

# Course Project 2

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## Analysing the NOAA storm database - A summary of health and economic impacts in the United States from 1950 - 2011.

Severe weather events cause major issues for public health and have acute economic effects. Such severe events result in fatalities, injuries, and generate massive property damages.

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 from the years 1950 to 2011, including when and where they took place with estimates of any fatalities, injuries, and property damage that occurred.

The data analysis addresses the following questions:

- Across the United States, which types of events (indicated as *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

### Loading libraries, importing dataset

The libraries used for this assessment are:

```
library(dplyr)
```

```
##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:stats':
##
##   filter, lag

## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union
```

```
library(ggplot2)
```

The dataset can be downloaded as a comma-separated-value file 47Mb compressed by the *bzip2* algorithm.

Documentation of the dataset are available from:

- National Weather Service Storm Data Documentation
- 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.

The following steps load the dataset:

```
#Load data
path <- getwd()
download.file(url = "https://d396qusza40orc.cloudfront.net/repdata%2Fdata%2FStormData.csv.bz2", destf

dir() #Check if data present

## [1] "Courseproject2.R"          "README.md"
## [3] "RepData_PeerAssessment2.Rproj" "StormData.csv.bz2"
## [5] "StormData.html"           "StormData.Rmd"

#Import dataset
StormData <- read.csv(bzfile("StormData.csv.bz2"), header = TRUE)

#Checking dataset
str(StormData)

## '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" ...
## $ CROPDMG      : 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 ...
```

The dataset consists of **37 columns** (variables) and **902,297 rows** (records).

## Extracting variables of interest for analysis of weather impact on health and economy

From the list of variables in *StormData*, the following columns of interest were selected:

### Health variables:

- FATALITIES: approx. number of deaths
- INJURIES: approx. number of injuries

### Economic variables:

- PROPDMG: approx. property damages
- PROPDMGEXP: the units for property damage value
- CROPDMG: approx. crop damages
- CROPDMGEXP: the units for crop damage value

### Events - target variable:

- EVTYPE: weather event (Tornados, Wind, Snow, Flood, etc..)

### Health impact

To evaluate the impact on health, the total fatalities and the total injuries for each event type (EVTYPE) are calculated. The codes for this calculation are shown as follows.

```
#Fatalities
storm.fatalities <- StormData %>% select(EVTYPE, FATALITIES) %>%
  group_by(EVTYPE) %>%
  summarise(total = sum(FATALITIES)) %>%
  arrange(-total)

## `summarise()` ungrouping output (override with `.groups` argument)
head(storm.fatalities, 10)

## # A tibble: 10 x 2
##   EVTYPE      total
##   <chr>      <dbl>
## 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
```

```
#Injuries
storm.injuries <- StormData %>% select(EVTYPE, INJURIES) %>%
  group_by(EVTYPE) %>%
```

```
summarise(total = sum(INJURIES)) %>%
arrange(-total)
```

```
## `summarise()` ungrouping output (override with `.groups` argument)
```

```
head(storm.injuries, 10)
```

```
## # A tibble: 10 x 2
##   EVTYPE      total
##   <chr>      <dbl>
## 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
```

## Economic impact

To evaluate the impact on the economy, the property and the crop damages for each event type (EVTYPE) are calculated. The codes for this calculation are shown as follows:

```
storm.damage <- StormData %>% select(EVTYPE, PROPDMG,PROPDMGEXP,CROPDMG,CROPDMGEXP)
head(storm.damage)
```

```
##   EVTYPE PROPDMG PROPDMGEXP CROPDMG CROPDMGEXP
## 1 TORNADO    25.0          K      0
## 2 TORNADO     2.5          K      0
## 3 TORNADO    25.0          K      0
## 4 TORNADO     2.5          K      0
## 5 TORNADO     2.5          K      0
## 6 TORNADO     2.5          K      0
```

