Untitled

September 30, 2020

Analysis the U.S. National Oceanic and Atmospheric Administration's (NOAA) storm database

This project explores the NOAA storm database, which tracks major storms and weather events, to address the most severe types of weather events in the USA, which caused greatest damage to human population in terms of fatalities/injuries and economic loss during the years 1950 - 2011.

There are two goals of this analysis:

- identify the weather events that are most harmful with respect to population health
- identify the weather events that have the greatest economic consequences.

Based on our analysis, we conclude that TORNADOS and FLOODS are most harmful weather events in the USA in terms of the risk to human health and economic impact.

Data processing

```
[13]: head(data)
```

```
STATE
                                   BGN DATE
                                                       BGN TIME
                                                                    TIME ZONE
                                                                                   COUNTY
                                                                                               COUN
                       <dbl>
                                   <fct>
                                                       < fct >
                                                                                    <dbl>
                                                                     <fct>
                                                                                               <fct>
                                                                                    97
                                   4/18/1950 0:00:00
                                                      0130
                                                                    CST
                                                                                               MOBII
                                   4/18/1950 0:00:00
                                                                    CST
                       1
                                                      0145
                                                                                    3
                                                                                               BALD
A data.frame: 6 \times 37
                                   2/20/1951 0:00:00
                       1
                                                       1600
                                                                    CST
                                                                                    57
                                                                                               FAYE
                                   6/8/1951 0:00:00
                                                                    CST
                                                                                    89
                                                                                               MADIS
                       1
                                                      0900
                    5
                                   11/15/1951 0:00:00
                                                                     CST
                                                                                               CULLI
                       1
                                                      1500
                                                                                    43
                                   11/15/1951 0:00:00
                    6
                       1
                                                                    CST
