Midterm 2 W24

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Instructions

Answer the following questions and complete the exercises in RMarkdown. Please embed all of your code and push your final work to your repository. Your code must be organized, clean, and run free from errors. Remember, you must remove the # for any included code chunks to run. Be sure to add your name to the author header above.

Your code must knit in order to be considered. If you are stuck and cannot answer a question, then comment out your code and knit the document. You may use your notes, labs, and homework to help you complete this exam. Do not use any other resources- including Al assistance.

Don't forget to answer any questions that are asked in the prompt. Some questions will require a plot, but others do not- make sure to read each question carefully.

For the questions that require a plot, make sure to have clearly labeled axes and a title. Keep your plots clean and professional-looking, but you are free to add color and other aesthetics.

Be sure to follow the directions and upload your exam on Gradescope.

Background

In the data folder, you will find data about shark incidents in California between 1950-2022. The data (https://catalog.data.gov/dataset/shark-incident-database-california-56167) are from: State of California-Shark Incident Database.

Load the libraries

library("tidyverse")
library("janitor")
library("naniar")

Load the data

Run the following code chunk to import the data.

sharks <- read csv("data/SharkIncidents 1950 2022 220302.csv") %>% clean names()

Questions

1. (1 point) Start by doing some data exploration using your preferred function(s). What is the structure of the data? Where are the missing values and how are they represented?

glimpse(sharks)

```
## Rows: 211
## Columns: 16
                    <chr> "1", "2", "3", "4", "5", "6", "7", "8", "9", "10", "1...
## $ incident num
                    <dbl> 10, 5, 12, 2, 8, 4, 10, 5, 6, 7, 10, 11, 4, 5, 5, 8, ...
## $ month
## $ day
                    <dbl> 8, 27, 7, 6, 14, 28, 12, 7, 14, 28, 4, 10, 24, 19, 21...
                    <dbl> 1950, 1952, 1952, 1955, 1956, 1957, 1958, 1959, 1959,...
## $ year
                    <chr> "12:00", "14:00", "14:00", "12:00", "16:30", "13:30",...
## $ time
## $ county
                    <chr> "San Diego", "San Diego", "Monterey", "Monterey", "Sa...
                    <chr> "Imperial Beach", "Imperial Beach", "Lovers Point", "...
## $ location
                    <chr> "Swimming", "Swimming", "Freediving", "Sw...
## $ mode
                    <chr> "major", "minor", "fatal", "minor", "major", "fatal",...
## $ injury
                    <chr> "surface", "surface", "surface", "surface"...
## $ depth
## $ species
                    <chr> "White", "White", "White", "White", "White", "White", "
                    <chr> "Body Surfing, bit multiple times on leg, thigh and b...
## $ comment
                    <chr> "-117.1466667", "-117.2466667", "-122.05", "-122.15",...
## $ longitude
                    <dbl> 32.58833, 32.58833, 36.62667, 36.62667, 35.13833, 35....
## $ latitude
## $ confirmed_source <chr> "Miller/Collier, Coronado Paper, Oceanside Paper", "G...
```

