

# US Weather Event Data Analysis

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## Human Health and Property Damage of Weather Event Types Across the United States

### Synopsis

Weather events can cause significant human harm and economic damage. Health effects and economic cost were studied by weather event type using data from the NOAA Storm Database.

### Data Processing

The database is downloaded from the source, extracted and loaded into R as a data object 'rawdata':

```
url <- "https://d396qusza40orc.cloudfront.net/repdata%2Fdata%2FStormData.csv.bz2"
download.file(url, destfile = "data.csv.bz2", method="curl")
rawdata <- read.csv("data.csv.bz2", sep=";", header=TRUE, na.strings = c("NA","", '#DIV/0!'))
```

Quick inspection of data file to ensure the above worked correctly and get an idea of what the data looks like:

```
dim(rawdata)
```

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

```
head(rawdata)
```

```
## STATE__ BGN_DATE BGN_TIME TIME_ZONE COUNTY COUNTYNAM 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
## 4 1 6/8/1951 0:00:00 0900 CST 89 MADISON AL
## 5 1 11/15/1951 0:00:00 1500 CST 43 CULLMAN AL
## 6 1 11/15/1951 0:00:00 2000 CST 77 LAUDERDALE AL
## EVTYPE BGN_RANGE BGN_AZI BGN_LOCATI END_DATE END_TIME COUNTY_END
## 1 TORNADO 0 <NA> <NA> <NA> <NA> 0
## 2 TORNADO 0 <NA> <NA> <NA> <NA> 0
## 3 TORNADO 0 <NA> <NA> <NA> <NA> 0
## 4 TORNADO 0 <NA> <NA> <NA> <NA> 0
## 5 TORNADO 0 <NA> <NA> <NA> <NA> 0
## 6 TORNADO 0 <NA> <NA> <NA> <NA> 0
## COUNTYENDN END_RANGE END_AZI END_LOCATI LENGTH WIDTH F MAG FATALITIES
## 1 NA 0 <NA> <NA> 14.0 100 3 0 0
## 2 NA 0 <NA> <NA> 2.0 150 2 0 0
## 3 NA 0 <NA> <NA> 0.1 123 2 0 0
## 4 NA 0 <NA> <NA> 0.0 100 2 0 0
## 5 NA 0 <NA> <NA> 0.0 150 2 0 0
## 6 NA 0 <NA> <NA> 1.5 177 2 0 0
## INJURIES PROPDGM PROPDMGEXP CROPDGM CROPDMGEXP WFO STATEOFFIC ZONENAMES
## 1 15 25.0 K 0 <NA> <NA> <NA> <NA>
## 2 0 2.5 K 0 <NA> <NA> <NA> <NA>
## 3 2 25.0 K 0 <NA> <NA> <NA> <NA>
## 4 2 2.5 K 0 <NA> <NA> <NA> <NA>
## 5 2 2.5 K 0 <NA> <NA> <NA> <NA>
## 6 6 2.5 K 0 <NA> <NA> <NA> <NA>
## LATITUDE LONGITUDE LATITUDE_E LONGITUDE_ REMARKS REFNUM
## 1 3040 8812 3051 8806 <NA> 1
## 2 3042 8755 0 0 <NA> 2
## 3 3340 8742 0 0 <NA> 3
## 4 3458 8626 0 0 <NA> 4
## 5 3412 8642 0 0 <NA> 5
## 6 3450 8748 0 0 <NA> 6
```

This study looks only at population health and economic consequences of weather events by event type. Therefore only a subset of the columns are required.

```
procdata <- rawdata[, c("EVTYPE", "FATALITIES", "INJURIES", "PROPDGM", "PROPDMGEXP", "CROPDGM", "CROPDMGEXP")]
head(procdata)
```

```
## EVTYPE FATALITIES INJURIES PROPDGM PROPDMGEXP CROPDGM CROPDMGEXP
## 1 TORNADO 0 15 25.0 K 0 <NA>
## 2 TORNADO 0 0 2.5 K 0 <NA>
## 3 TORNADO 0 2 25.0 K 0 <NA>
## 4 TORNADO 0 2 2.5 K 0 <NA>
## 5 TORNADO 0 2 2.5 K 0 <NA>
## 6 TORNADO 0 6 2.5 K 0 <NA>
```

