Adverse Health and Economic Impacts of US Storms

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1: Synopsis

The following analysis investigates which types of severe weather events are most harmful on:

- 1. Health (injuries and fatalities)
- 2. Property and crops (economic consequences)

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("D:/r-projects/datasciencecoursera", '/repdata%2Fdata%2FStormD
dat <- read.csv("D:/r-projects/datasciencecoursera/repdata%2Fdata%2FStormData.csv.bz2")
# Converting data.frame to data.table
data <- as.data.table(dat)</pre>
```

2.2: Examining Column Names

```
colnames(data)
```

```
[1] "STATE "
                      "BGN DATE"
                                    "BGN TIME"
                                                  "TIME_ZONE"
                                                                "COUNTY"
    [6] "COUNTYNAME"
                      "STATE"
                                    "EVTYPE"
                                                  "BGN_RANGE"
                                                                "BGN_AZI"
                                    "END_TIME"
## [11] "BGN_LOCATI"
                      "END_DATE"
                                                  "COUNTY_END"
                                                               "COUNTYENDN"
        "END_RANGE"
                                    "END_LOCATI" "LENGTH"
                      "END_AZI"
                                                                "WIDTH"
                      "MAG"
                                    "FATALITIES" "INJURIES"
                                                                "PROPDMG"
## [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(data[, !c("EVTYPE"</pre>
  , "FATALITIES"
  , "INJURIES"
 , "PROPDMG"
  , "PROPDMGEXP"
 , "CROPDMG"
  , "CROPDMGEXP")])
# Removing columns
data[, c(cols2Remove) := NULL]
# Only use data where fatalities or injuries occurred.
data <- data[(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>
data[, (cols) := c(lapply(.SD, toupper)), .SDcols = cols]
# Map property damage alphanumeric exponents to numeric values.
propDmgKey <- c("\""" = 10^0,
                  "-" = 10^{\circ}0,
                  "+" = 10^0,
                  "0" = 10^{\circ}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^{\circ}0,
                 "K" = 10^3,
                 "M" = 10^6,
                 "B" = 10^9
```

```
data[, PROPDMGEXP := propDmgKey[as.character(data[,PROPDMGEXP])]]
data[is.na(PROPDMGEXP), PROPDMGEXP := 10^0 ]
data[, CROPDMGEXP := cropDmgKey[as.character(data[,CROPDMGEXP])] ]
data[is.na(CROPDMGEXP), CROPDMGEXP := 10^0 ]
```

2.5: Making Economic Cost Columns

```
data <- data[, .(EVTYPE, FATALITIES, INJURIES, PROPDMG, PROPDMGEXP, propCost = PROPDMG * PROPDMGEXP, CR
```

2.6: Calcuating Total Property and Crop Cost

```
totalCostDT <- data[, .(propCost = sum(propCost), cropCost = sum(cropCost), Total_Cost = sum(propCost)
totalCostDT <- totalCostDT[order(-Total_Cost), ]
totalCostDT <- totalCostDT[1:10, ]
head(totalCostDT, 5)</pre>
## EVTYPE propCost cropCost Total_Cost
## 1. FIGURE 1446577709907 EGG1069450 150310678257
```

```
## 1: FLOOD 144657709807 5661968450 150319678257

## 2: HURRICANE/TYPHOON 69305840000 2607872800 71913712800

## 3: TORNADO 56947380677 414953270 57362333947

## 4: STORM SURGE 43323536000 5000 43323541000

## 5: HAIL 15735267513 3025954473 18761221986
```

2.7: Calcuating Total Fatalities and Injuries

```
totalInjuriesDT <- data[, .(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
##
## 1:
            TORNADO
                          5633
                                  91346 96979
## 2: EXCESSIVE HEAT
                          1903
                                   6525
                                          8428
## 3: FLASH FLOOD
                           978
                                   1777
                                          2755
                                   2100
## 4:
               HEAT
                           937
                                          3037
## 5:
         LIGHTNING
                           816
                                   5230
                                          6046
```

3: Results

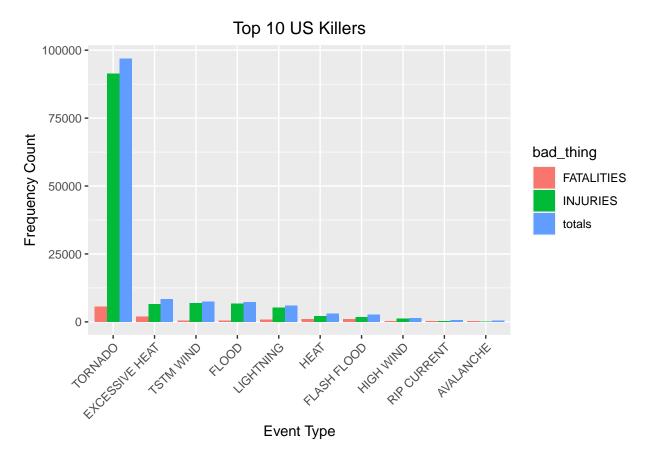
3.1: Events that are Most Harmful to Population Health

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")
head(bad_stuff, 5)</pre>
```

```
##
              EVTYPE bad_thing value
## 1:
             TORNADO FATALITIES
                                  5633
## 2: EXCESSIVE HEAT FATALITIES
                                  1903
         FLASH FLOOD FATALITIES
## 3:
                                   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</pre>
```



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)
##
                 EVTYPE Damage_Type
                                           value
## 1:
                           propCost 144657709807
## 2: HURRICANE/TYPHOON
                           propCost 69305840000
                TORNADO
                           propCost
                                     56947380677
## 4:
            STORM SURGE
                           propCost
                                    43323536000
## 5:
                           propCost 15735267513
                   HAIL
# Create chart
econChart <- ggplot(econ_consequences, aes(x=reorder(EVTYPE, -value), y=value))</pre>
# 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))
# Set chart title and center it
econChart = econChart + ggtitle("Top 10 US Storm Events causing Economic Consequences") + theme(plot.ti
econChart
```

Top 10 US Storm Events causing Economic Consequences