summary(sharks)

```
incident num
##
                           month
                                              dav
                                                              year
   Length:211
##
                       Min.
                              : 1.000
                                        Min.
                                                : 1.00
                                                         Min.
                                                                :1950
   Class :character
                                         1st Qu.: 7.50
##
                       1st Qu.: 6.000
                                                         1st Ou.:1985
   Mode :character
                       Median : 8.000
                                        Median :18.00
                                                         Median :2004
##
##
                       Mean
                              : 7.858
                                         Mean
                                                :16.54
                                                         Mean
                                                                :1998
                       3rd Qu.:10.000
                                         3rd Qu.:25.00
                                                         3rd Qu.:2014
##
##
                       Max.
                              :12.000
                                        Max.
                                                :31.00
                                                         Max.
                                                                :2022
##
##
        time
                          county
                                             location
                                                                  mode
                       Length:211
##
   Length:211
                                           Length:211
                                                              Length:211
   Class :character
                       Class :character
                                           Class :character
                                                              Class :character
##
   Mode :character
                       Mode :character
                                           Mode :character
                                                              Mode :character
##
##
##
##
##
                          depth
##
       injury
                                             species
                                                                comment
##
   Length:211
                       Length:211
                                           Length:211
                                                              Length:211
##
   Class :character
                       Class :character
                                           Class :character
                                                              Class :character
   Mode :character
                       Mode :character
                                           Mode :character
                                                              Mode :character
##
##
##
##
##
     longitude
##
                          latitude
                                        confirmed source
                                                           wfl case number
##
   Length:211
                       Min.
                              :32.59
                                        Length:211
                                                           Length:211
                                        Class :character
                                                           Class :character
##
   Class :character
                       1st 0u.:34.04
   Mode :character
                       Median :36.70
                                        Mode :character
                                                           Mode :character
##
##
                       Mean
                              :36.36
                       3rd Qu.:38.18
##
                              :41.56
##
                       Max.
##
                       NA's
                              :6
```

```
sharks%>%map_df(~ sum(is.na(.)))
```

```
## # A tibble: 1 × 16
     incident num month
##
                          day year time county location mode injury depth
##
            <int> <int> <int> <int> <int> <int>
                                                    <int> <int> <int> <int>
                                  0
                                        7
                                                                            0
## 1
                      0
                            0
                                               0
                                                        0
                                                              0
## # i 6 more variables: species <int>, comment <int>, longitude <int>,
       latitude <int>, confirmed source <int>, wfl case number <int>
## #
```

```
sharks %>% naniar::miss_var_summary()
```

```
## # A tibble: 16 × 3
##
      variable
                       n_miss pct_miss
##
      <chr>
                        <int>
                                  <dbl>
                                 95.7
## 1 wfl case number
                          202
## 2 time
                            7
                                 3.32
## 3 latitude
                            6
                                 2.84
## 4 longitude
                            5
                                 2.37
## 5 confirmed source
                            1
                                 0.474
## 6 incident_num
## 7 month
                            0
                                 0
## 8 day
                            0
                                 0
## 9 year
                            0
## 10 county
                            0
## 11 location
                            0
                                  0
## 12 mode
                            0
                                  0
## 13 injury
                            0
                                  0
## 14 depth
                            0
                                  0
## 15 species
                            0
                                  0
## 16 comment
                            0
```

2. (1 point) Notice that there are some incidents identified as "NOT COUNTED". These should be removed from the data because they were either not sharks, unverified, or were provoked. It's OK to replace the sharks object.

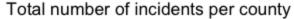
```
names(sharks)
```

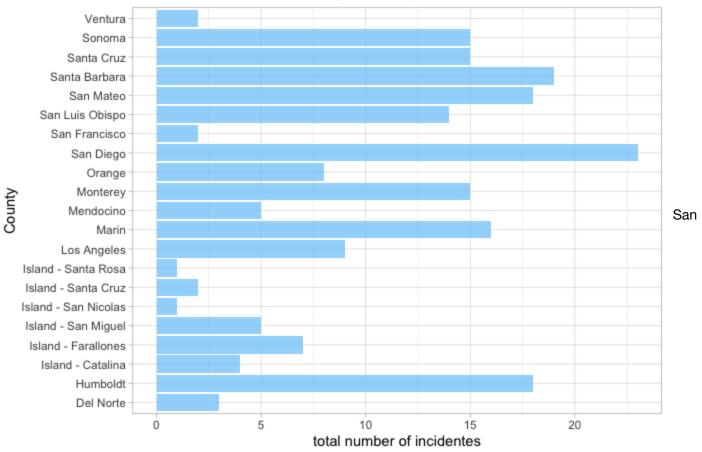
```
"dav"
                                                                    "vear"
##
    [1] "incident num"
                            "month"
                                                                    "mode"
   [5] "time"
                                                "location"
##
                            "county"
   [9] "injury"
                                                "species"
                                                                    "comment"
##
                            "depth"
## [13] "longitude"
                            "latitude"
                                                "confirmed_source" "wfl_case_number"
```

```
sharks <- sharks%>%
  filter(incident_num != "NOT COUNTED")
```

