

# **Assignment: Reproducible Research Week 4 Course Project 2**

## **OVERVIEW**

- Weather events cause public health and economic problems for communities and municipalities. Severe events result in fatalities, injuries, and damage. Predicting and/or preventing these outcomes is a primary objective.
- This analysis examines the damaging effects of severe weather conditions (e.g. hurricanes, tornadoes, thunderstorms, floods, etc.) on human populations and the economy in the U.S. from 1950 to 2011.
- As a result, the analysis will highlight the severe weather events associated with the greatest impact on the economy and population health.

## **SYNOPSIS**

- This is an exploration of the U.S. National Oceanic and Atmospheric Administration's (NOAA) storm database.
- This database tracks characteristics of major storms and weather events in the United States, including when and where they occur, which type of event, as well as the estimates of relevant fatalities, injuries, and various forms of damage.
- The dataset used in this project is provided by the U.S. National Oceanic and Atmospheric Administration (NOAA).
- This analysis discovered that tornados are responsible for a maximum number of fatalities and injuries.
- This analysis also discovered that floods are responsible for maximum property damage, while Droughts cause maximum crop damage.

Objective: Explore the NOAA Storm Database to help answer important questions about severe weather events.

## **DATA PROCESSING**

### **DATA PREP**

DP1.1 Install packages & Load libraries

Install packages,

```
# load libraries ...

library(R.utils)
library(rmarkdown)
library(knitr)
Sys.setlocale("LC_TIME", "English")
```

DP1.2 Download the storm data file into the designated working directory folder

```
temp <- tempfile()

##Performing the download
if(!file.exists("/stormData.csv.bz2")){
  download.file("https://d396qusza40orc.cloudfront.net/repdata%2Fdata%2F
StormData.csv.bz2", destfile="./stormData.csv.bz2")
}

##Uncompressing the file
if(!file.exists("stormdata.csv"))
{
  bunzip2("stormData.csv.bz2", "stormdata.csv", remove=F)
}

# DP1.3 loading the data & reading the file

storm <- read.csv("stormdata.csv", header=TRUE, sep=",")
summary(storm)
```

##	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
##
##              EVTYPE      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. : 0.0
000
## 1st Qu.:0.0 1st Qu.: 0.0 1st Qu.: 0.0000 1st Qu.: 0.0
000
## Median :1.0 Median : 50.0 Median : 0.0000 Median : 0.0
000
## Mean :0.9 Mean : 46.9 Mean : 0.0168 Mean : 0.1
557
## 3rd Qu.:1.0 3rd Qu.: 75.0 3rd Qu.: 0.0000 3rd Qu.: 0.0
000
## Max. :5.0 Max. :22000.0 Max. :583.0000 Max. :1700.0
000
## 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.:2255
75
## Trees down.\n : 1110 Median :4511
49
## Several trees were blown down.\n : 569 Mean :4511
49
## Trees were downed.\n : 446 3rd Qu.:6767
23
## Large trees and power lines were blown down.\n: 432 Max. :9022
97
## (Other) :588294

```

```
names(storm)
```

```

## [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"
```

## RESULTS

**QUESTION 1. Across the United States, which types of events (as indicated in the EVTYPE variable) are most harmful with respect to population health?**

Q1.1 Variable selection (reducing the data set to only needed columns and variables)

```
variables<-c("EVTYPE", "FATALITIES", "INJURIES", "PROPDMG", "PROPDMGEXP", "CROPDMG", "CROPDMGEXP")
strmdata<-storm[variables]

dim(strmdata)
```

```
## [1] 902297 7
```

```
## [1] "EVTYPE" "FATALITIES" "INJURIES" "PROPDMG" "PROPDMGEXP"
## [6] "CROPDMG" "CROPDMGEXP"
```

```
# Q1.2 Reviewing events that cause the most fatalities ( The Top-10 Fatalities by Weather Event )
```

