

Storm

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The Analysis the Weather Event Impact to Population Health and Properties

Qestions:

Across the United States,

- 1) Which types of events are most harmful with respect to population health?
- 2) Which types of events have the greatest economic consequences?

From the Storm Events data, answer the questions.

```
library(tidyverse)
library(lubridate)
options(digits=4)
options(warn=-1)
options(width=110)
options(scipen=3)
```

Loading and preprocessing the data

Data from:

National Climatic Data Center Storm Events

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

First time, Read data file each variavles as character.

```
Storm_wk <- read_csv('repdata_data_StormData.csv.bz2', col_names=T, col_types=str_dup('c',37))

dim( Storm_wk) %>% cat ('Dimension: ',.,'\n')

## Dimension:  902297 37

names(Storm_wk) %>% cat ('Col Names: ',.,'\n')

## Col Names:  STATE__ BGN_DATE BGN_TIME TIME_ZONE COUNTY COUNTYNAME STATE EVTYPE BGN_RANGE BGN_AZI BGN_LOCA
TI END_DATE END_TIME COUNTY_END COUNTYENDN END_RANGE END_AZI END_LOCATI LENGTH WIDTH F MAG FATALITIES INJURI
ES PROPDMG PROPDMGEXP CROPDMG CROPDMGEXP WFO STATEOFFIC ZONENAMES LATITUDE LONGITUDE LATITUDE_E LONGITUDE_ R
EMARKS REFNUM
```

Relational Valiables for the Questions in the file. EVTYPE - Event Type

FATALITIES

INJURIES

PROPDMG - Property Damage, PROPDMGEXP - Exponents

CROPDMG - Crop Damage, CROPDMGEXP - Exponents

Check the Variable Data characteristics..

```
map(c('EVTYPE', 'FATALITIES', 'INJURIES', 'PROPDMG', 'PROPDMGEXP', 'CROPDMG', 'CROPDMGEXP'),
function(x){
  cat('\n',x,'---\n')
```

```

wk1 <- Storm_wk %>% pull(x)
wk1 %>% is.na() %>% sum() %>% cat('Number of NAs: ',.,'\n')
wk2 <- wk1 %>% unique()
wk2 %>% length() %>% cat('Unique Length: ',.,'\n')
if(length(wk2)<=20){wk2 %>% sort() %>% cat('\n')}
else
    {wk2 %>% sort() %>% head(6) %>% cat('Head: ',.,'\n')
    wk2 %>% sort() %>% tail(6) %>% cat('Tail: ',.,'\n')}

}) -> null

##
## EVTYPE ---
## Number of NAs: 0
## Unique Length: 977
## Head: ? ABNORMAL WARMTH ABNORMALLY DRY ABNORMALLY WET ACCUMULATED SNOWFALL AGRICULTURAL FREEZE
## Tail: WINTER WEATHER/MIX WINTERY MIX Wintry mix Wintry Mix WINTRY MIX WND
##
## FATALITIES ---
## Number of NAs: 0
## Unique Length: 52
## Head: 0.00 1.00 10.00 11.00 114.00 116.00
## Tail: 74.00 75.00 8.00 9.00 90.00 99.00
##
## INJURIES ---
## Number of NAs: 0
## Unique Length: 200
## Head: 0.00 1.00 10.00 100.00 101.00 102.00
## Tail: 93.00 94.00 95.00 96.00 97.00 98.00
##
## PROPDGM ---
## Number of NAs: 0
## Unique Length: 1390
## Head: 0.00 0.01 0.02 0.03 0.04 0.05
## Tail: 99.00 99.39 99.97 990.00 995.00 996.00
##
## PROPDMGEXP ---
## Number of NAs: 465934
## Unique Length: 19
## - ? + 0 1 2 3 4 5 6 7 8 B h H K m M
##
## CROPDGM ---
## Number of NAs: 0
## Unique Length: 432
## Head: 0.00 0.01 0.02 0.03 0.05 0.10
## Tail: 97.00 975.00 978.00 985.00 99.00 990.00
##
## CROPDMGEXP ---
## Number of NAs: 618413
## Unique Length: 9
## ? 0 2 B k K m M

Storm_wk <- Storm_wk %>%
  select(BGN_DATE,EVTYPE,FATALITIES,INJURIES,PROPDGM,PROPDMGEXP,CROPDGM,CROPDMGEXP) %>%
  mutate(BGN_DATE = mdy_hms(BGN_DATE),
         EVTYPE = str_to_upper(EVTYPE),
         FATALITIES = as.integer(FATALITIES),
         INJURIES = as.integer(INJURIES),
         PROPDGM = as.double(PROPDGM),
         PROPDMGEXP = str_to_upper(PROPDMGEXP),
         CROPDGM = as.double(CROPDGM),
         CROPDMGEXP = str_to_upper(CROPDMGEXP)) %>%
  filter(!str_detect(EVTYPE,'SUMMARY'))
dim(Storm_wk) # [1] 902221      8

## [1] 902221      8