3. (3 points) Are there any "hotspots" for shark incidents in California? Make a plot that shows the total number of incidents per county. Which county has the highest number of incidents?

```
sharks%>%
  ggplot(aes(x = county))+
  geom_bar(fill = "lightskyblue", alpha = 0.8)+
  coord_flip()+
  labs(title = "Total number of incidents per county", y = "total number of incidentes",
x = "County")+
  theme_light()
```



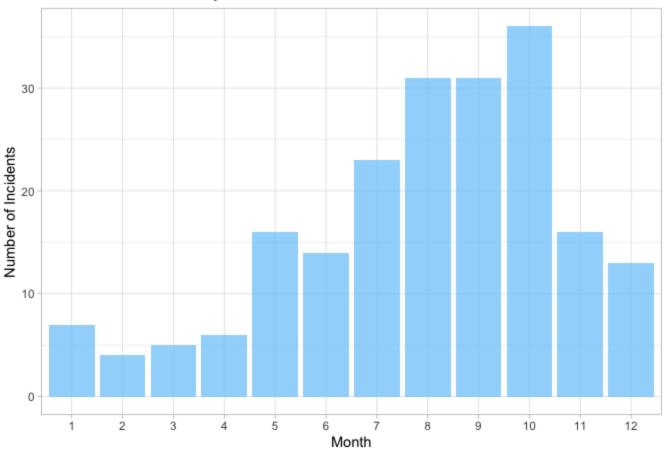


Diego has the highest number of incidents

4. (3 points) Are there months of the year when incidents are more likely to occur? Make a plot that shows the total number of incidents by month. Which month has the highest number of incidents?

```
sharks%>%
  ggplot(aes(x = as.factor(month)))+
  geom_bar(fill = "lightskyblue", alpha = 0.8)+
  labs(title = "Number of Indicants by Months", x="Month", y = "Number of Incidents")+
  theme_light()
```

Number of Indicants by Months



October has the highest number of incidents

5. (3 points) How do the number and types of injuries compare by county? Make a table (not a plot) that shows the number of injury types by county. Which county has the highest number of fatalities?

sharks%>%
 tabyl(county, injury)

```
county fatal major minor none
##
##
                 Del Norte
                                               2
                                                    9
##
                  Humboldt
                                 0
                                        7
                                               2
        Island - Catalina
                                 0
                                               1
                                                    3
##
                                        0
##
     Island - Farallones
                                 0
                                        7
                                               0
                                        2
##
     Island - San Miguel
                                 1
                                               2
                                                    0
##
    Island - San Nicolas
                                 0
                                        0
                                               1
                                                    0
##
     Island - Santa Cruz
                                 0
                                        0
                                               1
                                                    1
##
     Island - Santa Rosa
                                 0
                                        1
                                                    0
                                                    2
##
              Los Angeles
                                 1
                                        0
                                               6
                     Marin
                                 0
                                        9
                                               4
                                                    3
##
                 Mendocino
                                 1
                                        3
                                               1
                                                    0
##
                                 2
                                        8
                                               2
                                                    3
##
                  Monterey
                                                    5
                    0range
                                 0
                                        1
                                               2
##
##
                 San Diego
                                 2
                                        4
                                               8
                                                    9
##
            San Francisco
                                 1
                                        0
                                                    1
          San Luis Obispo
                                 3
                                                    7
##
                                        3
                                               1
##
                 San Mateo
                                 1
                                        1
                                               4
                                                   12
            Santa Barbara
                                 2
                                        2
                                               6
                                                    9
##
                                 1
                                        3
                                                    8
##
                Santa Cruz
                                               3
                                        8
##
                    Sonoma
                                 0
                                               1
                                                    6
##
                                 0
                                        0
                                               2
                                                    0
                   Ventura
```

San Luis Obispo has the largest number of fatalities.