Assessing the parameters for PROPDMGEXP and CROPDMGEXP:

```
sort(table(storm.damage$PROPDMGEXP), decreasing = TRUE)[1:10]
```

```
##
##           K           M           0           B           5           1           2           ?           m
## 465934 424665 11330    216    40    28    25    13    8    7
```

```
sort(table(storm.damage$CROPDMGEXP), decreasing = TRUE)[1:10]
```

```
##
##           K           M           k           0           B           ?           2           m   <NA>
## 618413 281832 1994    21    19    9    7    1    1
```

The dataset contains of two types of economic impact; the property damage (PROPDMG) and crop damage (CROPDMG). The actual damage in USD is indicated by PROPDMGEXP and CROPDMGEXP parameters. Based on information from this source, the parameters in the PROPDMGEXP and CROPDMGEXP columns are interpreted as follows:

- blank -> \*0
- (-) -> \*0

- (?) -> \*0
- (+) -> \*1
- H, h -> hundreds (\*100)
- K, k -> kilos (\*1,000)
- M, m -> millions (\*1,000,000)
- B, b -> billions (\*1,000,000,000)

The total damage caused by each event type is calculated by the following codes:

```
#Transfrom character values to numeric values in PROPDMGEXP & CROPDMGEXP
storm.damage <- StormData %>% select(EVTYPE, PROPDMG, PROPDMGEXP, CROPDMG, CROPDMGEXP)

symbols <- sort(unique(as.character(storm.damage$PROPDMGEXP)))
symbols

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

#Set a multiplier per symbol in the order of the symbols shown
multiplier <- c(0, 0, 0, 1, 10, 10, 10, 10, 10, 10, 10, 10, 10, 10, 10^9, 10^2, 10^2, 10^3, 10^6, 10^6)

#Create dataset for symbol values
storm.convert <- data.frame(symbols, multiplier)
storm.convert

##      symbols multiplier
## 1          0e+00
## 2          -      0e+00
## 3          ?      0e+00
## 4          +      1e+00
## 5          0      1e+01
## 6          1      1e+01
## 7          2      1e+01
## 8          3      1e+01
## 9          4      1e+01
## 10         5      1e+01
## 11         6      1e+01
## 12         7      1e+01
## 13         8      1e+01
## 14         B      1e+09
## 15         h      1e+02
## 16         H      1e+02
## 17         K      1e+03
## 18         m      1e+06
## 19         M      1e+06

#Fuse symbols with numeric values
storm.damage$property.multiplier <- storm.convert$multiplier[match(storm.damage$PROPDMGEXP,
                                                                    storm.convert$symbols)]
storm.damage$crop.multiplier <- storm.convert$multiplier[match(storm.damage$CROPDMGEXP,
                                                                storm.convert$symbols)]

#Property damages
storm.property <- storm.damage %>%
  mutate(DMG = PROPDMG*property.multiplier) %>%
  select(EVTYPE, DMG) %>%
  arrange(-DMG)
```

```
head(storm.property)
```

```
##           EVTYPE           DMG
## 1           FLOOD 1.150e+11
## 2      STORM SURGE 3.130e+10
## 3 HURRICANE/TYPHOON 1.693e+10
## 4      STORM SURGE 1.126e+10
## 5 HURRICANE/TYPHOON 1.000e+10
## 6 HURRICANE/TYPHOON 7.350e+09
```

```
#Crop damages
```

```
storm.crop <- storm.damage %>%
  mutate(DMG = CROPDMG*crop.multiplier) %>%
  select(EVTYPE, DMG) %>%
  arrange(-DMG)
```

```
head(storm.crop)
```

```
##           EVTYPE           DMG
## 1    RIVER FLOOD 5000000000
## 2      ICE STORM 5000000000
## 3 HURRICANE/TYPHOON 1510000000
## 4      DROUGHT 1000000000
## 5    EXTREME COLD 596000000
## 6      DROUGHT 578850000
```

