

Data Wrangling Using Dplyr

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Loading R packages

- dplyr: dataframe manipulation
- ggplot2: visualization

```
#install packages only if you have not already done so
list.of.packages <- c( "dplyr", "tidyverse")
new.packages <- list.of.packages[!(list.of.packages %in% installed.packages()[,"Package"])]
if(length(new.packages)) install.packages(new.packages)

#library packages
```

```
for (pkg in c("dplyr", "tidyverse")) {
  library(pkg, character.only = TRUE)
}
```

```
# read in data
load("surgery_data.RData")
```

I. Mutate Function

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

Instead of abbreviation for the gender, “F” and “M”, we want them to be “Female” and “Male”.

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 numerical 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    3
```

```
#step 3.
table(surgery_data$gender) #check how many non-NA levels are there in the gender variable
```

```
##
##      F      M
## 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 separate 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 separate 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  
##           3790           26488           1243
```

```
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      3
```

```
anyNA(surgery_data$race)
```

```
## [1] FALSE
```

```
anyNA(surgery_data$gender)
```

```
## [1] FALSE
```

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

Currently age is a numeric variable, we want to form five age groups for analysis: less than 20, 20-40, 40-60, 60-80, 80+.

```
#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
```

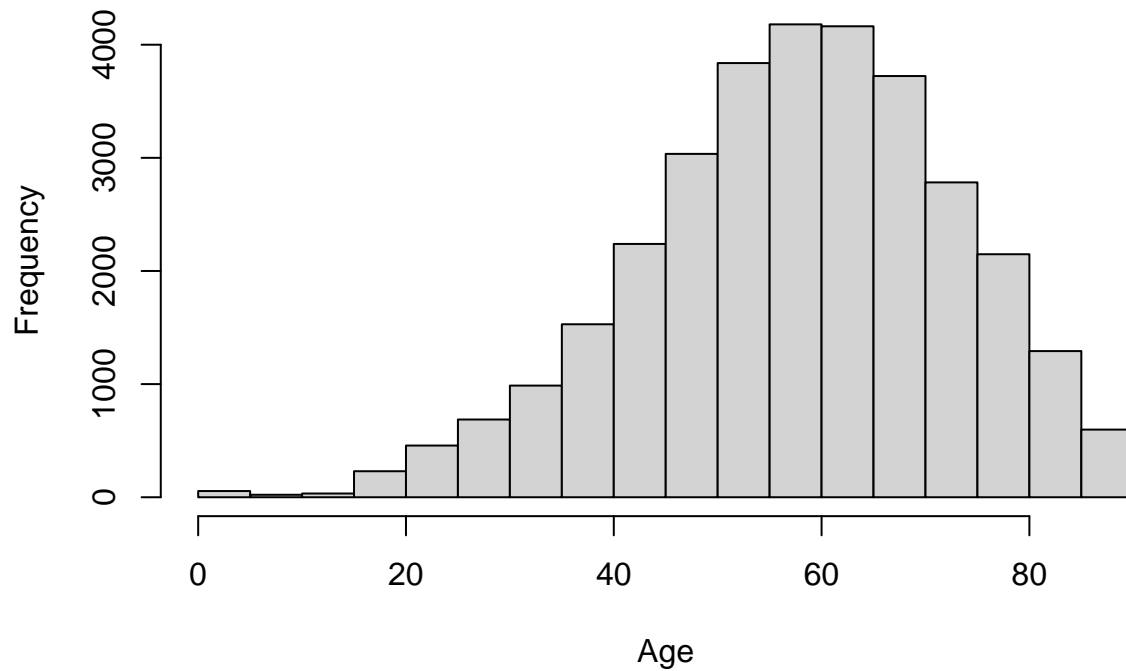
```
##  
## FALSE TRUE  
## 31999    2
```

```
#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         2
```

```
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



```
surgery_data <- surgery_data%>%
  mutate(age_group = if_else(age < 20, "less than 20",
    if_else(age < 40, "20-40 yrs",
      if_else(age < 60, "40-60 yrs",
        if_else(age < 80, "60-80 yrs",
          "80+")))))
table(surgery_data$age_group)
```

```
##
##      20-40 yrs      40-60 yrs      60-80 yrs      80+ less than 20
##      3628         13255         12857         1924         335
```

