Analysis of the Adverse Health and Economic Impacts of US Storm Data

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10/11/2021

Github repo for the Course: Reproducible Research

1: Synopsis

The objective of this assignment is to explore the NOAA Storm Database and explore the effects of severe weather events on both population and economy. The database covers the time period between 1950 and November 2011.

The following analysis determines which types of weather events are most harmful with respect to:

- 1. Population Health
- 2. Economic Consequences

Documentation of the Data: Documentation

2: Data Processing

2.1: Data Loading

Download the raw data file and extract the data into a dataframe. Then convert to a data.table

```
library("data.table")
library("ggplot2")
fileUrl <- "https://d396qusza40orc.cloudfront.net/repdata%2Fdata%2FStormData.csv.bz2"
download.file(fileUrl, destfile = paste0("", '/repdata%2Fdata%2FStormData.csv.bz2'))
stormDF <- read.csv("repdata_data_StormData.csv.bz2")
# Converting data.frame to data.table
stormDT <- as.data.table(stormDF)</pre>
```

2.2: Examining Column Names

```
colnames(stormDT)

## [1] "STATE__" "BGN_DATE" "BGN_TIME" "TIME_ZONE" "COUNTY"

## [6] "COUNTYNAME" "STATE" "EVTYPE" "BGN_RANGE" "BGN_AZI"

## [11] "BGN_LOCATI" "END_DATE" "END_TIME" "COUNTY_END" "COUNTYENDN"
```

```
## [16] "END_RANGE"
                     "END AZI"
                                   "END LOCATI" "LENGTH"
                                                              "WIDTH"
                                                              "PROPDMG"
## [21] "F"
                     "MAG"
                                   "FATALITIES" "INJURIES"
## [26] "PROPDMGEXP" "CROPDMG"
                                   "CROPDMGEXP" "WFO"
                                                              "STATEOFFIC"
## [31] "ZONENAMES"
                     "LATITUDE"
                                   "LONGITUDE" "LATITUDE_E" "LONGITUDE_"
## [36] "REMARKS"
                     "REFNUM"
```

2.3: Data Subsetting

Subset the dataset on the parameters of interest. Basically, we remove the columns we don't need for clarity.

```
# Finding columns to remove
cols2Remove <- colnames(stormDT[, !c("EVTYPE"</pre>
  , "FATALITIES"
   "INJURIES"
   "PROPDMG"
  , "PROPDMGEXP"
   "CROPDMG"
 , "CROPDMGEXP")])
# Removing columns
stormDT[, c(cols2Remove) := NULL]
# Only use data where fatalities or injuries occurred.
stormDT <- stormDT[(EVTYPE != "?" &</pre>
              (INJURIES > 0 | FATALITIES > 0 | PROPDMG > 0 | CROPDMG > 0)), c("EVTYPE"
                                                                                , "FATALITIES"
                                                                                  "INJURIES"
                                                                                  "PROPDMG"
                                                                                 "PROPDMGEXP"
                                                                                , "CROPDMG"
                                                                                  "CROPDMGEXP") ]
```

2.4: Converting Exponent Columns into Actual Exponents instead of (-,+, H, K, etc)

Making the PROPDMGEXP and CROPDMGEXP columns cleaner so they can be used to calculate property and crop cost.

```
# Change all damage exponents to uppercase.
cols <- c("PROPDMGEXP", "CROPDMGEXP")</pre>
stormDT[, (cols) := c(lapply(.SD, toupper)), .SDcols = cols]
# Map property damage alphanumeric exponents to numeric values.
propDmgKey <- c("\"" = 10^0,
                  "-" = 10^0,
                 "+" = 10^0,
                  "0" = 10^0,
                  "1" = 10^1,
                  "2" = 10^2,
                  "3" = 10^3,
                  "4" = 10^4,
                  "5" = 10^5,
                 "6" = 10^6,
                  "7" = 10^7.
                 "8" = 10^8,
                  "9" = 10^9,
```

```
"H" = 10^2,

"K" = 10^3,

"M" = 10^6,

"B" = 10^9)

# Map crop damage alphanumeric exponents to numeric values

cropDmgKey <- c("\"\"" = 10^0,

"?" = 10^0,

"0" = 10^0,

"K" = 10^3,

"M" = 10^6,

"B" = 10^9)

stormDT[, PROPDMGEXP := propDmgKey[as.character(stormDT[,PROPDMGEXP])]]

stormDT[is.na(PROPDMGEXP), PROPDMGEXP := 10^0]

stormDT[is.na(CROPDMGEXP), CROPDMGEXP := 10^0]
```

2.5: Making Economic Cost Columns

```
stormDT <- stormDT[, .(EVTYPE, FATALITIES, INJURIES, PROPDMG, PROPDMGEXP, propCost = PROPDMG * PROPDMGE
```