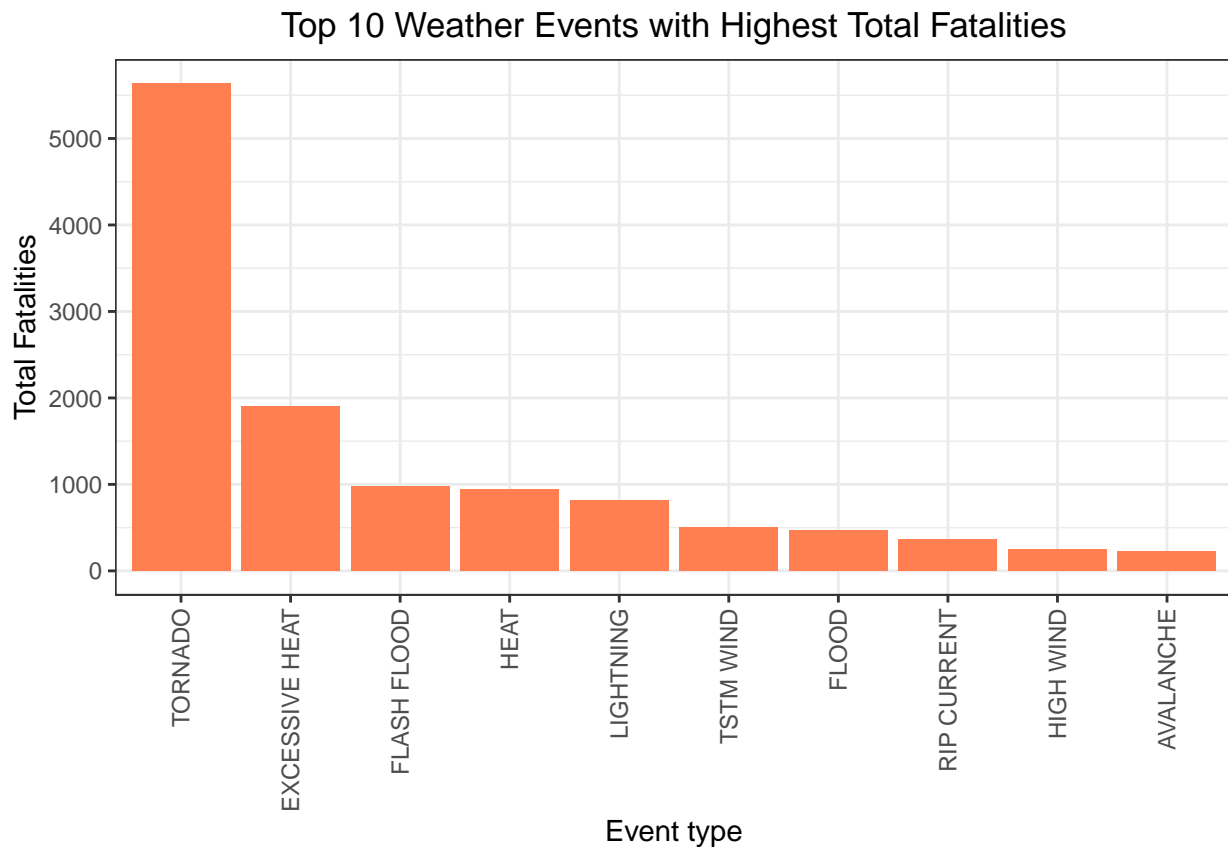
## Results

### Health impact

The top 10 weather events that caused the highest fatalities and injuries are presented individually and as a combined dataset:

