

Exploring the NOAA Storm Database

R Markdown

```
## Introduction
# Severe storms and other weather events can cause significant health and economic impacts to communities.

# The dataset used for this project is the Storm Database of the National Oceanic and Atmospheric Administration.

# The purpose of this project is to analyze the NOAA Storm Database to identify trends and patterns in severe weather events.

# The analysis will focus on the following questions:
# - What are the most common types of severe weather events in the United States?
# - Where do severe weather events occur most frequently?
# - What are the most significant health and economic impacts of severe weather events?
# - The analysis will use a variety of statistical methods to answer these questions.

# The results of the analysis will be presented in a report that includes a discussion of the findings,
# -----
```

how the data were loaded into R and processed for analysis

```
#dataset<-download.file("https://d396qusza40orc.cloudfront.net/repdata%2Fdata%2FStormData.csv.bz2", "E:/Data Science Foundations using R/5 Reproducible Research/Woche 4/Projekt 1/dataset.csv")
dataset <- read.csv(bzfile("E:/Data Science Foundations using R/5 Reproducible Research/Woche 4/Projekt 1/dataset.csv"))
```

some content about the data

```
head(dataset)
```

```
##   STATE__      BGN_DATE BGN_TIME TIME_ZONE COUNTY COUNTYNAMES STATE  EVTYPE
## 1      1  4/18/1950 0:00:00    0130      CST     97      MOBILE    AL  TORNADO
## 2      1  4/18/1950 0:00:00    0145      CST      3      BALDWIN   AL  TORNADO
## 3      1  2/20/1951 0:00:00    1600      CST     57      FAYETTE    AL  TORNADO
## 4      1   6/8/1951 0:00:00    0900      CST     89      MADISON    AL  TORNADO
## 5      1 11/15/1951 0:00:00    1500      CST     43      CULLMAN     AL  TORNADO
## 6      1 11/15/1951 0:00:00    2000      CST     77 LAUDERDALE    AL  TORNADO
##   BGN_RANGE BGN_AZI BGN_LOCATI END_DATE END_TIME COUNTY_END COUNTYNENDN
## 1         0      0          0    0000    0000         0         NA
## 2         0      0          0    0000    0000         0         NA
```

```

## 3      0      0      NA
## 4      0      0      NA
## 5      0      0      NA
## 6      0      0      NA
##   END_RANGE END_AZI END_LOCATI LENGTH WIDTH F MAG FATALITIES INJURIES PROPDMG
## 1      0      0      14.0   100 3   0      0      15    25.0
## 2      0      0      2.0   150 2   0      0      0     2.5
## 3      0      0      0.1   123 2   0      0      2    25.0
## 4      0      0      0.0   100 2   0      0      2     2.5
## 5      0      0      0.0   150 2   0      0      2     2.5
## 6      0      0      1.5   177 2   0      0      6     2.5
##   PROPDMGEXP CROPDMG CROPDMGEXP WFO STATEOFFIC ZONENAMES LATITUDE LONGITUDE
## 1      K      0      3040      8812
## 2      K      0      3042      8755
## 3      K      0      3340      8742
## 4      K      0      3458      8626
## 5      K      0      3412      8642
## 6      K      0      3450      8748
##   LATITUDE_E LONGITUDE_ REMARKS REFNUM
## 1      3051      8806      1
## 2      0      0      2
## 3      0      0      3
## 4      0      0      4
## 5      0      0      5
## 6      0      0      6

```

```
str(dataset)
```

```

## '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 ...
```

```
names(dataset)
```

```
## [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"
```

Injuries sorting, top 10

```
total_injuries <- aggregate(INJURIES~EVTYPE, dataset, sum)
total_injuries <- arrange(total_injuries, desc(INJURIES))
total_injuries <- total_injuries[1:20, ]
total_injuries
```

```
##           EVTYPE INJURIES
## 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
## 11        WINTER STORM    1321
## 12 HURRICANE/TYPHOON    1275
## 13           HIGH WIND    1137
## 14          HEAVY SNOW    1021
## 15           WILDFIRE     911
## 16 THUNDERSTORM WINDS     908
## 17          BLIZZARD     805
## 18           FOG       734
## 19    WILD/FOREST FIRE     545
## 20          DUST STORM     440
```

