

Midterm Mapping with more data

Zhiwei Liang, Chi Zhang

11/10/2020

1. Brief Description

The Public Assistance Funded Projects Details dataset contains obligated (financial obligation to grantee) Public Assistance projects, lists public assistance recipients designated as applicants in the data, and a list of every funded, individual project, called project worksheets. Open projects still under pre-obligation processing are not represented.

The dataset is from “<https://www.fema.gov/openfema-data-page/public-assistance-funded-projects-details-v1>”

2. Data Organization

In this project, we will focus on the hurricane, coastal storm during 2009-2018.

The following are the variables we need:

- **disasterNumber:** Sequentially assigned number used to designate an event or incident declared as a disaster.
- **declarationDate:** Date the disaster was declared.
- **incidentType:** Type of incident
- **projectAmount:** The estimated total cost of the Public Assistance grant project in dollars, without administrative costs.
- **totalObligated:** The federal share of the Public Assistance grant eligible project amount in dollars, plus grantee (State) and sub-grantee (applicant) administrative costs.
- **obligatedDate:** Date the grant was obligated.
- **dcc:** Damage category code. Category A: Debris removal; Category B: Emergency protective measures; Category C: Roads and bridges; Category D: Water control facilities; Category E: Public buildings and contents; Category F: Public utilities; Category G: Parks, recreational, and other facilities
- **projectSize:** Projects are designated as Large or Small.

Filter & Select

We filtered Hurricane and storm from disaster types and chose time and variables in this part.

```
data <- read.csv('PublicAssistanceFundedProjectsDetails.csv')

#choose disaster type
df <- data%>%filter(incidentType %in% c("Hurricane","Coastal Storm"))
df <- na.omit(df)

#choose time
df$declarationDate <- as_datetime(df$declarationDate)
df$declarationYear <- year(df$declarationDate)
df$obligatedDate <- as_datetime(df$obligatedDate)
df$obligatedYear <- year(df$obligatedDate)

#sum(ifelse(df$declarationYear==df$obligatedYear,0,1))==0
df <- df %>% filter(declarationYear > 2009 & declarationYear < 2018)

#choose variables
df <- df %>% select(-c("pwNumber","applicationTitle","applicantId","damageCategoryCode","damageCategory","hash","lastRefresh","id"))

df$county %<>% tolower()
df$state %<>% tolower()

#write.csv(df,"hur_stm.csv",row.names=FALSE)
```

Add fips,long and lat

To better find the corresponding area more conveniently, we added Fips code, latitude and longitude. Also, we separate our data into statewide and countywide by looking at the countyCode is 0 or not.

```
##### Add Fips
df= read.csv("hur_stm.csv")
df$fips=NA
for (i in 1:dim(df)[1]){
  if(df$countyCode[i]> 99){
    df$fips[i] <- str_c(as.character(df$stateNumberCode[i]),as.character(df$countyCode[i]))
  }
  else if(df$countyCode[i]>=10){
    df$fips[i] <- str_c(as.character(df$stateNumberCode[i]),"0",as.character(df$countyCode[i]))
  }
  else if(df$countyCode[i]<10){
    df$fips[i] <- str_c(as.character(df$stateNumberCode[i]),"0","0",as.character(df$countyCode[i]))
  }
}
```

```

}

#unique(df$stateNumberCode)
#there are 0 value for statewide disasters
#unique(df$countyCode)

df_statewide = df %>% filter(countyCode == 0)
df_countywide = df %>% filter(countyCode != 0)
#write.csv(df_countywide,"df_countywide.csv",row.names = FALSE)

##### GroupBy & Summarise
#df_countywide=read.csv("df_countywide.csv")
df_map=df_countywide %>%
  group_by(declarationYear,state,county,fips,incidentType,disasterNumber)%>%
  summarise(projectAmount=sum(projectAmount),totalObligated=sum(totalObligated),.groups="drop")%>%
  mutate(fips=as.numeric(fips))

MainStates <- map_data("state")
AllCounty <- map_data("county")
df_map_new = left_join(df_map,AllCounty, by=c("state"="region","county"="subregion"))
#write.csv(df_map_new,"df_map_new.csv",row.names = FALSE)
# for shiny

