

Impacts on the economy and public health by severe weather events in the US

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Set up

```
options(scipen = 1) # remove scientific notations
```

Synopsis

This work aims to carry out an analysis of the impacts of extreme climatic events on the economy and public health in the United States. The data used in this analysis are data collected by the US National Oceanic and Atmospheric Administration's (NOAA) and cover the 1950 period - 2011. In this the variables data set taken into consideration in this analysis can be cited: fatalities, injuries and property damage (in US dollars). The data used is available here (<https://d396qusza40orc.cloudfront.net/repdata%2Fdata%2FStormData.csv.bz2>) and the documentation is available here (https://d396qusza40orc.cloudfront.net/repdata%2Fpeer2_doc%2Fpd01016005curr.pdf)

Data Processing

This section consists of the loading of data and its preprocessing, going through a quick exploratory analysis of data in order to conceptualize the dataset, thus providing the presentation of the variables utilized in the study.

Data Load

After downloading the data by the link provided above, the data is loaded into the variable `DataStorm`.

```
dataStorm <- read.csv(bzfile("repdata_data_StormData.csv.bz2"))
```

Exploratory Data Analysis

As can be seen below the dataset has 902,297 rows and 37 columns, and is also available a sample of the dataset.

```
dim(dataStorm)
```

```
## [1] 902297    37
```

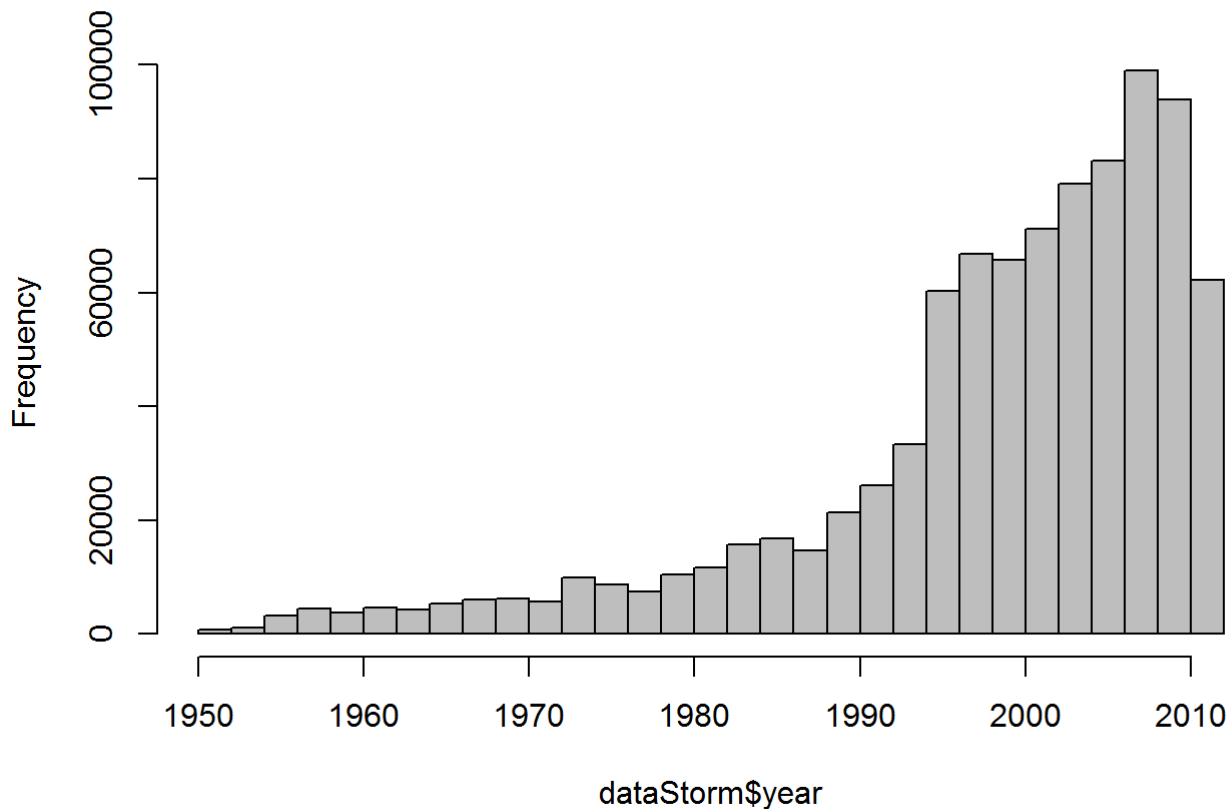
```
head(dataStorm, n = 3)
```

```
## STATE__ BGN_DATE BGN_TIME TIME_ZONE COUNTY COUNTYNAME STATE
## 1 1 4/18/1950 0:00:00 0130 CST 97 MOBILE AL
## 2 1 4/18/1950 0:00:00 0145 CST 3 BALDWIN AL
## 3 1 2/20/1951 0:00:00 1600 CST 57 FAYETTE AL
## EVTYPE BGN_RANGE BGN_AZI BGN_LOCATI END_DATE END_TIME COUNTY_END
## 1 TORNADO 0 0 0
## 2 TORNADO 0 0 0
## 3 TORNADO 0 0 0
## COUNTYENDN END_RANGE END_AZI END_LOCATI LENGTH WIDTH F MAG FATALITIES
## 1 NA 0 14.0 100 3 0 0
## 2 NA 0 2.0 150 2 0 0
## 3 NA 0 0.1 123 2 0 0
## INJURIES PROPDMG PROPDMGEXP CROPDMG CROPDMGEXP WFO STATEOFFIC ZONENAMES
## 1 15 25.0 K 0
## 2 0 2.5 K 0
## 3 2 25.0 K 0
## LATITUDE LONGITUDE LATITUDE_E LONGITUDE_ REMARKS REFNUM
## 1 3040 8812 3051 8806 1
## 2 3042 8755 0 0 2
## 3 3340 8742 0 0 3
```