Total fatalities

```
total_fatalities <- aggregate(FATALITIES~EVTYPE,dataset, sum)
total_fatalities <- arrange(total_fatalities, desc(FATALITIES))
total_fatalities <- total_fatalities[1:20, ]
total_fatalities
```

```
##           EVTYPE FATALITIES
## 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
## 11    WINTER STORM          206
## 12    RIP CURRENTS          204
## 13      HEAT WAVE          172
## 14    EXTREME COLD          160
## 15 THUNDERSTORM WIND          133
## 16      HEAVY SNOW          127
## 17 EXTREME COLD/WIND CHILL          125
## 18      STRONG WIND          103
## 19      BLIZZARD          101
## 20      HIGH SURF          101
```

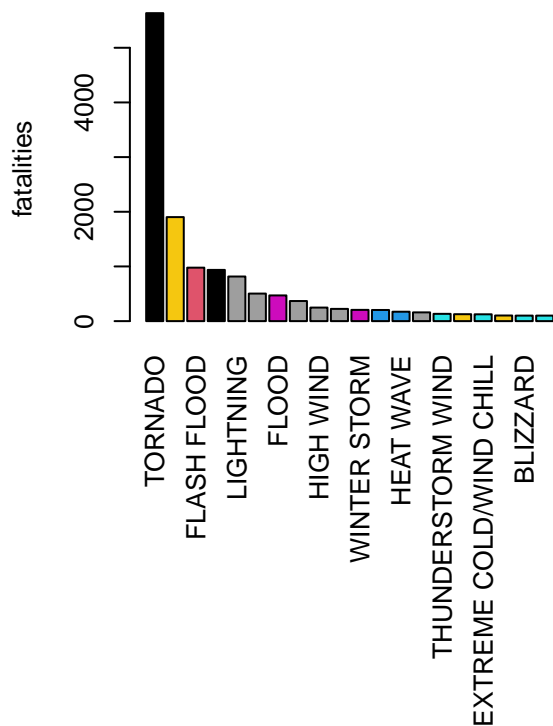
```
str(total_fatalities)
```

```
## 'data.frame':   20 obs. of  2 variables:
##  $ EVTYPE      : chr  "TORNADO" "EXCESSIVE HEAT" "FLASH FLOOD" "HEAT" ...
##  $ FATALITIES: num  5633 1903 978 937 816 ...
```

Results: Weather events, highest fatalities, top 10

```
totals <- total_fatalities
par(mfrow = c(1, 2), mar = c(15, 4, 3, 2), mgp = c(3, 1, 0), cex = 0.8)
barplot(totals$FATALITIES, las = 3, names.arg = totals$EVTYPE, main = "Weather events, highest fatalities")
```

Weather events, highest fatalities, top 1



Sad stuff - top 5

```
totals<- merge(total_fatalities, total_injuries, by.x = "EVTYPE", by.y = "EVTYPE")
totals<-arrange(totals,desc(FATALITIES+INJURIES))

sad_stuff <- melt(totals, id.vars="EVTYPE", variable.name = "bad_thing")
tail(sad_stuff, 5)
```

