

Assignment3

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#1 redo 13-26

```
setwd("C:/Users/student/Documents/Fall2019/")
getwd
```

```
## function ()
## .Internal(getwd())
## <bytecode: 0x0000000015138d78>
## <environment: namespace:base>
```

```
#install.packages("dplyr")
```

```
library(dplyr)
```

```
##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:stats':
##
##   filter, lag

## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union
```

```
Titanic<-read.csv(file= 'titanic.csv')
```

#1 redo 13-26

```
Titanic %>%
  select(Age,Sex)%>%
  #13
  filter(Sex=="female",Age)%>%
  summarise(mean_female_age = mean(Age))
```

```
##   mean_female_age
## 1           27.91571
```

#14 Calculate the median fare of the passengers in Class 1

```
Titanic %>%
  select(Fare,Pclass)%>%
  filter(Pclass == 1) %>%
  summarise(median_fare_class1 = median(Fare))
```

```
## median_fare_class1
## 1 60.2875
```

```
median(Titanic$Fare[Titanic$Pclass=='1'])
```

```
## [1] 60.2875
```

```
#15. Calculate the median fare of the female passengers that are not in Class 1
Titanic %>%
  select(Fare,Pclass,Sex)%>%
  filter(Sex=='female', Pclass!= 1)%>%
  summarise(median(Fare))
```

```
## median(Fare)
## 1 14.45625
```

```
#16. Calculate the median age of survived passengers who are female and Class 1 or Class 2,
Titanic %>%
  select(Age,Survived,Sex,Pclass)%>%
  filter(Survived==1, Sex=="female",Pclass !=3, Age != 'NA' ) %>%
  summarise(median(Age))
```

```
## median(Age)
## 1 31
```

```
#17. Calculate the mean fare of female teenagers survived passengers
Titanic %>%
  select(Age,Survived,Sex,Fare)%>%
  filter(Survived==1, Sex=="female",Age < 18) %>%
  summarise(mean(Fare))
```

```
## mean(Fare)
## 1 33.17226
```

```
#18. Calculate the mean fare of female teenagers survived passengers for each class
Titanic %>%
  group_by(Pclass) %>%
  summarise(m=mean(Fare))
```

```
## # A tibble: 3 x 2
##   Pclass     m
##   <int> <dbl>
## 1     1  84.2
## 2     2  20.7
## 3     3  13.7
```

```
#19. Calculate the ratio of Survived and not Survived for passengers who are who pays more than the
#average fare
Titanic %>%
  group_by(Survived)%>%
  summarize(mean(Fare))
```

```
## # A tibble: 2 x 2
##   Survived `mean(Fare)`
##   <int>     <dbl>
## 1     0     22.1
## 2     1     48.4
```

```
48.39541/22.11789
```

```
## [1] 2.188066
```

#20. Add column that standardizes the fare (subtract the mean and divide by standard deviation) and name it sfare

```
T =Titanic %>%
  mutate(S_fare= (Fare-mean(Fare))/sd(Fare))
```

#21. Add categorical variable named cfare that takes value cheap for passengers paying less the average fare and takes value expensive for passengers paying more than the average fare.

```
T =Titanic %>%
  mutate(cfare= ifelse(Fare>mean(Fare), 'expensive', 'cheap'))
```

#22. Add categorical variable named cage that takes value 0 for age 0-10, 1 for age 10-20, 2 for age 20-30 and so on

```
T =Titanic %>%
  mutate(cage = case_when(Age<10~'0',
                          (Age>=10)&(Age<20)~'1',
                          (Age>=20)&(Age<30)~'2',
                          (Age>=30)&(Age<40)~'3',
                          (Age>=40)&(Age<50)~'4',
                          (Age>=50)&(Age<60)~'5',
                          (Age>=60)&(Age<70)~'6',
                          (Age>=70)&(Age<80)~'7',
                          (Age>=80)~'8'))
```

#23. Show the frequency of Ports of Embarkation. It appears that there are two missing values in the Embarked variable. Assign the most frequent port to the missing ports. Hint: Use the levels function to modify the categories of categorical variables.

```
Titanic %>%
  count(Embarked)%>%
  mutate(Embarked=replace(Embarked,Embarked=='', 'S'))
```

```
## # A tibble: 4 x 2
##   Embarked    n
##   <fct>    <int>
## 1 S         2
## 2 C       168
## 3 Q        77
## 4 S       644
```

```
#2 assignment2 question 4
```

```
#install.packages('readxl') # install the library
library(readxl) # load the library
```

