

Course Project 2

Peiyu Xiao

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An Analysis on the U.S. National Oceanic and Atmospheric Administration's Storm Dataset

Summary

This analysis employs a storm dataset collected by the U.S. National Oceanic and Atmospheric Administration (NOAA) to answer two fundamental questions:

1. Across the United States, which types of events are most harmful with respect to population health?
2. Across the United States, which types of events have the greatest economic consequences?

The results show that tornadoes cause the most damage in terms of population health while flood and drought have the most negative economic consequences in terms of property and crop damage respectively. The following sections discuss the whole analysis in details, which includes 1) data importing and processing, 2) data transformation and visualization, and 3) results.

Data Importing and Processing

```
# download data download.file(url = "https://d396qusza40orc.cloudfront.net/repdata%2Fdata%2FstormData.csv.bz2",
"E:/Data Science Specialization/Reproducible Research/Course Project 2/storm_da

# read data storm_data <- read.csv(file = "E:/Data Science Specialization/Reproducible Research/Course Project 2/storm_data.csv",
sep = ",")

# glimpse data str(storm_data)
```

```
## '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 ...
## $ BGN_AZI      : chr "" "" "" "" "" ...
## $ BGN_LOCATI: chr "" "" "" "" "" ...
## $ END_DATE : chr "" "" "" "" "" ...
## $ END_TIME : chr "" "" "" "" "" ...
## $ 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      : 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" ...
## $ CROPDGMG     : num 0 0 0 0 0 0 0 0 0 0 ...
## $ CROPDGMGEXP: 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 ...

```

`head(storm_data)`

```

##   STATE__      BGN_DATE BGN_TIME TIME_ZONE COUNTY COUNTYNAME STATE EVTYPE
## 1      1  1 4/18/1950 0:00:00      0130      CST      97      MOBILE      AL TORNADO
## 2      1  1 4/18/1950 0:00:00      0145      CST       3      BALDWIN      AL TORNADO
## 3      1  1 2/20/1951 0:00:00      1600      CST      57      FAYETTE      AL TORNADO
## 4      1  1 6/8/1951 0:00:00      0900      CST      89      MADISON      AL TORNADO
## 5      1  1 11/15/1951 0:00:00      1500      CST      43      CULLMAN      AL TORNADO
## 6      1  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 COUNTYENDN
## 1         0         0         0         0         0         0         NA
## 2         0         0         0         0         0         0         NA
## 3         0         0         0         0         0         0         NA
## 4         0         0         0         0         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			14.0	100 3	0 0 15 25.0
## 2	0			2.0	150 2	0 0 0 2.5
## 3	0			0.1	123 2	0 0 2 25.0
## 4	0			0.0	100 2	0 0 2 2.5
## 5	0			0.0	150 2	0 0 2 2.5
## 6	0			1.5	177 2	0 0 6 2.5
##	PROPDMGEXP	CROPDGMG	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		

[tail\(storm_data\)](#)

##	STATE__	BGN_DATE	BGN_TIME	TIME_ZONE	COUNTY
## 902292	47	11/28/2011 0:00:00	03:00:00 PM	CST	21
## 902293	56	11/30/2011 0:00:00	10:30:00 PM	MST	7
## 902294	30	11/10/2011 0:00:00	02:48:00 PM	MST	9
## 902295	2	11/8/2011 0:00:00	02:58:00 PM	AKS	213
## 902296	2	11/9/2011 0:00:00	10:21:00 AM	AKS	202
## 902297	1	11/28/2011 0:00:00	08:00:00 PM	CST	6
##	COUNTYNAME	STATE	EVTYPE	BGN_RANGE	

## 902292	TNZ001>004 - 019>021 - 048>055 - 088	TN	WINTER WEATHER	0
## 902293	WYZ007 - 017	WY	HIGH WIND	0
## 902294	MTZ009 - 010	MT	HIGH WIND	0
## 902295	AKZ213	AK	HIGH WIND	0
## 902296	AKZ202	AK	BLIZZARD	0
## 902297	ALZ006	AL	HEAVY SNOW	0

##	BGN_AZI	BGN_LOCATI	END_DATE	END_TIME	COUNTY_END	COUNTYENDN
## 902292			11/29/2011 0:00:00	12:00:00 PM	0	NA
## 902293			11/30/2011 0:00:00	10:30:00 PM	0	NA
## 902294			11/10/2011 0:00:00	02:48:00 PM	0	NA
## 902295			11/9/2011 0:00:00	01:15:00 PM	0	NA
## 902296			11/9/2011 0:00:00	05:00:00 PM	0	NA
## 902297			11/29/2011 0:00:00	04:00:00 AM	0	NA

##	END_RANGE	END_AZI	END_LOCATI	LENGTH	WIDTH	F	MAG	FATALITIES	INJURIES
## 902292	0			0	0	NA	0	0	0
## 902293	0			0	0	NA	66	0	0
## 902294	0			0	0	NA	52	0	0
## 902295	0			0	0	NA	81	0	0
## 902296	0			0	0	NA	0	0	0
## 902297	0			0	0	NA	0	0	0

##	PROPD	MDMG	PROPD	MGEXP	CROP	MDMG	CROP	MDMG	MGEXP	WFO	STATE	OFFIC
## 902292	0		K	0		K	MEG				TENNESSEE,	West
## 902293	0		K	0		K	RIW				WYOMING,	Central and West
## 902294	0		K	0		K	TFX				MONTANA,	Central
## 902295	0		K	0		K	AFG				ALASKA,	Northern
## 902296	0		K	0		K	AFG				ALASKA,	Northern
## 902297	0		K	0		K	HUN				ALABAMA,	North

##

902292 LAKE - LAKE - OBION - WEAKLEY - HENRY - DYER - GIBSON - CARROLL - LAUDERDALE - TIPTON - HAYWOO

902293 OWL CREEK & BRIDG

902294 NORTH ROCK

902295

902296

902297

##	LATITUDE	LONGITUDE	LATITUDE_E	LONGITUDE_
## 902292	0	0	0	0
## 902293	0	0	0	0
## 902294	0	0	0	0
## 902295	0	0	0	0
## 902296	0	0	0	0
## 902297	0	0	0	0

