

Reproducible Research Course Project 2

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Project Title

U.S. National Oceanic and Atmospheric Administration's (NOAA) Storm Database Analysis

Synopsis

Storms and other severe weather events can cause both public health and economic problems for communities and municipalities. Many severe events can result in fatalities, injuries, and property damage, and preventing such outcomes to the extent possible is a key concern.

This project involves exploring 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, as well as estimates of any fatalities, injuries, and property damage.

Questions to Answer

- Across the United States, which types of events (as indicated in the EVTYPE variable) are most harmful with respect to population health?
- Across the United States, which types of events have the greatest economic consequences?

Data Processing

[<https://d396qusza40orc.cloudfront.net/repdata%2Fdata%2FStormData.csv.bz2>

(<https://d396qusza40orc.cloudfront.net/repdata%2Fdata%2FStormData.csv.bz2>)] (Storm Data)

[https://d396qusza40orc.cloudfront.net/repdata%2Fpeer2_doc%2Fpd01016005curr.pdf

(https://d396qusza40orc.cloudfront.net/repdata%2Fpeer2_doc%2Fpd01016005curr.pdf)] (National Weather Service Storm Data Documentation)

[https://d396qusza40orc.cloudfront.net/repdata%2Fpeer2_doc%2FNCLC%20Storm%20Events-FAQ%20Page.pdf

(https://d396qusza40orc.cloudfront.net/repdata%2Fpeer2_doc%2FNCLC%20Storm%20Events-FAQ%20Page.pdf)] (National Climatic Data Center Storm Events FAQ)

The events in the database start in the year 1950 and end in November 2011. In the earlier years of the database there are generally fewer events recorded, most likely due to a lack of good records. More recent years should be considered more complete.

Download, extract and read CSV file

```
download.file("https://d396qusza40orc.cloudfront.net/repdata%2Fdata%2FStormData.csv.bz2",  
             dest="stormdata.bz2",  
             method="curl")  
  
data <- read.csv(bzfile("stormdata.bz2"),  
                header = TRUE,  
                sep = ",",  
                stringsAsFactors = FALSE)
```

