

Casualties and Economic Impacts of Disaster Events from 1950 to 2011

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Synopsis

Disaster events(storms, tornados, flood, ...) can have significant effects on lives and economy. In this report, we use the data originated from the U.S. National Oceanic and Atmospheric Administration (NOAA) to explore this problem.

Data Processing

Data source loading

```
ReadData <- function() {  
  dataSource <- "https://d396qusza40orc.cloudfront.net/repdata%2Fdata%2FStormData.csv.bz2"  
  dataZip <- "StormData.csv.bz2"  
  
  if (!file.exists(dataZip)) {  
    download.file(dataSource, dataZip)  
  }  
  
  read.csv(dataZip, header=TRUE, strip.white = TRUE, na.strings = c("NA","", "?"))  
}  
  
rawData <- ReadData()
```

This is a big data source which contains 902297 rows and 37 columns. Out of these columns, only 7 that are of our interests: EVTYPE, FATALITIES, INJURIES, PROPDMG, PROPDMGEXP, CROPDMG, CROPDMGEXP. Unfortunately, we still have to deal with all the rows.

Data Pre-Processing Summary

The most time consuming task in this project is actually pre-processing the data to obtain the clean data for analysis. The raw data available from NOAA contains many conflicting encodings and misspellings that makes obtaining the clean data a challenging task. For example, sometimes the character “/”, sometimes “-”, is used to separate the types of events within one records. Sometimes character “-” does not act as a separation. Misspellings and short hand codes are pervasive. For example all “FLO”, “FLOO”, “FLD”, “FLDG” ... actually mean the same as “FLOOD”. Should “BITTER COLD” and “EXTREME COLD” mean the same thing? Even though I did a fairly thorough renormalizing task, different grouping methods will result in slightly different results.

Data Pre-Processing Step-by-Step

Combine the damages (property and crop) into one

```
Exp2Numeric <- function(x) {
  if (is.na(x)) 0
  else if (x %in% c(0:9)) as.numeric(x)
  else if (x == 'H' | x == 'h') 10^2
  else if (x == 'K' | x == 'k') 10^3
  else if (x == 'M' | x == 'm') 10^6
  else if (x == 'B' | x == 'b') 10^9
  else 0
}

ComputeDamage <- function(data) {
  mapply(FUN=function(x,y) {x*Exp2Numeric(y)}, data$PROPDMG, data$PROPDMGEXP) +
  mapply(FUN=function(x,y) {x*Exp2Numeric(y)}, data$CROPDMG, data$CROPDMGEXP)
}

rawData$DAMAGE <- ComputeDamage(rawData)
rawData$EVTYPE[is.na(rawData$EVTYPE)] <- "OTHER" # adjust one case
```

