

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

### 2.2: Examining Column Names

```
colnames(data)
```

##	[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"
##	[21]	"F"	"MAG"	"FATALITIES"	"INJURIES"	"PROPDMG"
##	[26]	"PROPDMGEXP"	"CROPDGMG"	"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", "FATALITIES", "INJURIES", "PROPDMG", "PROPDMGEXP", "CROPDMGEXP", "CROPDMG")])
# Removing columns
data[, c(cols2Remove) := NULL]
# Only use data where fatalities or injuries occurred.
data <- data[(EVTYPE != "?" &
              (INJURIES > 0 | FATALITIES > 0 | PROPDMG > 0 | CROPDMG > 0)), c("EVTYPE", "FATALITIES", "INJURIES", "PROPDMG", "PROPDMGEXP", "CROPDMGEXP", "CROPDMG")]

```

## 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.

## 2.5: Making Economic Cost Columns

```

data <- data[, .(EVTYPE, FATALITIES, INJURIES, PROPDMG, PROPDMGEXP, propCost = PROPDMG * PROPDMGEXP, CROPDMGEXP, CROPDMG)]

```

## 2.6: Calculating Total Property and Crop Cost

```

totalCostDT <- data[, .(propCost = sum(propCost), cropCost = sum(cropCost), Total_Cost = sum(propCost) + sum(cropCost))]
totalCostDT <- totalCostDT[order(-Total_Cost), ]
totalCostDT <- totalCostDT[1:10, ]
head(totalCostDT, 5)

```

##	EVTYPE	propCost	cropCost	Total_Cost
## 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: Calculating Total Fatalities and Injuries

```

totalInjuriesDT <- data[, .(FATALITIES = sum(FATALITIES), INJURIES = sum(INJURIES), totals = sum(FATALITIES) + sum(INJURIES))]
totalInjuriesDT <- totalInjuriesDT[order(-FATALITIES), ]
totalInjuriesDT <- totalInjuriesDT[1:10, ]
head(totalInjuriesDT, 5)

```

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

### 3: Results

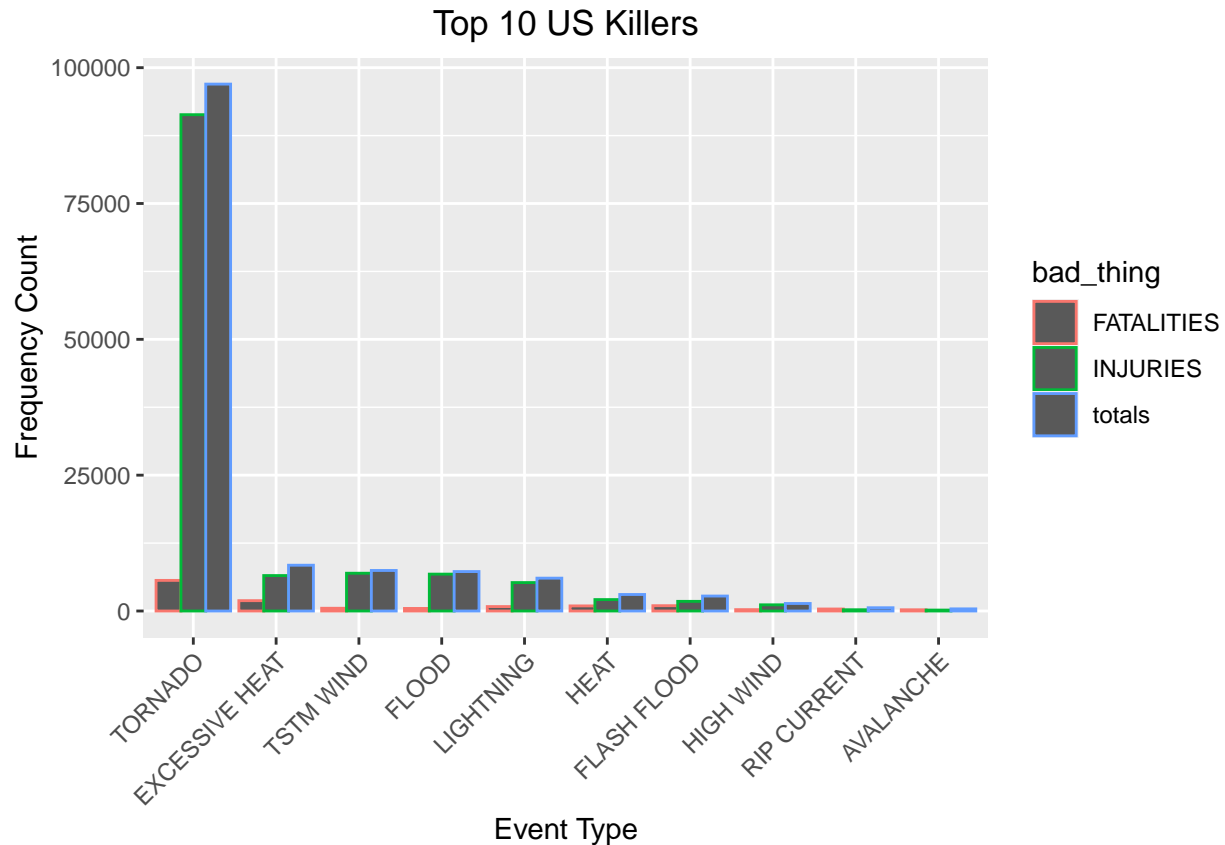
#### 3.1: Events that are Most Harmful to Population Health

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)
```

```
##           EVTYPE bad_thing value
## 1:    TORNADO FATALITIES  5633
## 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(col=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)
```

```
##           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(col=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
```