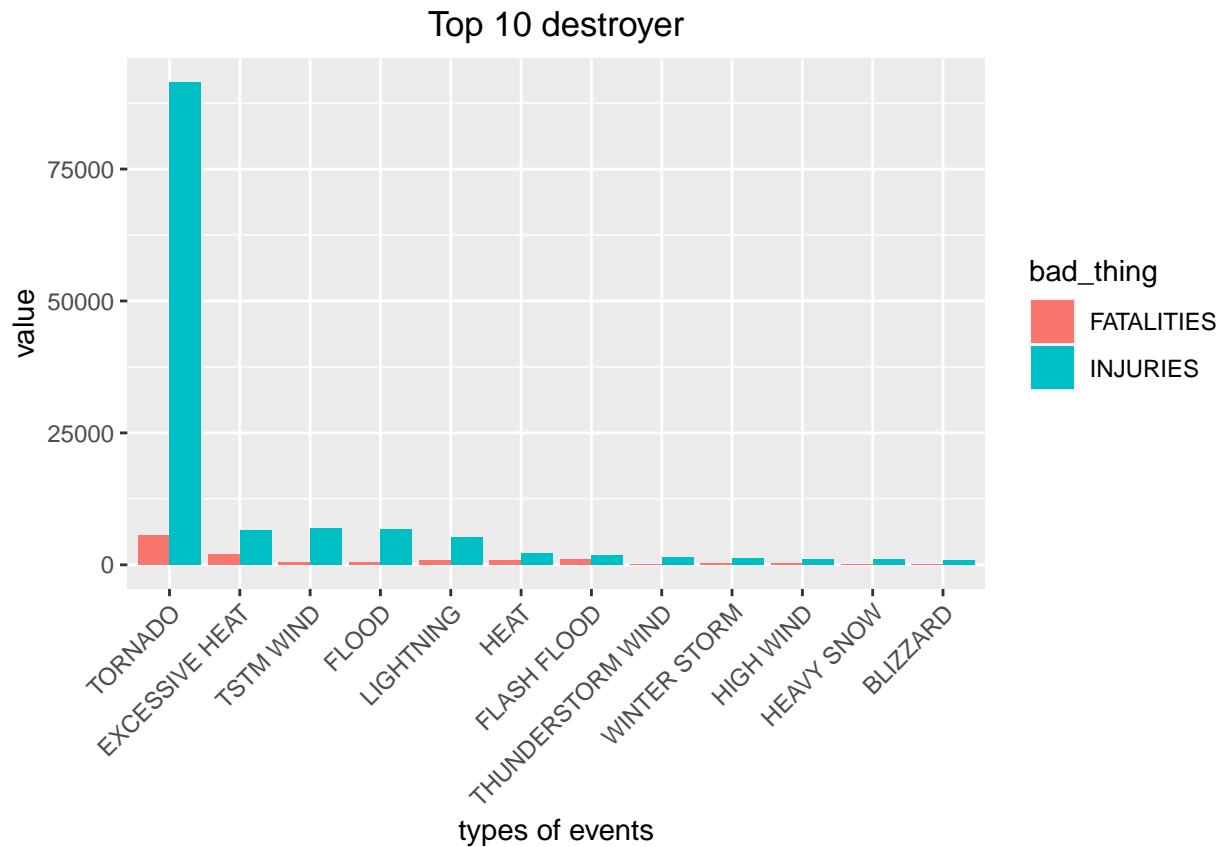
```
##           EVTYPE bad_thing value
## 20 THUNDERSTORM WIND INJURIES 1488
## 21   WINTER STORM INJURIES 1321
## 22     HIGH WIND INJURIES 1137
## 23   HEAVY SNOW INJURIES 1021
## 24     BLIZZARD INJURIES  805
```

Results: Top 10 destroyer

```
healthChart <- ggplot(sad_stuff, aes(x=reorder(EVTYPE, -value), y=value))

healthChart = healthChart + geom_bar(stat="identity", aes(fill=bad_thing), position="dodge")
```

```
healthChart = healthChart + xlab("types of events")
healthChart = healthChart + theme(axis.text.x = element_text(angle=45, hjust=1))
healthChart = healthChart + ggtitle("Top 10 destroyer") + theme(plot.title = element_text(hjust = 0.5))
healthChart
```