```
##
## 902292
## 902293
## 902294
## 902295 EPISODE NARRATIVE: A 960 mb low over the southern Aleutians at 0300AKST on the 8th intensified
## 902296 EPISODE NARRATIVE: A 960 mb low over the southern Aleutians at 0300AKST on the 8th intensified
## 902297 EPISODE NARRATIVE: An intense upper level low developed on the 28th
## REFNUM
## 902292 902292
## 902293 902293
## 902294 902294
## 902295 902295
## 902296 902296 ##
902297 902297
```

summary(storm_data)

```
## STATE__ BGN_DATE BGN_TIME TIME_ZONE
## Min. : 1.0 Length:902297 Length:902297 Length:902297
## 1st Qu.:19.0 Class :character Class :character Class :character
## Median :30.0 Mode :character Mode :character Mode :character
## Mean :31.2
## 3rd Qu.:45.0
## Max. :95.0
##
## COUNTY COUNTYNAME STATE EVTYPE
## Min. : 0.0 Length:902297 Length:902297 Length:902297
## 1st Qu.: 31.0 Class :character Class :character Class :character
## Median : 75.0 Mode :character Mode :character Mode :character
## Mean :100.6
## 3rd Qu.:131.0
## Max. :873.0
##
## BGN_RANGE BGN_AZI BGN_LOCATI END_DATE
## Min. : 0.000 Length:902297 Length:902297 Length:902297
## 1st Qu.: 0.000 Class :character Class :character Class :character
## Median : 0.000 Mode :character Mode :character Mode :character
## Mean : 1.484
## 3rd Qu.: 1.000
## Max. :3749.000
##
## END_TIME COUNTY_END COUNTYENDN END_RANGE
## Length:902297 Min. :0 Mode:logical Min. : 0.0000
## Class :character 1st Qu.:0 NA's:902297 1st Qu.: 0.0000
## Mode :character Median :0 Median : 0.0000
## Mean :0 Mean : 0.9862
## 3rd Qu.:0 3rd Qu.: 0.0000
```