II. Select Function

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

```
surgery_data_subset <- surgery_data%>%
  select(age, gender, bmi, hour, race)
glimpse(surgery_data_subset)
```

```
## Rows: 32,001
## Columns: 5
## $ age      <dbl> 67.8, 39.5, 56.5, 71.0, 56.3, 57.7, 56.6, 64.2, 66.2, 20.1, ...
## $ gender   <chr> "Male", "Female", "Female", "Male", "Male", "Female", "Male"...
## $ bmi      <dbl> 28.04, 37.85, 19.56, 32.22, 24.32, 40.30, 64.57, 43.20, 28.0...
## $ hour     <dbl> 9.03, 18.48, 7.88, 8.80, 12.20, 7.67, 9.53, 7.52, 16.35, 16....
## $ race     <chr> "Caucasian", "Caucasian", "Caucasian", "Caucasian", "African..."
```

III. Filter Function

Example III.i: Identify only African American patients

```
table(surgery_data$race)
```

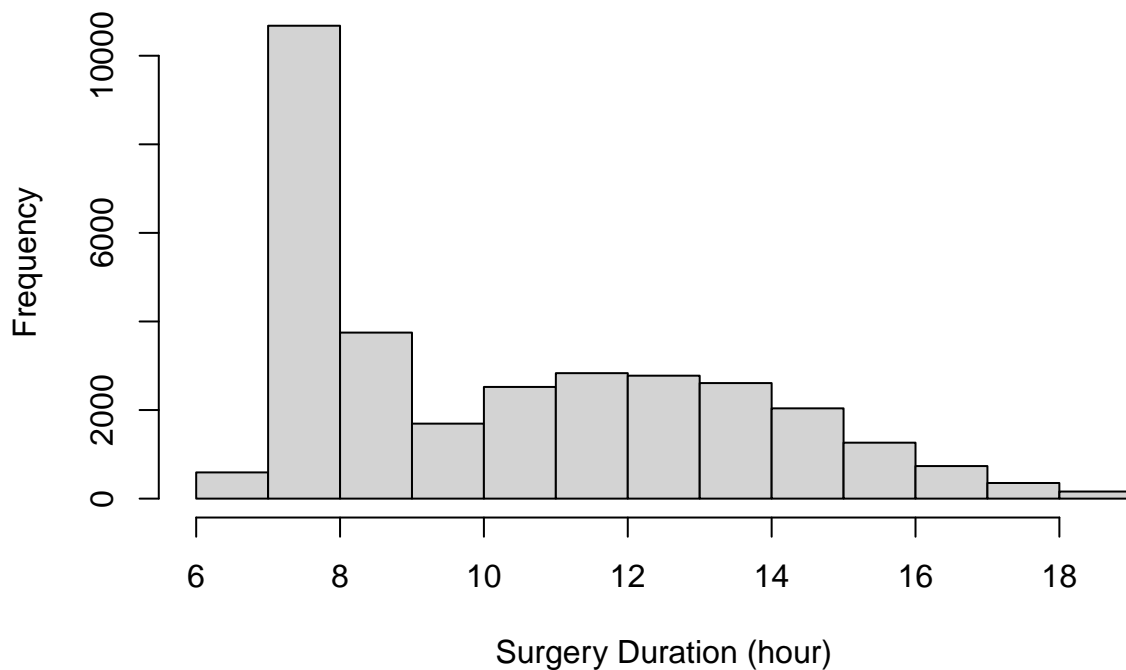
```
##  
## African American      Caucasian      Other      Unknown  
##           3790           26488           1243           480
```