2.6: Calcuating Total Property and Crop Cost

5000 43323541000

HAIL 15735267513 3025954473 18761221986

2.7: Calcuating Total Fatalities and Injuries

STORM SURGE 43323536000

4:

5:

```
totalInjuriesDT <- stormDT[, .(FATALITIES = sum(FATALITIES), INJURIES = sum(INJURIES), totals = sum(FATALITIES), totalInjuriesDT <- totalInjuriesDT[order(-FATALITIES), ]
totalInjuriesDT <- totalInjuriesDT[1:10, ]
head(totalInjuriesDT, 5)</pre>
```

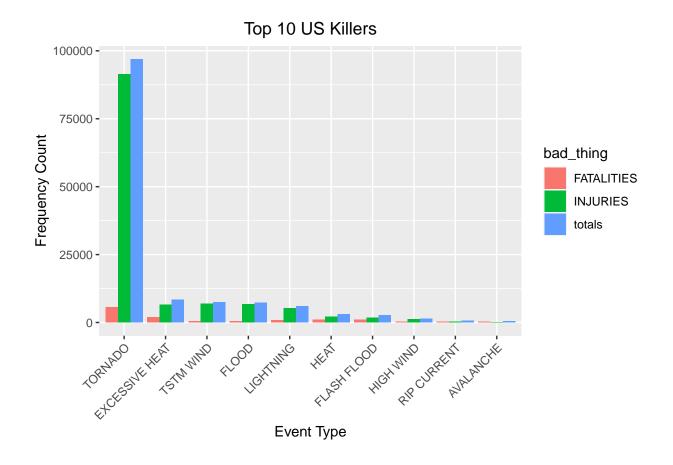
```
##
             EVTYPE FATALITIES INJURIES totals
                                 91346 96979
## 1:
            TORNADO
                          5633
## 2: EXCESSIVE HEAT
                          1903
                                  6525
                                         8428
## 3: FLASH FLOOD
                          978
                                  1777
                                         2755
## 4:
                           937
                                  2100
                                         3037
               HEAT
        LIGHTNING
                                  5230
                                         6046
## 5:
                          816
```

3: Results

3.1: Events that are Most Harmful to Population Health

Melting data.table so that it is easier to put in bar graph format

```
bad_stuff <- melt(totalInjuriesDT, id.vars="EVTYPE", variable.name = "bad_thing")</pre>
head(bad_stuff, 5)
##
              EVTYPE bad_thing value
             TORNADO FATALITIES 5633
## 1:
## 2: EXCESSIVE HEAT FATALITIES
                                 1903
## 3: FLASH FLOOD FATALITIES
                                  978
## 4:
               HEAT FATALITIES
                                  937
## 5:
           LIGHTNING FATALITIES
                                  816
# Create chart
healthChart <- ggplot(bad_stuff, aes(x=reorder(EVTYPE, -value), y=value))
# Plot data as bar chart
healthChart = healthChart + geom_bar(stat="identity", aes(fill=bad_thing), position="dodge")
# Format y-axis scale and set y-axis label
healthChart = healthChart + ylab("Frequency Count")
# Set x-axis label
healthChart = healthChart + xlab("Event Type")
# Rotate x-axis tick labels
healthChart = healthChart + theme(axis.text.x = element_text(angle=45, hjust=1))
# Set chart title and center it
healthChart = healthChart + ggtitle("Top 10 US Killers") + theme(plot.title = element_text(hjust = 0.5)
healthChart
```



3.2: Events that have the Greatest Economic Consequences

Melting data.table so that it is easier to put in bar graph format

```
econ_consequences <- melt(totalCostDT, id.vars="EVTYPE", variable.name = "Damage_Type")
head(econ_consequences, 5)</pre>
```

```
##
                 EVTYPE Damage_Type
                                           value
## 1:
                  FLOOD
                           propCost 144657709807
## 2: HURRICANE/TYPHOON
                           propCost 69305840000
## 3:
                TORNADO
                           propCost 56947380677
## 4:
            STORM SURGE
                           propCost
                                    43323536000
## 5:
                   HAIL
                           propCost
                                     15735267513
```

```
# Create chart
econChart <- ggplot(econ_consequences, aes(x=reorder(EVTYPE, -value), y=value))
# Plot data as bar chart
econChart = econChart + geom_bar(stat="identity", aes(fill=Damage_Type), position="dodge")
# Format y-axis scale and set y-axis label
econChart = econChart + ylab("Cost (dollars)")
# Set x-axis label
econChart = econChart + xlab("Event Type")
# Rotate x-axis tick labels
econChart = econChart + theme(axis.text.x = element_text(angle=45, hjust=1))</pre>
```