```

##                               Max.      :0                               Max.      :925.0000
##
##      END_AZI                END_LOCATI                LENGTH                WIDTH
## Length:902297              Length:902297              Min.      :    0.0000              Min.      :    0.000
## Class :character            Class :character            1st Qu.:    0.0000              1st Qu.:    0.000
## Mode :character            Mode :character            Median :    0.0000              Median :    0.000
##                               Mean      :    0.2301              Mean      :    7.503
##                               3rd Qu.:    0.0000              3rd Qu.:    0.000
##                               Max.      :2315.0000              Max.      :4400.000
##
##      F                      MAG                      FATALITIES                INJURIES
## Min.      :0.0              Min.      :    0.0              Min.      : 0.0000              Min.      :    0.0000
## 1st Qu.:0.0              1st Qu.:    0.0              1st Qu.: 0.0000              1st Qu.:    0.0000
## Median :1.0              Median :    50.0              Median : 0.0000              Median :    0.0000
## Mean      :0.9              Mean      :   46.9              Mean      : 0.0168              Mean      :    0.1557
## 3rd Qu.:1.0              3rd Qu.:    75.0              3rd Qu.: 0.0000              3rd Qu.:    0.0000
## Max.      :5.0              Max.      :22000.0              Max.      :583.0000              Max.      :1700.0000
## NA's      :843563
##      PROPDMG                PROPDMGEXP                CROPDMG                CROPDMGEXP
## Min.      :    0.00 Length:902297              Min.      : 0.000 Length:902297
## 1st Qu.:    0.00 Class :character            1st Qu.: 0.000 Class :character
## Median :    0.00 Mode :character            Median : 0.000 Mode :character
## Mean      : 12.06              Mean      : 1.527
## 3rd Qu.:    0.50              3rd Qu.: 0.000
## Max.      :5000.00              Max.      :990.000
##
##      WFO                      STATEOFFIC                ZONENAMES                LATITUDE
## Length:902297              Length:902297              Length:902297              Min.      :    0
## Class :character            Class :character            Class :character            1st Qu.:2802
## Mode :character            Mode :character            Mode :character            Median :3540
##                               Mean      :2875
##                               3rd Qu.:4019
##                               Max.      :9706
##                               NA's      :47
##
##      LONGITUDE                LATITUDE_E                LONGITUDE_                REMARKS
## Min.      :-14451              Min.      :    0              Min.      :-14455              Length:902297
## 1st Qu.: 7247              1st Qu.:    0              1st Qu.:    0              Class :character
## Median : 8707              Median :    0              Median :    0              Mode :character
## Mean      : 6940              Mean      :1452              Mean      : 3509
## 3rd Qu.: 9605              3rd Qu.:3549              3rd Qu.: 8735
## Max.      :17124              Max.      :9706              Max.      :106220

```

```
##          NA's      :40
## REFNUM
## Min.      :      1
## 1st Qu.:225575
## Median :451149
## Mean      :451149
## 3rd Qu.:676723
## Max.      :902297
##
```

Data Transformation and Visualization

Disaster Events and Population Health

I reorganized the storm data by disaster events types and summarized the total number of fatalities and injuries by each type of events. Then, I plotted two barcharts to visualize the top 5 most harmful events by total fatalities and injuries respectively. Last, I combined the two charts into Figure 1.

grouping data by EVTYPE and summarizing population health related variables [library\(tidyverse\)](#)

```
## -- Attaching packages ----- tidyverse 1.3.0 --

## v ggplot2 3.3.2      v purrr      0.3.4
## v tibble 3.0.3       v dplyr      1.0.1
## v tidyr      1.1.1    v stringr 1.4.0
## v readr      1.3.1    v forcats 0.5.0
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()

storm_data_ph <- storm_data %>% mutate(event_type =
  as.factor(EVTYPE)) %>% group_by(event_type) %>%
  summarize(total_fatalities = sum(FATALITIES, na.rm = TRUE),
            total_injuries = sum(INJURIES, na.rm = TRUE))