Initial remap of event types The task of renormalization of event types are divided into 2 sub-tasks. In the first phase here, we try to correct various pervasive conflicting encodings, and change all plurals to singulars. It does not make sense to have “TORNADO”, “TORNADOS”, or even “TORNADOES” as 3 separated categories.

```
InitialRemap <- function(x) {
  require(stringr)
```

```

result <- toupper(str_trim(x)) # make it all capital

result <- gsub('&',replacement="/", result) # substitute '/' for '&'
result <- gsub('\\\\', replacement="/", result) # substitute '/' for '\\'
result <- gsub('\\sAND\\s',replacement="/", result) # substitute '/' for 'AND'
result <- gsub(';',replacement="/", result) # replace '/' for ';'
result <- gsub('\\s*\\/\\s*',replacement="/", result) # remove space on either sides of '/'

result <- gsub('\\s*-\\s*',replacement="-", result) # remove space on either sides of '-'
## remove all character '-' that doesn't separate terms so that the only character
## that separate terms are '/'
result <- gsub('BLOW-OUT',replacement="BLOWOUT", result)
result <- gsub('HURRICANE-GENERATED',replacement="HURRICANE", result)
result <- gsub('LAKE-EFFECT',replacement="LAKE EFFECT", result)
result <- gsub('LATE-SEASON',replacement="LATE SEASON", result)
result <- gsub('NON-',replacement="NON", result)
result <- gsub('LATE-SEASON',replacement="LATE SEASON", result)
result <- gsub('-',replacement="/", result)

result <- gsub('\\s+{2}',replacement=" ", result) # remove redundant space
result <- gsub('[( )]',replacement="", result)

result <- sub('\\W+$',replacement="", result) # remove all non-word at the end

result <- gsub('\\bSNOWFALL\\b', replacement="SNOW", result)
result <- gsub('\\bRAINFALL\\b', replacement="RAIN", result)
result <- gsub('\\bASHFALL\\b', replacement="ASH", result)
result <- gsub('\\bFLOODING\\b', replacement="FLOOD", result)

result <- gsub('\\bTEMPERATURES\\b', replacement="TEMPERATURE", result)
result <- gsub('\\bTIDES\\b', replacement="TIDE", result)
result <- gsub('\\bFIRES\\b', replacement="FIRE", result)
result <- gsub('\\bFUNNELS\\b', replacement="FUNNEL", result)
result <- gsub('\\bCONDITIONS\\b', replacement="CONDITION", result)
result <- gsub('\\bWINDSS\\b', replacement="WIND", result)
result <- gsub('\\bWINDS\\b', replacement="WIND", result)
result <- gsub('\\bCHILLS\\b', replacement="CHILL", result)
result <- gsub('\\bJAMS\\b', replacement="JAM", result)
result <- gsub('\\bFLOODS\\b', replacement="FLOOD", result)
result <- gsub('\\bCLOUDS\\b', replacement="CLOUD", result)
result <- gsub('\\bSTORMS\\b', replacement="STORM", result)
result <- gsub('\\b\\bSEAS\\b', replacement="SEA", result)
result <- gsub('\\bSHOWERS\\b', replacement="SHOWER", result)
result <- gsub('\\bTREES\\b', replacement="TREE", result)
result <- gsub('\\bSQUALLS\\b', replacement="SQUALL", result)

```

```

result <- gsub('\\\\bTORNADOS\\\\b', replacement="TORNADO", result)
result <- gsub('\\\\bTORNADOES\\\\b', replacement="TORNADO", result)
result <- gsub('\\\\bWATERSPOUTS\\\\b', replacement="WATERSPOUT", result)
result <- gsub('\\\\bPELLETS\\\\b', replacement="PELLET", result)
result <- gsub('\\\\bLIGHTS\\\\b', replacement="LIGHT", result)
result <- gsub('\\\\bEFFECTS\\\\b', replacement="EFFECT", result)
result <- gsub('\\\\bSLIDES\\\\b', replacement="SLIDE", result)
result <- gsub('\\\\bROADS\\\\b', replacement="ROAD", result)
result <- gsub('\\\\bWAVES\\\\b', replacement="WAVE", result)
result <- gsub('\\\\bFLURRIES\\\\b', replacement="FLURRY", result)
result <- gsub('\\\\bSWELLS\\\\b', replacement="SWELL", result)
result <- gsub('\\\\bGUSTS\\\\b', replacement="GUST", result)
result <- gsub('\\\\bCURRENTS\\\\b', replacement="CURRENT", result)
result <- gsub('\\\\bTEMPS\\\\b', replacement="TEMPERATURE", result)
result <- gsub('\\\\bRAINS\\\\b', replacement="RAIN", result)
result <- gsub('\\\\bSNOWS\\\\b', replacement="SNOW", result)
result <- gsub('\\\\bADVISORIES\\\\b', replacement="ADVISORY", result)
result <- gsub('\\\\bLANDSLIDES\\\\b', replacement="LANDSLIDE", result)
result <- gsub('\\\\bTHUNDERSTORMS\\\\b', replacement="THUNDERSTORM", result)
result <- gsub('\\\\bMUDSLIDES\\\\b', replacement="MUDSLIDE", result)
result <- gsub('\\\\bHAILSTORMS\\\\b', replacement="HAILSTORM", result)
result <- gsub('\\\\bWILDFIRES\\\\b', replacement="WILDFIRE", result)
result <- gsub('\\\\bTHUNDERSTORMWINDS\\\\b', replacement="THUNDERSTORM WIND", result)

result <- gsub('^SUMMARY.*', replacement="OTHER", result)
result <- gsub('^NONE$', replacement="OTHER", result)

result <- gsub('\\\\bWINS\\\\b', replacement="WIND", result)
result <- gsub('W INDS', replacement="WIND", result)
result <- gsub('\\\\bWND\\\\b', replacement="WIND", result)
result <- gsub('\\\\bWINTRY\\\\b', replacement="WINTER", result)
result <- gsub('\\\\bWINTERY\\\\b', replacement="WINTER", result)
}

