

Analysis of Severe Weather Events on USA based on NOAA Data

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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.

Source Data

- The storm data can be downloaded from this link
- Documentation on the data can be found from this link
- National Climatic Data Center Storm Events FAQ

Data Processing

To process the data we need to download the data from the above mentioned link. let's start by downloading the data

Download Data

```
if (!"StormData.csv.bz2" %in% dir("./")) {  
  download.file(  
    "http://d396qusza40orc.cloudfront.net/repdata%2Fdata%2FStormData.csv.bz2",  
    destfile = "StormData.csv.bz2"  
  )  
}
```

Now Let's read the downloaded data

```
if (!"stormdata" %in% ls()) {  
  stormdata <- read.csv(bzfile("StormData.csv.bz2"), sep = ",", header = TRUE, stringsAsFactors = FALSE)  
}  
dim(stormdata)  
  
## [1] 902297      37
```

Extact Storm Event Type

```

Stormevents <-
  c(
    "Astronomical Low Tide",
    "Avalanche",
    "Blizzard",
    "Coastal Flood",
    "Cold/Wind Chill",
    "Debris Flow",
    "Dense Fog",
    "Dense Smoke",
    "Drought",
    "Dust Devil",
    "Dust Storm",
    "Excessive Heat",
    "Extreme cold/Wind Chill",
    "Flash Flood",
    "Flood",
    "Freezing",
    "Frost/Freeze",
    "Funnel Cloud",
    "Hail",
    "Heat",
    "Heavy Rain",
    "Heavy Snow",
    "High Surf",
    "High Wind",
    "Hurricane/Typhoon",
    "Ice Storm",
    "Lakeshore Flood",
    "Lake-Effect Snow",
    "Lightning",
    "Marine Hail",
    "Marine High Wind",
    "Marine Strong Wind",
    "Marine Thunderstorm Wind",
    "Rip Current",
    "Seiche",
    "Sleet",
    "Storm Tide",
    "Strong Wind",
    "Thunderstorm Wind",
    "Tornado",
    "Tropical Depression",
    "Tropical Storm",
    "Tsunami",
    "Volcanic Ash",
    "Waterspout",
    "Wildfire",
    "Winter Storm",
    "Winter Weather"
  )

```

Some events are having characters ie. '\', so we will use regex to identify that

```
Storamevents_regex <-  
  c(  
    "Astronomical Low Tide|Low Tide",  
    "Avalanche",  
    "Blizzard",  
    "Coastal Flood",  
    "Cold/Wind Chill",  
    "Debris Flow",  
    "Dense Fog",  
    "Dense Smoke",  
    "Drought",  
    "Dust Devil",  
    "Dust Storm",  
    "Excessive Heat",  
    "Extreme cold/Wind Chill|Extreme Cold|Wind Chill",  
    "Flash Flood",  
    "Flood",  
    "Freezing",  
    "Frost/Freeze|Frost|Freeze",  
    "Funnel Cloud",  
    "Hail",  
    "Heat",  
    "Heavy Rain",  
    "Heavy Snow",  
    "High Surf",  
    "High Wind",  
    "Hurricane/Typhoon|Hurricane|Typhoon",  
    "Ice Storm",  
    "Lakeshore Flood",  
    "Lake-Effect Snow",  
    "Lightning",  
    "Marine Hail",  
    "Marine High Wind",  
    "Marine Strong Wind",  
    "Marine Thunderstorm Wind|Marine tstm Wind",  
    "Rip Current",  
    "Seiche",  
    "Sleet",  
    "Storm Tide",  
    "Strong Wind",  
    "Thunderstorm Wind|tstm wind",  
    "Tornado",  
    "Tropical Depression",  
    "Tropical Storm",  
    "Tsunami",  
    "Volcanic Ash",  
    "Waterspout",  
    "Wildfire",  
    "Winter Storm",  
    "Winter Weather"  
  )
```

We keep only the fields needed to our analysis:

- EVTYPE: the type of event
- FATALITIES: number of fatalities
- INJURIES: number of injuries
- PROPDMG: damage to properties in USD
- PROPDMGEXP: magnitude for properties damage (K for thousands, M for millions, B for billions)
- CROPDMG: damage to crops in USD
- CROPDMGEXP: magnitude for crops damage (K for thousands, M for millions, B for billions)

```
options(scipen = 999)
cleandata <- data.frame(EVTYPE = character(0), FATALITIES = numeric(0), INJURIES = numeric(0), PROPDMG = numeric(0), CROPDMG = numeric(0))
for (i in 1:length(Stormevents)) {
  rows <-
    stormdata[grepl(Stormevents_regex[i], ignore.case = TRUE, stormdata$EVTYPE),]
  rows <-
    rows[, c(
      "EVTYPE",
      "FATALITIES",
      "INJURIES",
      "PROPDMG",
      "PROPDMGEXP",
      "CROPDMG",
      "CROPDMGEXP"
    )]
  CLEANNAME <- c(rep(Stormevents[i], nrow(rows)))
  rows <- cbind(rows, CLEANNAME)
  cleandata <- rbind(cleandata, rows)
}
```