Note that both PROPDMG and CROPDMG have a multiplier column. These multipliers are codified, so they first must be converted to real numbers.

```
unique(procdata$PROPDMGEXP)
```

```
## [1] K      M      <NA> B      m      +      0      5      6      ?      4      2      3      h
## [15] 7      H      -      1      8
## Levels: - ? + 0 1 2 3 4 5 6 7 8 B h H K m M
```

```
unique(procdata$CROPDMGEXP)
```

```
## [1] <NA> M      K      m      B      ?      0      k      2
## Levels: ? 0 2 B k K m M
```

```
procdata$PDMULTIPLIER <- 0

procdata$PDMULTIPLIER[procdata$PROPDMGEXP %in% c('h','H')] <- 100
procdata$PDMULTIPLIER[procdata$PROPDMGEXP %in% c('k','K')] <- 1000
procdata$PDMULTIPLIER[procdata$PROPDMGEXP %in% c('m','M')] <- 1e+06
procdata$PDMULTIPLIER[procdata$PROPDMGEXP %in% c('b','B')] <- 1e+09
procdata$PDMULTIPLIER[procdata$PROPDMGEXP == 1] <- 1e+1
procdata$PDMULTIPLIER[procdata$PROPDMGEXP == 2] <- 1e+2
procdata$PDMULTIPLIER[procdata$PROPDMGEXP == 3] <- 1e+3
procdata$PDMULTIPLIER[procdata$PROPDMGEXP == 4] <- 1e+4
procdata$PDMULTIPLIER[procdata$PROPDMGEXP == 5] <- 1e+5
procdata$PDMULTIPLIER[procdata$PROPDMGEXP == 6] <- 1e+6
procdata$PDMULTIPLIER[procdata$PROPDMGEXP == 7] <- 1e+7
procdata$PDMULTIPLIER[procdata$PROPDMGEXP == 8] <- 1e+8
procdata$PDMULTIPLIER[is.na(procdata$PROPDMGEXP)] <- 1

procdata$PROPDAMAGEVALUE <- procdata$PROPDMG*procdata$PDMULTIPLIER

procdata$CDMULTIPLIER[procdata$CROPDMGEXP %in% c('h','H')] <- 100
procdata$CDMULTIPLIER[procdata$CROPDMGEXP %in% c('k','K')] <- 1000
procdata$CDMULTIPLIER[procdata$CROPDMGEXP %in% c('m','M')] <- 1e+06
procdata$CDMULTIPLIER[procdata$CROPDMGEXP %in% c('b','B')] <- 1e+09
procdata$CDMULTIPLIER[procdata$CROPDMGEXP == 0] <- 1
procdata$CDMULTIPLIER[procdata$CROPDMGEXP == 2] <- 1e+2
procdata$CDMULTIPLIER[is.na(procdata$CROPDMGEXP)] <- 1

procdata$CROPDAMAGEVALUE <- procdata$CROPDMG*procdata$CDMULTIPLIER
```

```
summary(procdata)
```

```

##          EVTYPE          FATALITIES          INJURIES
## HAIL          :288661  Min.   : 0.0000  Min.   : 0.0000
## TSTM WIND      :219940  1st Qu.: 0.0000  1st Qu.: 0.0000
## THUNDERSTORM WIND: 82563  Median : 0.0000  Median : 0.0000
## TORNADO        : 60652  Mean    : 0.0168  Mean    : 0.1557
## FLASH FLOOD    : 54277  3rd Qu.: 0.0000  3rd Qu.: 0.0000
## FLOOD          : 25326  Max.    :583.0000  Max.    :1700.0000
## (Other)        :170878
##  PROPDMG      PROPDMGEXP      CROPDMG      CROPDMGEXP
## Min.   : 0.00  K          :424665  Min.   : 0.000  K          :281832
## 1st Qu.: 0.00  M          : 11330  1st Qu.: 0.000  M          : 1994
## Median : 0.00  0          : 216    Median : 0.000  k          : 21
## Mean   : 12.06  B          : 40     Mean   : 1.527  0          : 19
## 3rd Qu.: 0.50  5          : 28     3rd Qu.: 0.000  B          : 9
## Max.   :5000.00 (Other): 84     Max.   :990.000 (Other): 9
##          NA's   :465934          NA's   :618413
##  PDMULTIPLIER  PROPDAMAGEVALUE  CDMULTIPLIER
## Min.   :0.000e+00  Min.   :0.000e+00  Min.   :1.00e+00
## 1st Qu.:1.000e+00  1st Qu.:0.000e+00  1st Qu.:1.00e+00
## Median :1.000e+00  Median :0.000e+00  Median :1.00e+00
## Mean   :5.754e+04  Mean   :4.746e+05  Mean   :1.25e+04
## 3rd Qu.:1.000e+03  3rd Qu.:5.000e+02  3rd Qu.:1.00e+03
## Max.   :1.000e+09  Max.   :1.150e+11  Max.   :1.00e+09
##          NA's   :8
##  CROPDAMAGEVALUE
## Min.   :0.000e+00
## 1st Qu.:0.000e+00
## Median :0.000e+00
## Mean   :5.442e+04
## 3rd Qu.:0.000e+00
## Max.   :5.000e+09
##  NA's   :8