                                                                                               LAUD
                                                      2000
                                                                                    77
```

```
[60]: # subsetting by data
main_data<-data
```

```
[61]: main_data$BGN_DATE <- strptime(data$BGN_DATE, "%m/%d/%Y %H:%M:%S")
```

```
[62]: main_data <- subset(data,main_data$BGN_DATE>1996-01-01,)
[63]: str(main data)
     'data.frame': 866041 obs. of 37 variables:
      $ STATE : num 1 1 1 1 1 1 1 1 1 ...
      $ BGN_DATE : Factor w/ 16335 levels "1/1/1966 0:00:00",..: 3852 3852 5704 4834
     4834 5092 5185 5185 5185 5185 ...
      $ BGN_TIME : Factor w/ 3608 levels "00:00:00 AM",..: 2623 2865 2584 2732 2735
     3090 1143 1555 1683 3555 ...
      $ TIME_ZONE : Factor w/ 22 levels "ADT", "AKS", "AST", ...: 7 7 7 7 7 7 7 7 7 7 ...
                : num 97 63 129 67 5 57 65 73 125 17 ...
      $ COUNTYNAME: Factor w/ 29601 levels "", "5NM E OF MACKINAC BRIDGE TO PRESQUE
     ISLE LT MI",..: 13513 5653 27256 5778 1887 4598 5714 8485 24418 2808 ...
                 : Factor w/ 72 levels "AK", "AL", "AM", ...: 2 2 2 2 2 2 2 2 2 2 ...
      $ STATE
                  : Factor w/ 985 levels " HIGH SURF ADVISORY",..: 856 856 834 856
      $ EVTYPE
     856 244 834 244 856 834 ...
      \ BGN_RANGE: num \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ \dots
      $ BGN_AZI : Factor w/ 35 levels ""," N"," NW",..: 1 1 1 1 1 1 1 1 1 1 ...
      $ BGN_LOCATI: Factor w/ 54429 levels ""," Christiansburg",..: 1 1 1 1 1 1 1 1 1
      $ END_DATE : Factor w/ 6663 levels "","1/1/1993 0:00:00",..: 1 1 1 1 1 1 1 1 1
      $ END TIME : Factor w/ 3647 levels ""," 0900CST",..: 1 1 1 1 1 1 1 1 1 1 ...
      $ COUNTY_END: num 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 ...
      $ END_AZI : Factor w/ 24 levels "", "E", "ENE", "ESE", ...: 1 1 1 1 1 1 1 1 1 1
      $ END LOCATI: Factor w/ 34506 levels ""," CANTON"," TULIA",..: 1 1 1 1 1 1 1
     1 1 ...
      $ LENGTH
                 : num 000001002 ...
      $ WIDTH
                 : num 0 0 33 0 0 0 33 0 0 200 ...
      $ F
                  : int NA NA 2 NA NA NA 2 NA NA 3 ...
      $ MAG
                 : num 51 0 0 0 0 75 0 100 0 0 ...
      $ FATALITIES: num 0 0 0 0 0 0 0 0 2 ...
      $ INJURIES : num 0 0 0 0 0 0 0 0 14 ...
      $ PROPDMG
                : num 0 0 25 0 0 0 25 0 0 250 ...
      : num 00000000000...
      $ CROPDMGEXP: Factor w/ 9 levels "","?","0","2",...: 1 1 1 1 1 1 1 1 1 1 ...
                : Factor w/ 542 levels ""," CI","%SD",..: 1 1 1 1 1 1 1 1 1 1 ...
      $ STATEOFFIC: Factor w/ 250 levels "", "ALABAMA, Central", ..: 1 1 1 1 1 1 1 1 1
      $ ZONENAMES : Factor w/ 25112 levels "","
     "| __truncated__,..: 1 1 1 1 1 1 1 1 1 1 ...
      $ LATITUDE : num 3042 3247 3128 3136 3142 ...
```

```
$ LONGITUDE : num 8817 8754 8815 8517 8524 ...
      $ LATITUDE_E: num 0 0 0 0 0 0 0 0 0 ...
      $ LONGITUDE_: num 0 0 0 0 0 0 0 0 0 ...
      $ REMARKS : Factor w/ 436781 levels "","\t","\t\t",..: 1 1 1 1 1 1 1 1 1 1
      $ REFNUM
                 : num 799 800 801 802 803 804 805 806 807 808 ...
[64]: main_data <- subset(main_data, select = c(EVTYPE, FATALITIES, INJURIES, PROPDMG,
      PROPDMGEXP, CROPDMG, CROPDMGEXP))
[65]: #cleaning event types names
      main_data$EVTYPE <- toupper(main_data$EVTYPE)</pre>
[66]: # eliminating zero data
      #main_data <- main_data[main_data$FATALITIES!=0 |main_data$INJURIES !=0 |__
       →main_data$PROPDMG !=0 | main_data$CROPDMG !=0, ]
     Population health data processing
[68]: health_data <- aggregate(cbind(FATALITIES, INJURIES) ~ EVTYPE, data =_
       →main data, FUN=sum)
[70]: health_data$PEOPLE_LOSS <-health_data$FATALITIES + health_data$INJURIES
[71]: health_data <- health_data[order(health_data$PEOPLE_LOSS, decreasing =
```