As can be noted in the histogram, the frequency distribution of the data inputs have a higher intensity of the late '90s to now.

```
if (dim(dataStorm)[2] == 37) {dataStorm$year <- as.numeric(format(as.Date(dataStorm$B
GN_DATE, format = "%m/%d/%Y %H:%M:%S"), "%Y")) }
hist(dataStorm$year, breaks = 30, col = "gray")
```

Histogram of dataStorm\$year



Normalizing names of events.

```
x <- dataStorm[,c("EVTYPE", "FATALITIES", "INJURIES", "PROPDMG")]

x$EVTYPE <- gsub("^HEAT$", "EXCESSIVE HEAT", x$EVTYPE)
x$EVTYPE <- gsub("^TSTM WIND$", "THUNDERSTORM WIND", x$EVTYPE)
x$EVTYPE <- gsub("^THUNDERSTORM WIND$", "THUNDERSTORM WINDS", x$EVTYPE)
```

Aggregating the top 10 fatalities causes.

```
fatal<- aggregate(x$FATALITIES, by=list(x$EVTYPE), FUN=sum, na.rm=TRUE)
colnames(fatal) = c("event.type", "fatality.total")
f <- fatal[order(-fatal$fatality.total),]
t10fatal <- f[1:10,]
t10fatal$event.type <- factor(t10fatal$event.type, levels = t10fatal$event.type, orde
red=TRUE)
t10fatal
```

##	event.type	fatality.total
## 832	TORNADO	5633
## 130	EXCESSIVE HEAT	2840
## 153	FLASH FLOOD	978
## 463	LIGHTNING	816
## 784	THUNDERSTORM WINDS	701
## 170	FLOOD	470
## 584	RIP CURRENT	368
## 358	HIGH WIND	248
## 19	AVALANCHE	224
## 969	WINTER STORM	206

Aggregating the top 10 injuries causes.

```
injur<- aggregate(x$INJURIES, by=list(x$EVTYPE), FUN=sum, na.rm=TRUE)
colnames(injur) = c("event.type", "injury.total")
injuries.sorted <- injur[order(-injur$injury.total),]
t10inj <- injuries.sorted[1:10,]
t10inj$event.type <- factor(t10inj$event.type, levels=t10inj$event.type, ordered=TRUE
)
t10inj
```

##	event.type	injury.total
## 832	TORNADO	91346
## 784	THUNDERSTORM WINDS	9353
## 130	EXCESSIVE HEAT	8625
## 170	FLOOD	6789
## 463	LIGHTNING	5230
## 426	ICE STORM	1975
## 153	FLASH FLOOD	1777
## 244	HAIL	1361
## 969	WINTER STORM	1321
## 410	HURRICANE/TYPHOON	1275

Aggregating the top 10 property damage causes.

```
propdmg <- aggregate( x$PROPDMG, by=list(x$EVTYPE), FUN=sum, na.rm=TRUE)
colnames(propdmg) = c("event.type", "prop.dmg.total")
prop.dmg.sorted <- propdmg[order(-propdmg$prop.dmg.total),]
t10pd <- prop.dmg.sorted[1:10,]
t10pd$event.type <- factor(t10pd$event.type, levels=t10pd$event.type, ordered=TRUE)
t10pd
```

```
##           event.type prop.dmg.total
## 832          TORNADO    3212258.16
## 784 THUNDERSTORM WINDS    2659102.96
## 153        FLASH FLOOD    1420124.59
## 170           FLOOD     899938.48
## 244           HAIL     688693.38
## 463        LIGHTNING    603351.78
## 358        HIGH WIND    324731.56
## 969    WINTER STORM    132720.59
## 309        HEAVY SNOW    122251.99
## 954        WILDFIRE     84459.34
```