```

```
Storm_wk %>% head()
```

```
## # A tibble: 6 x 8
##   BGN_DATE      EVTYPE FATALITIES INJURIES PROPDMG PROPDMGEXP CROPDMG CROPDMGEXP
##   <dtm>      <chr>      <int>    <int>    <dbl> <chr>      <dbl> <chr>
## 1 1950-04-18 00:00:00 TORNADO      0      15      25 K      0 <NA>
## 2 1950-04-18 00:00:00 TORNADO      0      0      2.5 K      0 <NA>
## 3 1951-02-20 00:00:00 TORNADO      0      2      25 K      0 <NA>
## 4 1951-06-08 00:00:00 TORNADO      0      2      2.5 K      0 <NA>
## 5 1951-11-15 00:00:00 TORNADO      0      2      2.5 K      0 <NA>
## 6 1951-11-15 00:00:00 TORNADO      0      6      2.5 K      0 <NA>
```

Check Summary of Data

```
Storm_wk %>% pull(EVTYPE) %>% unique() %>% length()
```

```
## [1] 823
```

```
summary(Storm_wk[,c(3,4,5,7)])
```

```
##   FATALITIES      INJURIES      PROPDMG      CROPDMG
##   Min.   : 0   Min.   : 0.0   Min.   : 0   Min.   : 0.0
##   1st Qu.: 0   1st Qu.: 0.0   1st Qu.: 0   1st Qu.: 0.0
##   Median : 0   Median : 0.0   Median : 0   Median : 0.0
##   Mean   : 0   Mean   : 0.2   Mean   : 12   Mean   : 1.5
##   3rd Qu.: 0   3rd Qu.: 0.0   3rd Qu.: 0   3rd Qu.: 0.0
##   Max.   :583   Max.   :1700.0   Max.   :5000   Max.   :990.0
```

```
Storm_wk %>% pull(PROPDMGEXP) %>% unique() %>% sort()
```

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

```
Storm_wk %>% pull(CROPDMGEXP) %>% unique() %>% sort()
```

```
## [1] "?" "0" "2" "B" "K" "M"
```