```
## Procedure = aggregate the top 10 fatalities by the event type and sort the output in descending order
```

```
Fatalities <- aggregate(FATALITIES ~ EVTYPE, data = strmdata, FUN = sum)
Top10_Fatalities <- Fatalities[order(-Fatalities$FATALITIES), ][1:10, ]
```

## Top10\_Fatalities

##	EVTYPE	FATALITIES
## 834	TORNADO	5633
## 130	EXCESSIVE HEAT	1903
## 153	FLASH FLOOD	978
## 275	HEAT	937
## 464	LIGHTNING	816
## 856	TSTM WIND	504
## 170	FLOOD	470
## 585	RIP CURRENT	368
## 359	HIGH WIND	248
## 19	AVALANCHE	224

*# Q1.3 Reviewing events that cause the most injuries ( The Top-10 Injuries by Weather Event )*

*## Procedure = aggregate the top 10 injuries by the event type and sort the output in descending order*

```
Injuries <- aggregate(INJURIES ~ EVTYPE, data = strmdata, FUN = sum)
Top10_Injuries <- Injuries[order(-Injuries$INJURIES), ][1:10, ]
Top10_Injuries
```

##	EVTYPE	INJURIES
## 834	TORNADO	91346
## 856	TSTM WIND	6957
## 170	FLOOD	6789
## 130	EXCESSIVE HEAT	6525
## 464	LIGHTNING	5230
## 275	HEAT	2100
## 427	ICE STORM	1975
## 153	FLASH FLOOD	1777

## 760	THUNDERSTORM WIND	1488
## 244	HAIL	1361

```
# Q1.4 Plot of Top 10 Fatalities & Injuries for Weather Event Types ( Population Health Impact )
```

```
## Procedure = plot graphs showing the top 10 fatalities and injuries
```

```
par(mfrow=c(1,2),mar=c(10,3,3,2))
```

```
barplot(Top10_Fatalities$FATALITIES,names.arg=Top10_Fatalities$EVTYPE,las=2,col="purple",ylab="fatalities",main="Top 10 fatalities")
```

```
barplot(Top10_Injuries$INJURIES,names.arg=Top10_Injuries$EVTYPE,las=2,col="purple",ylab="injuries",main="Top 10 Injuries")
```

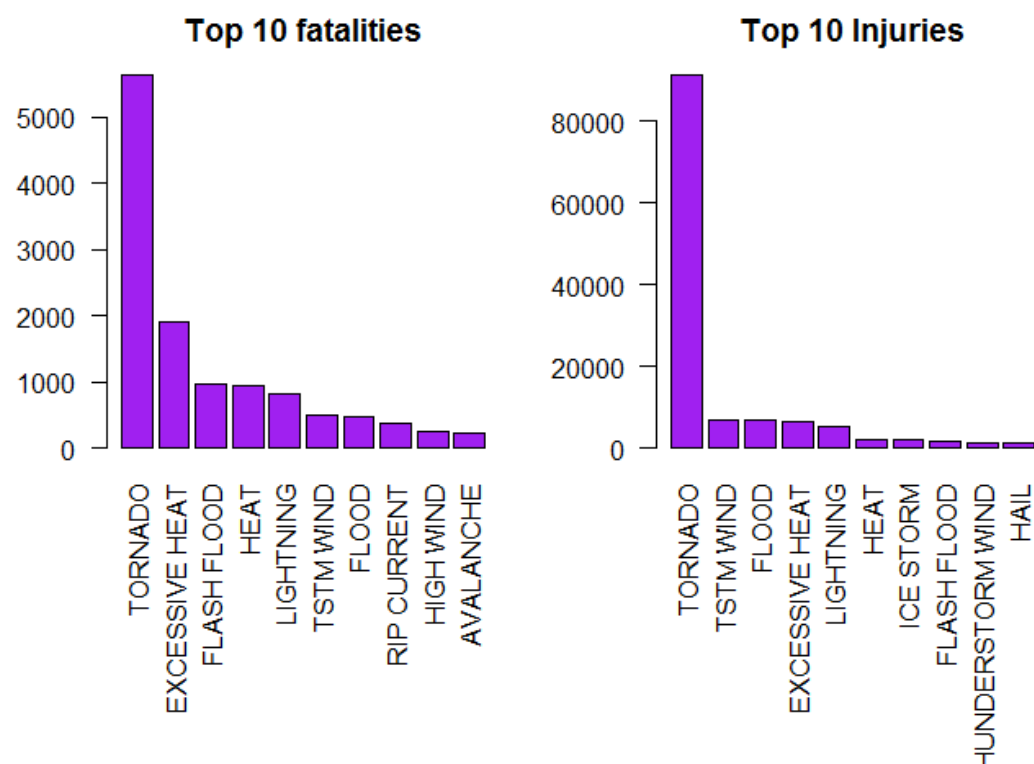


Figure 1-The weather event responsible for the highest fatalities and injuries is the ‘Tornado’