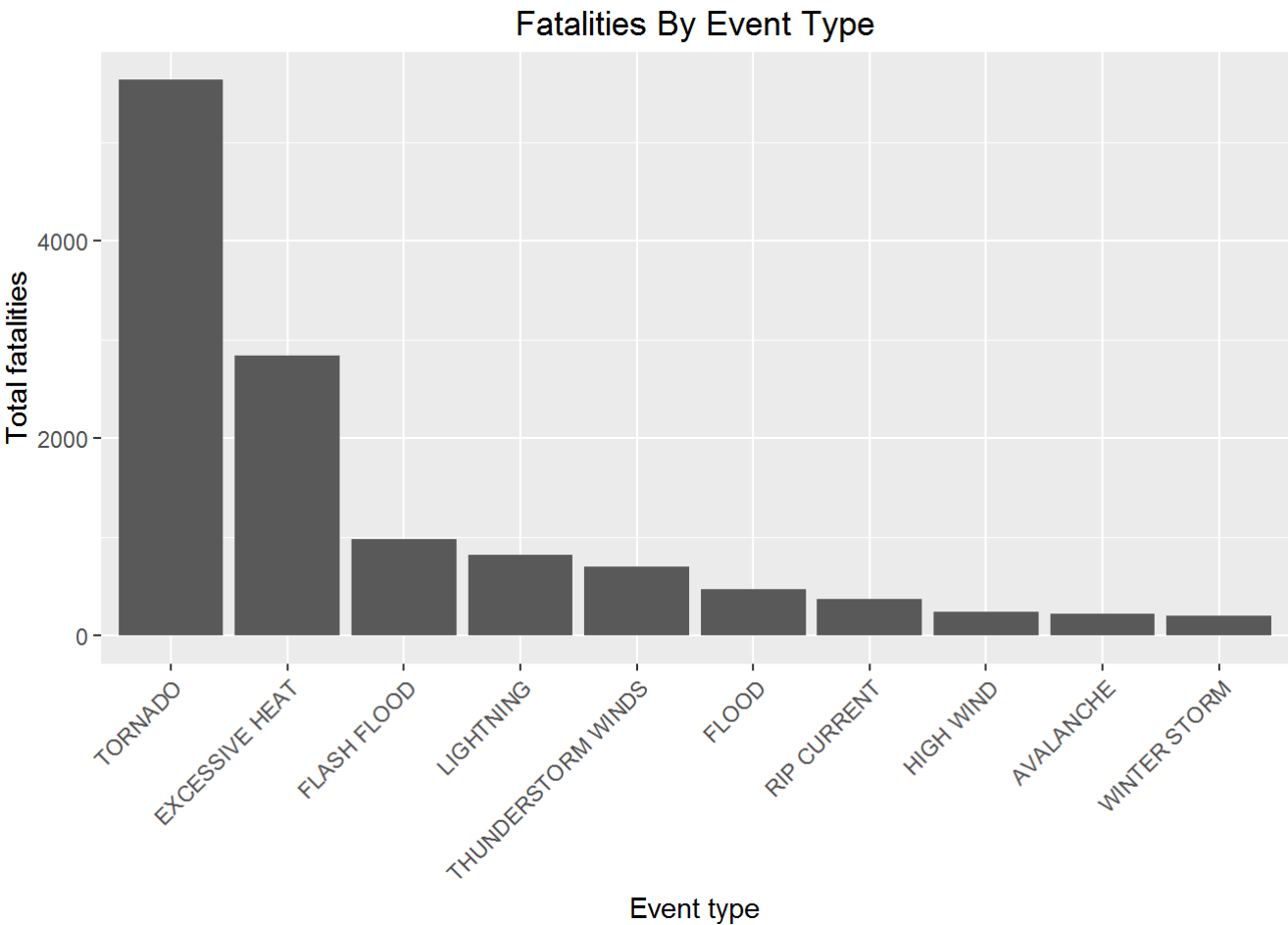
Results

The damages in the area of public health are the following:

```
library(ggplot2)
```

