

Reproducible Research

Assessment 2: Impact of severe weather events on US population health and economy

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Synopsis

Aim: The overarching aim of this report is to understand the impact of severe weather events on the US population health and economy.

Data source: Findings from this report is based on the NOAA Storm Database (period: 1950 to 2011)

Settings

```
#Make code visible, set working directory and read library
echo = TRUE
setwd("~/RepData_PeerAssessment2")
library(ggplot2)
library(plyr)
library(utils)
require(gridExtra)
```

```
## Loading required package: gridExtra
## Loading required package: grid
```

Data exploration

```
data <- read.csv(file="data/StormData.csv", header=T, sep="," , quote="", as.is=TRUE)
summary(data)
```

```
##   X.STATE__ .      X.BGN_DATE .      X.BGN_TIME .
## Length:1773320    Length:1773320    Length:1773320
## Class :character  Class :character  Class :character
## Mode  :character  Mode  :character  Mode  :character
##   X.TIME_ZONE .      X.COUNTY .      X.COUNTYNAME .
## Length:1773320      Length:1773320    Length:1773320
## Class :character    Class :character  Class :character
## Mode  :character    Mode  :character  Mode  :character
##   X.STATE .      X.EVTYPE .      X.BGN_RANGE .
```

##	Length:1773320	Length:1773320	Length:1773320
##	Class :character	Class :character	Class :character
##	Mode :character	Mode :character	Mode :character
##	X.BGN_AZI.	X.BGN_LOCATI.	X.END_DATE.
##	Length:1773320	Length:1773320	Length:1773320
##	Class :character	Class :character	Class :character
##	Mode :character	Mode :character	Mode :character
##	X.END_TIME.	X.COUNTY_END.	X.COUNTYENDN.
##	Length:1773320	Length:1773320	Length:1773320
##	Class :character	Class :character	Class :character
##	Mode :character	Mode :character	Mode :character
##	X.END_RANGE.	X.END_AZI.	X.END_LOCATI.
##	Length:1773320	Length:1773320	Length:1773320
##	Class :character	Class :character	Class :character
##	Mode :character	Mode :character	Mode :character
##	X.LENGTH.	X.WIDTH.	X.F.
##	Length:1773320	Length:1773320	Length:1773320
##	Class :character	Class :character	Class :character
##	Mode :character	Mode :character	Mode :character
##	X.MAG.	X.FATALITIES.	X.INJURIES.
##	Length:1773320	Length:1773320	Length:1773320
##	Class :character	Class :character	Class :character
##	Mode :character	Mode :character	Mode :character
##	X.PROPDMG.	X.PROPDMGEXP.	X.CROPDMG.
##	Length:1773320	Length:1773320	Length:1773320
##	Class :character	Class :character	Class :character
##	Mode :character	Mode :character	Mode :character
##	X.CROPDMGEXP.	X.WFO.	X.STATEOFFIC.
##	Length:1773320	Length:1773320	Length:1773320
##	Class :character	Class :character	Class :character
##	Mode :character	Mode :character	Mode :character
##	X.ZONENAMES.	X.LATITUDE.	X.LONGITUDE.
##	Length:1773320	Length:1773320	Length:1773320
##	Class :character	Class :character	Class :character
##	Mode :character	Mode :character	Mode :character
##	X.LATITUDE_E.	X.LONGITUDE_.	X.REMARKS.
##	Length:1773320	Length:1773320	Length:1773320
##	Class :character	Class :character	Class :character
##	Mode :character	Mode :character	Mode :character
##	X.REFNUM.		
##	Length:1773320		
##	Class :character		
##	Mode :character		

```

if (!"StormData" %in% ls()) {
  Data <- read.csv("data/StormData.csv", sep = ",")
}
dim(Data)