#2. Using Dplyr and in Assignment 2, redo 4 using sample_n function, redo 5 using glimpse, redo 11, 12 and 13. For 11, 12 and 13, you may want to use the combo group_by and summarise

```
read_excel("C:/Users/student/Documents/Fall2019/c2015.xlsx")
```

```
## # A tibble: 80,587 x 28
##   STATE ST_CASE VEH_NO PER_NO COUNTY DAY MONTH HOUR MINUTE AGE SEX
##   <chr>   <dbl>   <dbl>   <dbl>   <dbl> <dbl> <chr> <dbl>   <dbl> <chr> <chr>
## 1 Alab~   10001     1     1    127    1 Janu~    2    40 68 Male
## 2 Alab~   10002     1     1     83    1 Janu~   22    13 49 Male
## 3 Alab~   10003     1     1     11    1 Janu~    1    25 31 Male
## 4 Alab~   10003     1     2     11    1 Janu~    1    25 20 Fema~
## 5 Alab~   10004     1     1     45    4 Janu~    0    57 40 Male
## 6 Alab~   10005     1     1     45    7 Janu~    7     9 24 Male
## 7 Alab~   10005     2     1     45    7 Janu~    7     9 60 Male
## 8 Alab~   10006     1     1    111    8 Janu~    9    59 64 Male
## 9 Alab~   10006     1     2    111    8 Janu~    9    59 17 Male
## 10 Alab~  10007     1     1     89    8 Janu~   18    33 80 Male
## # ... with 80,577 more rows, and 17 more variables: PER_TYP <chr>,
## #   INJ_SEV <chr>, SEAT_POS <chr>, DRINKING <chr>, YEAR <dbl>,
## #   MAN_COLL <chr>, OWNER <chr>, MOD_YEAR <chr>, TRAV_SP <chr>,
## #   DEFORMED <chr>, DAY_WEEK <chr>, ROUTE <chr>, LATITUDE <dbl>,
## #   LONGITUD <dbl>, HARM_EV <chr>, LGT_COND <chr>, WEATHER <chr>
```

```
c2015=read_excel("C:/Users/student/Documents/Fall2019/c2015.xlsx")

set.seed(2019)
y<- sample_n(c2015,1000)
```

*#5. Use summary function to have a quick look at the data. You will notice there is one variable is act
#5a constant. Remove that variable from the data.*

```
glimpse(y)
```

```
## Observations: 1,000
## Variables: 28
## $ STATE <chr> "New Jersey", "Arizona", "Tennessee", "Minnesota", "M...
## $ ST_CASE <dbl> 340336, 40327, 470789, 270119, 290576, 62865, 330095,...
## $ VEH_NO <dbl> 1, 1, 1, 2, 1, 1, 0, 0, 2, 5, 1, 2, 1, 0, 1, 1, 2, 1,...
## $ PER_NO <dbl> 1, 1, 1, 4, 1, 1, 1, 1, 4, 1, 1, 1, 5, 1, 1, 2, 1, 1,...
## $ COUNTY <dbl> 27, 13, 163, 59, 201, 19, 15, 127, 13, 115, 29, 141, ...
## $ DAY <dbl> 19, 7, 2, 16, 2, 6, 3, 30, 17, 30, 19, 12, 9, 30, 9, ...
## $ MONTH <chr> "September", "May", "December", "May", "October", "Ju...
## $ HOUR <dbl> 3, 22, 8, 21, 15, 15, 14, 20, 7, 14, 14, 17, 18, 6, 4...
## $ MINUTE <dbl> 17, 15, 26, 59, 38, 20, 32, 20, 41, 36, 15, 50, 55, 4...
## $ AGE <chr> "Unknown", "47", "23", "15", "55", "56", "26", "63", ...
## $ SEX <chr> "Unknown", "Female", "Male", "Female", "Male", "Male"...
## $ PER_TYP <chr> "Driver of a Motor Vehicle In-Transport", "Driver of ...
## $ INJ_SEV <chr> "Unknown", "No Apparent Injury (0)", "Unknown", "Susp...
## $ SEAT_POS <chr> "Front Seat, Left Side", "Front Seat, Left Side", "Fr...
```

```
## $ DRINKING <chr> "Not Reported", "No (Alcohol Not Involved)", "Unknown...
## $ YEAR      <dbl> 2015, 2015, 2015, 2015, 2015, 2015, 2015, 2015, 2015,...
## $ MAN_COLL <chr> "Not a Collision with Motor Vehicle In-Transport", "N...
## $ OWNER     <chr> "Unknown", "Driver (in this crash) Not Registered Own...
## $ MOD_YEAR  <chr> "Unknown", "2003", "1994", "2011", "2000", "2013", NA...
## $ TRAV_SP   <chr> "Unknown", "048 MPH", "Not Rep", "055 MPH", "055 MPH"...
## $ DEFORMED  <chr> "Unknown", "Functional Damage", "Minor Damage", "Disa...
## $ DAY_WEEK  <chr> "Saturday", "Thursday", "Wednesday", "Saturday", "Fri...
## $ ROUTE     <chr> "State Highway", "Local Street", "County Road", "Stat...
## $ LATITUDE  <dbl> 40.95270, 33.41048, 36.57834, 45.42841, 37.13481, 36....
## $ LONGITUD  <dbl> -74.59644, -112.06459, -82.27889, -93.36788, -89.5946...
## $ HARM_EV   <chr> "Pedestrian", "Pedestrian", "Pedalcyclist", "Motor Ve...
## $ LGT_COND  <chr> "Dark - Not Lighted", "Dark - Lighted", "Dark - Not L...
## $ WEATHER   <chr> "Clear", "Clear", "Clear", "Rain", "Cloud", "Clear", ...
```