```

## Results

### Population Harm

Population harm, in terms of both fatalities and injuries, was investigated by event type.

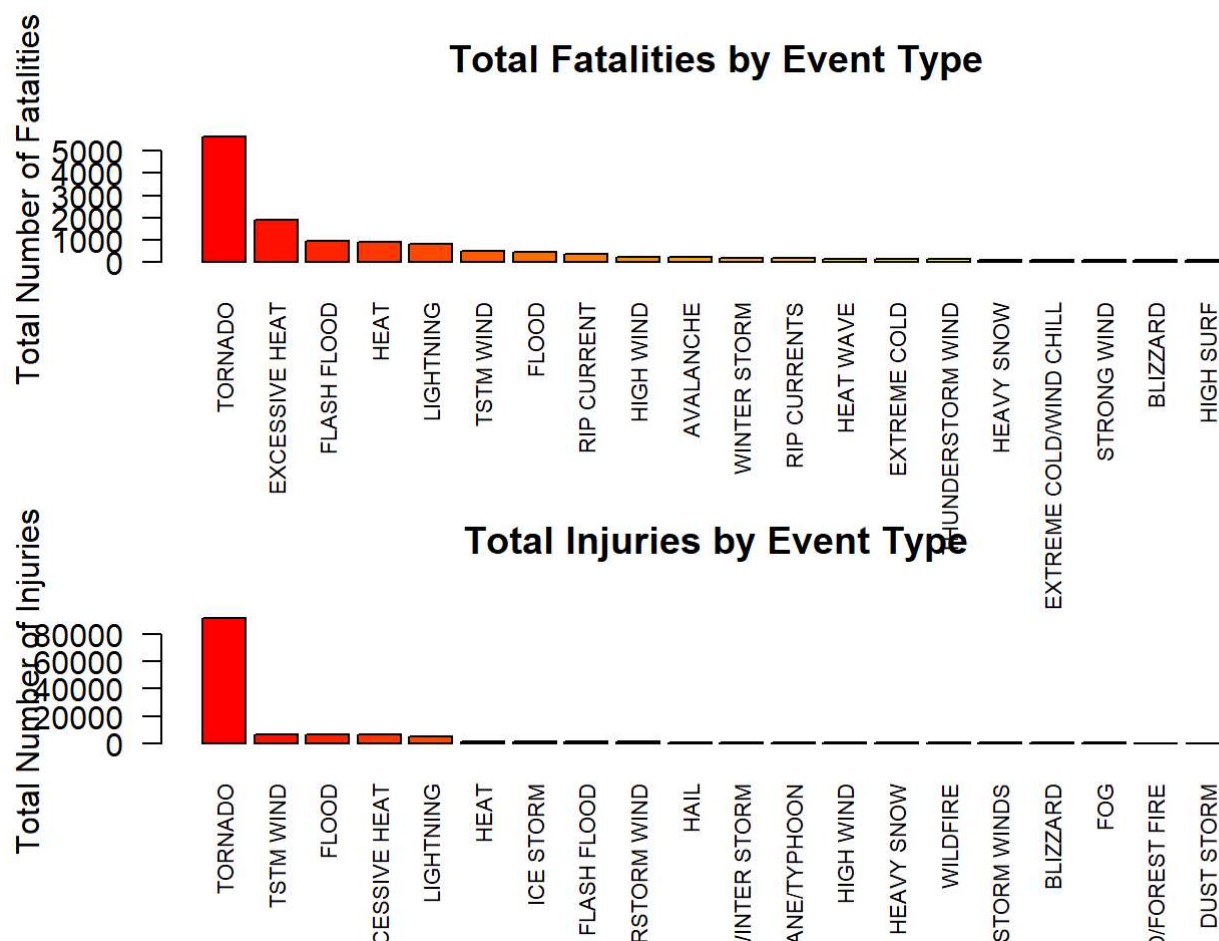
```
dt <- data.table(procdata)

totFatalities <- aggregate(FATALITIES ~ EVTYPE, data=procdata, FUN=sum)
totFatalities <- arrange(totFatalities, desc(FATALITIES))

totInjuries <- aggregate(INJURIES ~ EVTYPE, data=procdata, FUN=sum)
totInjuries <- arrange(totInjuries, desc(INJURIES))

par(mfrow=c(2,1))
barplot(totFatalities$FATALITIES[1:20], col =heat.colors(20), names.arg=totFatalities$EVTYPE[1:20],
cex.names=0.7, las=2, main="Total Fatalities by Event Type", ylab="Total Number of Fatalities")

barplot(totInjuries$INJURIES[1:20], col =heat.colors(20), names.arg=totInjuries$EVTYPE[1:20], cex.names=0.7,
las=2, main="Total Injuries by Event Type", ylab="Total Number of Injuries")
```