```

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

```
head(Data, n = 2)
```

```

##      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
##      EVTYPE BGN_RANGE BGN_AZI BGN_LOCATI END_DATE END_TIME COUNTY_END
## 1 TORNADO          0              0              0
## 2 TORNADO          0              0              0
##      COUNTYENDN END_RANGE END_AZI END_LOCATI LENGTH WIDTH F MAG FATALITIES
## 1          NA          0              14    100 3   0          0
## 2          NA          0              2    150 2   0          0
##      INJURIES PROPDMG PROPDMGEXP CROPDMG CROPDMGEXP WFO STATEOFFIC ZONENAMES
## 1         15     25.0           K         0
## 2          0      2.5           K         0
##      LATITUDE LONGITUDE LATITUDE_E LONGITUDE_ REMARKS REFNUM
## 1        3040        8812        3051        8806          1
## 2        3042        8755          0          0          2

```

The total number of rows and columns are 902297 and 37 respectively.

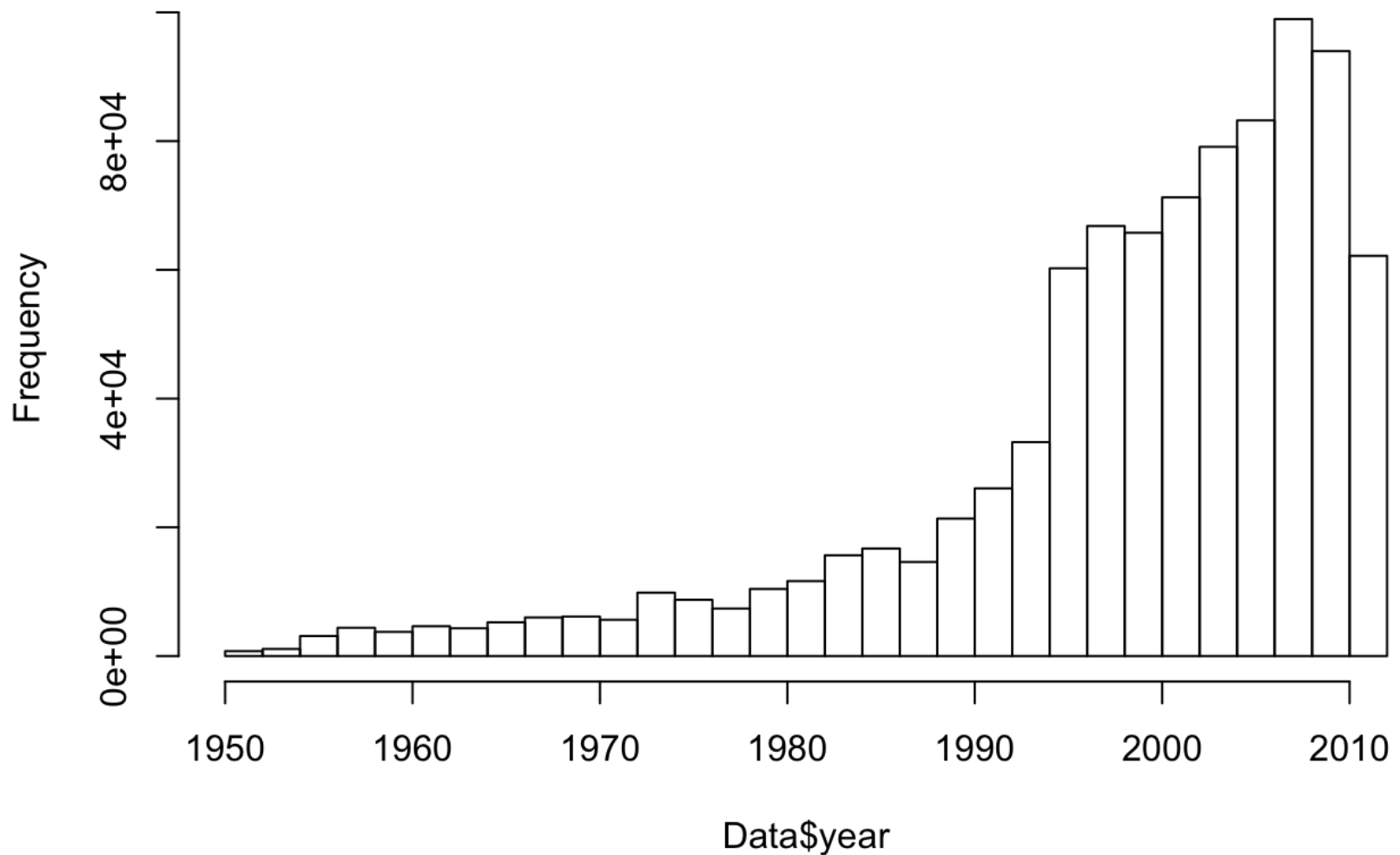
The events started from 1950 and end in 2011. There were few recorded events in the earlier years due to the lack of good records. While the records were more complete in the recent years.

```

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