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

```

# Top 5 most harmful events according to total number of fatalities across US top_fatalities <-
storm_data_ph %>% arrange(desc(total_fatalities)) %>% filter(row_number()<=5) %>% select(1:2)

plot1 <- top_fatalities %>% ggplot(aes(reorder(event_type, - total_fatalities), total_fatalities)) +
  geom_bar(stat = "identity") + labs(x = "", y = "Number of Fatalities",
    title = "Top 5 Most Harmful Events by Total Number of Fatalities Across US") +
  theme_bw()

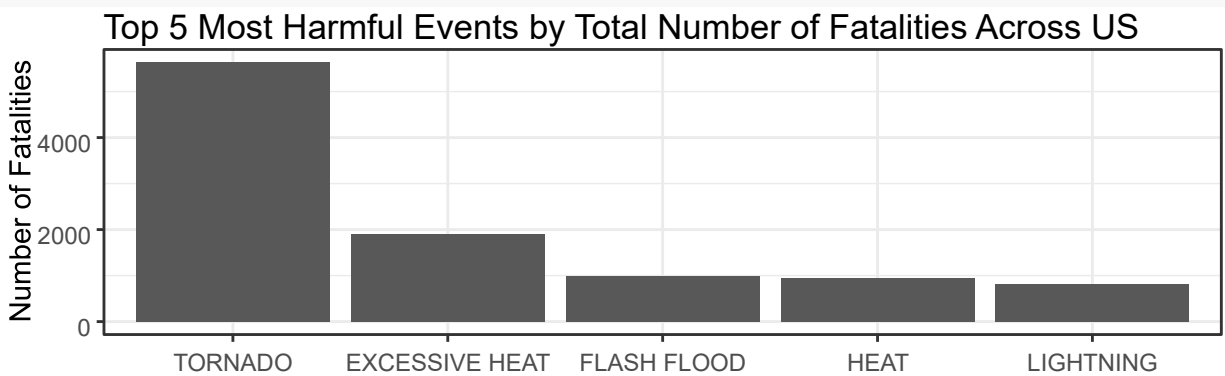
# Top 5 most harmful events according to total number of injuries across US top_injuries <-
storm_data_ph %>% arrange(desc(total_injuries)) %>% filter(row_number()<=5) %>%

select(1,3)

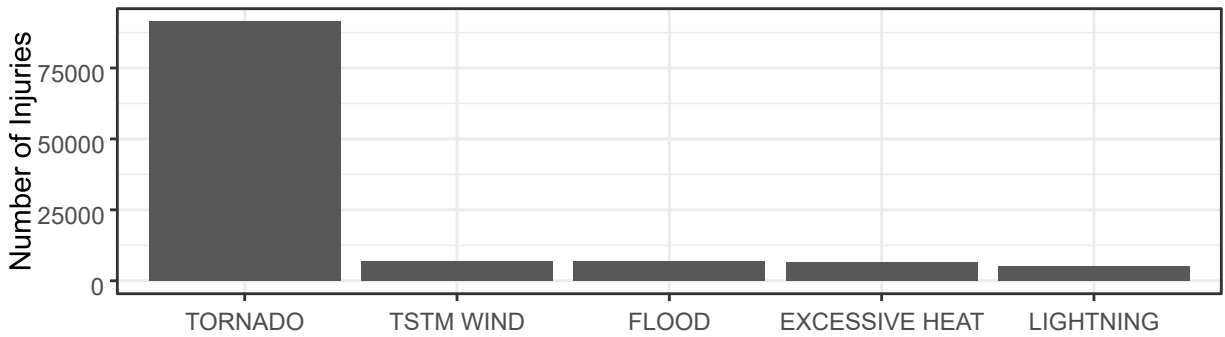
plot2 <- top_injuries %>% ggplot(aes(reorder(event_type, - total_injuries), total_injuries)) + geom_bar(stat =
  "identity") + labs(x = "", y = "Number of Injuries", title = "Top 5 Most Harmful Events by Total Number of
  Injuries Across US") +
  theme_bw()

# Figure 1 (Top 5 most harmful events with respect to population health) library("ggpubr")
ggarrange(plot1, plot2, nrow = 2)