Event type - sorting PROPDMG

```
propdmg <- aggregate(PROPDMG ~ EVTYPE, data = dataset, FUN = sum)
propdmg <- propdmg[order(propdmg$PROPDMG, decreasing = TRUE), ]
propdmgMax <- propdmg[1:10, ]
print(propdmgMax)
```

```
##           EVTYPE  PROPDMG
## 834      TORNADO 3212258.2
## 153  FLASH FLOOD 1420124.6
## 856     TSTM WIND 1335965.6
## 170        FLOOD  899938.5
## 760 THUNDERSTORM WIND 876844.2
## 244          HAIL  688693.4
## 464     LIGHTNING  603351.8
## 786 THUNDERSTORM WINDS 446293.2
## 359        HIGH WIND  324731.6
## 972    WINTER STORM  132720.6
```

Event type - sorting CROPDMG

```
cropdmg <- aggregate(CROPDMG ~ EVTYPE, data = dataset, FUN = sum)
cropdmg <- cropdmg[order(cropdmg$CROPDMG, decreasing = TRUE), ]
cropdmgMax <- cropdmg[1:10, ]
print(cropdmgMax)
```

```
##           EVTYPE  CROPDMG
## 244          HAIL 579596.28
## 153    FLASH FLOOD 179200.46
## 170          FLOOD 168037.88
## 856    TSTM WIND 109202.60
## 834    TORNADO 100018.52
## 760 THUNDERSTORM WIND 66791.45
## 95      DROUGHT 33898.62
## 786 THUNDERSTORM WINDS 18684.93
## 359    HIGH WIND 17283.21
## 290    HEAVY RAIN 11122.80
```

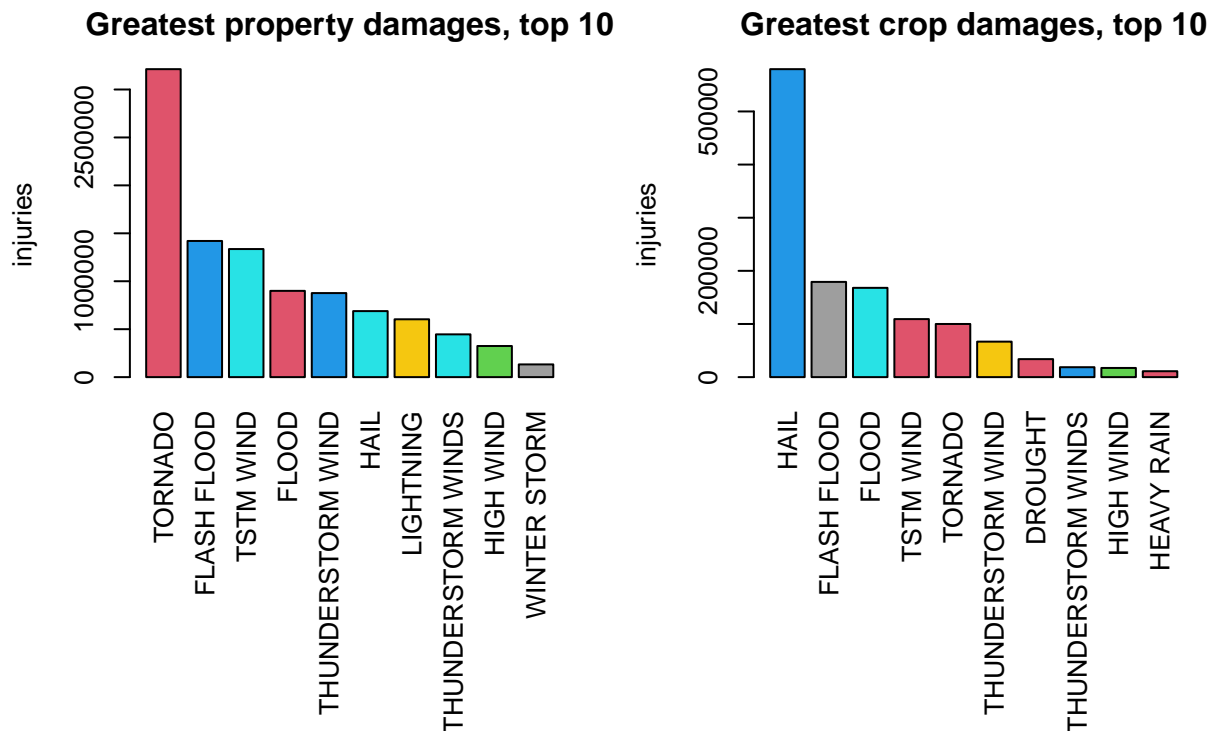
Greatest property damages & greatest crop damages, top10

```
par(mfrow = c(1, 2), mar = c(15, 4, 3, 2), mgp = c(3, 1, 0), cex = 0.8)

options(scipen = 999)

barplot(propdmgMax$PROPDMG, las = 3, names.arg = propdmgMax$EVTYPE,
        main = "Greatest property damages, top 10",
        ylab = "injuries", col = propdmgMax$PROPDMG)

barplot(cropdmgMax$CROPDMG, las = 3, names.arg = cropdmgMax$EVTYPE,
        main = "Greatest crop damages, top 10",
        ylab = "injuries", col = cropdmgMax$CROPDMG)
```



