

Project 2

1: Synopsis

The goal of the assignment is to explore the NOAA Storm Database and explore the effects of severe weather events on both population and economy. The time period for the database is between 1950 and November 2011.

Following analysis investigates which types of severe weather events are most harmful on:

1. Health (injuries and fatalities)
2. Property and crops (economic consequences)

2: Data Processing

2.1: Data Loading

Download the raw data file and extract the data into a dataframe. Then convert to a data.table

```
## -- Attaching packages -----  
  
## v ggplot2 3.3.2      v purrr  0.3.4  
## v tibble  3.0.3      v dplyr  1.0.2  
## v tidyr   1.1.1      v stringr 1.4.0  
## v readr   1.3.1      v forcats 0.5.0  
  
## -- Conflicts -----  
## x dplyr::filter() masks stats::filter()  
## x dplyr::lag()    masks stats::lag()  
  
##  
## Attaching package: 'reshape2'  
  
## The following object is masked from 'package:tidyr':  
##  
## smiths
```

2.2: Examining Column Names

```
colnames(storm)
```

```
## [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" "CROPDGMG" "CROPDGMGEXP" "WFO" "STATEOFFIC"
## [31] "ZONENAMES" "LATITUDE" "LONGITUDE" "LATITUDE_E" "LONGITUDE_"
## [36] "REMARKS" "REFNUM"
```

2.3: Data Subsetting

Subset the dataset on the parameters of interest. Basically, we remove the columns we don't need for clarity.

```
# Only use data where fatalities or injuries occurred.
storm <- storm %>%
  filter(EVTYPE != "?" &
         (INJURIES > 0 | FATALITIES > 0 |
          PROPDMG > 0 | CROPDGMG > 0)) %>%
  select( c("EVTYPE", "FATALITIES", "INJURIES", "PROPDMG",
            "PROPDMGEXP", "CROPDGMG", "CROPDGMGEXP"))
```

2.4: Converting Exponent Columns into Actual Exponents instead of (-,+, H, K, etc)

Making the PROPDMGEXP and CROPDGMGEXP columns cleaner so they can be used to calculate property and crop cost.

```
# Map property damage alphanumeric exponents to numeric values.
storm$PROPDMGEXP <- plyr::mapvalues(storm$PROPDMGEXP,
  from = c("K", "M", "", "B", "m", "+", "0", "5",
            "6", "?", "4", "2", "3", "h", "7",
            "H", "-", "1", "8"),
  to = c(10^3, 10^6, 1, 10^9, 10^6, 0, 1, 10^5, 10^6,
         0, 10^4, 10^2, 10^3, 10^2, 10^7, 10^2,
         0, 10, 10^8))
```

The following 'from' values were not present in 'x': ?, 1, 8

```
storm$PROPDMGEXP <- as.numeric(as.character(storm$PROPDMGEXP))
storm$PROPDMGTOTAL <- (storm$PROPDMG * storm$PROPDMGEXP)/1000000000

# Map crop damage alphanumeric exponents to numeric values
storm$CROPDGMGEXP <- plyr::mapvalues(storm$CROPDGMGEXP,
  from = c("", "M", "K", "m", "B",
            "?", "0", "k", "2"),
  to = c(1, 10^6, 10^3, 10^6, 10^9,
         0, 1, 10^3, 10^2))
```

The following 'from' values were not present in 'x': 2

```
storm$CROPDMGEXP <- as.numeric(as.character(storm$CROPDMGEXP))
storm$CROPDMGTOTAL <- (storm$CROPDMG * storm$CROPDMGEXP)/1000000000
```

2.5: Making Economic Cost Columns

```
storm <- storm %>%
  mutate(propCost = PROPDMG * PROPDMGEXP,
         cropCost = CROPDMG * CROPDMGEXP)
```

2.6: Calculating Total Property and Crop Cost

```
totalCost <- storm %>%
  group_by(EVTYPE) %>%
  summarize(propCost = sum(propCost),
           cropCost = sum(cropCost))
```

