

Analysis of Storms, and other severe weather events that cause both public health and economic problems in United States.

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1. Synopsis

The National Oceanic and Atmospheric Administration (NOAA) maintains a public database for storm event. The data contains the type of storm event, details like location, date, estimates for damage to property as well as the number of human victims of the storm. In this report we investigate which type of events are the most harmful to the population and financially.

The conclusion is that the impact on humans, be it injuries or fatalities, isn't directly correlated to the economic damage weather events cause. Tornado is the highest cause for injuries and fatalities, whilst flood and drought cause the greatest cost of damages on the cops and properties.

2. Data Processing

2.1 Load all the libraried used for data loading, computation and plotting

```
library(R.utils)
library(data.table)
```

2.2 Unzip the file and process it:-

The data file is downloaded and copied to the working directory.

```
bunzip2("repdata-data-StormData.csv.bz2", overwrite=T, remove=F)
data <- read.csv("repdata-data-StormData.csv")
```

We will perform the basic data analysis:

```
summary(data)
```

##	STATE__		BGN_DATE		BGN_TIME	
##	Min. : 1.0	5/25/2011 0:00:00:	1202	12:00:00 AM:	10163	
##	1st Qu.:19.0	4/27/2011 0:00:00:	1193	06:00:00 PM:	7350	
##	Median :30.0	6/9/2011 0:00:00 :	1030	04:00:00 PM:	7261	
##	Mean :31.2	5/30/2004 0:00:00:	1016	05:00:00 PM:	6891	
##	3rd Qu.:45.0	4/4/2011 0:00:00 :	1009	12:00:00 PM:	6703	
##	Max. :95.0	4/2/2006 0:00:00 :	981	03:00:00 PM:	6700	
##	(Other)	:895866	(Other)	:857229		
##	TIME_ZONE	COUNTY	COUNTYNAME		STATE	
##	CST :547493	Min. : 0.0	JEFFERSON : 7840	TX	: 83728	
##	EST :245558	1st Qu.: 31.0	WASHINGTON: 7603	KS	: 53440	
##	MST : 68390	Median : 75.0	JACKSON : 6660	OK	: 46802	
##	PST : 28302	Mean :100.6	FRANKLIN : 6256	MO	: 35648	
##	AST : 6360	3rd Qu.:131.0	LINCOLN : 5937	IA	: 31069	
##	HST : 2563	Max. :873.0	MADISON : 5632	NE	: 30271	
##	(Other): 3631	(Other)	:862369	(Other):	621339	
##		EVTTYPE	BGN_RANGE		BGN_AZI	
##	HAIL	:288661	Min. : 0.000		:547332	
##	TSTM WIND	:219940	1st Qu.: 0.000	N	: 86752	
##	THUNDERSTORM WIND:	82563	Median : 0.000	W	: 38446	
##	TORNADO	: 60652	Mean : 1.484	S	: 37558	
##	FLASH FLOOD	: 54277	3rd Qu.: 1.000	E	: 33178	
##	FLOOD	: 25326	Max. :3749.000	NW	: 24041	
##	(Other)	:170878	(Other):	134990		
##	BGN_LOCATI		END_DATE		END_TIME	
##	:287743		:243411		:238978	
##	COUNTYWIDE : 19680	4/27/2011 0:00:00:	1214	06:00:00 PM:	9802	
##	Countywide : 993	5/25/2011 0:00:00:	1196	05:00:00 PM:	8314	
##	SPRINGFIELD : 843	6/9/2011 0:00:00 :	1021	04:00:00 PM:	8104	
##	SOUTH PORTION: 810	4/4/2011 0:00:00 :	1007	12:00:00 