Clean Event types

There are variations in expression in adjective for events and event names. So, remove adjectives and be just events. and aggregate event names.

```
Storm <- Storm_wk %>%
  mutate(EVTYPE = str_replace(EVTYPE, ' ', ' '),
         EVTYPE = str_replace(EVTYPE,
                               '^BITTER|^RECORD|^EXCESSIVE|^EXTREME|^SMALL|^SEVERE|^HEAVY ', ''),
         EVTYPE = str_replace(EVTYPE,
                               '^HIGH|^STRONG|^PROLONG|^UNSEASONABLY|^MODERATE|^DENSE ', ''),
         EVTYPE = str_replace(EVTYPE, '^BLIZZARD.*', 'BLIZZARD'),
         EVTYPE = str_replace(EVTYPE, '^HAIL.*', 'HAIL'),
         EVTYPE = str_replace(EVTYPE, '^HURRICANE.*|^TYPHOON.*', 'HURRICANE'),
         EVTYPE = str_replace(EVTYPE, '^LIGHTNING.*', 'LIGHTNING'),
         EVTYPE = str_replace(EVTYPE, '^THUNDERSTORM.*|^TSTM.*', 'THUNDERSTORM WIND'),
         EVTYPE = str_replace(EVTYPE, '^COASTAL.*|^BEACH.*', 'COASTAL FLOOD'),
         EVTYPE = str_replace(EVTYPE, '^FLASH FLOOD.*', 'FLASH FLOOD'),
         EVTYPE = str_replace(EVTYPE, '^FLOOD.*', 'FLOOD'),
         EVTYPE = str_replace(EVTYPE, '^URBAN/SML STREAM FLD', 'FLOOD'),
         EVTYPE = str_replace(EVTYPE, '^URBAN FLOOD', 'FLOOD'),
         EVTYPE = str_replace(EVTYPE, '^RIVER FLOOD.*', 'FLOOD'),
         EVTYPE = str_replace(EVTYPE, '^FUNNEL CLOUD.*', 'FUNNEL CLOUD'),
         EVTYPE = str_replace(EVTYPE, '^COLD.*', 'COLD/WIND CHILL'),
         EVTYPE = str_replace(EVTYPE, '^UN.+ COLD', 'COLD/WIND CHILL'),
         EVTYPE = str_replace(EVTYPE, '^HEAT.*|^WARM.*', 'HEAT'),
         EVTYPE = str_replace(EVTYPE, '^RAIN.*', 'RAIN'),
         EVTYPE = str_replace(EVTYPE, '^MARINE TSTM.*', 'MARINE THUNDERSTORM WIND'),
```

```

EVTTYPE = str_replace(EVTTYPE, '^MARINE THUNDERSTORM.*', 'MARINE THUNDERSTORM WIND'),
EVTTYPE = str_replace(EVTTYPE, '^SNOW.*', 'SNOW'),
EVTTYPE = str_replace(EVTTYPE, '^STORM SURGE.*', 'STORM SURGE/TIDE'),
EVTTYPE = str_replace(EVTTYPE, '^SURF.*', 'SURF'),
EVTTYPE = str_replace(EVTTYPE, '^TORNADO.*', 'TORNADO'),
EVTTYPE = str_replace(EVTTYPE, '^WIND.*', 'WIND'),
EVTTYPE = str_replace(EVTTYPE, '^WATERSPOUT.*', 'WATERSPOUT'),
EVTTYPE = str_replace(EVTTYPE, '^WILD/FOREST FIRE.*', 'WILDFIRE'),
EVTTYPE = str_replace(EVTTYPE, '^WINTER WEATHER.*', 'WINTER WEATHER')
)

```

Data Processing for Q1

Which types of events are most harmful with respect to population health?