6. (2 points) In the data, mode refers to a type of activity. Which activity is associated with the highest number of incidents?

```
sharks%>%
  count(mode, sort = T)
```

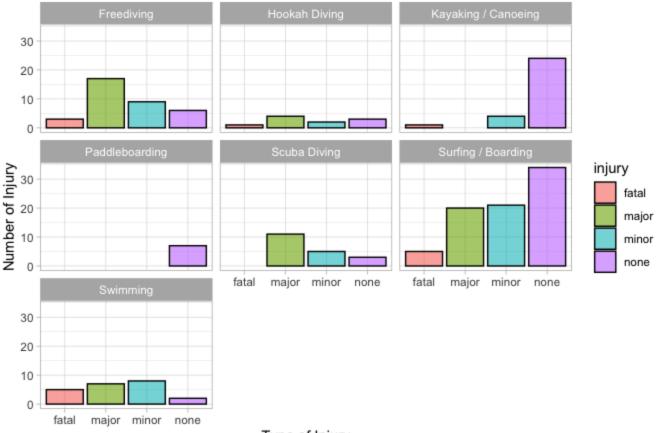
```
## # A tibble: 7 × 2
##
     mode
                              n
##
     <chr>
                          <int>
## 1 Surfing / Boarding
                             80
## 2 Freediving
                             35
## 3 Kayaking / Canoeing
                             29
## 4 Swimming
                             22
                             19
## 5 Scuba Diving
## 6 Hookah Diving
                             10
## 7 Paddleboarding
                              7
```

Surfing/Boarding is associated with the highest number of incidents

7. (4 points) Use faceting to make a plot that compares the number and types of injuries by activity. (hint: the x axes should be the type of injury)

```
sharks%>%
  ggplot(aes(x=injury, fill=injury))+
  geom_bar( color="black", alpha=0.6, na.rm=T)+
  facet_wrap(.~ mode)+
  labs(title="Comparison of Injury by Activity", x="Type of Injury", y="Number of Injury")+
  theme_light()
```

Comparison of Injury by Activity



Type of Injury

8. (1 point) Which shark species is involved in the highest number of incidents?

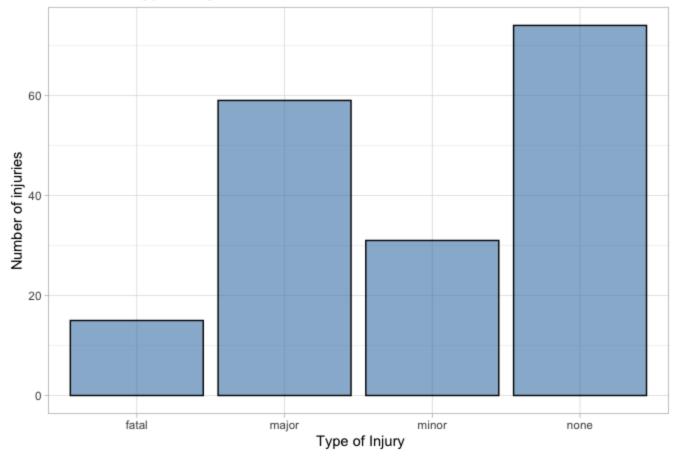
```
sharks%>%
  count(species, sort = T)
```

```
## # A tibble: 8 × 2
     species
##
##
     <chr>
                 <int>
## 1 White
                   179
## 2 Unknown
                    13
                     3
## 3 Hammerhead
## 4 Blue
                     2
## 5 Leopard
                     2
## 6 Salmon
                     1
## 7 Sevengill
                     1
## 8 Thresher
                     1
```

White sharks are involved in the highest number of incidents

9. (3 points) Are all incidents involving Great White's fatal? Make a plot that shows the number and types of injuries for Great White's only.