PM:	7483	
##	NORTH PORTION: 784	5/30/2004 0:00:00:	998	11:59:00 PM:	7184	
##	(Other) :591444	(Other)	:653450	(Other)	:622432	
##	COUNTY_END	COUNTYENDN	END_RANGE		END_AZI	
##	Min. :0	Mode:logical	Min. : 0.0000		:724837	
##	1st Qu.:0	NA's:902297	1st Qu.: 0.0000	N	: 28082	
##	Median :0		Median : 0.0000	S	: 22510	
##	Mean :0		Mean : 0.9862	W	: 20119	
##	3rd Qu.:0		3rd Qu.: 0.0000	E	: 20047	
##	Max. :0		Max. :925.0000	NE	: 14606	
##			(Other):	72096		
##	END_LOCATI		LENGTH		WIDTH	
##	:499225	Min. : 0.0000	Min. : 0.000			
##	COUNTYWIDE : 19731	1st Qu.: 0.0000	1st Qu.: 0.000			
##	SOUTH PORTION : 833	Median : 0.0000	Median : 0.000			
##	NORTH PORTION : 780	Mean : 0.2301	Mean : 7.503			
##	CENTRAL PORTION: 617	3rd Qu.: 0.0000	3rd Qu.: 0.000			
##	SPRINGFIELD : 575	Max. :2315.0000	Max. :4400.000			
##	(Other)	:380536				
##	F	MAG	FATALITIES		INJURIES	
##	Min. :0.0	Min. : 0.0	Min. : 0.0000	Min. : 0.0000		
##	1st Qu.:0.0	1st Qu.: 0.0	1st Qu.: 0.0000	1st Qu.: 0.0000		
##	Median :1.0	Median : 50.0	Median : 0.0000	Median : 0.0000		
##	Mean :0.9	Mean : 46.9	Mean : 0.0168	Mean : 0.1557		
##	3rd Qu.:1.0	3rd Qu.: 75.0	3rd Qu.: 0.0000	3rd Qu.: 0.0000		

```

## Max. :5.0 Max. :22000.0 Max. :583.0000 Max. :1700.0000
## NA's :843563
## PROPDMG PROPDMGEXP CROPDMG CROPDMGEXP
## Min. : 0.00 :465934 Min. : 0.000 :618413
## 1st Qu.: 0.00 K :424665 1st Qu.: 0.000 K :281832
## Median : 0.00 M : 11330 Median : 0.000 M : 1994
## Mean : 12.06 0 : 216 Mean : 1.527 k : 21
## 3rd Qu.: 0.50 B : 40 3rd Qu.: 0.000 0 : 19
## Max. :5000.00 5 : 28 Max. :990.000 B : 9
## (Other): 84 (Other): 9
## WFO STATEOFFIC
## :142069 :248769
## OUN : 17393 TEXAS, North : 12193
## JAN : 13889 ARKANSAS, Central and North Central: 11738
## LWX : 13174 IOWA, Central : 11345
## PHI : 12551 KANSAS, Southwest : 11212
## TSA : 12483 GEORGIA, North and Central : 11120
## (Other):690738 (Other) :595920
##
ZONENAMES
##
:594029
##
:205988
## GREATER RENO / CARSON CITY / M - GREATER RENO / CARSON CITY / M
: 639
## GREATER LAKE TAHOE AREA - GREATER LAKE TAHOE AREA
: 592
## JEFFERSON - JEFFERSON
: 303
## MADISON - MADISON
: 302
## (Other)
:100444
## LATITUDE LONGITUDE LATITUDE_E LONGITUDE_
## Min. : 0 Min. : -14451 Min. : 0 Min. : -14455
## 1st Qu.:2802 1st Qu.: 7247 1st Qu.: 0 1st Qu.: 0
## Median :3540 Median : 8707 Median : 0 Median : 0
## Mean :2875 Mean : 6940 Mean :1452 Mean : 3509
## 3rd Qu.:4019 3rd Qu.: 9605 3rd Qu.:3549 3rd Qu.: 8735
## Max. :9706 Max. : 17124 Max. :9706 Max. :106220
## NA's :47 NA's :40
## REMARKS REFNUM
## :287433 Min. : 1
## : 24013 1st Qu.:225575
## Trees down.\n : 1110 Median :451149
## Several trees were blown down.\n : 569 Mean :451149
## Trees were downed.\n : 446 3rd Qu.:676723
## Large trees and power lines were blown down.\n: 432 Max. :902297
## (Other) :588294