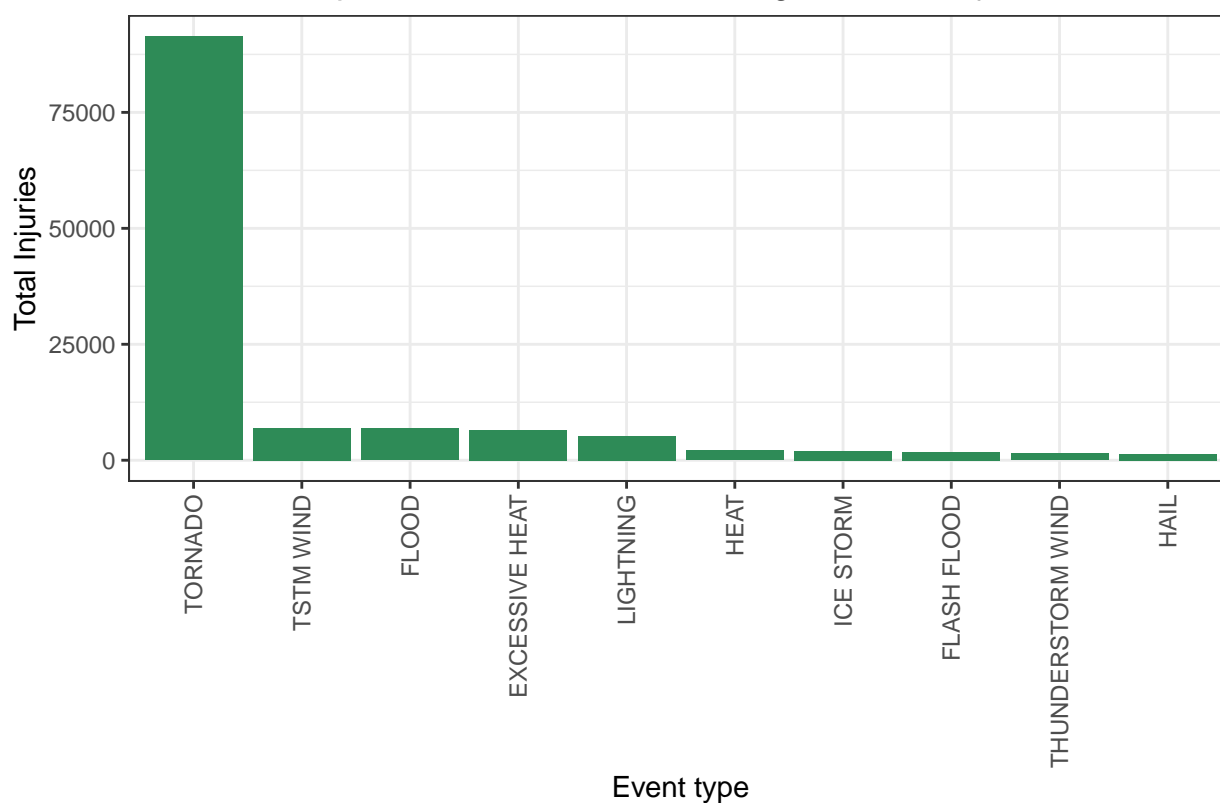
```
#Fatalities
```

```
ggplot(storm.fatalities[1:10,], aes(x = reorder(EVTYPE, -total), y = total)) +
  geom_bar(stat="identity", fill = "coral") +
  scale_y_continuous(breaks=seq(0,6000,1000)) +
  ggtitle("Top 10 Weather Events with Highest Total Fatalities")+
  labs(x="Event type", y="Total Fatalities")+
  theme_bw() +
  theme(plot.title = element_text(hjust = 0.5)) +
  theme(axis.text.x = element_text(angle=90, vjust=0.5, hjust=1))
```



```
#Injuries
ggplot(storm.injuries[1:10,], aes(x = reorder(EVTYPE, -total), y = total)) +
  geom_bar(stat="identity", fill = "seagreen") +
  scale_y_continuous(breaks=seq(0,100000,25000)) +
  ggtitle("Top 10 Weather Events with Highest Total Injuries")+
  labs(x="Event type", y="Total Injuries") +
  theme_bw() +
  theme(plot.title = element_text(hjust = 0.5)) +
  theme(axis.text.x = element_text(angle=90, vjust=0.5, hjust=1))
```