Number & Type of injuries for Great Whites



Not all incidents involving Great White's fatal

Background

Let's learn a little bit more about Great White sharks by looking at a small dataset that tracked 20 Great White's in the Fallaron Islands. The data (https://link.springer.com/article/10.1007/s00227-007-0739-4) are from: Weng et al. (2007) Migration and habitat of white sharks (*Carcharodon carcharias*) in the eastern Pacific Ocean.

Load the data

```
white_sharks <- read_csv("data/White sharks tracked from Southeast Farallon Island, CA, USA, 1999 2004.csv", na = c("?", "n/a")) %>% clean_names()
```

10. (1 point) Start by doing some data exploration using your preferred function(s). What is the structure of the data? Where are the missing values and how are they represented?

glimpse(white_sharks)

```
## Rows: 20
## Columns: 10
## $ shark
                    <chr> "1-M", "2-M", "3-M", "4-M", "5-F", "6-M", "7-F", "8-M"...
                    <chr> "19-0ct-99", "30-0ct-99", "16-0ct-00", "5-Nov-01", "5-...
## $ tagging_date
## $ total length cm <dbl> 402, 366, 457, 457, 488, 427, 442, 380, 450, 530, 427,...
                    ## $ sex
                    <chr> "Mature", "Adolescent", "Mature", "Mature", "Mature", ...
## $ maturity
                    <chr> "2-Nov-99", "25-Nov-99", "16-Apr-01", "6-May-02", "19-...
## $ pop_up_date
                    <dbl> 14, 26, 182, 182, 256, 275, 35, 60, 209, 91, 182, 240,...
## $ track days
## $ longitude
                    <dbl> -124.49, -125.97, -156.80, -141.47, -133.25, -138.83, ...
                    <dbl> 38.95, 38.69, 20.67, 26.39, 21.13, 26.50, 37.07, 34.93...
## $ latitude
## $ comment
                    <chr> "Nearshore", "Nearshore", "To Hawaii", "To Hawaii", "O...
```

summary(white_sharks)

```
##
       shark
                       tagging_date
                                           total_length_cm
                                                                sex
   Length:20
                       Length:20
                                           Min.
                                                  :360.0
                                                            Length:20
##
   Class :character
                       Class :character
                                           1st Qu.:400.5
                                                            Class :character
##
   Mode :character
                       Mode :character
                                           Median :434.5
                                                           Mode :character
##
##
                                           Mean
                                                  :436.1
                                           3rd Ou.:457.0
##
                                                  :530.0
##
                                           Max.
##
     maturity
                       pop_up_date
                                             track_days
                                                              longitude
##
                                                  : 14.0
   Length:20
                       Length:20
                                                           Min.
##
                                           Min.
                                                                   :-156.8
   Class :character
                       Class :character
                                           1st Qu.: 85.0
                                                            1st Qu.:-137.8
##
   Mode :character
                       Mode :character
                                           Median :182.0
                                                           Median :-133.2
##
                                           Mean
                                                  :166.8
                                                           Mean :-120.3
##
                                           3rd Qu.:216.8
                                                            3rd Qu.:-124.3
##
##
                                           Max.
                                                  :367.0
                                                           Max.
                                                                  : 131.7
##
                                                           NA's
                                                                   :1
##
       latitude
                      comment
##
   Min.
           :20.67
                    Length:20
   1st 0u.:22.48
##
                    Class :character
   Median :26.39
                    Mode :character
##
   Mean
          :28.24
##
##
   3rd Qu.:36.00
           :38.95
##
   Max.
   NA's
           :1
##
```

```
white_sharks%>%map_df(~ sum(is.na(.)))
```

```
## # A tibble: 1 × 10
##
     shark tagging_date total_length_cm sex maturity pop_up_date track_days
##
     <int>
                  <int>
                                   <int> <int>
                                                  <int>
                                                               <int>
                                                                          <int>
                                             3
                                                      1
## 1
                      0
                                       0
                                                                              0
## # i 3 more variables: longitude <int>, latitude <int>, comment <int>
```

```
white_sharks %>% naniar::miss_var_summary()
```

```
## # A tibble: 10 × 3
##
      variable
                       n_miss pct_miss
##
      <chr>
                        <int>
                                  <dbl>
##
   1 sex
                            3
                                     15
                                      5
##
    2 maturity
                            1
##
    3 longitude
                            1
                                      5
   4 latitude
                            1
                                      5
##
##
   5 shark
                                      0
## 6 tagging_date
                            0
                                      0
   7 total length cm
                                      0
##
##
   8 pop_up_date
                                      0
## 9 track_days
                            0
                                      0
## 10 comment
                            0
                                      0
```

```
names(white_sharks)
```

```
## [1] "shark" "tagging_date" "total_length_cm" "sex"
## [5] "maturity" "pop_up_date" "track_days" "longitude"
## [9] "latitude" "comment"
```