```
### QUESTION 2. Across the United States, which types of events have the greatest economic consequences?
```

```
# An analysis of the weather events responsible for the greatest economic consequences
```

```
## Hypothesis: Economic consequences means damages. The two significant types of damage typically caused by weather events include 'properties and crops'
```

```
# Q2.1 Data Exploration & Findings ...
```

```
# Upon reviewing the column names, the property damage (PROPDMG) and crop damage (CROPDMG) columns both have another related column titled 'exponents' (i.e - PROPDMGEXP and CROPDMGEXP respectively).
```

```
# As a result, let's convert the exponent columns into numeric data for the calculation of total property and crop damages encountered.
```

```
# Q2.2 Defining & Calculating [ Property Damage ]
```

```
## Property damage exponents for each level listed out & assigned those values for the property exponent data.
```

```
## Invalid data was excluded by assigning the value as '0'.
```

```
## Then, the property damage value was calculated by multiplying the property damage and property exponent value.
```

```
unique(strmdata$PROPDMGEXP)
```

```
## [1] K M B m + 0 5 6 ? 4 2 3 h 7 H - 1 8
```

```
## Levels: - ? + 0 1 2 3 4 5 6 7 8 B h H K m M
```

```
# Assigning values for the property exponent strmdata
```

```
strmdata$PROPEXP[strmdata$PROPDMGEXP == "K"] <- 1000
```

```
strmdata$PROPEXP[strmdata$PROPDMGEXP == "M"] <- 1e+06
```

```
strmdata$PROPEXP[strmdata$PROPDMGEXP == ""] <- 1
```

```
strmdata$PROPEXP[strmdata$PROPDMGEXP == "B"] <- 1e+09
```

```

strmdata$PROPEXP[strmdata$PROPDMGEXP == "m"] <- 1e+06
strmdata$PROPEXP[strmdata$PROPDMGEXP == "0"] <- 1
strmdata$PROPEXP[strmdata$PROPDMGEXP == "5"] <- 1e+05
strmdata$PROPEXP[strmdata$PROPDMGEXP == "6"] <- 1e+06
strmdata$PROPEXP[strmdata$PROPDMGEXP == "4"] <- 10000
strmdata$PROPEXP[strmdata$PROPDMGEXP == "2"] <- 100
strmdata$PROPEXP[strmdata$PROPDMGEXP == "3"] <- 1000
strmdata$PROPEXP[strmdata$PROPDMGEXP == "h"] <- 100
strmdata$PROPEXP[strmdata$PROPDMGEXP == "7"] <- 1e+07
strmdata$PROPEXP[strmdata$PROPDMGEXP == "H"] <- 100
strmdata$PROPEXP[strmdata$PROPDMGEXP == "1"] <- 10
strmdata$PROPEXP[strmdata$PROPDMGEXP == "8"] <- 1e+08

