

PHW251 Problem Set 7

Teaching Team

2021

Due date: Monday, November 15th

Part 1

For part 1 of this problem set we will work with motor vehicle crash data from New York City. You can read more about this publicly available data set on their website.

The data file is called “Motor_Vehicle_Collisions_Crashes.csv”. We want you to perform the following:

1. Rename the column names to lower-case and replace spaces with an underscore.
2. Select only:
 - crash_date
 - number_of_persons_injured
 - contributing_factor_vehicle_1
 - vehicle_type_code_1
3. Drop all rows that contain an NA value.
4. Make the values in the vehicle_type_code_1 variable all lowercase and replace the spaces with a dash.
5. Filter the data for vehicles that have a count of at least 500 (appear in the data set 500 times or more)
 - Hints: group_by(), mutate(), n(), filter()
6. Calculate the percentage of accidents by vehicle type
7. Which vehicle group accounted for 1.55% (0.0155) of the accidents?

We have grouped the questions below to push you to perform commands with less code. As you’re building your code we recommend going line by line to test, then combining to perform multiple steps in one command.

Questions 1-3

YOUR CODE HERE

```
df_motor <- df_motor %>%  
  # lower case and remove spaces  
  rename_with(~ tolower(gsub(" ","_", .x, fixed=TRUE))) %>%  
  # select certain columns  
  select(crash_date,  
         number_of_persons_injured,  
         contributing_factor_vehicle_1,  
         vehicle_type_code_1) %>%  
  # drop NA rows  
  drop_na()  
  
dim(df_motor)
```

```
## [1] 188989      4
```

```
head(df_motor)
```

```
## # A tibble: 6 x 4  
##   crash_date number_of_persons_injured contributing_factor_ve~ vehicle_type_cod~  
##   <chr>          <dbl> <chr>          <chr>  
## 1 9/11/19              0 Unspecified      Sedan  
## 2 12/7/19              1 Unspecified      Sedan  
## 3 12/7/19              0 Passing or Lane Usage ~ Sedan  
## 4 12/7/19              1 Unsafe Speed      Sedan  
## 5 12/7/19              0 Passing or Lane Usage ~ Sedan  
## 6 12/9/19              0 Oversized Vehicle  Ambulance
```

Questions 4-5

```
# YOUR CODE HERE
```

```
# lower case vehicles and add dash between spaces
```

```
df_motor <- df_motor %>%
```

```
  mutate(vehicle_type_code_1 =
```

```
    gsub(" ", "-", ignore.case=T, tolower(vehicle_type_code_1))) %>%
```

```
  # organize by vehicles
```

```
  group_by(vehicle_type_code_1) %>%
```

```
  # create a variable for counts
```

```
  mutate(count = n()) %>%
```

```
  # filter counts > 500
```

```
  filter(count > 500)
```

```
head(df_motor)
```

```
## # A tibble: 6 x 5
```

```
## # Groups:   vehicle_type_code_1 [2]
```

```
##   crash_date number_of_persons_i~ contributing_factor_v~ vehicle_type_cod~ count
```

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

```
## 1 9/11/19           0 Unspecified      sedan           85181
```

```
## 2 12/7/19           1 Unspecified      sedan           85181
```

```
## 3 12/7/19           0 Passing or Lane Usage~ sedan           85181
```

```
## 4 12/7/19           1 Unsafe Speed      sedan           85181
```

```
## 5 12/7/19           0 Passing or Lane Usage~ sedan           85181
```

```
## 6 12/9/19           0 Oversized Vehicle ambulance         692
```

```
min(df_motor$count)
```

```
## [1] 543
```

```
unique(df_motor$vehicle_type_code_1)
```

```
## [1] "sedan"
```

```
"ambulance"
```

```
## [3] "taxi"
```

```
"station-wagon/sport-utility-vehicle"
```

```
## [5] "motorcycle"
```

```
"box-truck"
```

```
## [7] "pick-up-truck"
```

```
"van"
```

```
## [9] "tractor-truck-diesel"
```

```
"bike"
```

```
## [11] "dump"
```

```
"bus"
```

```
## [13] "convertible"
```