Examine data

```
str(data)
```

```
## 'data.frame':    902297 obs. of  37 variables:
## $ STATE__ : num  1 1 1 1 1 1 1 1 1 1 ...
## $ BGN_DATE : chr   "4/18/1950 0:00:00" "4/18/1950 0:00:00" "2/20/1951 0:00:00" "6/8/
1951 0:00:00" ...
## $ BGN_TIME : chr   "0130" "0145" "1600" "0900" ...
## $ TIME_ZONE : chr   "CST" "CST" "CST" "CST" ...
## $ COUNTY : num  97 3 57 89 43 77 9 123 125 57 ...
## $ COUNTYNAME: chr   "MOBILE" "BALDWIN" "FAYETTE" "MADISON" ...
## $ STATE : chr   "AL" "AL" "AL" "AL" ...
## $ EVTYPE : chr   "TORNADO" "TORNADO" "TORNADO" "TORNADO" ...
## $ BGN_RANGE : num  0 0 0 0 0 0 0 0 0 0 ...
## $ BGN_AZI : chr   "" "" "" "" ...
## $ BGN_LOCATI: chr   "" "" "" "" ...
## $ END_DATE : chr   "" "" "" "" ...
## $ END_TIME : chr   "" "" "" "" ...
## $ COUNTY_END: num  0 0 0 0 0 0 0 0 0 0 ...
## $ COUNTYENDN: logi  NA NA NA NA NA NA ...
## $ END_RANGE : num  0 0 0 0 0 0 0 0 0 0 ...
## $ END_AZI : chr   "" "" "" "" ...
## $ END_LOCATI: chr   "" "" "" "" ...
## $ LENGTH : num  14 2 0.1 0 0 1.5 1.5 0 3.3 2.3 ...
## $ WIDTH : num  100 150 123 100 150 177 33 33 100 100 ...
## $ F : int  3 2 2 2 2 2 2 1 3 3 ...
## $ MAG : num  0 0 0 0 0 0 0 0 0 0 ...
## $ FATALITIES: num  0 0 0 0 0 0 0 0 1 0 ...
## $ INJURIES : num  15 0 2 2 2 6 1 0 14 0 ...
## $ PROPDMG : num  25 2.5 25 2.5 2.5 2.5 2.5 2.5 25 25 ...
## $ PROPDMGEXP: chr   "K" "K" "K" "K" ...
## $ CROPDGMG : num  0 0 0 0 0 0 0 0 0 0 ...
## $ CROPDMGEXP: chr   "" "" "" "" ...
## $ WFO : chr   "" "" "" "" ...
## $ STATEOFFIC: chr   "" "" "" "" ...
## $ ZONENAMES : chr   "" "" "" "" ...
## $ LATITUDE : num  3040 3042 3340 3458 3412 ...
## $ LONGITUDE : num  8812 8755 8742 8626 8642 ...
## $ LATITUDE_E: num  3051 0 0 0 0 ...
## $ LONGITUDE_: num  8806 0 0 0 0 ...
## $ REMARKS : chr   "" "" "" "" ...
## $ REFNUM : num  1 2 3 4 5 6 7 8 9 10 ...
```

```
summary(data)
```

```

##      STATE__      BGN_DATE      BGN_TIME      TIME_ZONE
## Min.      : 1.0    Length:902297    Length:902297    Length:902297
## 1st Qu.:19.0    Class :character    Class :character    Class :character
## Median :30.0    Mode  :character    Mode  :character    Mode  :character
## Mean      :31.2
## 3rd Qu.:45.0
## Max.      :95.0
##
##      COUNTY      COUNTYNAME      STATE      EVTYPE
## Min.      : 0.0    Length:902297    Length:902297    Length:902297
## 1st Qu.: 31.0    Class :character    Class :character    Class :character
## Median : 75.0    Mode  :character    Mode  :character    Mode  :character
## Mean      :100.6
## 3rd Qu.:131.0
## Max.      :873.0
##
##      BGN_RANGE      BGN_AZI      BGN_LOCATI
## Min.      : 0.000    Length:902297    Length:902297
## 1st Qu.: 0.000    Class :character    Class :character
## Median : 0.000    Mode  :character    Mode  :character
## Mean      : 1.484
## 3rd Qu.: 1.000
## Max.      :3749.000
##
##      END_DATE      END_TIME      COUNTY_END COUNTYENDN
## Length:902297    Length:902297    Min.      :0    Mode:logical
## Class :character    Class :character    1st Qu.:0    NA's:902297
## Mode  :character    Mode  :character    Median :0
##                               Mean      :0
##                               3rd Qu.:0
##                               Max.      :0
##
##      END_RANGE      END_AZI      END_LOCATI
## Min.      : 0.0000    Length:902297    Length:902297
## 1st Qu.: 0.0000    Class :character    Class :character
## Median : 0.0000    Mode  :character    Mode  :character
## Mean      : 0.9862
## 3rd Qu.: 0.0000
## Max.      :925.0000
##
##      LENGTH      WIDTH      F      MAG
## Min.      : 0.0000    Min.      : 0.000    Min.      :0.0    Min.      : 0.0
## 1st Qu.: 0.0000    1st Qu.: 0.000    1st Qu.:0.0    1st Qu.: 0.0
## Median : 0.0000    Median : 0.000    Median :1.0    Median : 50.0
## Mean      : 0.2301    Mean      : 7.503    Mean      :0.9    Mean      : 46.9
## 3rd Qu.: 0.0000    3rd Qu.: 0.000    3rd Qu.:1.0    3rd Qu.: 75.0
## Max.      :2315.0000    Max.      :4400.000    Max.      :5.0    Max.      :22000.0
##                               NA's      :843563
##      FATALITIES      INJURIES      PROPDMG
## Min.      : 0.0000    Min.      : 0.0000    Min.      : 0.00
## 1st Qu.: 0.0000    1st Qu.: 0.0000    1st Qu.: 0.00
## Median : 0.0000    Median : 0.0000    Median : 0.00
## Mean      : 0.0168    Mean      : 0.1557    Mean      : 12.06

