# Data Wrangling Using Dplyr

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| Loading R packages   |       |
| • dplyr: dataframe manipulation  |       |
| • ggplot2: visualization   |       |
| <pre>#install packages only if you have not already done so list.of.packages &lt;- c( "dplyr", "tidyverse") new.packages &lt;- list.of.packages[!(list.of.packages %in% installed.packages()[,"Package</pre> | ;"])] |

```
if(length(new.packages)) install.packages(new.packages)
#library packages
for (pkg in c("dplyr", "tidyverse")) {
   library(pkg, character.only = TRUE)
   }
```

```
load("surgery_data.RData")
```

#### I. Mutate Function

#### Example I.i: Change the label for a categorical variable

Before any data wrangling, we will always perform the 3-step procedure for the variable we are interested in:

- 1. check the data type (character or integer or others)
- 2. check whether there are NAs, how many NAs are there in the variable
- 3. what values are there in the variable. Use table() function for categorical variables, use summary() function for numerica variables.

```
#step 1.
class(surgery_data$gender) #check the data type
## [1] "character"
#step 2.
anyNA(surgery_data$gender) #check whether there are NA values
## [1] TRUE
table(is.na(surgery_data$gender)) #gives the count of NA values: 3
##
## FALSE TRUE
## 31998
#step 3.
table(surgery_data$gender) #check how many non-NA levels are there in the gender variable
##
##
       F
             Μ
## 17230 14768
#overwirte gender variable
surgery_data <- surgery_data%>%
 mutate(gender = if_else(gender == "F", "Female",
                  if_else(gender == "M", "Male", "Unknown")))
```

#### Example I.ii: Group patients whose race, gender are NA into a seperate group

There are 480 patients who have NA values for race. We don't want to exclude these sample from our data, let's treat them as a seperate group called "Unknown"

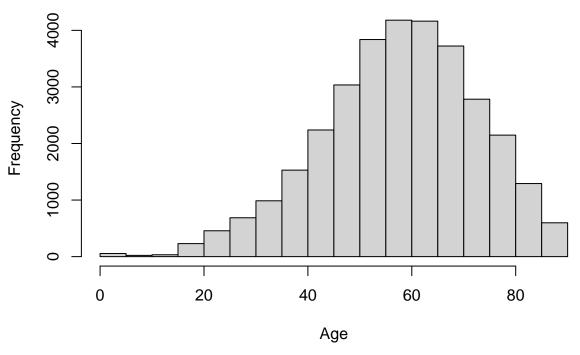
```
#step 1.
class(surgery_data$race)
## [1] "character"
#step 2.
anyNA(surgery_data$race) #check whether there are NA values
## [1] TRUE
table(is.na(surgery_data$race)) #gives the count of NA values: 3
##
## FALSE TRUE
## 31521
           480
#step 3.
table(surgery_data$race)
##
## African American
                           Caucasian
                                                 Other
                                                  1243
               3790
                                26488
##
surgery_data <- surgery_data%>%
 mutate(race = if_else(is.na(race), "Unknown", race))%>%
 mutate(gender = if_else(is.na(gender), "Unknown", gender))
table(surgery_data$race)
##
## African American
                           Caucasian
                                                 Other
                                                                 Unknown
               3790
                                26488
                                                  1243
                                                                     480
table(surgery_data$gender)
##
##
    Female
              Male Unknown
##
     17230
             14768
anyNA(surgery_data$race)
## [1] FALSE
```

```
anyNA(surgery_data$gender)
## [1] FALSE
```

#### Example I.iii: Create age groups from a numeric age variable

```
#step 1
class(surgery_data$age) #check the data type
## [1] "numeric"
#step 2
anyNA(surgery_data$age) #check whether there are NA values
## [1] TRUE
table(is.na(surgery_data$age)) #gives the count of NA values: 3
## FALSE TRUE
## 31999
#step 3.
summary(surgery_data$age) #check the range of the variable
##
     Min. 1st Qu. Median Mean 3rd Qu.
                                            Max.
                                                   NA's
      1.00 48.20 58.60 57.66 68.30 90.00
##
hist(surgery_data$age,
    main = "Distribution of Age",
    xlab = "Age") #check the distribution of the variable, which helps us to seperate into groups
```

#### **Distribution of Age**



1924

335

12857

#### II. Select Function

3628

##

#### Example II.i: Only keep variables of interest in the dataframe

13255

#### III. Filter Function

#### Example III.i: Identify one African American patients

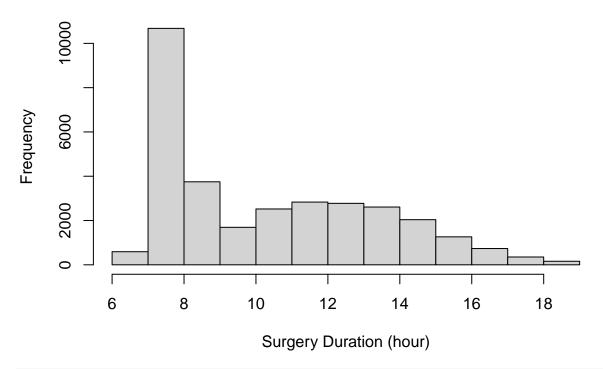
```
##
## African American Caucasian Other Unknown
## 3790 26488 1243 480

surgery_data_AfricanAmerican <- surgery_data%>%
filter(race == "African American")
```

#### Example III.ii: Identify patients who's surgery time is longer than 10 hours

```
hist(surgery_data$hour,
    main = "Distribution of Surgery Duration",
    xlab = "Surgery Duration (hour)")
```