Adjusting the valid values are K for thousands(3 exp), M for millions(6 exp) and B for billions(9 exp)

```
cleandata[(cleandata$PROPDMGEXP == "K" | cleandata$PROPDMGEXP == "k"), ]$PROPDMGEXP <- 3
cleandata[(cleandata$PROPDMGEXP == "M" | cleandata$PROPDMGEXP == "m"), ]$PROPDMGEXP <- 6
cleandata[(cleandata$PROPDMGEXP == "B" | cleandata$PROPDMGEXP == "b"), ]$PROPDMGEXP <- 9
cleandata[(cleandata$CROPDMGEXP == "K" | cleandata$CROPDMGEXP == "k"), ]$CROPDMGEXP <- 3
cleandata[(cleandata$CROPDMGEXP == "M" | cleandata$CROPDMGEXP == "m"), ]$CROPDMGEXP <- 6
cleandata[(cleandata$CROPDMGEXP == "B" | cleandata$CROPDMGEXP == "b"), ]$CROPDMGEXP <- 9
```

Calculate the properties and crop damages

```
cleandata$PROPDMG <- cleandata$PROPDMG * 10as.numeric(cleandata$PROPDMGEXP)
```

```
## Warning: NAs introduced by coercion
```

```
cleandata$CROPDMG <- cleandata$CROPDMG * 10as.numeric(cleandata$CROPDMGEXP)
```

```
## Warning: NAs introduced by coercion
```

Calculate the total damages

```
TOTECODMG <- cleandata$PROPDMG + cleandata$CROPDMG
cleandata <- cbind(cleandata, TOTECODMG)
```

Thus concludes the initial preparation of data.

Q.1: Across the United States, which types of events (as indicated in the EV-TYPE variable) are most harmful with respect to population health?

Calculating the number of FATALITIES

```
fatalities <- aggregate(FATALITIES ~ CLEANNAME, data = cleandata, FUN = sum)
fatalities <- fatalities[order(fatalities$FATALITIES, decreasing = TRUE), ]
```

Top 10 Events causing the most FATALITIES

```
MaxFatalities <- fatalities[1:10, ]
print(MaxFatalities)
```

##	CLEANNAME	FATALITIES
## 38	Tornado	5661
## 19	Heat	3138
## 11	Excessive Heat	1922
## 14	Flood	1525
## 13	Flash Flood	1035
## 28	Lightning	817
## 37	Thunderstorm Wind	753
## 33	Rip Current	577
## 12	Extreme cold/Wind Chill	382
## 23	High Wind	299

Calculating the number of INJURIES

```
injuries <- aggregate(INJURIES ~ CLEANNAME, data = cleandata, FUN = sum)
injuries <- injuries[order(injuries$INJURIES, decreasing = TRUE), ]
```

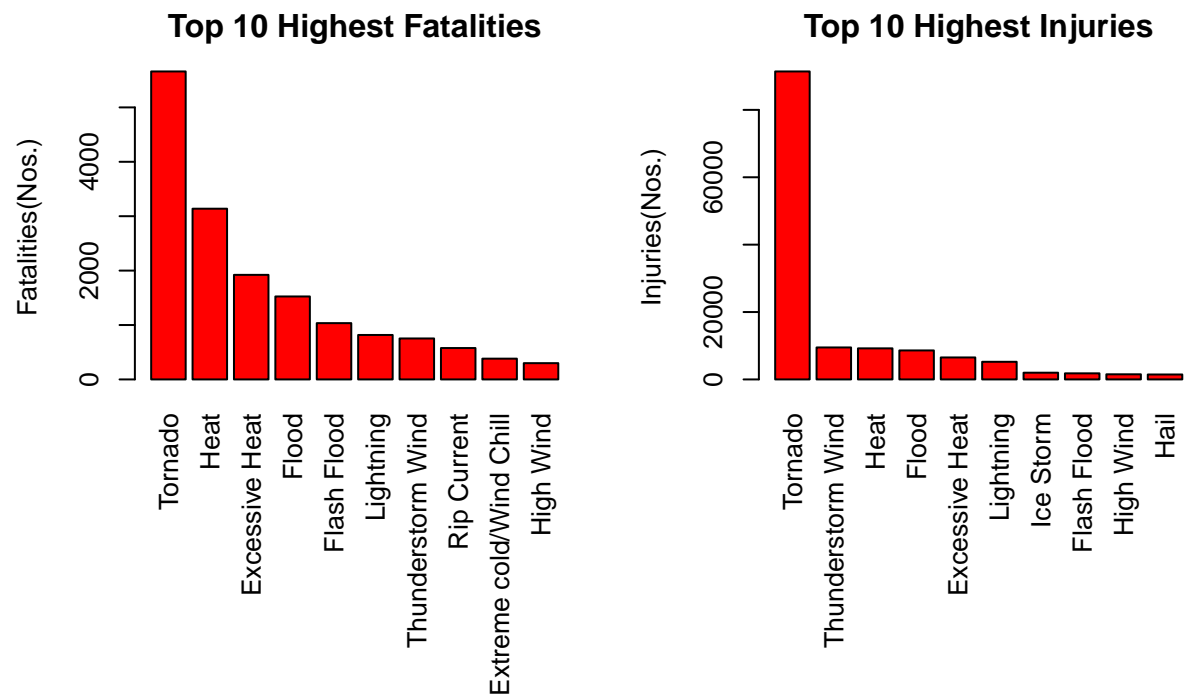
Top 10 Events causing the most INJURIES

```
MaxInjuries <- injuries[1:10, ]
print(MaxInjuries)
```