```

names(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"   "CROPDMG"       "CROPDMGEXP"    "WFO"           "STATEOFFIC"
## [31] "ZONENAMES"    "LATITUDE"      "LONGITUDE"     "LATITUDE_E"    "LONGITUDE_"
## [36] "REMARKS"      "REFNUM"
```

From all the columns, only several of them will be relevant to our analysis. These are: * EVTYPE: the type of weather event * FATALITIES: the number of fatalities * INJURIES: the number of injuries * PROPDMG: the amount of property damage (in US dollars) * PROPDMGEXP: a multiplier for PROPDMG * CROPDMG: the amount of crop damage (in US dollars) * CROPDMGEXP: a multiplier for CROPDMG

2.3 Remove the unused columns (not needed for analysis)

```
use_columns <- c("BGN_DATE", "EVTYPE", "FATALITIES", "INJURIES", "PROPDMG", "PROPDMGEXP", "CROPDMG", "CROPDMGEXP")
storm_data <- data[,use_columns]
storm_data$EVTYPE <- toupper(data$EVTYPE)
storm_data <- storm_data[!grep1("Summary", data$EVTYPE), ]
```

3. Results

3.1 Most harmful events with respect to population health

Find the top 10 weather events that cause fatalities and injuries

```
dt_storm_data <- as.data.table(storm_data)
dt_fatalities <- dt_storm_data[,sum(FATALITIES),by=EVTYPE]
dt_fatalities <- dt_fatalities[order(-rank(V1))]
dt_fatalities <- dt_fatalities[1:10,]
setnames(dt_fatalities, c('Event','Total'))
dt_fatalities
```

```
##           Event Total
## 1:    TORNADO  5633
## 2: EXCESSIVE HEAT 1903
## 3:   FLASH FLOOD  978
## 4:        HEAT  937
## 5:   LIGHTNING  816
## 6:   TSTM WIND  504
## 7:        FLOOD  470
## 8:   RIP CURRENT 368
## 9:    HIGH WIND  248
## 10:  AVALANCHE  224
```

```
dt_injuries <- dt_storm_data[,sum(INJURIES), by=EVTTYPE]
dt_injuries <- dt_injuries[order(-rank(V1))]
dt_injuries <- dt_injuries[1:10,]
setnames(dt_injuries, c('Event','Total'))
dt_injuries
```

```
##           Event Total
## 1:      TORNADO 91346
## 2:      TSTM WIND 6957
## 3:       FLOOD 6789
## 4: EXCESSIVE HEAT 6525
## 5:      LIGHTNING 5230
## 6:         HEAT 2100
## 7:      ICE STORM 1975
## 8:    FLASH FLOOD 1777
## 9: THUNDERSTORM WIND 1488
## 10:        HAIL 1361
```

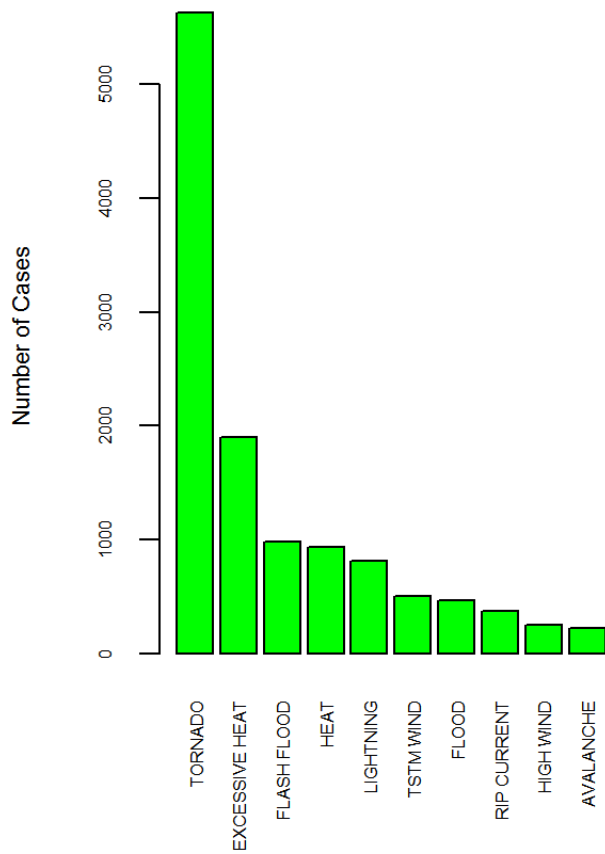
Plot graph for top 10 weather events classified by number of fatalities and injuries