```



Top 5 Most Harmful Events by Total Number of Injuries Across US



Disaster Events and Economic Consequences

To learn about the economic consequences of disaster events, I rearranged the storm data by disaster events types and summarized the total property and crop damage estimates by each type of events. Then, I plotted two barcharts to visualize the top 5 most harmful events by total property and crop damage estimates. The results are shown in Figure 2.

```
# grouping data by EVTYPE and summarizing economic consequences related variables storm_data_ec <-
storm_data %>% select(EVTYPE, PROPDMG, PROPDMGEXP, CROPDMG, CROPDMGEXP) %>%
```

```

mutate(property_damage_parameters = case_when(
  PROPDMGEXP == "" ~ 1,
  PROPDMGEXP == "-" ~ 1,
  PROPDMGEXP == "?" ~ 1,
  PROPDMGEXP == "+" ~ 1,
  PROPDMGEXP == "0" ~ 1,
  PROPDMGEXP == "1" ~ 10,
  PROPDMGEXP == "2" ~ 100,
  PROPDMGEXP == "3" ~ 1000,
  PROPDMGEXP == "4" ~ 10000,
  PROPDMGEXP == "5" ~ 100000,
  PROPDMGEXP == "6" ~ 1000000,
  PROPDMGEXP == "7" ~ 10000000,
  PROPDMGEXP == "8" ~ 100000000,
  PROPDMGEXP == "B" ~ 1000000000,
  PROPDMGEXP == "h" ~ 1,
  PROPDMGEXP == "H" ~ 1,
  PROPDMGEXP == "K" ~ 1000,
  PROPDMGEXP == "m" ~ 1000000, PROPDMGEXP
  == "M" ~ 1000000),
  crop_damage_parameters = case_when(
    CROPDMGEXP == "" ~ 1,
    CROPDMGEXP == "?" ~ 1,
    CROPDMGEXP == "0" ~ 1,
    CROPDMGEXP == "2" ~ 100,
    CROPDMGEXP == "B" ~ 1000000000,
    CROPDMGEXP == "k" ~ 1000,
    CROPDMGEXP == "K" ~ 1000,
    CROPDMGEXP == "m" ~ 1000000,
    CROPDMGEXP == "M" ~ 1000000),
  property_damage = PROPDMG*property_damage_parameters, crop_damage =
  CROPDMG*crop_damage_parameters, event_type = as.factor(EVTYPE)) %>%
  group_by(event_type) %>% summarize(total_property_damage =
  sum(property_damage, na.rm = TRUE),
  total_crop_damage = sum(crop_damage, na.rm = TRUE))
## 'summarise()' ungrouping output (override with 'groups' argument)