# Assigning '0' to invalid exponent strmdata
strmdata$PROPEXP[strmdata$PROPDMGEXP == "+"] <- 0
strmdata$PROPEXP[strmdata$PROPDMGEXP == "-"] <- 0
strmdata$PROPEXP[strmdata$PROPDMGEXP == "?"] <- 0

# Calculating the property damage value
strmdata$PROPDMGVAL <- strmdata$PROPDMG * strmdata$PROPEXP

# Q2.3 Defining & Calculating [ Crop Damage ]

## Crop damage exponents for each level listed out & assigned those values for the crop exponent data.
## Invalid data was excluded by assigning the value as '0'.
## Then, the crop damage value was calculated by multiplying the crop damage and crop exponent value.

unique(strmdata$CROPDMGEXP)

```

```

## [1]    M K m B ? 0 k 2
## Levels:  ? 0 2 B k K m M

```

```

# Assigning values for the crop exponent strmdata
strmdata$CROPEXP[strmdata$CROPDMGEXP == "M"] <- 1e+06
strmdata$CROPEXP[strmdata$CROPDMGEXP == "K"] <- 1000
strmdata$CROPEXP[strmdata$CROPDMGEXP == "m"] <- 1e+06
strmdata$CROPEXP[strmdata$CROPDMGEXP == "B"] <- 1e+09
strmdata$CROPEXP[strmdata$CROPDMGEXP == "0"] <- 1
strmdata$CROPEXP[strmdata$CROPDMGEXP == "k"] <- 1000
strmdata$CROPEXP[strmdata$CROPDMGEXP == "2"] <- 100
strmdata$CROPEXP[strmdata$CROPDMGEXP == ""] <- 1

# Assigning '0' to invalid exponent strmdata
strmdata$CROPEXP[strmdata$CROPDMGEXP == "?"] <- 0

# calculating the crop damage
strmdata$CROPDMGVAL <- strmdata$CROPDMG * strmdata$CROPEXP

# Q2.4 Property Damage Summary

## Procedure = aggregate the property damage by the event type and sort
the output it in descending order

prop <- aggregate(PROPDMGVAL~EVTYPE,data=strmdata,FUN=sum,na.rm=TRUE)
prop <- prop[with(prop,order(-PROPDMGVAL)),]
prop <- head(prop,10)
print(prop)

```

```

##           EVTYPE      PROPDMGVAL
## 170           FLOOD 144657709807
## 411 HURRICANE/TYPHOON 69305840000
## 834           TORNADO 56947380617
## 670          STORM SURGE 43323536000
## 153          FLASH FLOOD 16822673979
## 244              HAIL 15735267513
## 402          HURRICANE 11868319010
## 848    TROPICAL STORM 7703890550

```

## 972	WINTER STORM	6688497251
## 359	HIGH WIND	5270046260

```
# Q2.5 Crop Damage Summary
```

```
## Procedure = aggregate the crop damage by the event type and sort the
output it in descending order
```

```
crop <- aggregate(CROPDMGVAL~EVTYPE,data=strmdata,FUN=sum,na.rm=TRUE)
crop <- crop[with(crop,order(-CROPDMGVAL)),]
crop <- head(crop,10)
print(crop)
```

##	EVTYPE	CROPDMGVAL
## 95	DROUGHT	13972566000
## 170	FLOOD	5661968450
## 590	RIVER FLOOD	5029459000
## 427	ICE STORM	5022113500
## 244	HAIL	3025954473
## 402	HURRICANE	2741910000
## 411	HURRICANE/TYPHOON	2607872800
## 153	FLASH FLOOD	1421317100
## 140	EXTREME COLD	1292973000
## 212	FROST/FREEZE	1094086000

```
# Q2.6 Plot of Top 10 Property & Crop damages by Weather Event Types ( E
conomic Consequences )
```