##	CLEANNAME	INJURIES
## 38	Tornado	91407
## 37	Thunderstorm Wind	9493
## 19	Heat	9224
## 14	Flood	8604
## 11	Excessive Heat	6525
## 28	Lightning	5232
## 25	Ice Storm	1992
## 13	Flash Flood	1802
## 23	High Wind	1523
## 18	Hail	1467

Plotting the above information as a pair of graphs of total FATALITIES & INJURIES

```
par(
  mfrow = c(1, 2),
  mar = c(15, 4, 3, 2),
  mgp = c(3, 1, 0),
  cex = 0.8
)
barplot(
  MaxFatalities$FATALITIES,
  las = 3,
  names.arg = MaxFatalities$CLEANNAME,
  main = "Top 10 Highest Fatalities",
  ylab = "Fatalities(Nos.)",
  col = "RED"
)
barplot(
  MaxInjuries$INJURIES,
  las = 3,
  names.arg = MaxInjuries$CLEANNAME,
  main = "Top 10 Highest Injuries",
  ylab = "Injuries(Nos.)",
  col = "RED"
)
```



ANS for Qus 1

As you can see with respect to population health

* Fatalities : 1. Tornado & 2. Heat

* Injuries : Tornado

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

Calculating the total property damage

```
propdmg <- aggregate(PROPDMG ~ CLEANNAME, data = cleandata, FUN = sum)
propdmg <- propdmg[order(propdmg$PROPDMG, decreasing = TRUE), ]
```

Top 10 Events causing the most property damage

```
propdmgMax <- propdmg[1:10, ]
print(propdmgMax)
```

##	CLEANNAME	PROPDMG
## 14	Flood	168212215589
## 24	Hurricane/Typhoon	85356410010
## 38	Tornado	58603317864
## 18	Hail	17622990956
## 13	Flash Flood	17588791879
## 37	Thunderstorm Wind	11575228673
## 40	Tropical Storm	7714390550
## 45	Winter Storm	6749997251
## 23	High Wind	6166300000
## 44	Wildfire	4865614000

Calculating the total crop damage

```
croprdmg <- aggregate(CROPDGM ~ CLEANNAME, data = cleandata, FUN = sum)
croprdmg <- croprdmg[order(croprdmg$CROPDGM, decreasing = TRUE), ]
```

Top 10 Events causing the most crop damage

```
croprdmgMax <- croprdmg[1:10, ]
print(croprdmgMax)
```

##	CLEANNAME	CROPDGM
## 8	Drought	13972621780
## 14	Flood	12380109100
## 24	Hurricane/Typhoon	5516117800
## 25	Ice Storm	5022113500
## 18	Hail	3114212870
## 16	Frost/Freeze	1997061000
## 13	Flash Flood	1532197150

```
## 12 Extreme cold/Wind Chill 1313623000
## 37 Thunderstorm Wind 1255947980
## 19 Heat 904469280
```

Calculating the total economic damage

```
ecodmg <- aggregate(TOTECODMG ~ CLEANNAME, data = cleandata, FUN = sum)
ecodmg <- ecodmg[order(ecodmg$TOTECODMG, decreasing = TRUE), ]
```