```

```
## 3rd Qu.: 0.0000 3rd Qu.: 0.0000 3rd Qu.: 0.50
## Max. :583.0000 Max. :1700.0000 Max. :5000.00
##
## PROPDMGEXP      CROPDMG      CROPDMGEXP
## Length:902297   Min. : 0.000   Length:902297
## Class :character 1st Qu.: 0.000   Class :character
## Mode :character  Median : 0.000   Mode :character
##                  Mean  : 1.527
##                  3rd Qu.: 0.000
##                  Max. : 990.000
##
## WFO      STATEOFFIC      ZONENAMES      LATITUDE
## Length:902297 Length:902297 Length:902297 Min. : 0
## Class :character Class :character Class :character 1st Qu.:2802
## Mode :character Mode :character Mode :character Median :3540
##                  Mean :2875
##                  3rd Qu.:4019
##                  Max. :9706
##                  NA's :47
## LONGITUDE      LATITUDE_E      LONGITUDE_      REMARKS
## Min. : -14451 Min. : 0 Min. : -14455 Length:902297
## 1st Qu.: 7247 1st Qu.: 0 1st Qu.: 0 Class :character
## Median : 8707 Median : 0 Median : 0 Mode :character
## Mean : 6940 Mean :1452 Mean : 3509
## 3rd Qu.: 9605 3rd Qu.:3549 3rd Qu.: 8735
## Max. : 17124 Max. :9706 Max. :106220
##                  NA's :40
## REFNUM
## Min. : 1
## 1st Qu.:225575
## Median :451149
## Mean :451149
## 3rd Qu.:676723
## Max. :902297
##
```

View data

```
#View(data)
```

Lowercase variable names

```
colnames(data) <- tolower(colnames(data))
```

Change to factor for event types

```
data$evtype <- as.factor(data$evtype)
```

Change dates to POSIXct

```
data$bgn_date <- as.Date(data$bgn_date, "%m/%d/%Y")
data$end_date <- as.Date(data$end_date, "%m/%d/%Y")
```

Subset data to records that have fatalities, injuries related only

```
fData <- subset(data, subset = data$fatalities > 0)
iData <- subset(data, subset = data$injuries > 0)
phyData <- rbind(fData, iData)
```

Subset data to records that have property, crop damage related only

```
pData <- subset(data, subset = data$propdmg > 0)
cData <- subset(data, subset = data$cropdmg > 0)
ecoData <- rbind(pData, cData)
```

Subset data to records that have fatalities, injuries, property and crop damage

```
data <- rbind(phyData, ecoData)
```

View data

```
#View(data)
```

Upper property and crop exponent multipliers Map multipliers to numeric values

```

data$propdmgexp <- toupper(data$propdmgexp)
data$cropdmgexp <- toupper(data$cropdmgexp)

pDmgExp <- c("\\" = 10^0,
             "-" = 10^0,
             "+" = 10^0,
             "0" = 10^0,
             "1" = 10^1,
             "2" = 10^2,
             "3" = 10^3,
             "4" = 10^4,
             "5" = 10^5,
             "6" = 10^6,
             "7" = 10^7,
             "8" = 10^8,
             "9" = 10^9,
             "H" = 10^2,
             "K" = 10^3,
             "M" = 10^6,
             "B" = 10^9)

data$propdmgexp <- pDmgExp[as.character(data$propdmgexp)]
data$propdmgexp[is.na(data$propdmgexp)] <- 10^0

cDmgExp <- c("\\" = 10^0,
             "?" = 10^0,
             "0" = 10^0,
             "K" = 10^3,
             "M" = 10^6,
             "B" = 10^9)

data$cropdmgexp <- cDmgExp[as.character(data$cropdmgexp)]
data$cropdmgexp[is.na(data$cropdmgexp)] <- 10^0