```

```
rawData$TYPE <- sapply(levels(rawData$EVTYPE)[rawData$EVTYPE], InitialRemap)
```

```
## Loading required package: stringr
```

Create clean data with each row only contain one event type. In the original raw data, each row represent more than one event types. However, instead of having “FLOOD/HEAVY RAIN” and “HEAVY RAIN/FLOOD” as unrelated categories, we need to separate them out as “FLOOD” and “HEAVY RAIN”. This would mean double counting the total damages. Our purpose is to

count the damages associated with each event so double counting is expected in these specific cases, which consist of around 11600 rows (approximately 1.2% of the total database).

```
SeparateEventType <- function(data) {
  evtype <- data$TYPE

  require(reshape2)
  evtype <- colsplit(evtype, "/", c("TYPE1", "TYPE2", "TYPE3"))
  evtype$TYPE2[evtype$TYPE2==""] <- NA
  evtype$TYPE3[evtype$TYPE3==""] <- NA

  data2 <- data[,c("REFNUM", "BGN_DATE", "FATALITIES", "INJURIES", "DAMAGE")]
  data2 <- cbind(data2, evtype)

  data2 <- melt(data2, measure.vars=c("TYPE1", "TYPE2", "TYPE3"),
                value.name="TYPE",
                na.rm=TRUE)
  data2$variable <- NULL
  row.names(data2) <- NULL

  data2 # return new clean data
}

cleanData <- SeparateEventType(rawData)

## Loading required package: reshape2
```

The resulting clean data have 6 column. 4 columns that are of our interest (FATALITIES, INJURIES, DAMAGE, TYPE). The remaining 2 columns are used to identify the record in case we want to trace back to original raw data (REFNUM), or to further analyzing the disaster events (BGN_DATE) in the time dimension (which we do not analyze in this report)

Create a list of categories for use in phase 2

```
# make the raw map file
CreateRawMap <- function(data) {
  rawMapFile <- "RawMap.txt"
  evnames <- unique(sort(data$TYPE))
  write(evnames, rawMapFile)
}

CreateRawMap(cleanData)
```

In order to have refined remapping of the categories, we automatically build a list of singular categories. Originally, out of 902297 rows of raw data, there are 984 different complex categories. This number is first reduced after the initial remapping, and then further reduced in this step to create 555 singular categories. After finishing phase 2 of the renormalizing process, there will be only 80 categories left. It is possible to reduce it further but that means grouping more events into one group. With 80 categories, I *almost* do no grouping at all except *for example* putting various levels of thunderstorm wind speed to one category of “THUNDERSTORM WIND”.

Create a refined remapping of categories The **RawMap.txt** file created in previous step is copied to the **remap.txt** file and the manually edit to create a detail remapping of categories. This is still a reproducible process, however, because all information is recorded in the file **remap.txt** which can be read by the script and executed accordingly. Since there is no way to save this file in Rpubs, this **remap.txt** file can be downloaded from my github account: [remap.txt](#).

```
DetailRemap <- function(data) {
  remap <- read.table("remap.txt", fill=TRUE,header=TRUE,sep=">",
                     colClasses=c("character","character"),strip.white=TRUE)

  RemapEventName <- function(x) {
    res <- remap$new[remap$original==x]
    ifelse(res=="",x,res)
  }
  sapply(data$TYPE,RemapEventName)
}

cleanData$TYPE <- DetailRemap(cleanData)
cleanData$TYPE <- factor(cleanData$TYPE)
```

The final result of the remapping process is the 80 categories of events used in the data:

- STORM RELATED
 - THUNDERSTORM
 - THUNDERSTORM WIND
 - TROPICAL DEPRESSION
 - TROPICAL STORM
 - HURRICANE
 - WINTER STORM