[72]: Top10_events_people <- health_data[1:10,]

[73]: Top10_events_people

TRUE),]

		EVTYPE	FATALITIES	INJURIES	PEOPLE_LOSS
		<chr $>$	<dbl></dbl>	<dbl $>$	<dbl></dbl>
A data.frame: 10×4	758	TORNADO	3272	59611	62883
	116	EXCESSIVE HEAT	1903	6525	8428
	779	TSTM WIND	504	6957	7461
	154	FLOOD	470	6789	7259
	418	LIGHTNING	816	5230	6046
	243	HEAT	937	2100	3037
	138	FLASH FLOOD	978	1777	2755
	387	ICE STORM	89	1975	2064
	685	THUNDERSTORM WIND	133	1488	1621
	888	WINTER STORM	206	1321	1527

Economic consequences data processing

```
[74]: main_data$PROPDMGEXP <- gsub("[Hh]", "2", main_data$PROPDMGEXP)
main_data$PROPDMGEXP <- gsub("[Kk]", "3", main_data$PROPDMGEXP)
main_data$PROPDMGEXP <- gsub("[Mm]", "6", main_data$PROPDMGEXP)
```

```
main_data$PROPDMGEXP <- gsub("[Bb]", "9", main_data$PROPDMGEXP)</pre>
      main_data$PROPDMGEXP <- gsub("\\+", "1", main_data$PROPDMGEXP)</pre>
      main_data$PROPDMGEXP <- gsub("\\?|\\-|\\ ", "0", main_data$PROPDMGEXP)</pre>
      main_data$PROPDMGEXP <- as.numeric(main_data$PROPDMGEXP)</pre>
[76]: main_data$CROPDMGEXP <- gsub("[Hh]", "2", main_data$CROPDMGEXP)
      main_data$CROPDMGEXP <- gsub("[Kk]", "3", main_data$CROPDMGEXP)</pre>
      main_data$CROPDMGEXP <- gsub("[Mm]", "6", main_data$CROPDMGEXP)</pre>
      main_data$CROPDMGEXP <- gsub("[Bb]", "9", main_data$CROPDMGEXP)</pre>
      main_data$CROPDMGEXP <- gsub("\\+", "1", main_data$CROPDMGEXP)</pre>
      main_data$CROPDMGEXP <- gsub("\\-|\\?|\\ ", "0", main_data$CROPDMGEXP)</pre>
      main_data$CROPDMGEXP <- as.numeric(main_data$CROPDMGEXP)</pre>
      main_data$PROPDMGEXP[is.na(main_data$PROPDMGEXP)] <- 0</pre>
      main_data$CROPDMGEXP[is.na(main_data$CROPDMGEXP)] <- 0</pre>
[78]: #creating total damage values
      library(dplyr)
[80]: main data <- mutate(main data, PROPDMGTOTAL = PROPDMG * (10 ^ PROPDMGEXP), __
       → CROPDMGTOTAL = CROPDMG * (10 ^ CROPDMGEXP))
[81]: | Economic <- aggregate(cbind(PROPDMGTOTAL, CROPDMGTOTAL) ~ EVTYPE, data =___
       →main_data, FUN=sum)
[82]: Economic $ECONOMIC_LOSS <- Economic $PROPDMGTOTAL + Economic $CROPDMGTOTAL
      Economic <- Economic[order(Economic$ECONOMIC_LOSS, decreasing =TRUE), ]</pre>
[84]: Top10_events_economy <- Economic[1:10,]
[85]: Top10_events_economy
                              EVTYPE
                                                         PROPDMGTOTAL CROPDMGTOTAL
                                                                                                 ECONON
                                                         <dbl>
                                                                             <dbl>
                                                                                                 <dbl>
                              < chr >
                                                                                                 15\overline{031967}
                         154
                              FLOOD
                                                         144657709807
                                                                             5661968450
                              HURRICANE/TYPHOON
                         372
                                                         69305840000
                                                                             2607872800
                                                                                                 71913712
                         758
                              TORNADO
                                                                                                 52088388
                                                         51673434936
                                                                             414953270
                         599
                              STORM SURGE
                                                         43323536000
                                                                             5000
                                                                                                 43323541
     A data.frame: 10 \times 4
                         212
                              HAIL
                                                                                                 18761221
                                                         15735267513
                                                                             3025954473
                         138
                              FLASH FLOOD
                                                         16822673978
                                                                             1421317100
                                                                                                 18243991
                          84
                              DROUGHT
                                                                                                 15018672
                                                         1046106000
                                                                             13972566000
                         363
                              HURRICANE
                                                         11868319010
                                                                             2741910000
                                                                                                 14610229
```

Results

[86]: #plotting health loss
library(ggplot2)

529 | RIVER FLOOD

387 ICE STORM

5118945500

3944927860

5029459000

5022113500

10148404

89670413

```
g <- ggplot(data = Top10_events_people, aes(x = reorder(EVTYPE, PEOPLE_LOSS),__

y= PEOPLE_LOSS))

g <- g + geom_bar(stat = "identity", colour = "black") + labs(title = "Total__

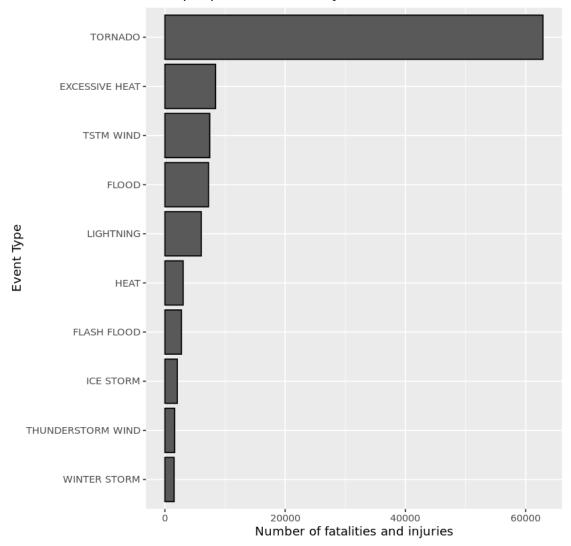
people loss in USA by weather events in 1996-2011") + theme(plot.title =__

element_text(hjust = 0.5)) + labs(y = "Number of fatalities and injuries",__

x = "Event Type") + coord_flip()
```

[87]: print(g)

Total people loss in USA by weather events in 1996-2011



```
g <- g + geom_bar(stat = "identity", colour = "black") + labs(title = "Total

→economic loss in USA by weather events in

1996-2011") + theme(plot.title = element_text(hjust = 0.5)) + labs(y = "Size of

→property and crop loss", x = "Event Type") + coord_flip()
```

[89]: print(g)

Total economic loss in USA by weather events in 1996-2011

