The U.S. NOAA Storm Data Analysis by type of harmful events, and by events with greatest economic consequences

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Synopsis

Storms and other severe weather events can cause both public health and economic problems for communities and municipalities. Many severe events can result in fatalities, injuries, and property damage, and preventing such outcomes to the extent possible is a key concern.

This project involves exploring the U.S. National Oceanic and Atmospheric Administration's (NOAA) storm database. This database tracks characteristics of major storms and weather events in the United States, including when and where they occur, as well as estimates of any fatalities, injuries, and property damage.

The goal of this report is to answer to the following questions about severe weather events:

- (i) Across the United States, which types of events (as indicated in the EVTYPE variable) are most harmful with respect to population health?
- (ii) Across the United States, which types of events have the greatest economic consequences?

This analysis shows that the tornadoes are the most harmful events for population health in the United States, floods caused worst damages for properties, and hurricane caused big damages on crops.

Data Processing

The data for this project are in the form of a comma-separated-value file compressed via the bzip2 algorithm. Data can be downloaded from the Storm Data. In addition, there is some documentation available at the National Weather Service Storm Data Documentation where there are provided details how some of the variables are constructed/defined.

First, clear R environment and load necessary libraries.

```
rm(list=ls())
library(dplyr); library(knitr); library(ggplot2); library(tidyr); library(car)
```

Download the file.

```
if(!file.exists("./StormData")){dir.create("./StormData")}
fileUrl <- "http://d396qusza40orc.cloudfront.net/repdata%2Fdata%2FStormData.csv.bz2"
download.file(fileUrl, destfile = "./StormData/repdataFdataFStormData.csv.bz2", method = "curl", mode =
```

Read the data.

```
storm <- read.csv(bzfile("./StormData/repdataFdataFStormData.csv.bz2"),</pre>
                  header = TRUE, sep = ",", stringsAsFactors = FALSE, na.strings="NA")
```

Extract variables which are relevant to our analysis:

```
relevant.storm.data <- storm %>%
       select(STATE, EVTYPE, FATALITIES, INJURIES, PROPDMG, PROPDMGEXP, CROPDMG, CROPDMGEXP)
str(relevant.storm.data)
## 'data.frame':
                  902297 obs. of 8 variables:
   $ STATE : chr "AL" "AL" "AL" "AL" ...
                     "TORNADO" "TORNADO" "TORNADO" ...
   $ EVTYPE : chr
   $ FATALITIES: num 0 0 0 0 0 0 0 1 0 ...
## $ INJURIES : num 15 0 2 2 2 6 1 0 14 0 ...
## $ PROPDMG
             : num 25 2.5 25 2.5 2.5 2.5 2.5 2.5 25 25 ...
## $ PROPDMGEXP: chr "K" "K" "K" "K" ...
## $ CROPDMG
             : num 0000000000...
## $ CROPDMGEXP: chr "" "" "" ...
```

Rename variables as follow:

- STATE (US state abbreviation) -> State
- EVTYPE (Type of event) -> Event
- FATALITIES (Number of fatalities) -> Fatalities
- INJURIES (Number of injuries) -> Injuries
- PROPDMG (Amount of property damage in orders of magnitude) -> Property.Damage.Magnitude
- \bullet PROPDMGEXP (Order of magnitude for property damage in \$H, K, M, B) -> Property. Damage.
Expenses
- CROPDMG (Amount of crop damage in orders of magnitude) -> Crop.Damage.Magnitude
- PROPDMGEXP (Order of magnitude for crop damage in \$H, K, M, B) -> Crop.Damage.Expenses

where abbreviations H, K, M, B represent H = hundreds, K = thousands, M = millions, B = billions.

```
names(relevant.storm.data) <- c("State", "Event", "Fatalities", "Injuries", "Property.Damage.Magnitude"</pre>
```

To avoid duplicates and various problems due to mixt uppercase/lowercase, spelling errors and events with many sub-categories, renames are performed for the events.

```
relevant.storm.data$Event <- gsub("^(HEAT).*", "Heat", relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(RECORD HEAT).*", "Heat", relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(EXTREME HEAT).*", "Heat", relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(Heat).*", "Heat", relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(EXCESSIVE HEAT).*", "Heat", relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(TSTM).*", "Thunder.Storm", relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(THUNDERSTORM).*", "Thunder.Storm",relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(LIGHTNING).*", "Lightning",relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(TROPICAL STORM).*", "Tropical.Storm", relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(FLASH FLOOD).*", "Flood", relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(FLOOD).*", "Flood", relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(WIND).*", "WIND", relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(STRONG WIND).*", "Wind", relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(HIGH WIND).*", "Wind", relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(HURRICANE).*", "Hurricane", relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(SNOW).*", "Snow", relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(HEAVY SNOW).*", "Snow", relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(FIRE).*", "Fire", relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(WILD/FOREST FIRE).*", "Fire", relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(WILDFIRE).*", "Fire", relevant.storm.data$Event)
```

```
relevant.storm.data$Event <- gsub("^(WILD FIRES).*", "Fire", relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(HAIL).*", "Hail", relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(BLIZZARD).*", "Blizzard", relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(COLD).*", "Cold", relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(WINTER WEATHER).*", "Cold", relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(EXTREME COLD).*", "Cold", relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(RIP).*", "Rip", relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(FOG).*", "Fog", relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(DENSE FOG).*", "Fog", relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(AVALANCHE).*", "Avalanche", relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(AVALANCE).*", "Avalanche", relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(RAIN).*", "Rain", relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(HEAVY RAIN).*", "Rain", relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(HIGH SURF).*", "Surf", relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(HEAVY SURF).*", "Surf", relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(SURF).*", "Surf", relevant.storm.data$Event)
relevant.storm.data$Event <- gsub("^(TORNADO).*", "Tornado", relevant.storm.data$Event)
```