- SNOW STORM
 - ICE STORM
 - BLIZZARD
- FLOOD RELATED
 - FLOOD
 - FLASH FLOOD
 - LANDSLIDE
- MASSIVE DISASTER
 - AVALANCHE
 - DAM FAILURE
 - WILDFIRE
 - VOLCANIC ACTIVITY
 - TSUNAMI
- PRECIPITATION RELATED
 - EXCESSIVE PRECIPITATION
 - RAIN
 - HEAVY RAIN
 - FREEZING RAIN
 - FREEZE (FROST/FREEZE)
 - HAIL
 - SLEET
 - SNOW
 - LAKE EFFECT SNOW
 - SNOW DROUGHT (lack of snow)
 - HEAVY SNOW
 - FREEZING FOG
 - DENSE FOG
 - DENSE SMOKE
 - ICE
 - ICE FLOES
 - ICE JAM
- MARINE RELATED
 - COASTAL EROSION
 - COASTAL FLOOD
 - COASTAL STORM
 - COASTAL SURGE
 - HIGH TIDE

- HIGH WAVE
- EXTREME HIGH TIDE
- EXTREME LOW TIDE
- RIP CURRENT
- STORM SURGE
- SEICHE
- MARINE HAIL
- MARINE HIGH WIND
- MARINE THUNDERSTORM WIND
- MARINE ACCIDENT
- WEATHER CONDITIONS
 - DRY
 - DROUGHT
 - HEAT
 - COOL
 - WET
 - COLD
 - EXTREME COLD
 - WIND CHILL
 - EXTREME WIND CHILL
 - UNSEASONABLY COOL
 - UNSEASONABLY COLD
 - UNSEASONABLY DRY
 - UNSEASONABLY HOT
 - UNSEASONABLY WET
 - WINTER WEATHER
- WIND RELATED
 - WIND
 - HIGH WIND
 - WHIRLWIND
 - DOWNBURST
 - GUSTNADO
 - FUNNEL CLOUD
 - WALL CLOUD
 - WATERSPOUT
 - TORNADO
- DUST RELATED
 - DUST DEVIL

- DUST STORM
- SAHARAN DUST
- RARE CONDITIONS
 - DROWNING
 - EXPOSURE
 - LIGHTNING
 - OTHER

Results

Create summary data for each type of events

```
CreateSummaryData <- function(data) {
  aggregate(data[3:5], by=data["TYPE"],sum)
}
```

```
(summaryData <- CreateSummaryData(cleanData))
```

##	TYPE	FATALITIES	INJURIES	DAMAGE
## 1	AVALANCHE	225	171	8.722e+06
## 2	BLIZZARD	101	806	7.770e+08
## 3	COASTAL EROSION	0	5	5.210e+07
## 4	COASTAL FLOOD	6	7	4.489e+08
## 5	COASTAL STORM	4	2	5.000e+04
## 6	COASTAL SURGE	0	0	5.000e+05
## 7	COLD	166	65	1.876e+08
## 8	COOL	0	0	5.000e+06
## 9	DAM FAILURE	0	0	1.002e+06
## 10	DENSE FOG	81	1077	2.283e+07
## 11	DENSE SMOKE	0	0	1.000e+05
## 12	DOWNBURST	3	29	7.315e+06
## 13	DROUGHT	6	19	1.502e+10
## 14	DROWNING	1	0	0.000e+00
## 15	DRY	29	0	0.000e+00
## 16	DUST DEVIL	2	43	7.186e+05
## 17	DUST STORM	22	440	9.219e+06
## 18	EXCESSIVE PRECIPITATION	0	0	1.805e+06
## 19	EXPOSURE	17	0	0.000e+00
## 20	EXTREME COLD	290	255	1.446e+09
## 21	EXTREME HIGH TIDE	0	0	9.425e+06
## 22	EXTREME LOW TIDE	0	0	3.200e+05
## 23	EXTREME WIND CHILL	17	5	1.780e+07
## 24	FLASH FLOOD	1035	1802	1.844e+10