Top 10 Events causing the most economic damage

```
ecodmgMax <- ecodmg[1:10, ]
print(ecodmgMax)
```

```
##          CLEANNAME    TOTECODMG
## 14          Flood 157764680787
## 24 Hurricane/Typhoon 44330000800
## 38          Tornado 18172843863
## 18          Hail 11681050140
## 13      Flash Flood  9224527227
## 37 Thunderstorm Wind  7098296330
## 25          Ice Storm  5925150850
## 44          Wildfire  3685468370
## 23          High Wind  3472442200
## 8           Drought  1886667000
```

Plotting the above information as a pair of graphs of total property, crop & economic damage

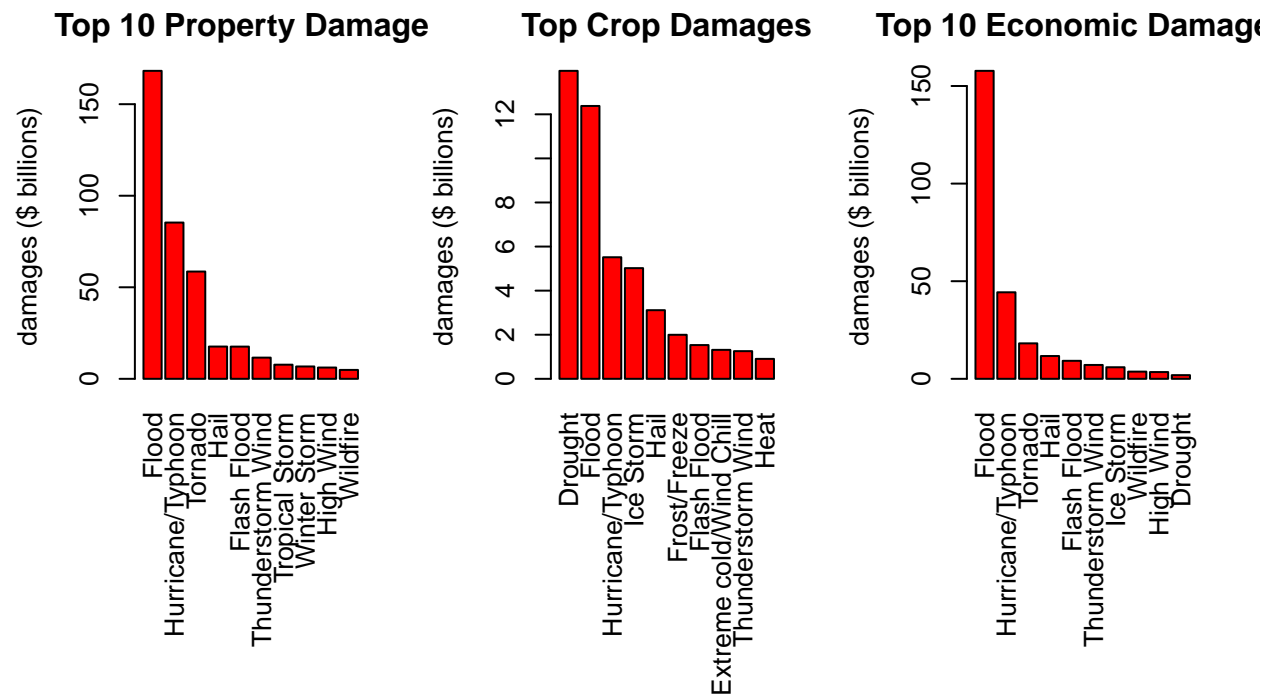
```
par(
  mfrow = c(1, 3),
  mar = c(15, 4, 3, 2),
  mgp = c(3, 1, 0),
  cex = 0.8
)
barplot(
  propdmgMax$PROPDMG / (10 ^ 9),
  las = 3,
  names.arg = propdmgMax$CLEANNAME,
  main = "Top 10 Property Damages",
  ylab = "damages ($ billions)",
  col = "RED"
)
barplot(
  cropdmgMax$CROPDGM / (10 ^ 9),
  las = 3,
  names.arg = cropdmgMax$CLEANNAME,
  main = "Top Crop Damages",
  ylab = "damages ($ billions)",
  col = "RED"
)
barplot(
  ecodmgMax$TOTECODMG / (10 ^ 9),
```



```

las = 3,
names.arg = ecodmgMax$CLEANNAME,
main = "Top 10 Economic Damages",
ylab = "damages ($ billions)",
col = "RED"
)

```



ANS for Qus 2

As you can see with respect to economic consequences causing damage

* Property : Flood

* Crop : Drought and Flood

* Economic : Flood followed by Hurricane/Typhoon

Conclusion

From the above analysis we can see that Tornado and Heat are most harmful with respect to Population Health, while Flood, Drought and Hurricane/Typhoon have the greatest Economic Consequences.

Project Codes have been uploaded to Github