```
##plot the graph showing the top 10 property and crop damages
```

```
par(mfrow=c(1,2),mar=c(11,3,3,2))
```

```
barplot(prop$PROPDMGVAL/(10^9),names.arg=prop$EVTYPE,las=2,col="gold",yl
ab="Prop.damage (billions)",main="Top10 Prop.Damages")
```

```
barplot(crop$CROPDMGVAL/(10^9),names.arg=crop$EVTYPE,las=2,col="gold",ylab="Crop damage (billions)",main="Top10 Crop.Damages")
```

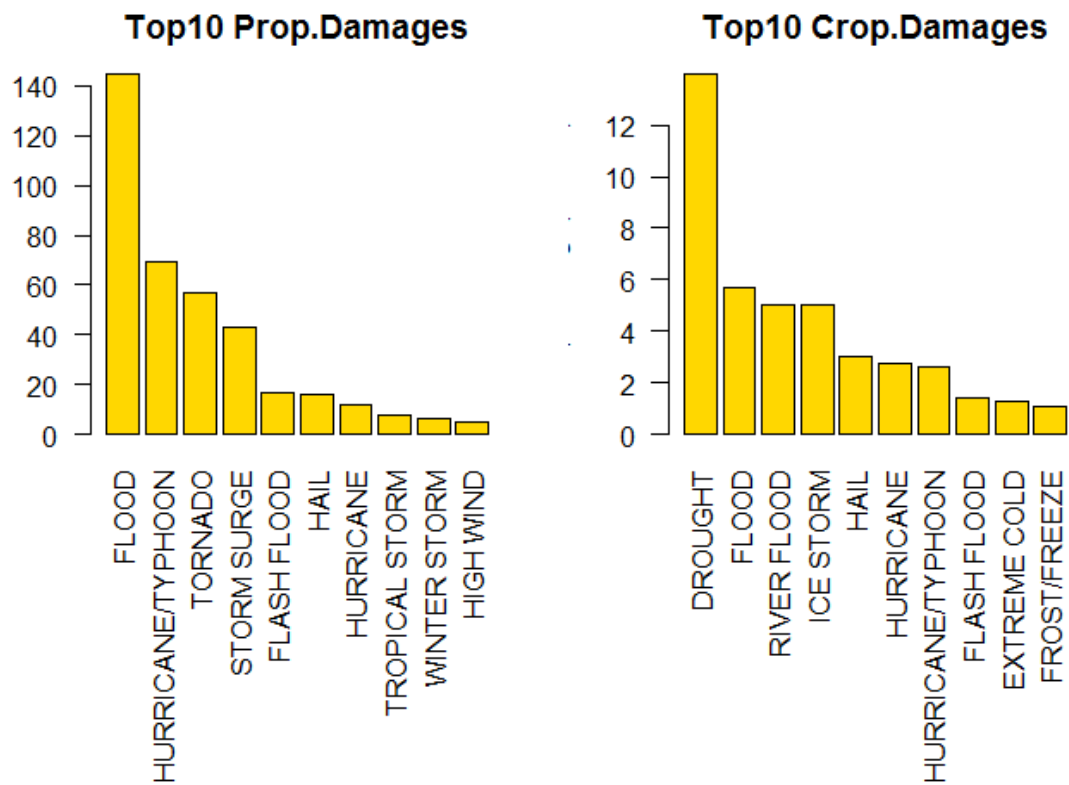


Figure 2-‘Floods’ are responsible for the highest property damage while ‘droughts’ cause the greatest crop damage.

## Summary of Conclusions

- Tornadoes are responsible for the maximum number of fatalities and injuries, followed by Excessive Heat for fatalities and Thunderstorm wind for injuries.
- Floods are responsible for maximum property damage, while Droughts cause maximum crop damage. Second major events that caused the maximum damage was Hurricanes/Typhoons for property damage and Floods for crop damage.