## 25	FLOOD	550	6889	1.616e+11
## 26	FREEZE	4	18	3.122e+09
## 27	FREEZING FOG	0	0	2.182e+06
## 28	FREEZING RAIN	39	121	2.218e+07
## 29	FUNNEL CLOUD	0	3	1.996e+05
## 30	GUSTNADO	0	0	1.036e+05
## 31	HAIL	20	1467	1.913e+10
## 32	HEAT	3134	9211	9.245e+08
## 33	HEAVY RAIN	101	280	4.044e+09
## 34	HEAVY SNOW	129	1037	1.092e+09
## 35	HIGH TIDE	3	0	5.054e+05
## 36	HIGH WAVE	226	309	1.274e+08
## 37	HIGH WIND	414	1820	9.550e+09
## 38	HURRICANE	199	2608	1.628e+11
## 39	ICE	25	436	2.350e+07
## 40	ICE FLOES	0	0	1.000e+05
## 41	ICE JAM	0	0	5.000e+03
## 42	ICE STORM	89	1992	8.968e+09
## 43	LAKE EFFECT SNOW	0	0	4.018e+07
## 44	LANDSLIDE	44	55	3.474e+08
## 45	LIGHTNING	817	5232	9.509e+08
## 46	MARINE ACCIDENT	8	7	5.000e+04
## 47	MARINE HAIL	0	0	4.000e+03
## 48	MARINE HIGH WIND	15	23	1.715e+06
## 49	MARINE THUNDERSTORM WIND	19	34	5.907e+06
## 50	OTHER	64	109	4.711e+09
## 51	RAIN	5	2	1.289e+08
## 52	RIP CURRENT	577	529	1.630e+05
## 53	SAHARAN DUST	0	0	0.000e+00
## 54	SEICHE	0	0	9.800e+05
## 55	SLEET	2	0	2.000e+06
## 56	SNOW	36	93	2.980e+07
## 57	SNOW DROUGHT	0	0	0.000e+00
## 58	SNOW STORM	4	36	1.465e+06
## 59	STORM SURGE	24	43	4.797e+10
## 60	THUNDERSTORM	1	12	1.226e+09
## 61	THUNDERSTORM WIND	710	9497	1.102e+10
## 62	TORNADO	5661	91407	5.901e+10
## 63	TROPICAL DEPRESSION	0	0	1.737e+06
## 64	TROPICAL STORM	66	383	8.409e+09
## 65	TSUNAMI	33	129	1.441e+08
## 66	UNSEASONABLY COLD	2	0	3.014e+07
## 67	UNSEASONABLY COOL	0	0	0.000e+00
## 68	UNSEASONABLY DRY	0	0	0.000e+00
## 69	UNSEASONABLY HOT	40	17	1.000e+04
## 70	UNSEASONABLY WET	0	0	0.000e+00

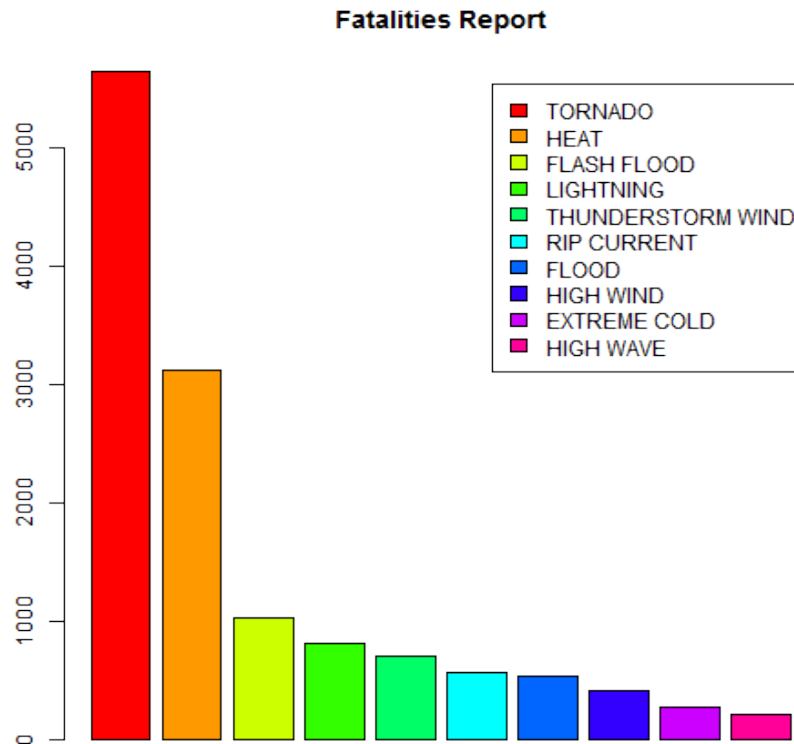
## 71	VOLCANIC ACTIVITY	0	0	5.000e+05
## 72	WALL CLOUD	0	0	0.000e+00
## 73	WATERSPOUT	6	71	6.069e+07
## 74	WET	0	0	2.130e+08
## 75	WHIRLWIND	1	0	1.200e+04
## 76	WILDFIRE	102	2153	1.201e+10
## 77	WIND	29	95	1.248e+08
## 78	WIND CHILL	220	36	1.129e+07
## 79	WINTER STORM	217	1421	6.783e+09
## 80	WINTER WEATHER	62	547	4.225e+07