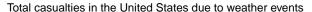
Results

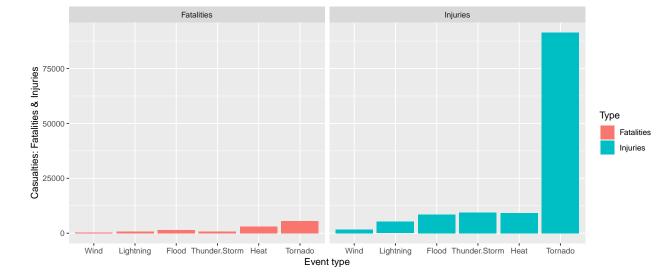
1. Across the United States, which types of events are most harmful with respect to population health?

Determine how many casualties (injuries + fatalities) were caused by type of event, and arrange them in descending order:

```
## # A tibble: 6 x 4
                Fatalities Injuries Total
##
   Event
##
    <chr>
                       <dbl>
                                <dbl> <dbl>
## 1 Tornado
                        5658
                                91364 97022
## 2 Heat
                                 9224 12343
                        3119
## 3 Thunder.Storm
                         710
                                 9508 10218
## 4 Flood
                        1513
                                 8591 10104
## 5 Lightning
                         817
                                 5232 6049
## 6 Wind
                         403
                                 1772 2175
```

Plot the casualties (injuries & fatalities) by type of event:





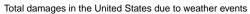
This analysis shows clearly that tornadoes are the most harmful events for population health in the United States, causing the highest number of fatalities and injuries.

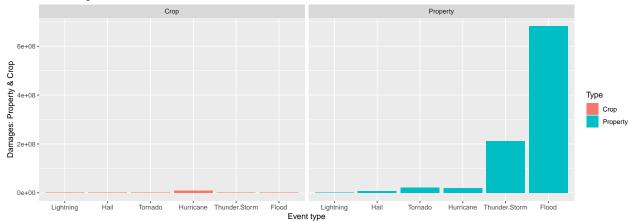
2. Across the United States, which types of events have the greatest economic consequences?

To determine numerical value of damages for both properties and crops, expenses are converted from charater to numeric:

```
relevant.storm.data$Property.Damage.Expenses <- as.numeric(recode(as.character(relevant.storm.data$Prop
    "'0'=1;'1'=10;'2'=10^2;'3'=10^3;'4'=10^4;'5'=10^5;'6'=10^6;'7'=10^7;'8'=10^8;'b'=10^9;'h'=10^2;'k'=
relevant.storm.data$Crop.Damage.Expenses <- as.numeric(recode(as.character(relevant.storm.data$Crop.Dam
    "'0'=1;'1'=10;'2'=10^2;'3'=10^3;'4'=10^4;'5'=10^5;'6'=10^6;'7'=10^7;'8'=10^8;'b'=10^9;'h'=10^2;'k'=
Calculate amount of the property and crop damages by type of event, and arrange them in descending order:
damages <- relevant.storm.data %>%
        group_by(Event) %>%
        summarize(Property = sum(Property.Damage.Magnitude * Property.Damage.Expenses, na.rm = TRUE),
                  Crop = sum(Crop.Damage.Magnitude * Crop.Damage.Expenses, na.rm = TRUE),
                  Total = Property + Crop) %>%
        arrange(desc(Total))
damages[1:6,]
## # A tibble: 6 x 4
##
     Event
                      Property
                                              Total
                                   Crop
                                              <dbl>
##
     <chr>>
                         <dbl>
                                  <dbl>
## 1 Flood
                    682475270.
                                  10000 682485270.
                                   4080 211141523.
## 2 Thunder.Storm 211137443.
## 3 Hurricane
                    20000000 10000000
                                         30000000
## 4 Tornado
                    21720194.
                                         21720354.
                                    160
## 5 Hail
                      7900236.
                                 417020
                                           8317256.
## 6 Lightning
                     1720146.
                                           1720146.
Plot damages for properties and crops by type of event:
damages<- gather(damages[1:6, 1:3], Type, Total, Property:Crop)</pre>
```

plot2 <- ggplot(damages, aes(x = reorder(Event, +Total), y = Total, fill = Type)) +</pre>





This analysis shows clearly that floods caused worst damages for properties, while hurricane caused big damages on crops. Overall floods represent the worst economic consequences damages in the United States.