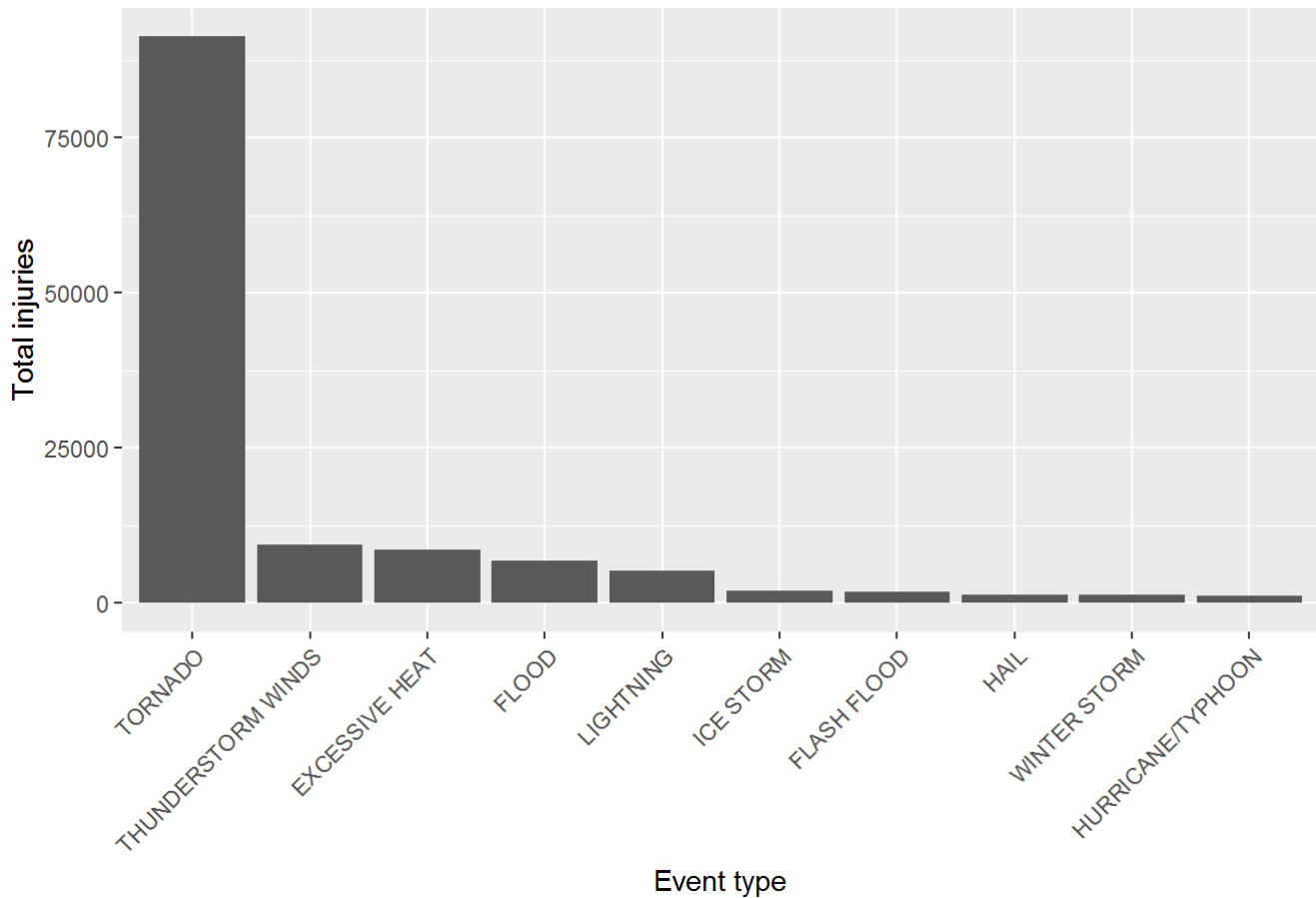
```
## Warning: package 'ggplot2' was built under R version 3.2.3
```

```
ggplot(data=t10fatal, aes(x=event.type, y=fatality.total)) +
  geom_bar(stat="identity") + xlab("Event type") + ylab("Total fatalities") +
  ggtitle("Fatalities By Event Type") +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```



```
ggplot(data=t10inj, aes(x=event.type, y=injury.total)) +  
  geom_bar(stat="identity") + xlab("Event type") + ylab("Total injuries") +  
  ggtitle("Injuries By Event Type") +  
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```

Injuries By Event Type



Finally the economic damage are:

```
ggplot(data=t10pd, aes(x=event.type, y=prop.dmg.total)) +
  geom_bar(stat="identity") + xlab("Event type") + ylab("Total property damage") +
  ggtitle("Property Damage By Event Type") +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```