Top 10 Weather Events with Highest Total Injuries



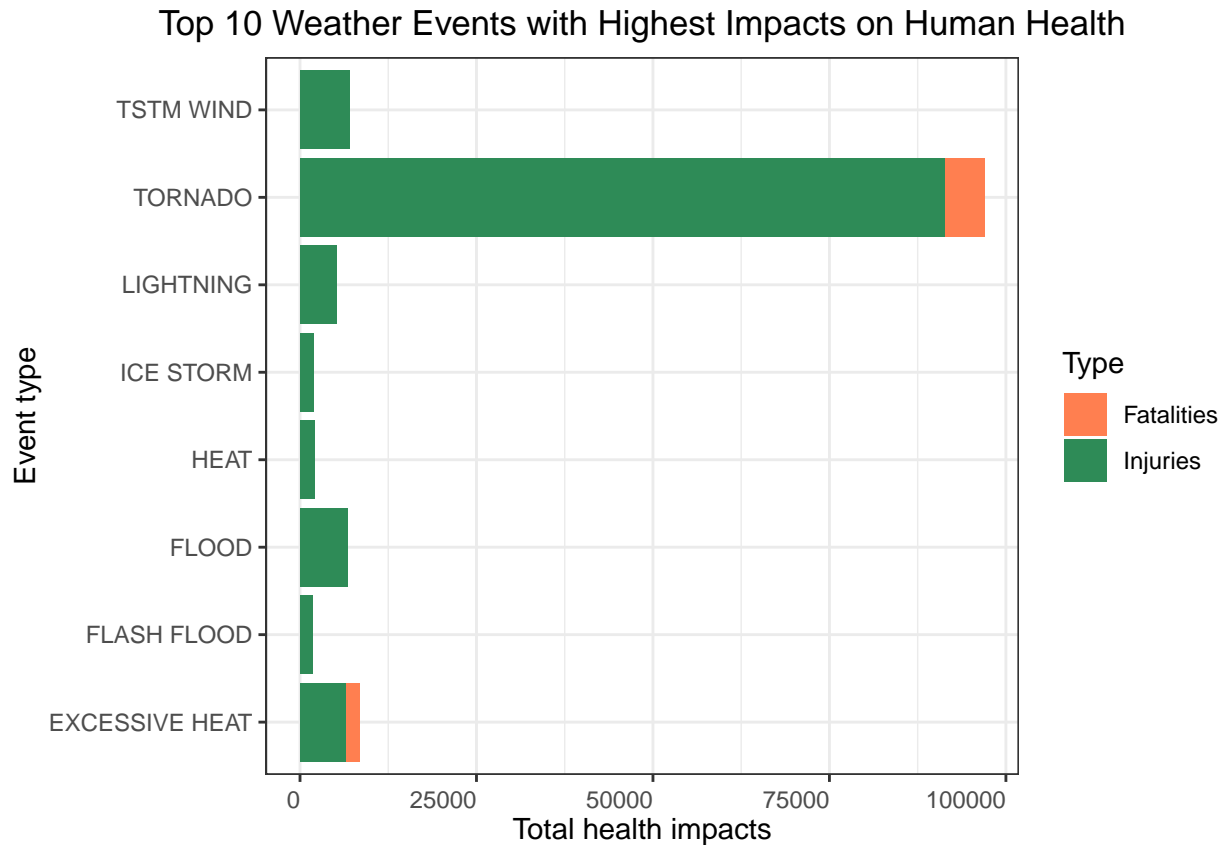
```
#Combine fatalities & injuries
storm.fatalities$Type <- "Fatalities"
storm.injuries$Type <- "Injuries"

storm.health <- rbind(storm.injuries, storm.fatalities)

storm.health <- storm.health %>%
  arrange(-total, EVTYPE)

ggplot(storm.health[1:10,], aes(x = EVTYPE, y = total, fill = Type)) +
  coord_flip() +
  geom_bar(stat="identity") +
  scale_fill_manual(values = c("coral", "seagreen")) +
  ggtitle("Top 10 Weather Events with Highest Impacts on Human Health") +
  labs(x="Event type", y="Total health impacts") +
  theme_bw() +
  theme(plot.title = element_text(hjust = 0.5)) +
  theme(axis.text.x = element_text(angle=0, vjust=0.5, hjust=1))
```