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

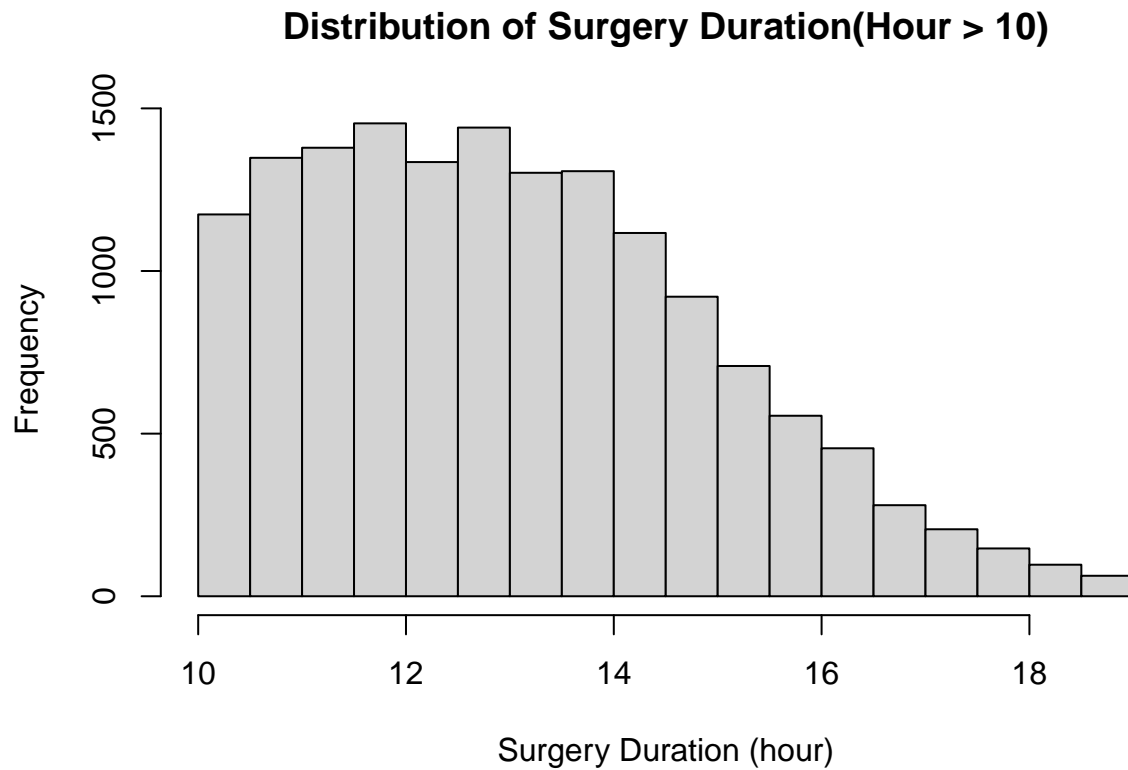
Example III.ii: Identify patients who's surgery time is longer than 10 hours

```
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)")
```



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
##   race          count hour_mean hour_median hour_sd perc
##   <chr>         <int>   <dbl>     <dbl>   <dbl> <dbl>
## 1 African American  3790     10.6      10.1    2.98  11.8
## 2 Caucasian        26488     10.4       9.6    2.91  82.8
## 3 Other            1243     10.3       9.28   2.94   3.88
## 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

```
table <- surgery_data%>%
  mutate(asa_status = if_else(is.na(asa_status), "Unknown", asa_status))%>%
  group_by(race, asa_status)%>%
  summarize(count = n(),
            hour_mean = mean(hour),
            hour_median= median(hour),
            hour_sd = sd(hour))%>%
  filter(count > 5)
```

`summarise()` regrouping output by 'race' (override with `.`groups` argument)

table

```
## # A tibble: 13 x 6
## # Groups:   race [4]
##   race          asa_status count hour_mean hour_median hour_sd
##   <chr>          <chr>    <int>    <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]
##   race          asa_status count hour_mean hour_median hour_sd
##   <chr>          <chr>    <int>    <dbl>      <dbl>    <dbl>
## 1 Caucasian       Unknown      7     11.1      13.0     3.38
## 2 Unknown        IV-VI      20     11.4      12.0     2.82
## 3 African American IV-VI      165     11.1      11.1     2.93
## 4 Other          IV-VI       33     11.5      11.0     3.18
## 5 Caucasian       IV-VI      837     11.1      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 |