```

```

# Top 5 most harmful events according to total property damage across US top_pd <-
storm_data_ec %>% arrange(desc(total_property_damage)) %>%
filter(row_number()<=5) %>% select(1:2)

plot3 <- top_pd %>% ggplot(aes(reorder(event_type, - total_property_damage), total_property_damage)) +
  geom_bar(stat = "identity") + labs(x = "", y = "Property Damage Estimates", title = "Top 5 Most Harmful
  Events by Total Property Damage Across US") +
  theme_bw()

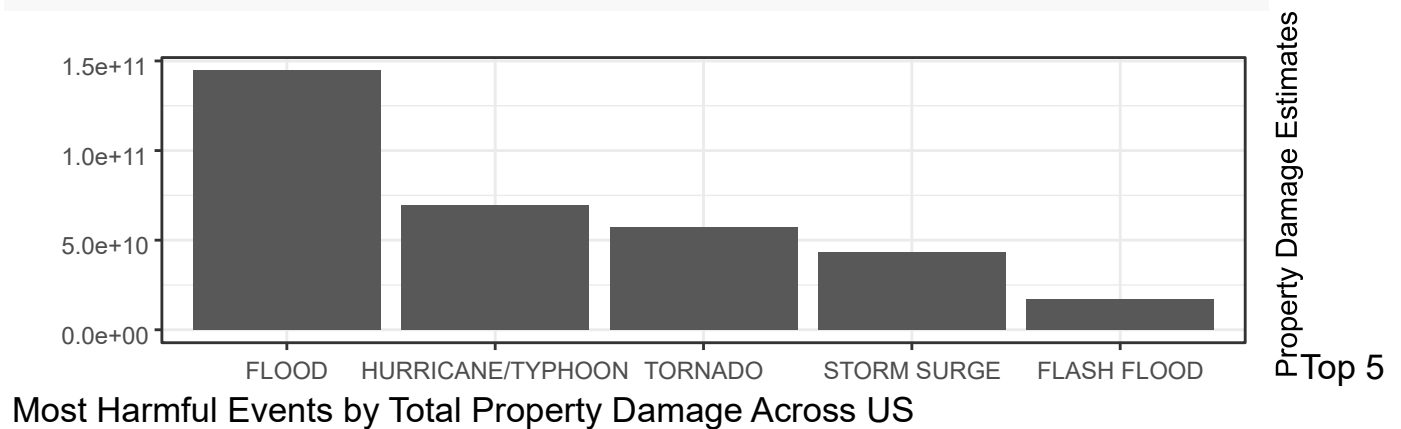
# Top 5 most harmful events according to total crop damage across US top_cd <-
storm_data_ec %>%

  arrange(desc(total_crop_damage)) %>%
  filter(row_number()<=5) %>% select(1,3)

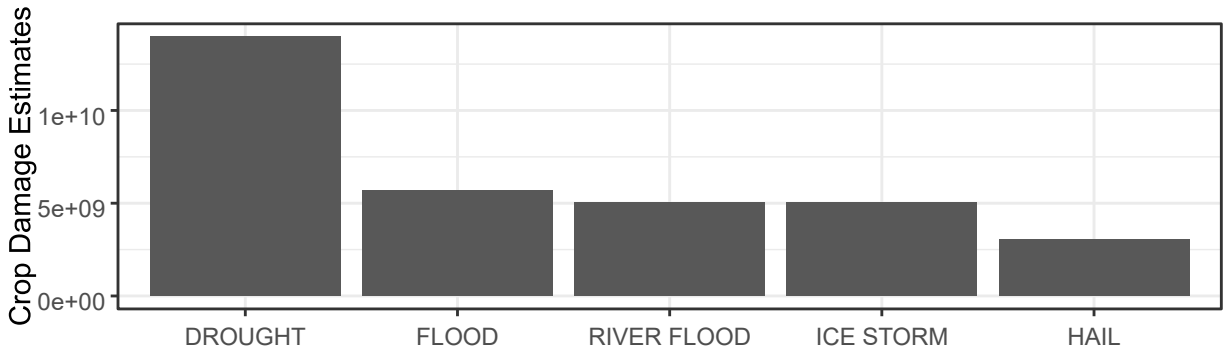
plot4 <- top_cd %>% ggplot(aes(reorder(event_type, - total_crop_damage), total_crop_damage)) +
  geom_bar(stat = "identity") + labs(x = "", y = "Crop Damage Estimates", title = "Top 5 Most
  Harmful Events by Total Crop Damage Across US") +
  theme_bw()

# Figure 2 (Top 5 most harmful events with respect to economic consequences) ggarrange(plot3, plot4, nrow = 2)

```



Top 5 Most Harmful Events by Total Crop Damage Across US



Results

Based on Figure 1, it is clear that the most harmful events with respect to population health is the tornado, which caused about 5,600 deaths and 91,000 injuries during the period between 1950 to November 2011. Figure 2 further indicates that flood and drought are most detrimental events to the economy given that the flood has brought property damages for about 0.15 trillion dollars since 1950, whereas the crop damage triggered by the drought is about 14 billion dollars until 2011.