```

Histogram of Data\$year



Based on the graph, the number of events tracked start to increase significantly from 1995. Hence, 1990 to 2011 will be use to make most of the data.

```
storm <- Data[Data$year >= 1995, ]  
dim(storm)
```

```
## [1] 681500    38
```

With this, the total number of rows and column 681500 and 38 respectively.

Impact on Population Health

To identify the no. of **fatalities** and **injuries** that are caused by the severe weather events, we list the top 10 most severe types of weather events.

```

sortHelper <- function(fieldName, top = 10, dataset = Data) {
  index <- which(colnames(dataset) == fieldName)
  field <- aggregate(dataset[, index], by = list(dataset$EVTYPE), FUN = "sum")
  names(field) <- c("EVTYPE", fieldName)
  field <- arrange(field, field[, 2], decreasing = T)
  field <- head(field, n = top)
  field <- within(field, EVTYPE <- factor(x = EVTYPE, levels = field$EVTYPE))
  return(field)
}

fatalities <- sortHelper("FATALITIES", dataset = storm)
injuries <- sortHelper("INJURIES", dataset = storm)

```

Impact on Economy

We will convert the **property damage** and **crop damage** data into comparable numerical forms. Both `PROPDMGEXP` and `CROPDMGEXP` columns record a multiplier for each observation where we have Hundred (H), Thousand (K), Million (M) and Billion (B).

```

convertHelper <- function(dataset = storm, fieldName, newFieldName) {
  totalLen <- dim(dataset)[2]
  index <- which(colnames(dataset) == fieldName)
  dataset[, index] <- as.character(dataset[, index])
  logic <- !is.na(toupper(dataset[, index]))
  dataset[logic & toupper(dataset[, index]) == "B", index] <- "9"
  dataset[logic & toupper(dataset[, index]) == "M", index] <- "6"
  dataset[logic & toupper(dataset[, index]) == "K", index] <- "3"
  dataset[logic & toupper(dataset[, index]) == "H", index] <- "2"
  dataset[logic & toupper(dataset[, index]) == "", index] <- "0"
  dataset[, index] <- as.numeric(dataset[, index])
  dataset[is.na(dataset[, index]), index] <- 0
  dataset <- cbind(dataset, dataset[, index - 1] * 10^dataset[, index])
  names(dataset)[totalLen + 1] <- newFieldName
  return(dataset)
}

storm <- convertHelper(storm, "PROPDMGEXP", "propertyDamage")

```

```

## Warning in convertHelper(storm, "PROPDMGEXP", "propertyDamage"): NAs
## introduced by coercion

```

```

storm <- convertHelper(storm, "CROPDMGEXP", "cropDamage")

```

```

## Warning in convertHelper(storm, "CROPDMGEXP", "cropDamage"): NAs
## introduced by coercion

```

```
names(storm)
```

```
## [1] "STATE__" "BGN_DATE" "BGN_TIME" "TIME_ZONE"
## [5] "COUNTY" "COUNTYNAME" "STATE" "EVTYPE"
## [9] "BGN_RANGE" "BGN_AZI" "BGN_LOCATI" "END_DATE"
## [13] "END_TIME" "COUNTY_END" "COUNTYENDN" "END_RANGE"
## [17] "END_AZI" "END_LOCATI" "LENGTH" "WIDTH"
## [21] "F" "MAG" "FATALITIES" "INJURIES"
## [25] "PROPDMG" "PROPDMGEXP" "CROPDMG" "CROPDMGEXP"
## [29] "WFO" "STATEOFFIC" "ZONENAMES" "LATITUDE"
## [33] "LONGITUDE" "LATITUDE_E" "LONGITUDE_" "REMARKS"
## [37] "REFNUM" "year" "propertyDamage" "cropDamage"
```

```
options(scipen=999)
property <- sortHelper("propertyDamage", dataset = storm)
crop <- sortHelper("cropDamage", dataset = storm)
```