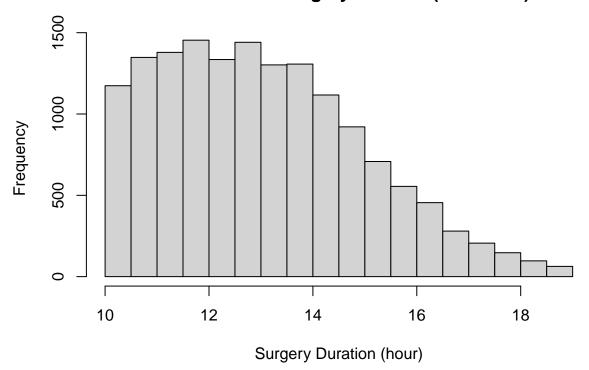
## **Distribution of Surgery Duration**



```
surgery_data_10hr<- surgery_data%>%
  filter(hour > 10)

hist(surgery_data_10hr$hour,
    main = "Distribution of Surgery Duration(Hour > 10)",
    xlab = "Surgery Duration (hour)")
```

## **Distribution of Surgery Duration(Hour > 10)**



#### IV. Summarize Function

Example IV.i: Identify the average surgery hour for each race group

```
surgery_data%>%
  group_by(race)%>%
  summarize(count = n(),
           hour_mean = mean(hour),
           hour_median= median(hour),
           hour_sd = sd(hour))%>%
  mutate(perc = count/sum(count) * 100)
## `summarise()` ungrouping output (override with `.groups` argument)
## # A tibble: 4 x 6
##
                      count hour_mean hour_median hour_sd perc
     race
     <chr>
                      <int>
                                <dbl>
                                            <dbl>
                                                    <dbl> <dbl>
## 1 African American 3790
                                            10.1
                                 10.6
                                                     2.98 11.8
## 2 Caucasian
                      26488
                                 10.4
                                             9.6
                                                     2.91 82.8
                      1243
                                             9.28
                                                     2.94 3.88
## 3 Other
                                 10.3
## 4 Unknown
                        480
                                 10.5
                                             9.45
                                                     2.91 1.50
```

Example IV.ii: Further investigate within each race, what's the average surgery hour for different asa statis

| ## # A tiddle: 13 x 6 |    |                  |              |               |                    |                      |             |  |
|-----------------------|----|------------------|--------------|---------------|--------------------|----------------------|-------------|--|
| ## # Groups: race [4] |    |                  |              |               |                    |                      |             |  |
| ##                    |    | race             | $asa_status$ | ${\tt count}$ | ${\tt hour\_mean}$ | ${\tt hour\_median}$ | hour_sd     |  |
| ##                    |    | <chr></chr>      | <chr></chr>  | <int></int>   | <dbl></dbl>        | <dbl></dbl>          | <dbl></dbl> |  |
| ##                    | 1  | African American | I-II         | 1839          | 10.5               | 9.83                 | 3.01        |  |
| ##                    | 2  | African American | III          | 1785          | 10.6               | 10.2                 | 2.94        |  |
| ##                    | 3  | African American | IV-VI        | 165           | 11.1               | 11.1                 | 2.93        |  |
| ##                    | 4  | Caucasian        | I-II         | 14443         | 10.2               | 9.22                 | 2.87        |  |
| ##                    | 5  | Caucasian        | III          | 11201         | 10.5               | 9.87                 | 2.92        |  |
| ##                    | 6  | Caucasian        | IV-VI        | 837           | 11.1               | 11.0                 | 3.04        |  |
| ##                    | 7  | Caucasian        | Unknown      | 7             | 11.1               | 13.0                 | 3.38        |  |
| ##                    | 8  | Other            | I-II         | 718           | 10.3               | 8.87                 | 2.98        |  |
| ##                    | 9  | Other            | III          | 492           | 10.3               | 9.30                 | 2.86        |  |
| ##                    | 10 | Other            | IV-VI        | 33            | 11.5               | 11.0                 | 3.18        |  |
| ##                    | 11 | Unknown          | I-II         | 261           | 10.4               | 8.92                 | 2.96        |  |
| ##                    | 12 | Unknown          | III          | 199           | 10.4               | 9.58                 | 2.85        |  |
| ##                    | 13 | Unknown          | IV-VI        | 20            | 11.4               | 12.0                 | 2.82        |  |

## V. Arrange Function

Arrange the median surgery hour in race+asa status group in descending order

```
table%>%
arrange(-hour_median)
```

```
## # A tibble: 13 x 6
## # Groups: race [4]
##
         asa_status count hour_mean hour_median hour_sd
   race
   <chr>
##
             <chr> <int> <dbl> <dbl> <dbl>
           Unknown
## 1 Caucasian
                     7
                           11.1
                                  13.0
                                        3.38
11.4
                                  12.0
                                        2.82
                           11.1
                                  11.1
                                        2.93
## 4 Other IV-VI
                      33
                           11.5
                                  11.0 3.18
## 5 Caucasian IV-VI
                            11.1
                     837
                                  11.0
                                        3.04
```

| ## | 6  | African American | III  | 1785  | 10.6 | 10.2 | 2.94 |
|----|----|------------------|------|-------|------|------|------|
| ## | 7  | Caucasian        | III  | 11201 | 10.5 | 9.87 | 2.92 |
| ## | 8  | African American | I-II | 1839  | 10.5 | 9.83 | 3.01 |
| ## | 9  | Unknown          | III  | 199   | 10.4 | 9.58 | 2.85 |
| ## | 10 | Other            | III  | 492   | 10.3 | 9.30 | 2.86 |
| ## | 11 | Caucasian        | I-II | 14443 | 10.2 | 9.22 | 2.87 |
| ## | 12 | Unknown          | I-II | 261   | 10.4 | 8.92 | 2.96 |
| ## | 13 | Other            | I-II | 718   | 10.3 | 8.87 | 2.98 |