'summarise()' ungrouping output (override with '.groups' argument)

```
head(totalCost)
```

```
## # A tibble: 6 x 3
##   EVTYPE                propCost cropCost
##   <chr>                <dbl>    <dbl>
## 1 "  HIGH SURF ADVISORY" 200000      0
## 2 " FLASH FLOOD"       50000      0
## 3 " TSTM WIND"         8100000      0
## 4 " TSTM WIND (G45)"    8000      0
## 5 "AGRICULTURAL FREEZE"      0 28820000
## 6 "APACHE COUNTY"       5000      0
```

```
totalCost <- storm %>%
  group_by(EVTYPE) %>%
  summarize(propCost = sum(propCost),
           cropCost = sum(cropCost)) %>%
  mutate(Total_Cost = propCost + cropCost) %>%
  select(EVTYPE, Total_Cost) %>%
  arrange(-Total_Cost)
```

'summarise()' ungrouping output (override with '.groups' argument)

```
totalCost <- totalCost[1:10, ]
head(totalCost, 5)
```

```
## # A tibble: 5 x 2
##   EVTYPE                Total_Cost
##   <chr>                <dbl>
## 1 FLOOD                150319678257
```

```
## 2 HURRICANE/TYPHOON 71913712800
## 3 TORNADO           57362333886.
## 4 STORM SURGE       43323541000
## 5 HAIL              18761221986.
```

2.7: Calculating Total Fatalities and Injuries

```
totalInjuries <- storm %>%
  group_by(EVTYPE) %>%
  summarize( FATALITIES = sum(FATALITIES),
             INJURIES = sum(INJURIES)) %>%
  mutate(total = FATALITIES + INJURIES) %>%
  select(EVTYPE, total ) %>%
  arrange(-total)
```

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

```
totalInjuries <- totalInjuries[1:10, ]
head(totalInjuries, 5)
```

```
## # A tibble: 5 x 2
##   EVTYPE      total
##   <chr>      <dbl>
## 1 TORNADO    96979
## 2 EXCESSIVE HEAT 8428
## 3 TSTM WIND   7461
## 4 FLOOD      7259
## 5 LIGHTNING   6046
```