Top 5 Event-Type - Damage Type =“PROPDMG”

```
totalDamage<- merge(propdmgMax,cropdmgMax,by.x = "EVTYPE", by.y = "EVTYPE")
totalDamage<-arrange(totalDamage,desc(PROPDMG + CROPDMG))
top10damages <- melt(totalDamage, id.vars="EVTYPE", variable.name = "Damage_Types")
head(top10damages, 5)
```

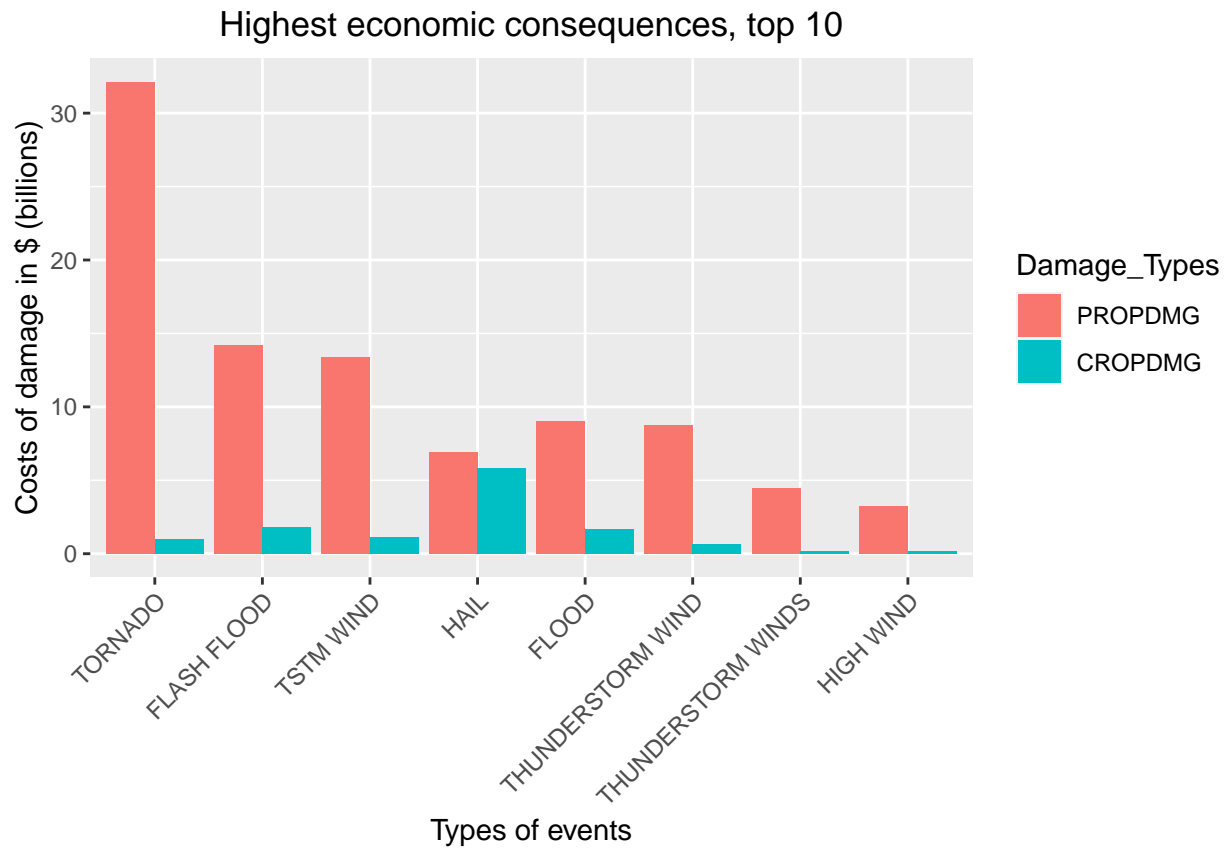
```
##      EVTYPE Damage_Types      value
## 1  TORNADO      PROPDMG 3212258.2
## 2 FLASH FLOOD      PROPDMG 1420124.6
## 3  TSTM WIND      PROPDMG 1335965.6
## 4    HAIL      PROPDMG  688693.4
## 5   FLOOD      PROPDMG  899938.5
```

