

Course 1 Section 2.24 - Pedestrian activity around the City of Melbourne 2

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```
library(tidyverse)
library(here)
```

Step 1: Create a new R Markdown and read in the data

```
ped <- read_csv(here("data", "melb_walk.csv"))
ped
```

```
## # A tibble: 31,992 x 5
##   Sensor                                Date_Time      Date      Time Count
##   <chr>                                <dtm>         <date>    <dbl> <dbl>
## 1 Bourke Street Mall (North)          2018-12-31 13:00:00 2019-01-01      0    918
## 2 Bourke Street Mall (South)          2018-12-31 13:00:00 2019-01-01      0    770
## 3 Melbourne Central                   2018-12-31 13:00:00 2019-01-01      0     NA
## 4 Town Hall (West)                    2018-12-31 13:00:00 2019-01-01      0   3025
## 5 Princes Bridge                      2018-12-31 13:00:00 2019-01-01      0    531
## 6 Flinders Street Station Underpass  2018-12-31 13:00:00 2019-01-01      0   3284
## 7 Birrarung Marr                      2018-12-31 13:00:00 2019-01-01      0   2733
## 8 Webb Bridge                        2018-12-31 13:00:00 2019-01-01      0    762
## 9 Southern Cross Station              2018-12-31 13:00:00 2019-01-01      0   1830
## 10 Victoria Point                    2018-12-31 13:00:00 2019-01-01      0   1217
## # ... with 31,982 more rows
```

Step 2: Count the sensors, and more

Q1: Use a wrangling verb, to count the number of sensors. Do all the sensors have the same number of measurements?

```
num_sensor <- ped %>%
  group_by(Sensor) %>%
  summarise(num_sensor = n())

num_sensor
```

```
## # A tibble: 43 x 2
##   Sensor                                num_sensor
```

```
##      <chr>                                <int>
## 1 Alfred Place                            744
## 2 Birrarung Marr                          744
## 3 Bourke St-Russell St (West)             744
## 4 Bourke Street Mall (North)              744
## 5 Bourke Street Mall (South)              744
## 6 Chinatown-Lt Bourke St (South)          744
## 7 Chinatown-Swanston St (North)           744
## 8 City Square                            744
## 9 Collins Place (North)                   744
## 10 Collins Place (South)                  744
## # ... with 33 more rows
```

```
unique(num_sensor$num_sensor)
```

```
## [1] 744
```

All the sensors have the same number of measurements.

Q2: For each sensor, compute the total count for January. Which sensor had the largest count? Which sensor had the smallest count?

```
count_by_sensor <- ped %>%
  group_by(Sensor) %>%
  summarise(sum = sum(Count, na.rm = TRUE)) %>%
  arrange(sum)
```

```
head(count_by_sensor)
```

```
## # A tibble: 6 x 2
##   Sensor          sum
##   <chr>          <dbl>
## 1 City Square      0
## 2 Flagstaff Station 0
## 3 Flinders St-Elizabeth St (East) 0
## 4 Tin Alley-Swanston St (West) 38773
## 5 Waterfront City 61481
## 6 Monash Rd-Swanston St (West) 66420
```

```
tail(count_by_sensor)
```

```
## # A tibble: 6 x 2
##   Sensor          sum
##   <chr>          <dbl>
## 1 The Arts Centre 884885
## 2 Bourke Street Mall (North) 895483
## 3 Spencer St-Collins St (North) 910109
## 4 Flinders Street Station Underpass 1015331
## 5 Town Hall (West) 1035715
## 6 Southbank 1395117
```

Q3: For each sensor, compute the total number of missing counts. Which sensor had the most missing counts? Why might this be?

```
ped %>%
  group_by(Sensor) %>%
  select(Count) %>%
  summarise_all(funs(sum(is.na(.)))) %>%
  rename(na_count = "Count") %>%
  arrange(desc(na_count))
```

```
## Warning: 'funs()' is deprecated as of dplyr 0.8.0.
## Please use a list of either functions or lambdas:
##
##   # Simple named list:
##   list(mean = mean, median = median)
##
##   # Auto named with 'tibble::lst()':
##   tibble::lst(mean, median)
##
##   # Using lambdas
##   list(~ mean(., trim = .2), ~ median(., na.rm = TRUE))
## This warning is displayed once every 8 hours.
## Call 'lifecycle::last_warnings()' to see where this warning was generated.
```

```
## # A tibble: 43 x 2
##   Sensor                na_count
##   <chr>                <int>
## 1 City Square          744
## 2 Flagstaff Station    744
## 3 Flinders St-Elizabeth St (East) 744
## 4 Birrarung Marr       416
## 5 Melbourne Central    127
## 6 Monash Rd-Swanston St (West)    50
## 7 Grattan St-Swanston St (West)   38
## 8 Tin Alley-Swanston St (West)    25
## 9 St Kilda Rd-Alexandra Gardens   24
## 10 Waterfront City        21
## # ... with 33 more rows
```

Q4: Create a new table that contains the counts for the Melbourne Central and State Library sensors, then use a tidying verb to create two new columns that contain their counts.

```
ped %>%
  filter(Sensor %in% c("Melbourne Central", "State Library")) %>%
  spread(key = Sensor, value = Count)
```

```
## # A tibble: 744 x 5
##   Date_Time      Date      Time 'Melbourne Central' 'State Library'
##   <dtm>         <date>    <dbl>          <dbl>          <dbl>
## 1 2018-12-31 13:00:00 2019-01-01      0              NA           1548
## 2 2018-12-31 14:00:00 2019-01-01      1              NA           1494
```

```
## 3 2018-12-31 15:00:00 2019-01-01      2      NA      878
## 4 2018-12-31 16:00:00 2019-01-01      3      NA      309
## 5 2018-12-31 17:00:00 2019-01-01      4      NA      133
## 6 2018-12-31 18:00:00 2019-01-01      5      NA      110
## 7 2018-12-31 19:00:00 2019-01-01      6      NA       42
## 8 2018-12-31 20:00:00 2019-01-01      7      NA       50
## 9 2018-12-31 21:00:00 2019-01-01      8      NA       83
## 10 2018-12-31 22:00:00 2019-01-01      9      NA      128
## # ... with 734 more rows
```

Q5: Create the following 100 percent chart to compare the foot traffic at Melbourne Central and the State Library during different hours of the day. Explain why the first 8 days of January appear this way.

```
ped %>%
  filter(Sensor %in% c("Melbourne Central", "State Library")) %>%
  ggplot(aes(x = Time , y = Count, fill = Sensor)) +
  geom_bar(stat = "identity", position = "fill") +
  facet_wrap(~Date, ncol = 7)
```

```
## Warning: Removed 127 rows containing missing values (position_stack).
```