3: Results

3.1: Events that are Most Harmful to Population Health

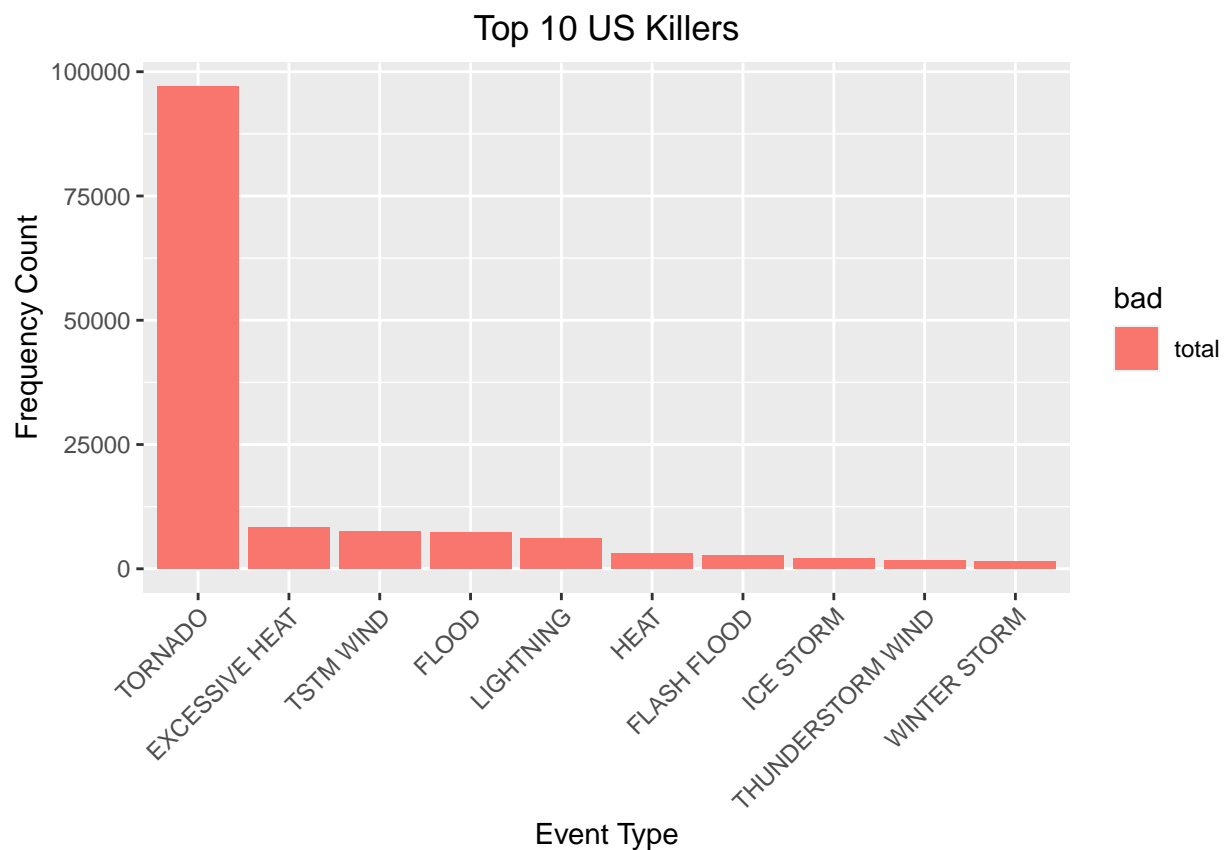
Melting data.table so that it is easier to put in bar graph format

```
bad_stuff <- melt(totalInjuries,
                  id.vars="EVTYPE",
                  variable.name = "bad")
head(bad_stuff, 5)
```

```
##           EVTYPE  bad value
## 1          TORNADO total 96979
## 2 EXCESSIVE HEAT total  8428
## 3          TSTM WIND total  7461
## 4           FLOOD total  7259
## 5    LIGHTNING total  6046
```

```
# Create chart
healthChart <- ggplot(bad_stuff,
                      aes(reorder(EVTYPE, -value), value)) +
  geom_bar(stat="identity",
          aes(fill=bad),
          position="dodge") +
  ylab("Frequency Count") +
  xlab("Event Type") +
  theme(axis.text.x = element_text(angle=45, hjust=1)) +
  ggtitle("Top 10 US Killers") +
  theme(plot.title = element_text(hjust = 0.5))
```

healthChart



3.2: Events that have the Greatest Economic Consequences

Melting data.table so that it is easier to put in bar graph format

```
econ_consequences <- melt(totalCost,
                          id.vars="EVTYPE",
                          variable.name = "Damage_Type")
head(econ_consequences, 5)
```

##	EVTYPE	Damage_Type	value
----	--------	-------------	-------

```
## 1          FLOOD  Total_Cost 150319678257
## 2 HURRICANE/TYPHOON Total_Cost 71913712800
## 3          TORNADO Total_Cost 57362333887
## 4      STORM SURGE Total_Cost 43323541000
## 5          HAIL  Total_Cost 18761221986
```

```
# Create chart
econChart <- ggplot(econ_consequences,
                    aes(reorder(EVTYPE, -value),
                        value)) +
  geom_bar(stat="identity",
          aes(fill=Damage_Type),
          position="dodge") +
  ylab("Cost (dollars)") +
  xlab("Event Type") +
  theme(axis.text.x = element_text(angle=45, hjust=1)) +
  ggtitle("Top 10 US Storm Events causing Economic Consequences") +
  theme(plot.title = element_text(hjust = 0.5))
```

```
econChart
```