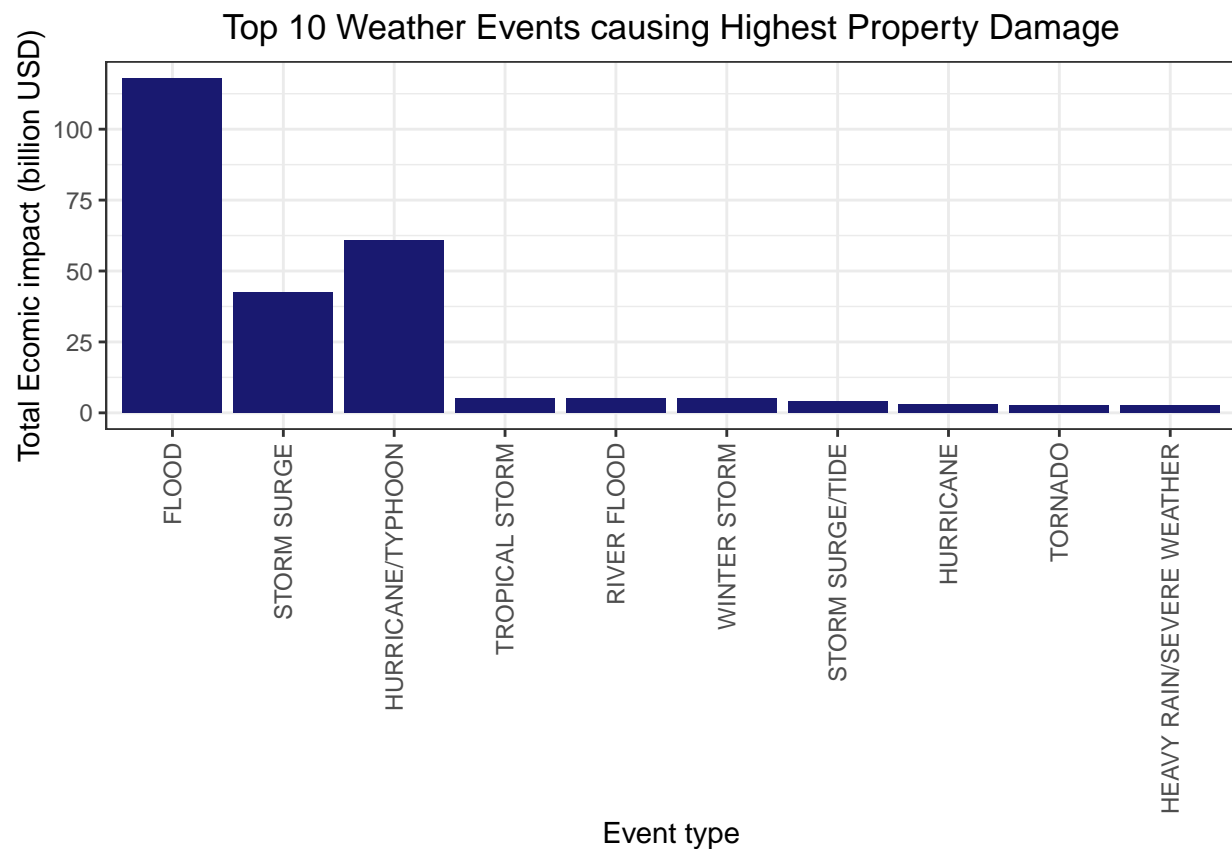


The figures show that most **Fatalities** and **Injuries** were caused by **Tornadoes**.