- 1) summarize sum of FATALITIES and INJURIES by each event.
- 2) Sort them in descending order of sum of FATALITIES and INJURIES.
- 3) Select events by top 10 and make pie chart.

```

event_FAIN <- Storm %>%
  mutate(FATAL_INJU = FATALITIES+INJURIES) %>%
  group_by(EVTTYPE) %>%
  summarise(FATAL_INJU=sum(FATAL_INJU)) %>%
  arrange(desc(FATAL_INJU))
event_FAIN <- event_FAIN %>%
  head(10) %>%
  bind_rows(tibble(EVTTYPE='others', FATAL_INJU=sum(event_FAIN$FATAL_INJU[11:nrow(event_FAIN)]))) %>%
  mutate(percentage=100*FATAL_INJU/sum(FATAL_INJU))
event_FAIN %>% print(n=Inf)

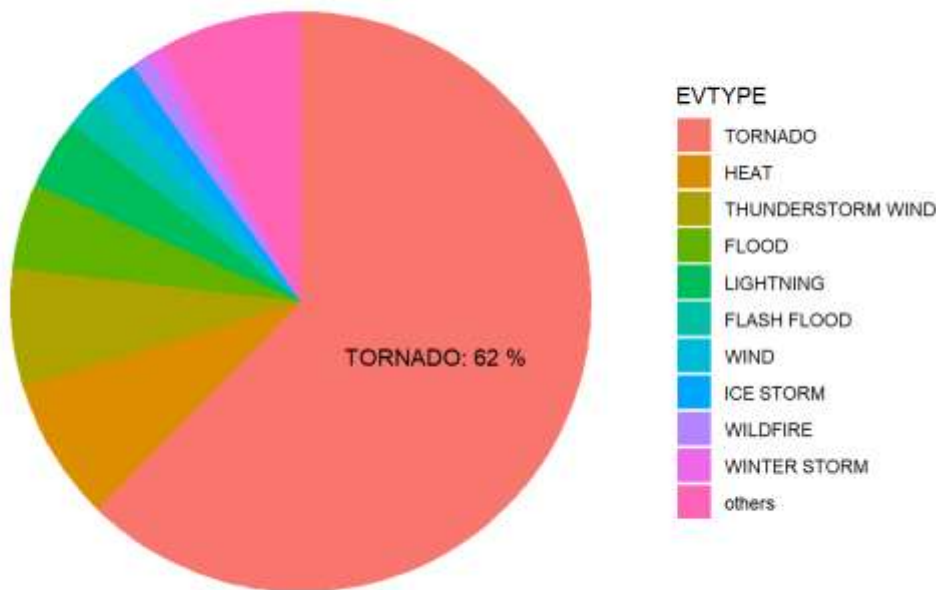
## # A tibble: 11 x 3
##   EVTTYPE          FATAL_INJU percentage
##   <chr>          <int>         <dbl>
## 1 TORNADO          97022          62.3
## 2 HEAT             12402           7.97
## 3 THUNDERSTORM WIND 10218           6.56
## 4 FLOOD             7415           4.76
## 5 LIGHTNING         6049           3.89
## 6 FLASH FLOOD       2803           1.80
## 7 WIND              2310           1.48
## 8 ICE STORM         2064           1.33
## 9 WILDFIRE          1543           0.991
## 10 WINTER STORM     1527           0.981
## 11 others          12320           7.91

event_FAIN <- event_FAIN %>%
  mutate(EVTTYPE=factor(EVTTYPE, levels=event_FAIN$EVTTYPE))

ggplot(event_FAIN) +
  geom_bar(aes(x=1,y=FATAL_INJU,fill=EVTTYPE),stat='identity') +
  annotate('text', x=1, y=sum(event_FAIN$FATAL_INJU)-event_FAIN$FATAL_INJU[1]/2,
    label=sprintf('%s: %d %%',
      event_FAIN$EVTTYPE[1],
      as.integer(100*event_FAIN$FATAL_INJU[1]/sum(event_FAIN$FATAL_INJU)))) +
  coord_polar(theta='y') + scale_y_reverse() + theme_void() +
  theme(axis.title.x=element_blank(), axis.title.y=element_blank()) +
  ggtitle('fig1. The top 10 events; Percentage of sum of FATALITIES and INJURIES')

```

fig1. The top 10 events; Percentage of sum of FATALITIES and INJURIES



Data Processing for Q2

Which types of events have the greatest economic consequences?

- 1) Create ECO_DAMAGE as variable from PROPDGMGEXP, CROPDMGEXP, PROPDMGEXP and CROPDMGEXP. Exponents means Power of 10, and "H" is 10^2 , "M" is Million, "B" is Billion, and others not numeric is 10^0 .
- 2) summarize ECO_DAMAGE by event.
- 3) Sort them in descending order of ECO_DAMAGE.
- 4) Select events bytop 10 ECO_DAMAGE and make pie chart.

```
event_DMG <- Storm %>%
  mutate(PROPDGMGEXP=if_else(is.na(PROPDGMGEXP), '0', PROPDGMGEXP),
         CROPDMGEXP=if_else(is.na(CROPDMGEXP), '0', CROPDMGEXP),
         PROPDMGEXP=recode(PROPDMGEXP, '+'='0', '-'='0', '?'='0', 'H'='2', 'K'='3', 'M'='6', 'B'='9'),
         CROPDMGEXP=recode(CROPDMGEXP, '+'='0', '-'='0', '?'='0', 'H'='2', 'K'='3', 'M'='6', 'B'='9'),
         ECO_DAMAGE= PROPDMG*(10^as.integer(PROPDMGEXP))+CROPDMG*(10^as.integer(CROPDMGEXP))) %>%
  group_by(EVTYPE) %>%
  summarise(ECO_DAMAGE=sum(ECO_DAMAGE)) %>%
  arrange(desc(ECO_DAMAGE))
event_DMG <- event_DMG %>%
  head(10) %>%
  bind_rows(tibble(EVTYPE='others', ECO_DAMAGE=sum(event_DMG$ECO_DAMAGE[11:nrow(event_DMG)]))) %>%
  mutate(percentage=100*ECO_DAMAGE/sum(ECO_DAMAGE))
event_DMG %>% print(n=Inf)
```

