ column that sums up all the plastic. Let's do a little quality check and see if the $grand_total$ column has all the plastic from column empty until column pvc summed up.

For the approach, we will be using the pivot_longer function from the tidyr package. We will demonstrate some other approaches later that perform some row-wise operations. However,

I have a better mental model of the data and what I want to accomplish when I am thinking column-wise.

Row-wise operation the column-wise way with pivot_longer()

```
plastic[[1]] %>%
  tidyr::pivot longer(empty:pvc) %>%
 dplyr::group by(country, year, parent company) %>%
 dplyr::mutate(grand total check = sum(value, na.rm = T)) %>%
 dplyr::ungroup() %>%
 tidyr::pivot wider() %>%
 dplyr::filter(grand total check != grand total)
## # A tibble: 2 x 15
## country year parent company grand total num events volunteers
##
## 1 Korea 2020 Dongsuh
                                                    26
                                          44
                                                                NA
                                        4037
## 2 United... 2020 null
                                                   134
                                                               511
## # ... with 9 more variables: grand_total_check , empty , hdpe ,
## # ldpe , o , pet , pp , ps , pvc
```

With the code above, we are pivoting longer on all the plastic categories and then are identifying the row by grouping by country, year, and parent_company. The we can do any operation we wish to do in the summarise function.

And we find some rows where the sum in the plastic columns does not match the total.

Row-wise Data Operations With transpose

```
plastic[[1]] %>%
 dplyr::mutate(
   grand total check = plastic[[1]] %>%
     dplyr::select(empty:pvc) %>%
     purrr::transpose() %>%
     purrr::map dbl(~ flatten dbl(.) %>%
                    sum(na.rm = T))
 dplyr::filter(grand total check != grand total)
## # A tibble: 2 x 15
   country year parent company empty hdpe ldpe o
                                                       pet
                                                             qq
ps pvc
##
## 1 Korea 2020 Dongsuh
                                 0 0
                                           32
                                                 4
                                                              \cap
   0
## 2 United... 2020 null
                         102 165 275 752 788 1668
243 41
\#\# \# ... with 4 more variables: grand total , num events ,
\#\# # volunteers , grand total check
```

It basically turns the data frame "inside-out" and then we can add up the rows with map_dbl.

Row-wise Operations with rowSums

We can also just simply with the base R rowSums function. However, this operation is specific to adding up numbers so if you would like to do other row-wise operations, you have to use another function.

```
plastic[[1]] %>%
  dplyr::mutate(
    grand_total_check = dplyr::select(
        ., empty:pvc
    ) %>% base::rowSums(na.rm = T)
    ) %>%
  dplyr::filter(grand_total_check != grand_total)

plastic[[1]] %>%
  dplyr::mutate(
    grand_total_check = dplyr::select(
        ., empty:pvc
    ) %>% base::rowSums(na.rm = T)
    ) %>%
  dplyr::filter(grand_total_check != grand_total)
```

Row-wise Operations With pmap and apply

Other ways to do row-wise operations are with purrr's pmap function and the base R apply function.

```
plastic[[1]] %>%
  dplyr::mutate(
   grand total check = dplyr::select(
      ., empty:pvc
   ) %>% purrr::pmap dbl(sum, na.rm = T)
 dplyr::filter(grand total check != grand total)
## # A tibble: 2 x 15
    country year parent company empty hdpe ldpe
                                                           pet
                                                                   pp
ps
##
                                                 32
## 1 Korea
            2020 Dongsuh
                                      0
                                            0
                                                        4
                                                                    0
## 2 United... 2020 null
                                   102
                                         165
                                                275
                                                      752
                                                          788 1668
243
      41
## # ... with 4 more variables: grand total , num events ,
## # volunteers , grand total check
plastic[[1]] %>%
 dplyr::mutate(
    grand total check = apply(dplyr::select(plastic[[1]], empty:pvc),
1,
                              function(x) \{sum(x, na.rm = TRUE)\})
  ) 응>응
 dplyr::filter(grand total check != grand total)
## # A tibble: 2 x 15
##
   country year parent company empty hdpe ldpe
                                                           pet
                                                                   pp
    pvc
ps
##
## 1 Korea
            2020 Dongsuh
                                            0
                                                 32
                                                                    0
     0
## 2 United... 2020 null
                                   102
                                         165
                                               275
                                                     752
                                                          788 1668
243
      41
\#\# \# ... with 4 more variables: grand total , num events ,
## # volunteers , grand total check
```

All these approaches work. However, lately, I have been finding myself using the pivot_longer and pivot_wider functions a lot from the tidyr package to do row-wise operations.

ID Columns for Doing Row-wise Operations the Column-wise Way

Sometimes, you have to first add an id to do row-wise operations column-wise. For example, when you would like to sum up all the rows where the columns are numeric in the mtcars data set, you can add an id, pivot_wider and then group by id (the row previously) and then sum up the value.

```
mtcars %>%
  dplyr::select_if(is.numeric) %>%
  dplyr::mutate(id = dplyr::row number()) %>%
```

```
tidyr::pivot_longer(-id) %>%
 dplyr::group_by(id) %>%
 dplyr::summarise(total = sum(value))
## # A tibble: 32 x 2
##
        id total
## *
## 1 1 329.
        2 330.
## 2
## 3
       3 260.
## 4 4 426.
## 5 5 590.
       6 386.
## 6
## 7
       7 657.
       8 271.
## 8
## 9 9 300.
## 10 10 350.
## # ... with 22 more rows
```

Additional Resources

- A tutorial about pivot wider() and pivot longer() with a #tidytuesday data set
- Row-wise data operation's with purrr and pmap
- Another purrr apply example