```

Aggregate data for physical data, calculating totals

```

aggregatedPhysicalData <- aggregate(cbind(fatalities, injuries) ~ evtype, data = data, FUN = sum)
aggregatedPhysicalData$total <- aggregatedPhysicalData$fatalities + aggregatedPhysicalData$injuries

```

Determine events with the highest physical impact, by aggregating data

```
# aggregated data greater than 0
aggregatedPhysicalData <- aggregatedPhysicalData[aggregatedPhysicalData$total > 0, ]
# sort in descending order
aggregatedPhysicalData <- aggregatedPhysicalData[order(aggregatedPhysicalData$total, decreasing = TRUE), ]

# renumber rows in descending order
rownames(aggregatedPhysicalData) <- 1:nrow(aggregatedPhysicalData)

# top 10
aggregatedPhysicalDataTop10 <- aggregatedPhysicalData[1:10, ]
```

Calculate property and crop loss based on exponent multipliers

```
data$propertyLoss <- data$propdmg * data$propdmgexp
data$cropLoss <- data$cropdmg * data$cropdmgexp
```

Aggregate data for economic data, calculating totals

```
aggregatedEconomicData <- aggregate(cbind(propertyLoss, cropLoss) ~ evtype, data = data, FUN = sum)
# create total field
aggregatedEconomicData$total <- aggregatedEconomicData$propertyLoss + aggregatedEconomicData$cropLoss
```

Determine events with the highest economic impact

```
# aggregated data greater than 0
aggregatedEconomicData <- aggregatedEconomicData[aggregatedEconomicData$total > 0, ]
# sort in descending order
aggregatedEconomicData <- aggregatedEconomicData[order(aggregatedEconomicData$total, decreasing = TRUE), ]

# renumber rows in descending order
rownames(aggregatedEconomicData) <- 1:nrow(aggregatedEconomicData)

# top 10
aggregatedEconomicDataTop10 <- aggregatedEconomicData[1:10, ]
```