Create more consise report on fatalities

```
(fatalities <- (head(summaryData[order(summaryData$FATALITIES, decreasing=TRUE),
c("TYPE", "FATALITIES")],n=10)))
```

##	TYPE	FATALITIES
## 62	TORNADO	5661
## 32	HEAT	3134
## 24	FLASH FLOOD	1035
## 45	LIGHTNING	817
## 61	THUNDERSTORM WIND	710
## 52	RIP CURRENT	577
## 25	FLOOD	550
## 37	HIGH WIND	414
## 20	EXTREME COLD	290
## 36	HIGH WAVE	226

```
barplot(fatalities[,2], col=rainbow(10), legend=fatalities[,1],
main="Fatalities Report")
```



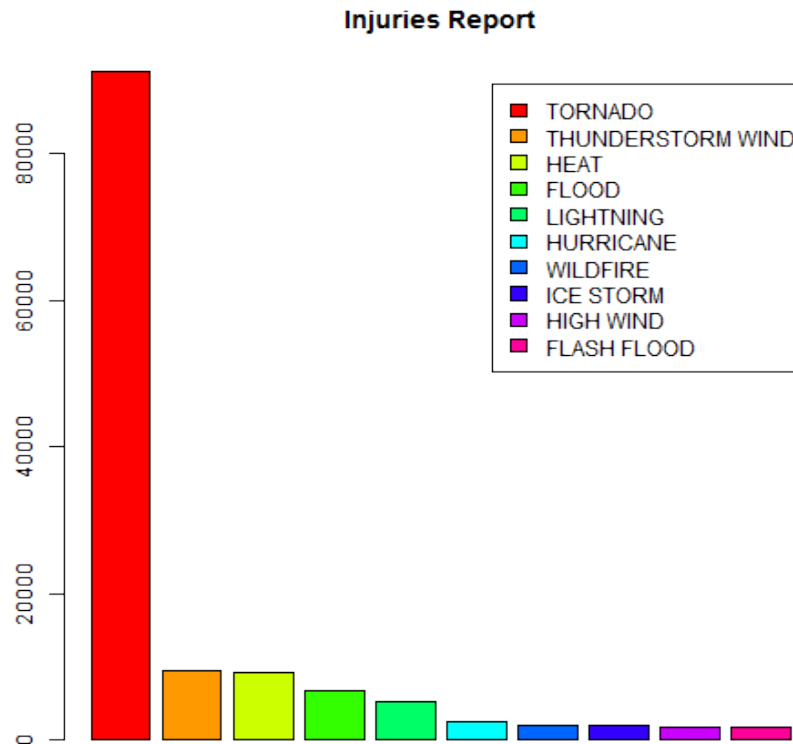
Create more consise report on injuries

```
(injuries <- (head(summaryData[order(summaryData$INJURIES, decreasing=TRUE),
                                     c("TYPE", "INJURIES")],n=10)))
```

##	TYPE	INJURIES
## 62	TORNADO	91407
## 61	THUNDERSTORM WIND	9497
## 32	HEAT	9211
## 25	FLOOD	6889
## 45	LIGHTNING	5232
## 38	HURRICANE	2608
## 76	WILDFIRE	2153
## 42	ICE STORM	1992
## 37	HIGH WIND	1820

```
## 24          FLASH FLOOD      1802
```

```
barplot(injuries[,2], col=rainbow(10), legend=injuries[,1],  
        main="Injuries Report")
```



