Report

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Cleaning Data

```
av_data_1 <- read.delim('../rawdata/aviation_data.txt', header = TRUE,</pre>
                         sep = '|' , stringsAsFactors = FALSE)
air_data_1 <- read.csv('../rawdata/airplane_crashes.csv',</pre>
                        stringsAsFactors = FALSE)
# This takes only the airplane data fromt the data frame as opposed to all
# aircrafts.
av_data_2 <- subset(av_data_1, Aircraft.Category == ' Airplane '|</pre>
                      Aircraft.Category == ' ')
air_data_2 <- air_data_1[!grepl('airship', air_data_1$Type), ]</pre>
# We will merge these two date tables so that air_data provides information
# from 1920-1981 and av_data information from 1982-2015. So now we extract
# only the years we want from each data frame.
acc_date <- av_data_2$Event.Date</pre>
av_data_2$year <- as.numeric(substr(acc_date, start = nchar(acc_date)-4,
                                     stop = nchar(acc_date)))
av_data_3 <- subset(av_data_2, year >= 1982)
ac_date <- air_data_2$Date</pre>
air_data_2$year <- as.numeric(substr(ac_date, start = nchar(ac_date)-3,
                                      stop = nchar(ac_date)))
air_data_3 <- subset(air_data_2, year <= 1981 & year >= 1920)
# In av_data the columns 'Location' and 'Country' should be combined as well
# as 'Make' and 'Model'
av_data_3$new_location <- paste0(av_data_3$Location, av_data_3$Country)
av_data_3$Type <- paste0(av_data_3$Make, av_data_3$Model)</pre>
# In air_data the order of data ia currently in ascending order, but to match
# av_data we want it to be in descending order.
air_data_4 <- air_data_3[nrow(air_data_3):1,]</pre>
# Now to extract the information (columns) that we want
av_data_4 <- av_data_3[c('Event.Date', 'new_location', 'Type')]</pre>
air_data_5 <- air_data_4[c('Date', 'Location', 'Type')]</pre>
# We don't want the crash information for crashes missing 'Type' information.
av_data_5 <- subset(av_data_4, Type != ' ')</pre>
air_data_6 <- subset(air_data_5, Type != '')</pre>
# Rename column names so that they are the same for both data frames.
names(av data 5) <- c('date', 'location', 'type')</pre>
names(air_data_6) <- c('date', 'location', 'type')</pre>
```

```
# Now we combine the two data frames.
data <- rbind(av_data_5, air_data_6)

# Changing the 'type' column to all lowercase
data$type <- tolower(data$type)

# Changing the 'location' column to all lowercase
data$location <- tolower(data$location)

# Showing the head of the cleaned data
head(data)</pre>
```

```
##
                                         location
            date
## 1 11/06/2015
                      gonzales, la united states
## 2 11/04/2015 mount pocono, pa united states
## 3 11/04/2015
                       wakeman, oh united states
## 4 11/04/2015
                        haines, ak united states
## 6 11/03/2015
                      tecumseh, mi united states
                fayetteville, ar united states
## 7 11/03/2015
##
                                type
## 1
                        cessna 120
## 2
                  schweizer 269c 1
## 3
     american autogyro sparrowhawk
## 4
                        cessna 180
## 6
                   piper pa 18-150
## 7
          cirrus design corp sr22t
```

```
# Exporting the 'data' file into the data folder.
write.csv(data, file = '../data/data.csv')
```

Part 1: Monthly Data

In this part of the project, we were looking to see which of the months had the most plane crashes. In order to do this, we will need to first load the necessary code and packages.

```
# packages needed
library(stringr)

# function needed
source('../code/num_name.R')
```