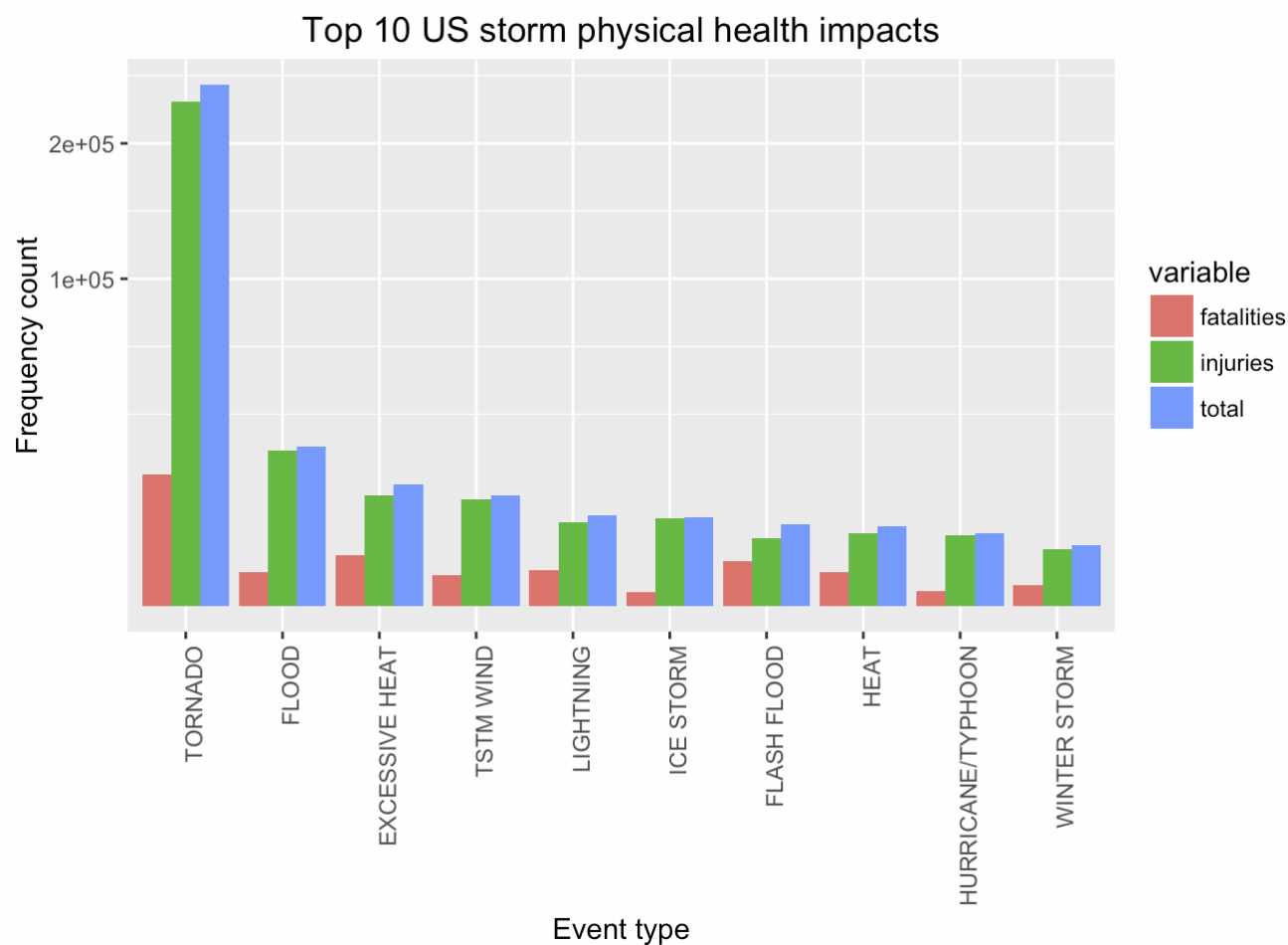
Results

Plot of the top ten event types with the highest fatality and injury counts


```
# melt id variables evtype
physicalDataTop10Melt <- melt(aggregatedPhysicalDataTop10, id.vars = "evtype")

# build ggplot of top 10
physicalChart <- ggplot(physicalDataTop10Melt, aes(x = reorder(evtype, -value), y = value)) +
  geom_bar(stat = "identity", aes(fill = variable), position = "dodge") +
  scale_y_sqrt("Frequency count") +
  xlab("Event type") +
  theme(axis.text.x = element_text(angle = 90, hjust = 1)) +
  ggtitle("Top 10 US storm physical health impacts")

print(physicalChart)
```

