Ananlysis of Storm dataset with respect to health and economic damage

Summary

The data comes from the U.S. National Oceanic and Atmospheric Administration's (NOAA) storm database and contains data from 1950 to November 2011. We are interested to study and compare the damage caused by storms and other weather events in the United States in the given period. The analysis involves reading the data in the software and applying necessary transformations (discussed in the data processing section) on the data and comparing the damage by means of graphical aid.

Data Processing

Steps used to format and group data.

- 1. The data is loaded into R using the read.csv function.
- 2. The columns named
 - "EVTYPE", "FATALITIES", "INJURIES", "PROPDMG", "CROPDMG", "PROPDMGEXP", "CROPDMGEXP" are extracted from the data as they are the data of interest.
- 3. The names of the columns are changed to lower case for ease of writing code.

```
# reading and formatting data

data=read.csv("C:\\Users\\HP\\R\\repdata_data_StormData.csv\\repdata_data_StormData.csv")
names(data)=tolower(names(data))
data=data[,c("evtype","fatalities","injuries","propdmg","cropdmg")]
head(data)
```

```
evtype fatalities injuries propdmg cropdmg
## 1 TORNADO 0 15 25.0 0
             ## 2 TORNADO
                         2.5
## 3 TORNADO
                       25.0
## 4 TORNADO
                    2
                               0
                         2.5
## 5 TORNADO
                    2
                         2.5
                                 0
## 6 TORNADO
                         2.5
```

- 4. We group the data by "EVTYPE" and calculate the following:
 - o sumh=stores the sum of total fatalities and injuries for each "EVTYPE".
 - · sump=stores the sum of total property damage for each "EVTYPE".
 - sumc=stores the sum of total crop damage for each "EVTYPE".

```
library(dplyr)
data_dmg=data %>%
  group_by(evtype) %>%
  summarise(sumh=sum(fatalities)+sum(injuries), sump=sum(propdmg), sumc=sum(cropdmg))
data_dmg
```

```
## # A tibble: 985 x 4
## 1 " HIGH SURF ADVISORY" 0 200 0
## 2 " COASTAL FLOOD"
                      0 0
                     0 50.0
## 3 " FLASH FLOOD"
                   ## 4 " LIGHTNING"
## 5 " TSTM WIND"
## 6 " TSTM WIND (G45)"
  7 " WATERSPOUT"
##
## 8 " WIND"
                     0 5.00 0
0 0 0
## 9 ?
## 10 ABNORMAL WARMTH
## # ... with 975 more rows
```

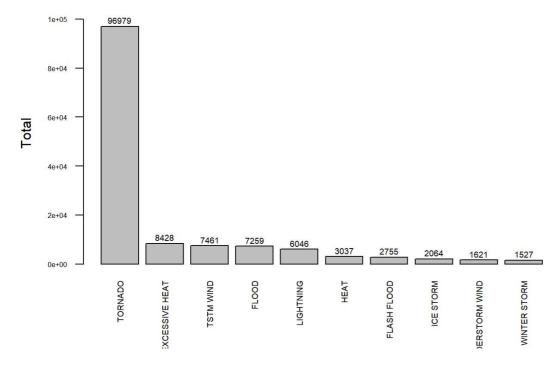
Results

Due to presence of large number of missing data and untidy data, we order the data in decreasing order of the number of fatalities+injuries and plot the first 10 values in a barplot.

```
data_health=data_dmg[order(data_dmg$sumh,decreasing = TRUE),]
data_health=data_health[1:10,]
data_health[,c(1,2)]
```

```
## # A tibble: 10 x 2
## evtype
                    sumh
   <fctr>
                   <dbl>
##
## 1 TORNADO 96979
## 1 TOKNADO
## 2 EXCESSIVE HEAT 8428
"" 2 TERM WIND 7461
## 3 TSTM WIND
                   7259
## 4 FLOOD
## 5 LIGHTNING
                   6046
                    3037
## 6 HEAT
                    2755
## 7 FLASH FLOOD
                2064
## 8 ICE STORM
## 9 THUNDERSTORM WIND 1621
## 10 WINTER STORM 1527
```