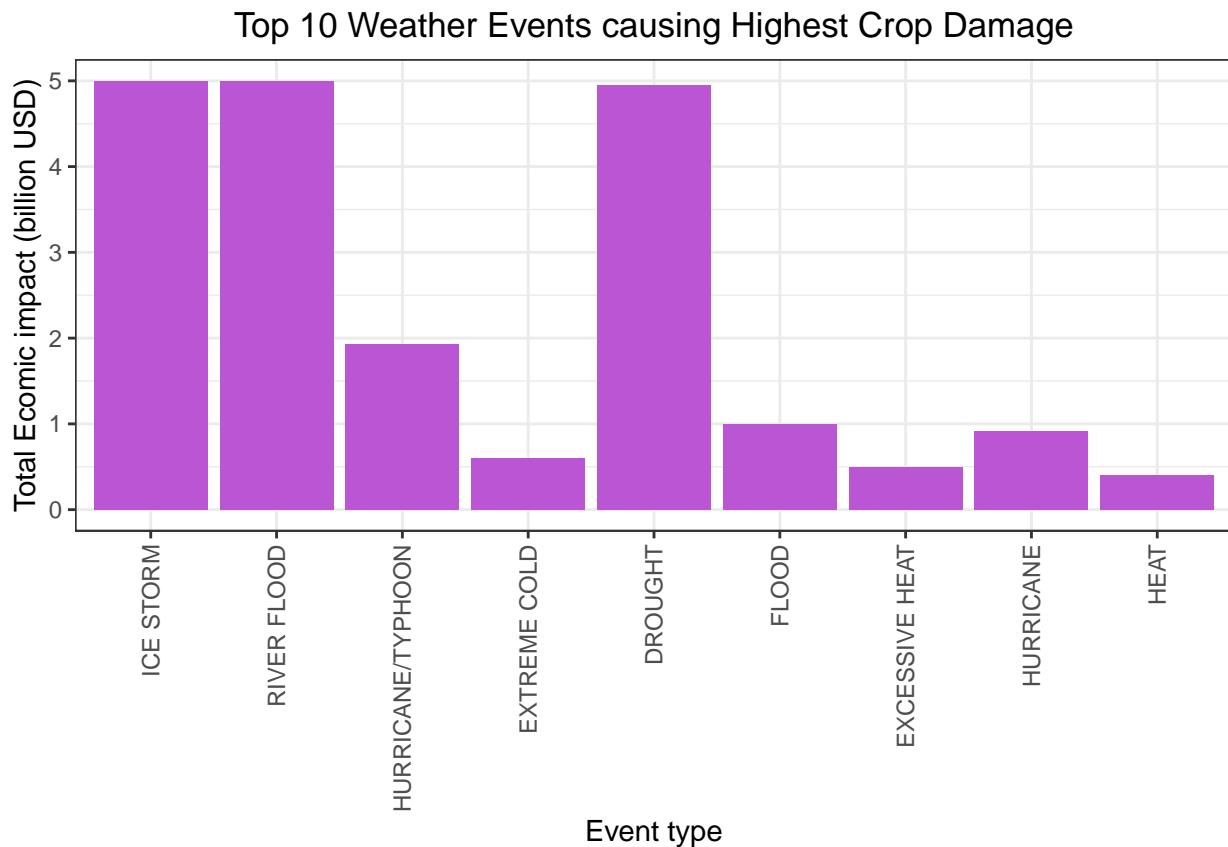
## Economic impact

The top 10 weather events that caused the highest economic damages on properties and crops are presented individually and as a combined dataset:

```
#Property
ggplot(storm.property[1:20,], aes(x=reorder(EVTYPE, -DMG), y=DMG/10^9)) +
  geom_bar(stat="identity", fill = "midnightblue") +
  scale_y_continuous(breaks=seq(0,150,25)) +
  ggtitle("Top 10 Weather Events causing Highest Property Damage") +
  labs(x="Event type", y="Total Ecomic impact (billion USD)") +
  theme_bw() +
  theme(plot.title = element_text(hjust = 0.5)) +
  theme(axis.text.x = element_text(angle=90, vjust=0.5, hjust=1))
```



```
#Crops
ggplot(storm.crop[1:20,], aes(x=reorder(EVTYPE, -DMG), y=DMG/10^9)) +
  geom_bar(stat="identity", fill = "mediumorchid") +
  ggtitle("Top 10 Weather Events causing Highest Crop Damage") +
  labs(x="Event type", y="Total Ecomic impact (billion USD)") +
  theme_bw() +
  theme(plot.title = element_text(hjust = 0.5)) +
  theme(axis.text.x = element_text(angle=90, vjust=0.5, hjust=1))
```