```
par(mfrow=c(1,2))
```

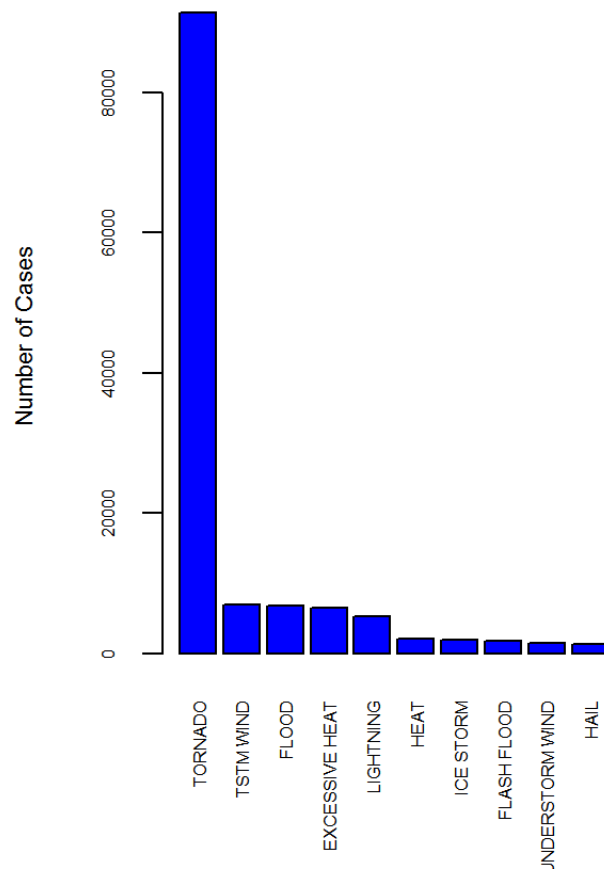
```
barplot(dt_fatalities$Total, las=3, names.arg=dt_fatalities$Event, col="green", main='Top 10 Event Causes Highest Fatalities',ylab='Number of Cases',cex.lab=0.7,cex.main=0.7,cex.axis=0.5,cex.name=0.5)
```

```
barplot(dt_injuries$Total, las=3, names.arg=dt_injuries$Event, col="blue", main='Top 10 Event Causes Highest Injuries',ylab='Number of Cases',cex.lab=0.7,cex.main=0.7,cex.axis=0.5,cex.name=0.5)
```

Top 10 Event Causes Highest Fatalities



Top 10 Event Causes Highest Injuries



From these plots, we see that tornadoes are the weather events that caused the most injuries and fatalities between 1950 and 2011 across the U.S. If we consider injuries, the 2nd and 3rd most harmful types of events are respectively thunderstorm winds and floods. In the case of fatalities, the equivalent are excessive heats and flash floods.

3.2 The events that had the greatest economic consequences

Calculate the costs of the damages on crops and properties