#11. Compare the average speed of those who had "No Apprent Injury" and the rest. What do you observe?

#12. Use the SEAT_POS variable to filter the data so that there is only drivers in the dataset. Compare average speed of man drivers and woman drivers. Comment on the results.

#13. Compare the average speed of drivers who drink and those who do not. Comment on the results. #Hint: This calculation can be done manually or by using the aggregate function or by function in #base R. For example:

```
library(stringr)
#11
y%>%
  mutate(TRAV_SP1=str_replace(TRAV_SP," MPH","")) %>%
  mutate(TRAV_SP1 = as.numeric(TRAV_SP1))%>%
  mutate(TRAV_SP1 = replace(TRAV_SP1,is.na(TRAV_SP1),mean(TRAV_SP1,na.rm=TRUE))) %>%
  group_by(INJ_SEV)%>%
  summarise(mean(TRAV_SP1))
```

Warning: NAs introduced by coercion

```
## # A tibble: 7 x 2
##   INJ_SEV                `mean(TRAV_SP1)`
##   <chr>                  <dbl>
## 1 Fatal Injury (K)      52.1
## 2 Injured, Severity Unknown 45.5
## 3 No Apparent Injury (O) 48.7
## 4 Possible Injury (C)    48.1
## 5 Suspected Minor Injury(B) 51.4
## 6 Suspected Serious Injury(A) 52.6
## 7 Unknown              46.8
```

#Obviously, no injury type of drivers who have slower speed comparing to other drivers who have injuries

```
#12
y%>%
  mutate(TRAV_SP1=str_replace(TRAV_SP," MPH","")) %>%
  mutate(TRAV_SP1 = as.numeric(TRAV_SP1))%>%
  mutate(TRAV_SP1 = replace(TRAV_SP1,is.na(TRAV_SP1),mean(TRAV_SP1,na.rm=TRUE))) %>%
  filter(SEAT_POS=="Front Seat, Left Side")%>%
  group_by(SEX) %>%
  summarise(mean(TRAV_SP1))
```

```
## Warning: NAs introduced by coercion
```

```
## # A tibble: 3 x 2
##   SEX      `mean(TRAV_SP1)`
##   <chr>      <dbl>
## 1 Female      49.0
## 2 Male        51.1
## 3 Unknown     45.5
```

```
#Male drivers tend to drive faster than female drivers
```

```
#13
y%>%
  mutate(TRAV_SP1=str_replace(TRAV_SP," MPH","")) %>%
  mutate(TRAV_SP1 = as.numeric(TRAV_SP1))%>%
  mutate(TRAV_SP1 = replace(TRAV_SP1,is.na(TRAV_SP1),mean(TRAV_SP1,na.rm=TRUE))) %>%
  group_by(DRINKING)%>%
  summarise(mean(TRAV_SP1))
```

```
## Warning: NAs introduced by coercion
```

```
## # A tibble: 4 x 2
##   DRINKING      `mean(TRAV_SP1)`
##   <chr>      <dbl>
## 1 No (Alcohol Not Involved)      48.8
## 2 Not Reported                  51.3
## 3 Unknown (Police Reported)     51.6
## 4 Yes (Alcohol Involved)        57.2
```

```
#Drinking alcohol drivers drive faster than not drinking drivers
```