Our cleaned data set does not have a column that just contains the months when each accident happens, so we will had to make our own by doing the following.

```
# importing data
data <- read.csv('../data/data.csv', stringsAsFactors = FALSE)

# removing first column due to repeated index
data$X <- NULL

# extracting months</pre>
```

```
data$month <- as.numeric(str_extract(data$date, '[0-9]+'))

# labeling months with words using the num_name function.
dates <- data$month
month <- num_name(dates)

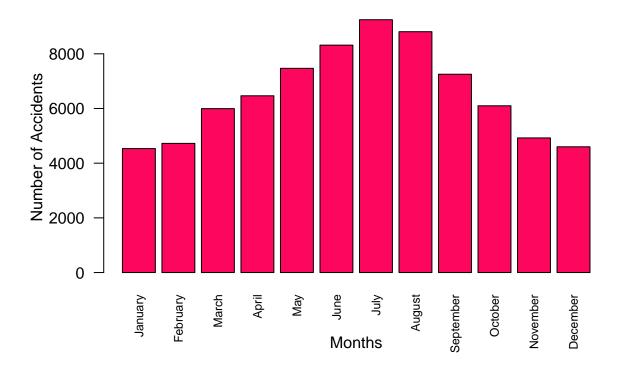
# seeing the new dataframe
head(data)</pre>
```

```
##
            date
                                         location
## 1 11/06/2015
                      gonzales, la united states
## 2 11/04/2015
                mount pocono, pa united states
## 3 11/04/2015
                       wakeman, oh united states
## 4 11/04/2015
                       haines, ak united states
## 5 11/03/2015
                      tecumseh, mi united states
## 6 11/03/2015 fayetteville, ar united states
##
                               type month
## 1
                        cessna 120
                                       11
## 2
                  schweizer 269c 1
                                       11
## 3 american autogyro sparrowhawk
                                       11
## 4
                        cessna 180
                                       11
## 5
                   piper pa 18-150
                                       11
## 6
          cirrus design corp sr22t
                                       11
```

With this column, we are now ready to graph to see our findings.

```
# Number of Entries Each Month
month_names <- c('January', 'February', 'March', 'April', 'May', 'June',</pre>
                  'July', 'August', 'September', 'October', 'November',
                  'December')
monthly data = NULL
for (name in month_names){
 total_per_month <- sum(month == name)</pre>
  monthly_data <- c(monthly_data, total_per_month)</pre>
# Graphing the months
barplot(monthly_data, xlab = 'Months',
        ylab = 'Number of Accidents',
        main = 'Number of Crashes Each Month',
        names.arg = month_names,
        las = 2,
        cex.names = .75,
        col = '\#FD075E')
```

Number of Crashes Each Month



Our original belief was that the number of accidents every month would be relatively uniform, but instead, we found that there are more accidents that occur during the middle of the year, with the most occurring in July. We found this idea to be interesting and surprising. The number of accidents could have increased due to there being more traveling being done during the summer due to kids being on summer vacation. However, during the summer seasons, there are also a lot of tropical storms, thunderstorms, and changes in the weather that could affect how safe it is to travel and increases the chance of getting into a plane accident. According to a New York Times article, the number of airplane accidents could be greater in the summer months because that is typically when people fly their private airplanes. A smaller, private aircraft is more likely to get into an accident than a commercial plane. The article states that, "Statistics from the N.T.S.B. show that general aviation aircraft average nearly seven accidents per 100,000 flight hours, compared with an average of 0.16 accidents per 100,000 hours for commercial airlines" (Fowler).

Finally, we will export our graphics.

pdf

```
## 2
```

Part 2: Data by Decades

```
# packages needed
library(stringr)

# function needed
source('../code/norm_decade.R')
```

We wanted to look at the number of accidents per decade to see if any particular decades differed from the rest. However, the data is not organized into decades. First, we used regular expressions to pull the year out of the date column and created a new column named "year".

```
# making a year only column
data$year <- as.numeric(str_extract(data$date, '[0-9]{4}'))</pre>
```

For graphing, we wanted the axis to display decades, for example "1980s", and the corresponding aggregated data for that decade. We created a loop that categorizes the year and would group all years in a certain decade into that decade.

```
# labeling by decades
decade = NULL
for(i in 1:length(data$year)){
  x <- data$year[i]
  if ((x \%in\% c(1920:1929))==TRUE) x = '1920s'
  if ((x \%in\% c(1930:1939)) == TRUE) x = '1930s'
  if ((x \%in\% c(1940:1949)) == TRUE) x = '1940s'
  if ((x \%in\% c(1950:1959))==TRUE) x = '1950s'
  if ((x \%in\% c(1960:1969)) == TRUE) x = '1960s'
  if ((x \%in\% c(1970:1979)) == TRUE) x = '1970s'
  if ((x \%in\% c(1980:1989)) == TRUE) x = '1980s'
  if ((x \%in\% c(1990:1999))==TRUE) x = '1990s'
  if ((x \%in\% c(2000:2009))==TRUE) x = '2000s'
  if ((x \%in\% c(2010:2019))==TRUE) x = '2010s'
  decade = c(decade, x)
}
data$decade <- decade
```