```

# Set to upper case
dt_storm_data$CROPDMGEXP <- toupper(dt_storm_data$CROPDMGEXP)
dt_storm_data$PROPDMGEXP <- toupper(dt_storm_data$PROPDMGEXP)

# Take the unique values
cropdmg_exp <- as.character(unique(dt_storm_data$CROPDMGEXP))
propdmg_exp <- as.character(unique(dt_storm_data$PROPDMGEXP))

# Combine the unique upper case values using union
damage_exp <- union(cropdmg_exp, propdmg_exp)

damage_values <- function(x) {
  e <- list(K=3, M=6, B=9,
            "+"=0, "-"=0, "?"=0,
            "0"=0, "1"=1, "2"=2, "3"=3, "4"=4, "5"=5, "6"=6, "7"=7, "8"=8, "9"=9)

  if (x %in% names(e)) {
    int_value <- e[[x]]
  }
  else {
    int_value <- 0
  }
  return(int_value)
}

calculate_cost <- function(damage, damage_exp) {
  damage*10^damage_values(damage_exp)
}

dt_storm_data$CROPDMG_COST = mapply(calculate_cost, dt_storm_data$CROPDMG, dt_storm_data$CROPDMGEXP)
dt_storm_data$PROPDMG_COST = mapply(calculate_cost, dt_storm_data$PROPDMG, dt_storm_data$PROPDMGEXP)

```

Find the top 10 events having the greatest cost of damages on the crops and properties

```

dt_crops <- dt_storm_data[,sum(CROPDMG_COST), by=EVTTYPE]
dt_crops <- dt_crops[order(-rank(V1))]
dt_crops<- dt_crops[1:10,]
setnames(dt_crops, c('Event', 'Total'))
dt_crops

```


##	Event	Total
## 1:	DROUGHT	13972566000
## 2:	FLOOD	5661968450
## 3:	RIVER FLOOD	5029459000
## 4:	ICE STORM	5022113500
## 5:	HAIL	3025954473
## 6:	HURRICANE	2741910000
## 7:	HURRICANE/TYPHOON	2607872800
## 8:	FLASH FLOOD	1421317100
## 9:	EXTREME COLD	1312973000
## 10:	FROST/FREEZE	1094186000

```
dt_properties <- dt_storm_data[,sum(PROPDMG_COST), by=EVTTYPE]
dt_properties <- dt_properties[order(-rank(V1))]
dt_properties <- dt_properties[1:10,]
setnames(dt_properties, c('Event','Total'))
dt_properties
```

##	Event	Total
## 1:	FLOOD	144657709807
## 2:	HURRICANE/TYPHOON	69305840000
## 3:	TORNADO	56947380677
## 4:	STORM SURGE	43323536000
## 5:	FLASH FLOOD	16822673979
## 6:	HAIL	15735267018
## 7:	HURRICANE	11868319010
## 8:	TROPICAL STORM	7703890550
## 9:	WINTER STORM	6688497251
## 10:	HIGH WIND	5270046295

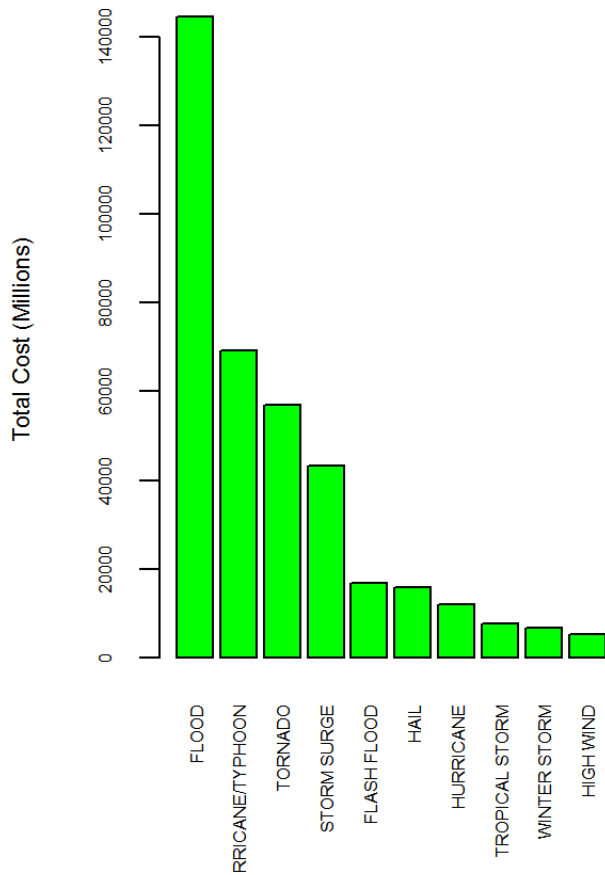
Plot graph for top 10 weather events having the greatest economic consequences