#3. Calculate the travel speed (TRAV_SP variable) by day. Compare the travel speed of the first 5 days of the last 5 days of months.

```
y%>%
  mutate(TRAV_SP1=str_replace(TRAV_SP," MPH","")) %>%
  mutate(TRAV_SP1 = as.numeric(TRAV_SP1))%>%
  mutate(TRAV_SP1 = replace(TRAV_SP1,is.na(TRAV_SP1),mean(TRAV_SP1,na.rm=TRUE))) %>%
  filter(DAY==1:5)%>%
  group_by(DAY)%>%
  summarise(mean = mean(TRAV_SP1))%>%
  summarise(fist_5_day_mean=mean(mean))
```

```
## Warning: NAs introduced by coercion
```

```
## # A tibble: 1 x 1
##   fist_5_day_mean
##   <dbl>
## 1          45.0
```

#continued question 3

```
y%>%
  mutate(TRAV_SP1=str_replace(TRAV_SP," MPH","")) %>%
  mutate(TRAV_SP1 = as.numeric(TRAV_SP1))%>%
  mutate(TRAV_SP1 = replace(TRAV_SP1,is.na(TRAV_SP1),mean(TRAV_SP1,na.rm=TRUE))) %>%
  filter(DAY==27:31)%>%
  group_by(DAY)%>%
  summarise(mean2 = mean(TRAV_SP1)) %>%
  summarise(last_5_day_mean=mean(mean2))
```

Warning: NAs introduced by coercion

```
## # A tibble: 1 x 1
##   last_5_day_mean
##             <dbl>
## 1             49.8
```

From calculated the mean of first 5 days and last 5 days mean speed, here comes conclusion: people wh

#4. Calculate the travel speed (TRAV_SP variable) by day of the week. Compare the travel speed of the weekdays and weekends.

```
y%>%
  mutate(TRAV_SP1=str_replace(TRAV_SP," MPH","")) %>%
  mutate(TRAV_SP1 = as.numeric(TRAV_SP1))%>%
  mutate(TRAV_SP1 = replace(TRAV_SP1,is.na(TRAV_SP1),mean(TRAV_SP1,na.rm=TRUE))) %>%
  group_by(DAY_WEEK=="Saturday" | DAY_WEEK=="Sunday")%>%
  summarise(mean(TRAV_SP1))
```

Warning: NAs introduced by coercion

```
## # A tibble: 2 x 2
##   `DAY_WEEK == "Saturday" | DAY_WEEK == "Sunday"` `mean(TRAV_SP1)`
##   <lgl>                                     <dbl>
## 1 FALSE                                     50.1
## 2 TRUE                                      52.0
```

#The comparasion calculated the mean of speed when day of week is weekend or not. The conclusion is tha

#5. Find the top 5 states with greatest travel speed.

```
y%>%
  mutate(TRAV_SP1=str_replace(TRAV_SP," MPH","")) %>%
  mutate(TRAV_SP1 = as.numeric(TRAV_SP1))%>%
  mutate(TRAV_SP1 = replace(TRAV_SP1,is.na(TRAV_SP1),mean(TRAV_SP1,na.rm=TRUE))) %>%
  group_by(STATE)%>%
  summarise(mean=mean(TRAV_SP1))%>%
  top_n(5,mean)
```

Warning: NAs introduced by coercion

```
## # A tibble: 5 x 2
##   STATE      mean
##   <chr>      <dbl>
## 1 Colorado    56.9
## 2 Nevada      62.1
## 3 North Dakota 62.2
## 4 South Dakota 69.5
## 5 Wyoming     61.3
```

#6. Rank the travel speed by MONTH

```
y%>%
  mutate(TRAV_SP1=str_replace(TRAV_SP, " MPH", "")) %>%
  mutate(TRAV_SP1 = as.numeric(TRAV_SP1))%>%
  mutate(TRAV_SP1 = replace(TRAV_SP1, is.na(TRAV_SP1), mean(TRAV_SP1, na.rm=TRUE))) %>%
  group_by(MONTH)%>%
  summarise(mean=mean(TRAV_SP1))%>%
  arrange(desc(mean))
```

Warning: NAs introduced by coercion

```
## # A tibble: 12 x 2
##   MONTH      mean
##   <chr>      <dbl>
## 1 December    52.8
## 2 April       52.4
## 3 September   52.3
## 4 June        51.5
## 5 November    51.4
## 6 October     51.4
## 7 August      50.1
## 8 May         49.9
## 9 February    49.8
## 10 July       49.1
## 11 March      48.9
## 12 January    48.7
```