Conclusion about the effects of diaster events on human lives

Tornado remains the most disastrous event in human lives in the US in term of both fatalities as well as injuries. Heat is the second biggest factor in fatalities but third in causing injuries after tornado and thunderstorm wind. Flooding, including flood and flash flood, is very high on the disaster scale. Separately, flash flood causes more fatalities, but flood cause more injuries. One of the most surprised result is the level of effects caused by lightning which placed 4th and 5th in the two lists. Lightning is supposed to be rare event, so perhaps the

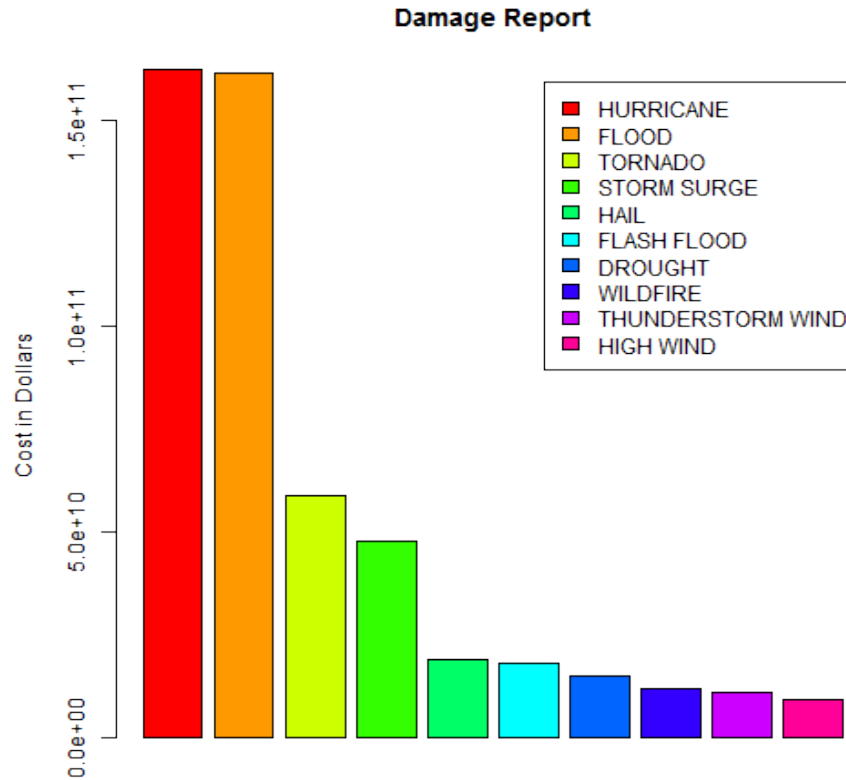
reason it is so high on the lists is because it is always recorded during 60 years of recording span.

Create more consise report on damages

```
(damages <- (head(summaryData[order(summaryData$DAMAGE, decreasing=TRUE),  
                                c("TYPE", "DAMAGE")],n=10)))
```

```
##           TYPE      DAMAGE  
## 38      HURRICANE 1.628e+11  
## 25         FLOOD 1.616e+11  
## 62       TORNADO 5.901e+10  
## 59    STORM SURGE 4.797e+10  
## 31         HAIL 1.913e+10  
## 24    FLASH FLOOD 1.844e+10  
## 13       DROUGHT 1.502e+10  
## 76       WILDFIRE 1.201e+10  
## 61 THUNDERSTORM WIND 1.102e+10  
## 37      HIGH WIND 9.550e+09
```

```
barplot(damages[,2], col=rainbow(10), legend=damages[,1],  
        main="Damage Report",  
        ylab="Cost in Dollars")
```



Hurricane and flood are *statistically* tied in first place in term of economic damages. However, if we include both flood and flash flood into flooding, it would easily be the highest damaging factor. Tornado drops to the 3rd rank, but lightning is no where in sight, understandably.

Conclusions

Tornado is the single biggest factor that affects human lives in the US both in term of fatalities as well as injuries. Heat is the second in this respect, with flooding, which include flash flood and flood, the third.

In term of property damages, flooding(flash flood and flood) is the biggest factor with hurricane 2nd and tornado 3rd.