11. (3 points) How do male and female sharks compare in terms of total length? Are males or females larger on average? Do a quick search online to verify your findings. (hint: this is a table, not a plot).

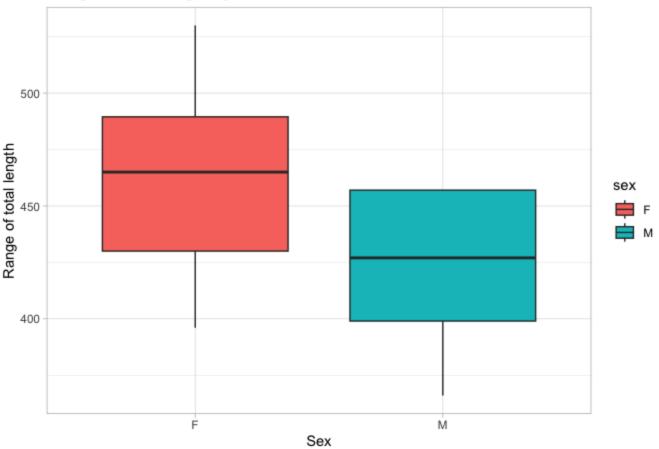
```
white_sharks%>%
  filter(sex != "NA")%>%
  group_by(sex)%>%
  summarize(mean_length = mean(total_length_cm))
```

In this data, Female white shark has a larger body length on average than male white sharks. I got the same results online (Females are generally bigger than males. Male great whites on average measure 3.4 to 4.0 m (11 to 13 ft) in length, while females measure 4.6 to 4.9 m (15 to 16 ft).)

12. (3 points) Make a plot that compares the range of total length by sex.

```
white_sharks%>%
  filter(sex != "NA")%>%
  group_by(sex)%>%
  ggplot(aes(x = sex, y = total_length_cm, fill = sex))+
  geom_boxplot()+
  labs(title = "Range of total length by sex", x = "Sex", y = "Range of total length")+
  theme_light()
```

Range of total length by sex



13. (2 points) Using the sharks or the white_sharks data, what is one question that you are interested in exploring? Write the question and answer it using a plot or table.

Question: make a plot that compares the number and types of injuries by shark speceis.

```
sharks %>%
  ggplot(aes(x=species, fill = injury)) +
  geom_bar(na.rm = T, position = "dodge", alpha=0.7)+
  coord_flip()+
  labs(title = "Number and types of Injuries by shark speceis", y = "Number of injurie
s", x = "Shark species")+
  theme_light()
```

Number and types of Injuries by shark speceis