Highest economic consequences, top 10

```
DamageChart <- ggplot(top10damages, aes(x=reorder(EVTYPE, -value/100000), y=value/100000))
DamageChart = DamageChart + geom_bar(stat="identity", aes(fill=Damage_Types), position="dodge")
DamageChart = DamageChart + xlab("Types of events") + ylab("Costs of damage in $ (billions)")
DamageChart = DamageChart + theme(axis.text.x = element_text(angle=45, hjust=1))
```



```
DamageChart = DamageChart + ggtitle("Highest economic consequences, top 10") + theme(plot.title = element_text(margin = 10))
DamageChart
```



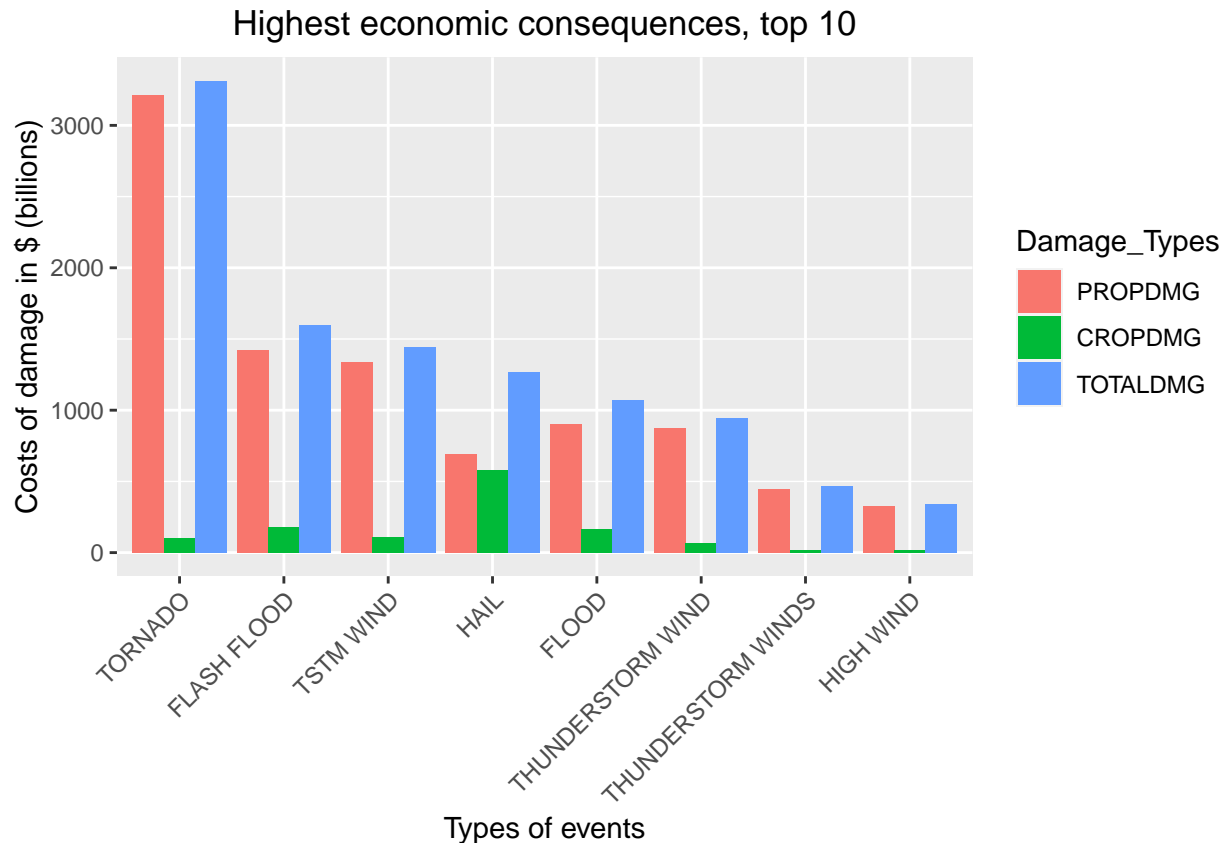
Top 5 Event-Type - Damage Type =“TOTALDMG”

```
totalDamage <- merge(propdmgMax, cropdmgMax, by.x = "EVTYPE", by.y = "EVTYPE")
totalDamage$TOTALDMG <- totalDamage$PROPDMG + totalDamage$CROPDMG
totalDamage <- arrange(totalDamage, desc(TOTALDMG))
top10damages <- melt(totalDamage, id.vars = "EVTYPE", variable.name = "Damage_Types")
tail(top10damages, 5)
```