Comparing the total number of deaths and injuries for various events



Comments

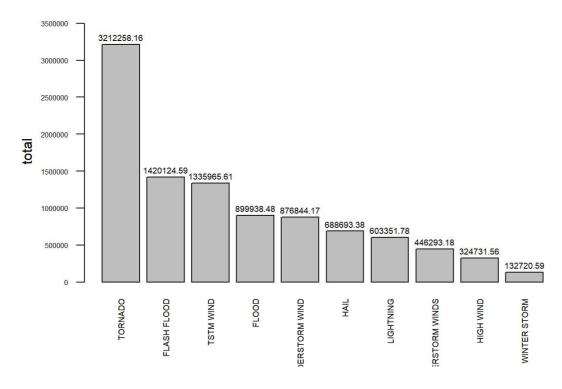
Based on the graph, we can say that Tornado is most hazardous with respect to population health with other events contributing very little to the population health hazard.

Due to presence of large number of missing data and untidy data, we order the data in decreasing order of the number of property damage and plot the first 10 values in a barplot.

```
data_prop=data_dmg[order(data_dmg$sump,decreasing=TRUE),]
data_prop=data_prop[1:10,]
data_prop[,c(1,3)]
```

```
## # A tibble: 10 x 2
##
   evtype
                       sump
    <fctr>
                       <dbl>
  1 TORNADO
                     3212258
## 2 FLASH FLOOD
                     1420125
## 3 TSTM WIND
                     1335966
##
  4 FLOOD
                      899938
## 5 THUNDERSTORM WIND 876844
##
   6 HAIL
                      688693
##
   7 LIGHTNING
                      603352
   8 THUNDERSTORM WINDS 446293
##
  9 HIGH WIND
##
                       324732
## 10 WINTER STORM
                      132721
```

Comparing the total property damage for various events



Comments

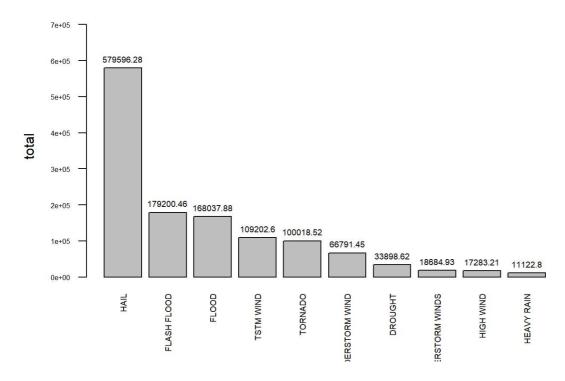
For the total property damage by an event, Tornadoes have the highest damage cost. Floods and TSTM winds having almost similar damage cost.

Due to presence of large number of missing data and untidy data, we order the data in decreasing order of the number of crop damage and plot the first 10 values in a barplot.

```
data_crop=data_dmg[order(data_dmg$sumc,decreasing=TRUE),]
data_crop=data_crop[1:10,]
data_crop[,c(1,4)]
```

```
## # A tibble: 10 x 2
##
   evtype
                      sumc
##
    <fctr>
                     <dbl>
                    579596
##
  1 HAIL
## 2 FLASH FLOOD
                   179200
  3 FLOOD
                    168038
##
  4 TSTM WIND
                    109203
##
  5 TORNADO
                    100019
  6 THUNDERSTORM WIND 66791
##
  7 DROUGHT
                     33899
##
   8 THUNDERSTORM WINDS 18685
##
  9 HIGH WIND
                      17283
## 10 HEAVY RAIN
                      11123
```

Comparing the total crop damage for various events



Comments

Hail caused the most crop damage over the years. Floods and Flash floods being the second and third most destructive respectively.

NOTE

- The data had many inconsistencies with respect to the names of the events and require further cleaning.
- The varaibles "propdmgexp" and "cropdmgexp" havent been taken into account while calculating the total damage due to lack of proper definitons.
- The actual interpretation might change when the above factors have been taken into account.