Findings

As for the impact on public health, we have the lists of severe weather events by the no. of people badly injured.

```
fatalities
```

```
##           EVTYPE FATALITIES
## 1  EXCESSIVE HEAT         1903
## 2      TORNADO         1545
## 3  FLASH FLOOD          934
## 4          HEAT          924
## 5  LIGHTNING           729
## 6      FLOOD           423
## 7  RIP CURRENT          360
## 8    HIGH WIND          241
## 9    TSTM WIND          241
## 10  AVALANCHE          223
```

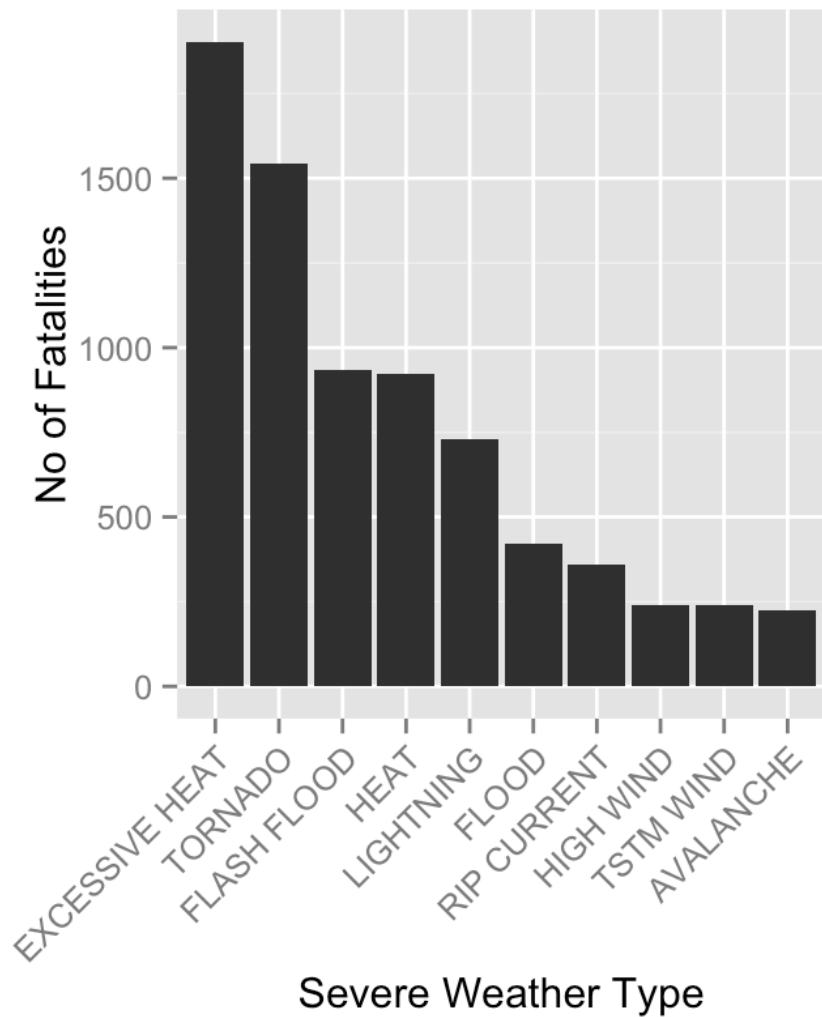
```
injuries
```

##	EVTTYPE	INJURIES
## 1	TORNADO	21765
## 2	FLOOD	6769
## 3	EXCESSIVE HEAT	6525
## 4	LIGHTNING	4631
## 5	TSTM WIND	3630
## 6	HEAT	2030
## 7	FLASH FLOOD	1734
## 8	THUNDERSTORM WIND	1426
## 9	WINTER STORM	1298
## 10	HURRICANE/TYPHOON	1275