Tornados were responsible for both the highest number of fatalities and the highest number of injuries over the studied period.

## Economic Cost:

As for population harm, economic cost was investigated by event type.

The dataset separates economic damage into property damage ('PROPDMG') and crop damage ('CROPDMG').

# Economic Damage

```
totPropDamage <- aggregate(PROPDAMAGEVALUE ~ EVTYPE, data=procdata, FUN=sum)
totPropDamage <- arrange(totPropDamage, desc(PROPDAMAGEVALUE))

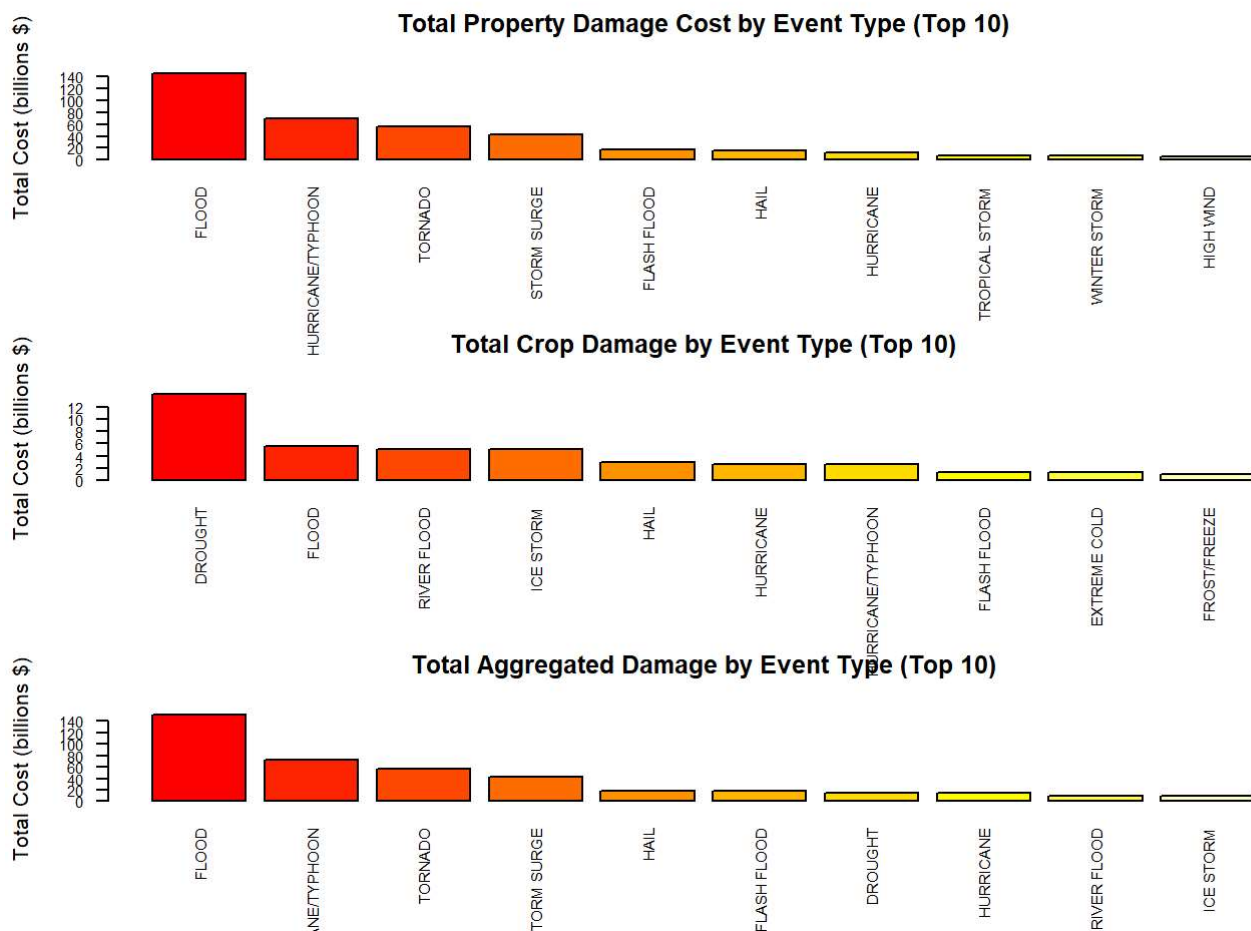
totCropDamage <- aggregate(CROPDAMAGEVALUE ~ EVTYPE, data=procdata, FUN=sum)
totCropDamage <- arrange(totCropDamage, desc(CROPDAMAGEVALUE))

totAggDamage <- aggregate(PROPDAMAGEVALUE+CROPDAMAGEVALUE ~ EVTYPE, data=procdata, FUN=sum)
names(totAggDamage)[2]="AGGDAMAGEVALUE"
totAggDamage <- arrange(totAggDamage, desc(AGGDAMAGEVALUE))

par(mfrow=c(3,1))
barplot(totPropDamage$PROPDAMAGEVALUE[1:10]/1e+9, col=heat.colors(10), names.arg=totPropDamage$EVTYPE[1:10], cex.names=0.7, cex.axis=0.7, las=2, main="Total Property Damage Cost by Event Type (Top 10)", ylab="Total Cost (billions $)")

barplot(totCropDamage$CROPDAMAGEVALUE[1:10]/1e+9, col=heat.colors(10), names.arg=totCropDamage$EVTYPE[1:10], cex.names=0.7, cex.axis=0.7, las=2, main="Total Crop Damage by Event Type (Top 10)", ylab="Total Cost (billions $)")

barplot(totAggDamage$AGGDAMAGEVALUE[1:10]/1e+9, col=heat.colors(10), names.arg=totAggDamage$EVTYPE[1:10], cex.names=0.7, cex.axis=0.7, las=2, main="Total Aggregated Damage by Event Type (Top 10)", ylab="Total Cost (billions $)")
```



It can be seen that floods produce the most property damage and cause the most economic damage overall, that drought causes the most damage to crops, and that property damage is of significantly greater cost than crop damage.