```
par(mfrow=c(1,2))
```

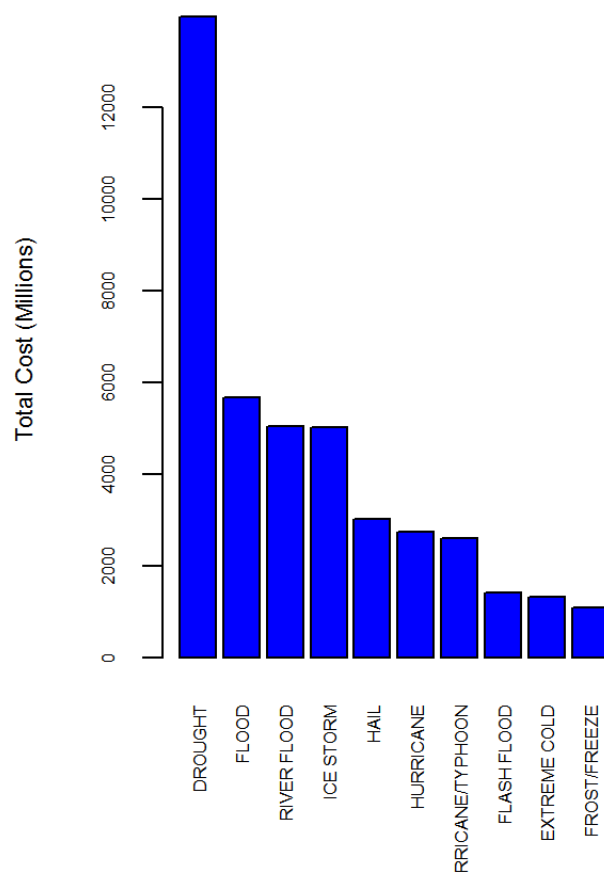
```
barplot(dt_properties$Total/1000000, las=3, names.arg=dt_properties$Event, col="green", main='Top 10 Event Causes Greatest Cost of the Damages on Properties',ylab='Total Cost (Millions)',cex.lab=0.7,cex.main=0.7,cex.axis=0.5,cex.name=0.5)
```

```
barplot(dt_crops$Total/1000000, las=3, names.arg=dt_crops$Event, col="blue", main='Top 10 Event Causes Greatest Cost of the Damages on Crops',ylab='Total Cost (Millions)',cex.lab=0.7,cex.main=0.7,cex.axis=0.5,cex.name=0.5)
```

Top 10 Event Causes Greatest Cost of the Damages on Proper



Top 10 Event Cause Greatest Cost of the Damages on Crop



Based on the histograms above, we find that flood and hurricane/typhoon cause most property damage; drought and flood causes most crop damage in the United States from 1995 to 2011.

4. Conclusion

From these data, we found that excessive heat and tornado are most harmful with respect to population health, while flood, drought, and hurricane/typhoon have the greatest economic impact.