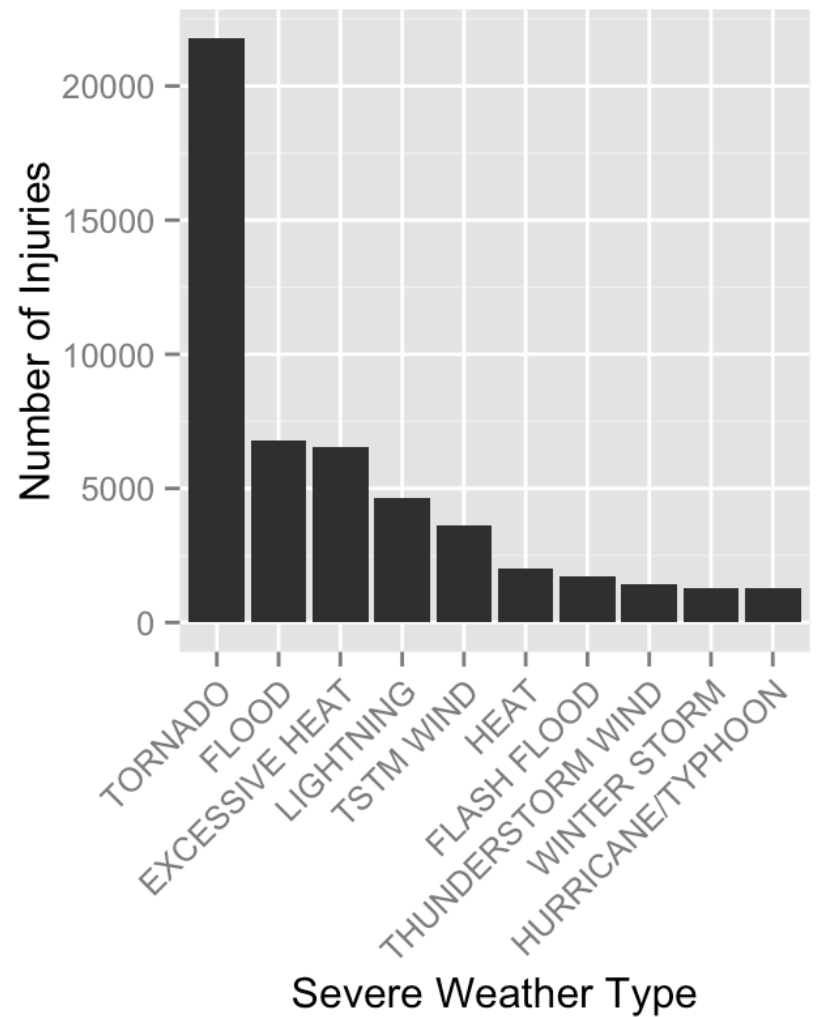
Below illustrates the total fatalities and injuries caused by the severe weather events.

```
fatalitiesPlot <- qplot(EVTTYPE, data = fatalities, weight = FATALITIES, geom = "bar", binwidth = 1) +
  scale_y_continuous("No of Fatalities") +
  theme(axis.text.x = element_text(angle = 45,
    hjust = 1)) + xlab("Severe Weather Type") +
  ggtitle("Total Fatalities by Severe Weather\n Events in the U.S.\n from 1995 - 2011"
)
injuriesPlot <- qplot(EVTTYPE, data = injuries, weight = INJURIES, geom = "bar", binwidth = 1) +
  scale_y_continuous("Number of Injuries") +
  theme(axis.text.x = element_text(angle = 45,
    hjust = 1)) + xlab("Severe Weather Type") +
  ggtitle("Total Injuries by Severe Weather\n Events in the U.S.\n from 1995 - 2011")
grid.arrange(fatalitiesPlot, injuriesPlot, ncol = 2)
```

Total Fatalities by Severe Weather Events in the U.S. from 1995 - 2011



Total Injuries by Severe Weather Events in the U.S. from 1995 - 2011



Excessive heat and **tornado** caused the highest fatalities, whereas **tornado** caused the most injuries in the US from 1995 to 2011.