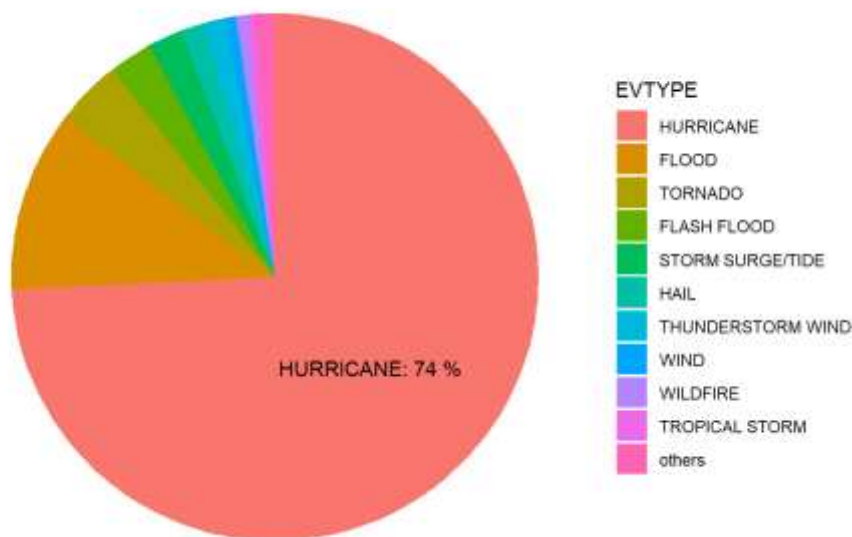
```
## # A tibble: 11 x 3
##   EVTYPE          ECO_DAMAGE percentage
##   <chr>          <dbl>         <dbl>
## 1 HURRICANE      1.64e12         74.3
## 2 FLOOD         2.45e11         11.1
## 3 TORNADO       8.93e10          4.05
## 4 FLASH FLOOD   5.76e10          2.61
```

```
## 5 STORM SURGE/TIDE      4.80e10      2.18
## 6 HAIL                  3.13e10      1.42
## 7 THUNDERSTORM WIND     2.83e10      1.28
## 8 WIND                  1.55e10      0.701
## 9 WILDFIRE              1.50e10      0.682
## 10 TROPICAL STORM       9.65e 9      0.438
## 11 others                2.78e10      1.26
```

```
event_DMG <- event_DMG %>%
  mutate(EVTYPE=factor(EVTYPE,levels=event_DMG$EVTYPE))

ggplot(event_DMG) +
  geom_bar(aes(x=1,y=ECO_DAMAGE,fill=EVTYPE),stat='identity') +
  annotate('text', x=1, y=sum(event_DMG$ECO_DAMAGE)-event_DMG$ECO_DAMAGE[1]/2,
    label=sprintf('%s: %d %%',
      event_DMG$EVTYPE[1],
      as.integer(100*event_DMG$ECO_DAMAGE[1]/sum(event_DMG$ECO_DAMAGE)))) +
  coord_polar(theta='y') + scale_y_reverse() + theme_void() +
  theme(axis.title.x=element_blank(), axis.title.y=element_blank()) +
  ggtitle('fig2. The top 10 events; Percentage of sum of Property and Crops Damates')
```

fig2. The top 10 events; Percentage of sum of Property and Crops Damates



The Results

For Question 1

The type of events most harmful with respect to population health is TORNADO. They occupy about 62% of the whole. (Refer fig1)

For Question 2

The types of events the greatest economic consequences is HURRICANE. They occupy about 74% of the whole. (Refer fig2)