#7. Find the average speed of teenagers in December.

```
y%>%
  mutate(TRAV_SP1=str_replace(TRAV_SP, " MPH", "")) %>%
  mutate(TRAV_SP1 = as.numeric(TRAV_SP1))%>%
  mutate(TRAV_SP1 = replace(TRAV_SP1, is.na(TRAV_SP1), mean(TRAV_SP1, na.rm=TRUE))) %>%
  filter(AGE<18 & MONTH=="December")%>%
  summarise(mean(TRAV_SP1))
```

Warning: NAs introduced by coercion

```
## # A tibble: 1 x 1
##   `mean(TRAV_SP1)`
##               <dbl>
## 1              58.1
```


#8. Find the month that female drivers drive fastest on average.

```
y%>%
  mutate(TRAV_SP1=str_replace(TRAV_SP," MPH","")) %>%
  mutate(TRAV_SP1 = as.numeric(TRAV_SP1))%>%
  mutate(TRAV_SP1 = replace(TRAV_SP1,is.na(TRAV_SP1),mean(TRAV_SP1,na.rm=TRUE))) %>%
  filter(SEX=="Female")%>%
  group_by(MONTH)%>%
  summarise(mean=mean(TRAV_SP1))%>%
  top_n(1,mean)
```

Warning: NAs introduced by coercion

```
## # A tibble: 1 x 2
##   MONTH      mean
##   <chr>    <dbl>
## 1 December  53.5
```

#9. Find the month that male driver drive slowest on average.

```
y%>%
  mutate(TRAV_SP1=str_replace(TRAV_SP," MPH","")) %>%
  mutate(TRAV_SP1 = as.numeric(TRAV_SP1))%>%
  mutate(TRAV_SP1 = replace(TRAV_SP1,is.na(TRAV_SP1),mean(TRAV_SP1,na.rm=TRUE))) %>%
  filter(SEX=="Male")%>%
  group_by(MONTH)%>%
  summarise(mean=mean(TRAV_SP1))%>%
  arrange(mean)
```

Warning: NAs introduced by coercion

```
## # A tibble: 12 x 2
##   MONTH      mean
##   <chr>    <dbl>
## 1 February  48.3
## 2 May      48.3
## 3 July     49.0
## 4 March    49.0
## 5 January  49.1
## 6 June     51.9
## 7 August   52.0
## 8 September 52.1
## 9 December 52.4
## 10 November 52.5
## 11 October  52.7
## 12 April    53.6
```

#It's February.

#10. Create a new column containing information about the season of the accidents. Compare the percenta

```
y1 = y%>%
  mutate(SEASON=case_when(
    MONTH == "March" | MONTH == "April" | MONTH == "May" ~ "Spring",
```

```
MONTH == "June" | MONTH == "July" | MONTH == "August" ~ "Summer",
MONTH == "September" | MONTH == "October" | MONTH == "November" ~ "Autumn",
MONTH == "December" | MONTH == "January" | MONTH == "February" ~ "Winter"))
```

#continue question10

```
y1 %>%
  group_by(SEASON) %>% summarize(prop=prop.table(table(INJ_SEV))[1])
```

```
## # A tibble: 4 x 2
##   SEASON prop
##   <chr>   <dbl>
## 1 Autumn 0.440
## 2 Spring 0.418
## 3 Summer 0.459
## 4 Winter 0.409
```

#11. Compare the percentage of fatal injuries for different type of deformations (DEFORMED variable)
y%>%

```
group_by(INJ_SEV, DEFORMED) %>%
  summarise(n = n()) %>%
  mutate(prop = n / sum(n))
```

```
## # A tibble: 33 x 4
## # Groups:   INJ_SEV [7]
##   INJ_SEV          DEFORMED      n  prop
##   <chr>         <chr>    <int> <dbl>
## 1 Fatal Injury (K) Disabling Damage  315 0.726
## 2 Fatal Injury (K) Functional Damage    9 0.0207
## 3 Fatal Injury (K) Minor Damage      7 0.0161
## 4 Fatal Injury (K) No Damage        2 0.00461
## 5 Fatal Injury (K) Not Reported     9 0.0207
## 6 Fatal Injury (K) Unknown          7 0.0161
## 7 Fatal Injury (K) <NA>           85 0.196
## 8 Injured, Severity Unknown Disabling Damage    3 1
## 9 No Apparent Injury (0) Disabling Damage   94 0.355
## 10 No Apparent Injury (0) Functional Damage   63 0.238
## # ... with 23 more rows
```