As for the impact on economy, we have got two sorted lists below by the amount of money cost by damages.

property

##	EVTTYPE	propertyDamage
## 1	FLOOD	144022037057
## 2	HURRICANE/TYPHOON	69305840000
## 3	STORM SURGE	43193536000
## 4	TORNADO	24935939545
## 5	FLASH FLOOD	16047794571
## 6	HAIL	15048722103
## 7	HURRICANE	11812819010
## 8	TROPICAL STORM	7653335550
## 9	HIGH WIND	5259785375
## 10	WILDFIRE	4759064000

crop

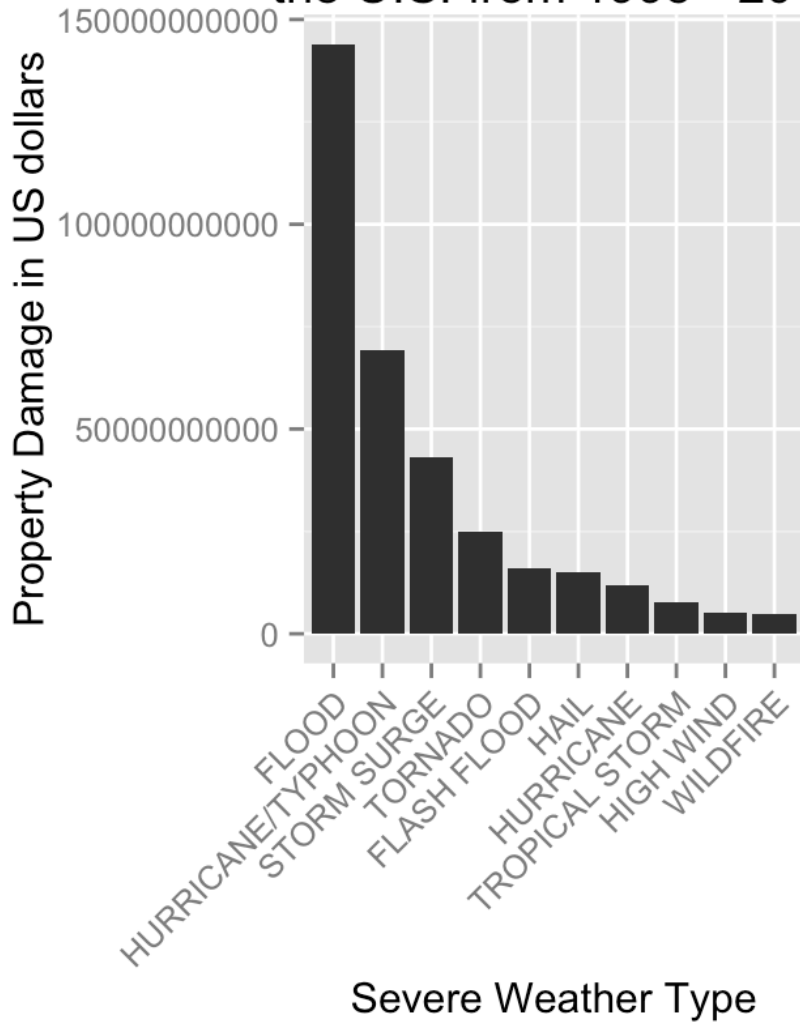
##	EVTTYPE	cropDamage
## 1	DROUGHT	13922066000
## 2	FLOOD	5422810400
## 3	HURRICANE	2741410000
## 4	HAIL	2614127070
## 5	HURRICANE/TYPHOON	2607872800
## 6	FLASH FLOOD	1343915000
## 7	EXTREME COLD	1292473000
## 8	FROST/FREEZE	1094086000
## 9	HEAVY RAIN	728399800
## 10	TROPICAL STORM	677836000

Below illustrates the total property and crop damage affected by the severe weather events.