Next, we looked at the frequency of accidents for each decade by calling upon the data through table().

```
# looking at freq of accidents
table(data$decade)
##
```

837 24422 22953 19419 8507

1920s 1930s 1940s 1950s 1960s 1970s 1980s 1990s 2000s 2010s

719

##

167

319

500

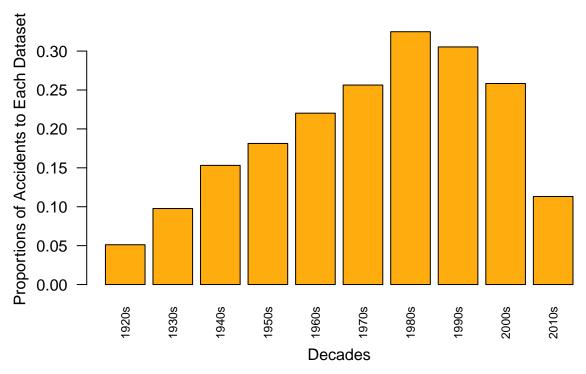
592

We started off by graphing a bar chart of the number of accidents per decade but found that the data was skewed due to the significantly larger size of the dataset for years 1980s and onwards. We then normalized the data by taking proportion of each decade's accidents to the total amount of accidents for the data set it came from. For example, 1920s proportion was (# of accidents in 1920s) / (total # of accidents of dataset1)

```
# size of each dataset
amt_dataset1 <- nrow(air_data_6)</pre>
amt_dataset2 <- nrow(av_data_5)</pre>
norm decade 1920 <- norm decade('1920s')
norm decade 1930 <- norm decade('1930s')
norm_decade_1940 <- norm_decade('1940s')</pre>
norm_decade_1950 <- norm_decade('1950s')</pre>
norm_decade_1960 <- norm_decade('1960s')</pre>
norm_decade_1970 <- norm_decade('1970s')</pre>
norm_decade_1980 <- norm_decade('1980s')</pre>
norm decade 1990 <- norm decade('1990s')
norm_decade_2000 <- norm_decade('2000s')</pre>
norm_decade_2010 <- norm_decade('2010s')</pre>
vec_norm_decade <- c(norm_decade_1920, norm_decade_1930, norm_decade_1940,
                       norm_decade_1950, norm_decade_1960, norm_decade_1970,
                       norm_decade_1980, norm_decade_1990, norm_decade_2000,
                       norm_decade_2010)
```

There seems to be a gradual increase in accidents till 1980s where it starts to drop off starting 1990s. Data from the 2000s show a decrease, which agrees with the FAA report.





According to the FAA: "Between 2001 and 2007, aviation witnessed one of its safest periods for scheduled air carriers. Not counting the terrorist activities of September 11, 2001, there were only three fatal accidents in 2001; none in 2002; two in 2003; one in 2004; three in 2005; two in 2006; and none in 2007. Fatal accidents became rare events with only .01 accidents per 100,000 flight hours or .018 accidents per 100,000 departures."

Another explanation could be the improvements of air traffic control since after the 1980s (start of the digital age).

According to Wikipedia: As computers became more sophisticated in the 2000s, they began to take over routine aspects of the air traffic controller's task. Up until then all air traffic in nearby airspace was tracked and displayed, with the air traffic controller responsible for monitoring its position and assessing any need for action. Modern computerised systems are capable of monitoring the flight paths of many more aircraft at a given time, allowing the controller to manage more aircraft and to focus on the decision-making and follow-up processes.

pdf

```
## 2
```

```
# PNG
png('../plots_and_graphics/normalized_number_of_crashes_each_decade.png',
    res = 96)
barplot(vec_norm_decade, xlab = 'Decades',
        ylab = 'Proportions of Accidents to Each Dataset',
        main = 'Normalized Number of Crashes Each Decade',
        names.arg = decade_names,
        las = 2,
        cex.names = .75,
        col = '#FFACOE')
dev.off()
```

pdf ## 2