Question 6

YOUR CODE HERE

```
# calculate percentage by vehicle type
df_motor %>%
  group_by(vehicle_type_code_1) %>%
  summarize(count = n(),
            perc = count/nrow(df_motor)) %>%
  arrange(perc)
```

```
## # A tibble: 13 x 3
##   vehicle_type_code_1      count    perc
##   <chr>              <int>   <dbl>
## 1 dump                543 0.00294
## 2 convertible         577 0.00313
## 3 ambulance           692 0.00375
## 4 van                1177 0.00638
## 5 motorcycle          1214 0.00658
## 6 tractor-truck-diesel 1434 0.00777
## 7 bike               1825 0.00989
## 8 bus                 2862 0.0155
## 9 box-truck           3830 0.0208
## 10 pick-up-truck      5411 0.0293
## 11 taxi               8104 0.0439
## 12 station-wagon/sport-utility-vehicle 71728 0.389
## 13 sedan             85181 0.461
```

Question 7

WRITE YOUR ANSWER HERE

Buses account for 1.55% of the accidents.

- count: 2862
- perc: 0.0155

Part 2

For this part we will work with four tables that are relational to each other. The following keys link the tables together:

- patient_id: patients, schedule
- visit_id: schedule, visits
- doctor_id: visits, doctors

Question 8

You've been asked to collect information on patients who are actually on the schedule. To start this task, you need to join the patient data to the schedule data, since we only want to keep the observations that are present in both the patient data AND the schedule data.

Which kind of join do you use?

WRITE YOUR ANSWER HERE **inner join**

How many observations do you see in your joined data set? Notice that some patients have multiple visits.

```
# YOUR CODE HERE

# inner join by patient_id
inner.join.patient <- patients %>%
  inner_join(schedule, by = "patient_id")

head(inner.join.patient)

## # A tibble: 6 x 8
##   patient_id age race_ethnicity gender_identity height weight visit_id date
##   <dbl> <dbl> <chr>          <chr>          <dbl> <dbl> <dbl> <chr>
## 1      1000   54 Asian            woman            163    57    17 7/5/~
## 2      1001   60 Hispanic, Latin~ woman            190    80     1 1/2/~
## 3      1001   60 Hispanic, Latin~ woman            190    80    37 2/7/~
## 4      1001   60 Hispanic, Latin~ woman            190    80    53 8/3/~
## 5      1001   60 Hispanic, Latin~ woman            190    80    80 3/7/~
## 6      1001   60 Hispanic, Latin~ woman            190    80    83 4/7/~
```

WRITE YOUR ANSWER HERE **124 observations**

Question 9

In the visits data, we have a variable called “follow_up” where Y means a follow-up is needed and N means a follow-up is not needed. How many patients require a follow-up? You will want to first make a join and then subset. Start with the data frame created in the previous question.

```
# YOUR CODE HERE
```

```
left.follow.up <- inner.join.patient %>%  
  left_join(visits, by = "visit_id")
```

```
# two ways we can filter:
```

```
follow.up <- left.follow.up %>% filter(follow_up == "Y")  
follow.up <- left.follow.up[which(left.follow.up$follow_up == "Y"), ]
```

```
# make sure we count unique patients who need follow-up  
length(unique(follow.up$patient_id))
```

```
## [1] 27
```

```
# or
```

```
follow.up %>% tally()
```

```
## # A tibble: 1 x 1
```

```
##       n
```

```
##   <int>
```

```
## 1     27
```

Which join did you use?

WRITE YOUR ANSWER HERE **left join**

How many patients need a follow-up?

WRITE YOUR ANSWER HERE **27**

Question 10

Which doctors do these patients need follow-up with? Print out each doctor's name.

YOUR CODE HERE

```
doctors.contact <- follow.up %>%  
  left_join(doctors, by = "doctor_id")  
  
unique(doctors.contact$doctor)
```

```
## [1] "Ariadne Anthony" "Millie Albert" "Ellesha Castaneda"  
## [4] "Bea Frame" "Vera Irwin" "Cade Gale"  
## [7] "Estelle Landry" "Wiktorja Travis" "Huzaifa Chung"  
## [10] "Jamie-Lee Wilder" "Jeremy Camacho" "Daanyaal Griffin"  
## [13] "Ammar Phelps" "Rabia Browning" "Amritpal Goodman"  
## [16] "Merlin Jacobs" "Tudor Moran"
```

Which join did you use?

WRITE YOUR ANSWER HERE `left join`