| ## | EVTYPE | Damage_Types | value |
|-------|--------------------|--------------|-----------|
| ## 20 | HAIL | TOTALDMG | 1268289.7 |
| ## 21 | FLOOD | TOTALDMG | 1067976.4 |
| ## 22 | THUNDERSTORM WIND | TOTALDMG | 943635.6 |
| ## 23 | THUNDERSTORM WINDS | TOTALDMG | 464978.1 |
| ## 24 | HIGH WIND | TOTALDMG | 342014.8 |

Highest economic consequences, top 10

```
DamageChart <- ggplot(top10damages, aes(x=reorder(EVTYPE, -value/1000), y=value/1000), fill=Damage_Types)
DamageChart = DamageChart + geom_bar(stat="identity", aes(fill=Damage_Types), position="dodge")
DamageChart = DamageChart + xlab("Types of events") + ylab("Costs of damage in $ (billions)")
DamageChart = DamageChart + theme(axis.text.x = element_text(angle=45, hjust=1))
DamageChart = DamageChart + ggtitle("Highest economic consequences, top 10") + theme(plot.title = element_text(hjust=0.5))
DamageChart
```



Summary: The analysis of the NOAA Storm Database has shown that the most common types of severe weather events in the United States are tornadoes, hurricanes, and storms. These events occur most frequently in the South and Southeast United States. The most serious health and economic impacts of severe weather events are fatalities, injuries, and property damage.

Discussion of Implications: The findings of the analysis have important implications for public policy and decision-making. The government should take action to protect the public from the consequences of severe weather events. This includes improved early warning systems, disaster preparedness plans, and recovery efforts. Businesses and individuals should also take action to protect themselves from the consequences of severe weather events. This includes insurance, emergency plans, and supplies.

Recommendations for Further Research: The findings of the analysis can be supported or expanded by further research. Areas that should be further investigated include:

- The impacts of severe weather events on different population groups
- The impacts of severe weather events on the environment
- The costs of severe weather events
- The effectiveness of disaster mitigation measures

Conclusion: The analysis of the NOAA Storm Database has provided important insights into the severity and extent of severe weather events in the United States. The findings of the analysis have important implications for public policy and decision-making. By implementing disaster mitigation measures, the public can be protected from the consequences of these events.