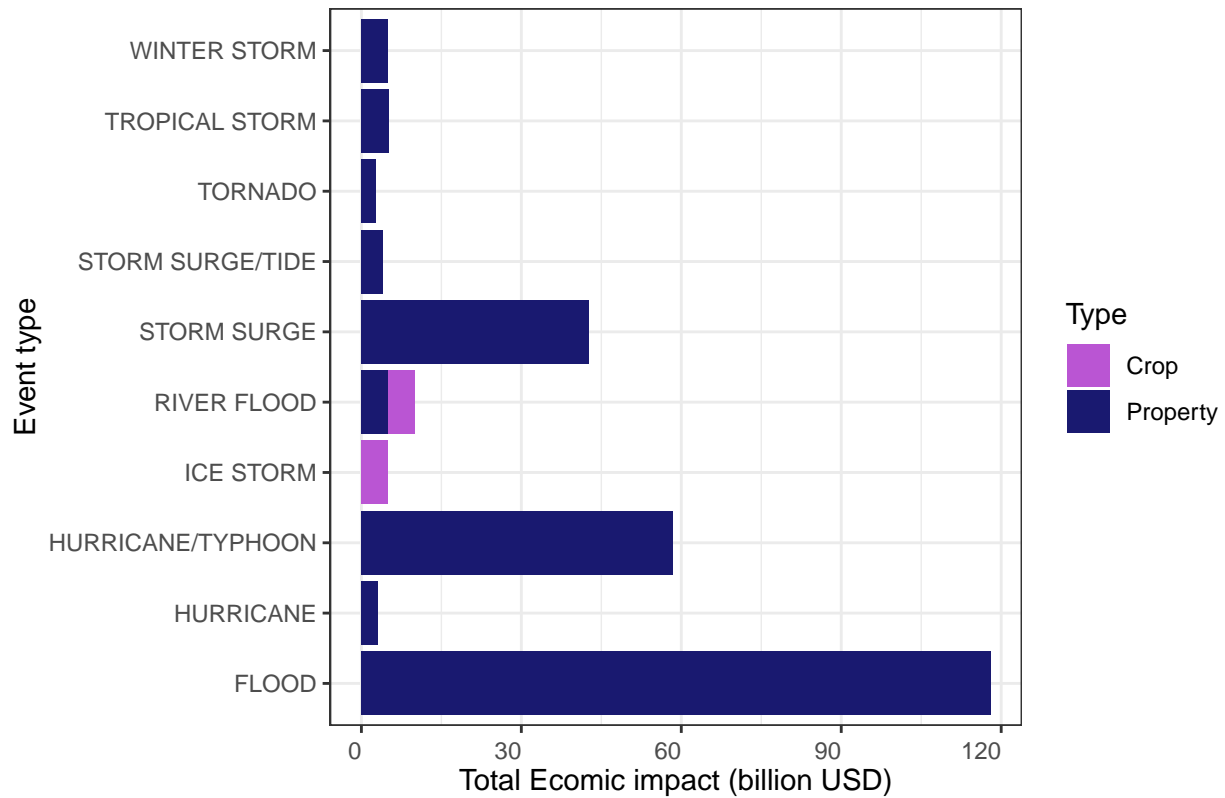
```
#Combine property & crop damages
storm.property$Type <- "Property"
storm.crop$Type <- "Crop"

storm.damage <- rbind(storm.property, storm.crop)

storm.damage <- storm.damage %>%
  arrange(-DMG, EVTYPE)

ggplot(storm.damage[1:20,], aes(x = EVTYPE, y = DMG/10^9, fill = Type)) +
  coord_flip() +
  geom_bar(stat="identity") +
  scale_fill_manual(values = c("mediumorchid", "midnightblue")) +
  ggtitle("Top 10 Weather Events with Greatest Economic Consequences") +
  labs(x="Event type", y="Total Economic impact (billion USD)") +
  theme_bw() +
  theme(plot.title = element_text(hjust = 0.5)) +
  theme(axis.text.x = element_text(angle=0, vjust=0.5, hjust=1))
```

## Top 10 Weather Events with Greatest Economic Consequences



The figures show that most **Property** damages were caused by **Floods** and highest **Crop** damages result from **Ice Storms**, **River floods** and **Droughts**.