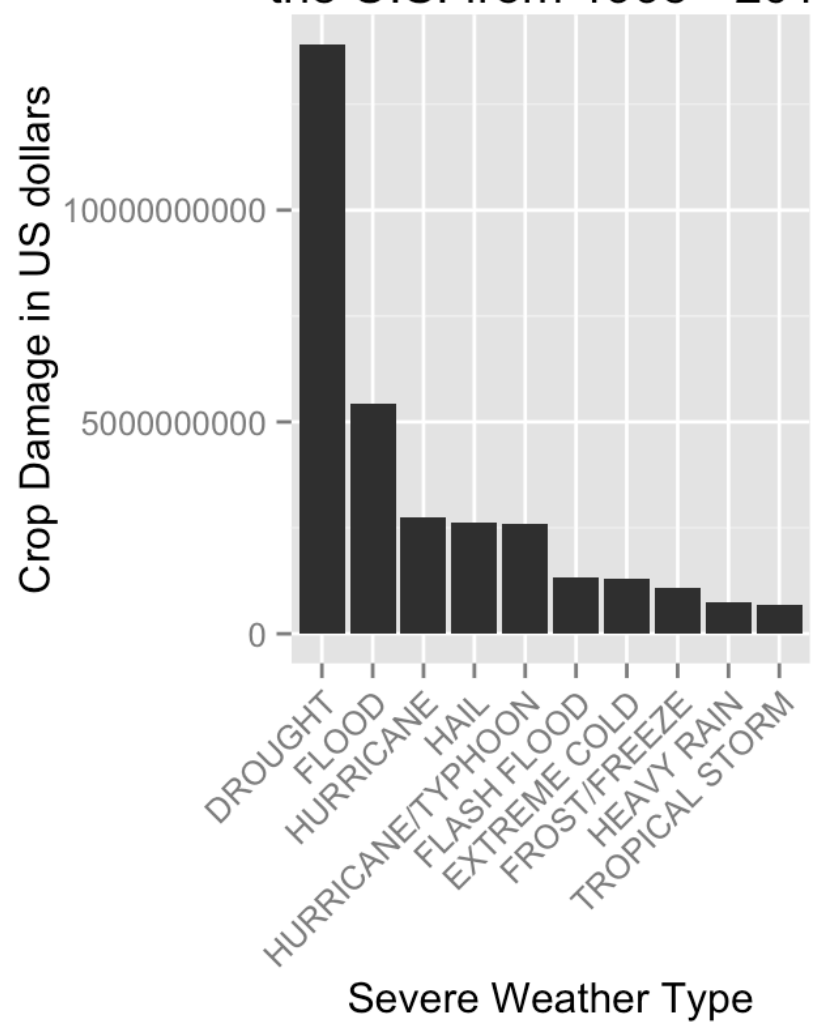
```
propertyPlot <- qplot(EVTTYPE, data = property, weight = propertyDamage, geom = "bar", binwidth = 1) +
  theme(axis.text.x = element_text(angle = 45, hjust = 1)) + scale_y_continuous("Property Damage in US dollars")+
  xlab("Severe Weather Type") + ggtitle("Total Property Damage by\n Severe Weather Events in\n the U.S. from 1995 - 2011")

cropPlot<- qplot(EVTTYPE, data = crop, weight = cropDamage, geom = "bar", binwidth = 1) +
  theme(axis.text.x = element_text(angle = 45, hjust = 1)) + scale_y_continuous("Crop Damage in US dollars") +
  xlab("Severe Weather Type") + ggtitle("Total Crop Damage by \nSevere Weather Events in\n the U.S. from 1995 - 2011")
grid.arrange(propertyPlot, cropPlot, ncol = 2)
```

Total Property Damage by Severe Weather Events in the U.S. from 1995 - 2011



Total Crop Damage by Severe Weather Events in the U.S. from 1995 - 2011



Both **flood** and **hurricane/typhoon** caused the most property damage. While **drought** and **flood** caused the most crop damage from 1995 to 2011.

Conclusion

Based on the findings, **excessive heat** and **tornado** had the greatest impact on the population health, whereas **flood**, **drought**, and **hurricane/typhoon** had the greatest economic impact.