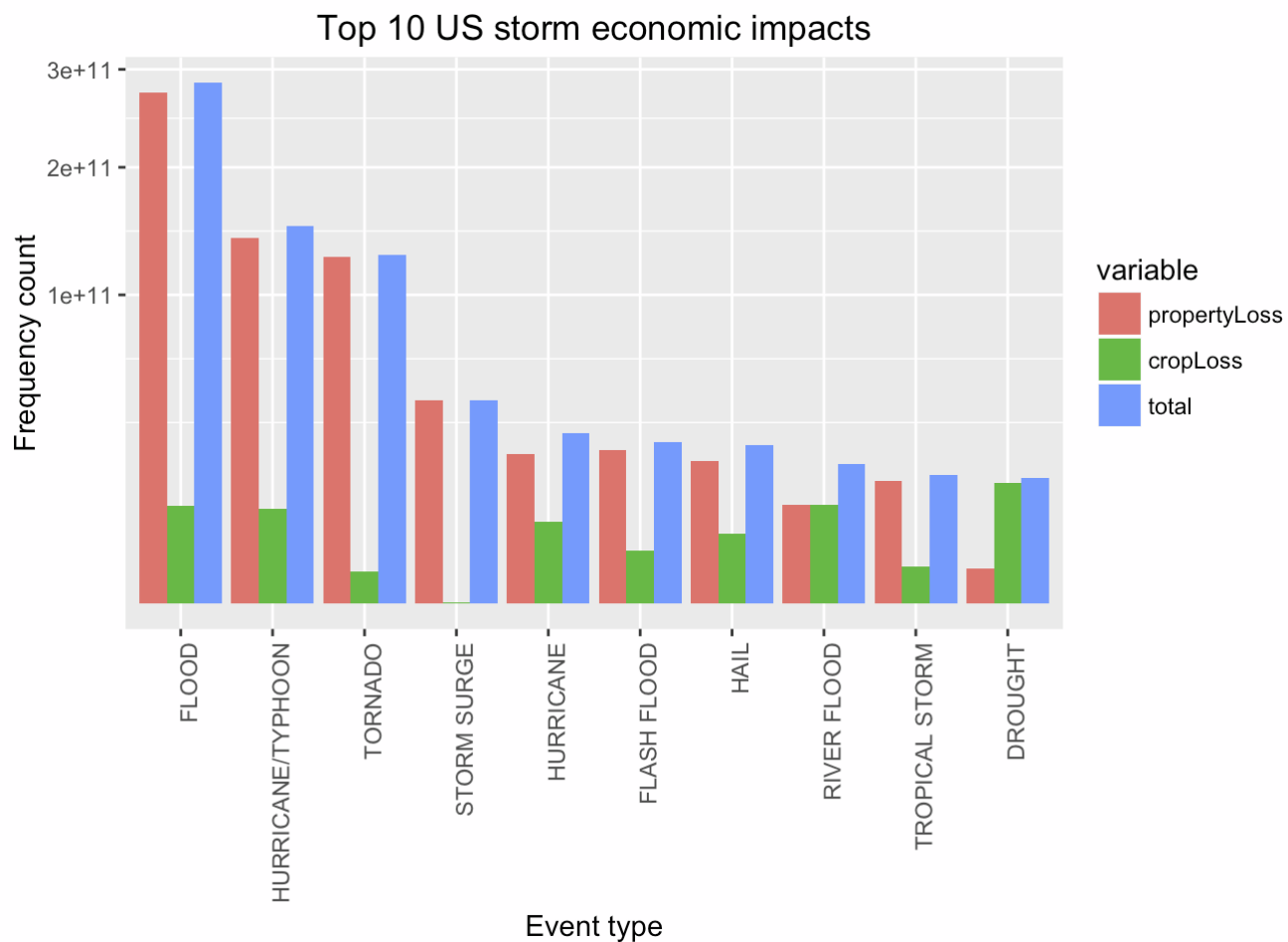


Plot of the top ten event types with the highest property and crop counts

```
# melt id variables evtype
economicDataTop10Melt <- melt(aggregatedEconomicDataTop10, id.vars = "evtype")

# build ggplot of top 10
economicChart <- ggplot(economicDataTop10Melt, aes(x = reorder(evtype, -value), y = value)) +
  geom_bar(stat = "identity", aes(fill = variable), position = "dodge") +
  scale_y_sqrt("Frequency count") +
  xlab("Event type") +
  theme(axis.text.x = element_text(angle = 90, hjust = 1)) +
  ggtitle("Top 10 US storm economic impacts")

print(economicChart)
```



Answers

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

Based on the results, tornados are the most harmful with to population health, fatalities and injuries.

Across the United States, which types of events have the greatest economic consequences?

Based on the results, floods have the greatest economic consequences.

References

Reference:

<http://rpubs.com/cneiderer/rrCourseProject2>

<https://rpubs.com/withgemini/25349>

<https://github.com/paul-reiners/reproducible-research-project-2/blob/master/StormReport.Rmd>

<http://54.225.166.221/Drarreg/rrcourseproject2>