```

3.Mapping

Total Obligated

For this map, it shows the total obligated amount of whole country and counties by states. User can choose specific year, state and even disaster type to get different maps. There are two examples. The first one displays the obligated amount caused by Coastal Storm, in Louisiana,2011. The second one displays the total obligated amount of U.S caused by Hurricane in 2011.

```

df_map_new <- read.csv("df_map_new.csv")
obligate_map=function(myyear,mystate,mytype){

  h=df_map_new%>%filter(declarationYear==myyear,state==mystate,incidentType==mytype)
  s=sum(h$totalObligated)
  my_states <- map_data("state", region =mystate)
  my_counties <- map_data("county", region =mystate)

  ggplot() +
    geom_polygon(data=my_counties,
                 aes(x=long, y=lat, group=group),

```

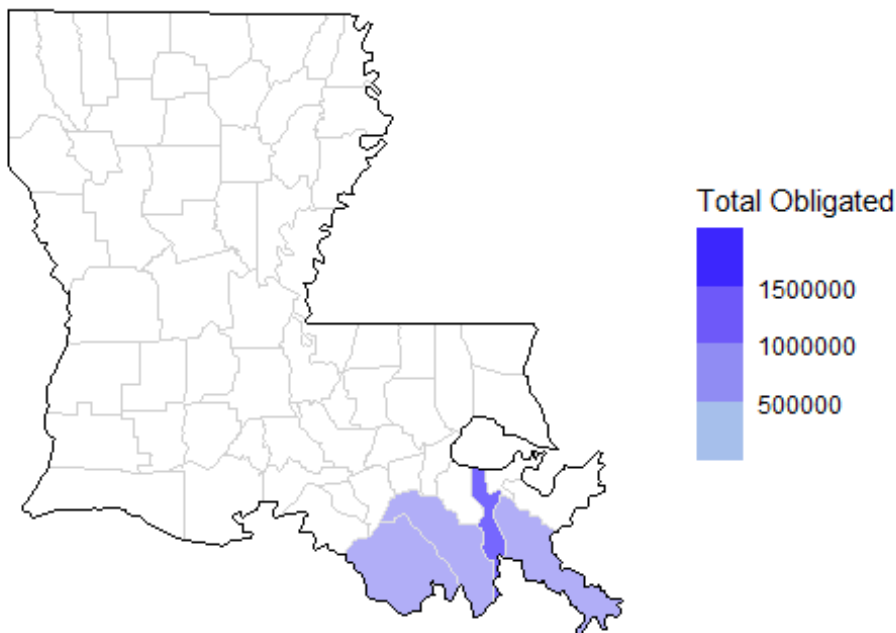
```

        color="gray", fill="white", size = .1 ) +
    geom_polygon(data = h,
        aes(x = long, y = lat, group = group, fill = `totalObligated`),
        color = "lightgrey", size = 0.2, alpha = 1.6,)+
    scale_fill_steps(low="lightblue",high="blue",name='Total Obligated',guide = "coloursteps")+
    geom_polygon(data=my_states,
        aes(x=long, y=lat, group=group),
        color="black", fill="white", size = .5, alpha = .3)+
    ggtitle(paste(myyear,mystate,mytype,"total obligated: $",s))+
    theme(plot.title=element_text(hjust=0.5),
        panel.background=element_blank(),
        panel.border=element_blank(),
        axis.title=element_blank(),
        axis.text=element_blank(),
        axis.ticks=element_blank())
}

obligate_map(2011,"louisiana","Coastal Storm")

```

louisiana Coastal Storm total obligated: \$ 345541951.2



```

## for whole country
country_obligate_map=function(myyear,mytype){

    h=df_map_new%>%filter(declarationYear==myyear,incidentType==mytype)
    s=sum(h$totalObligated)
}

```

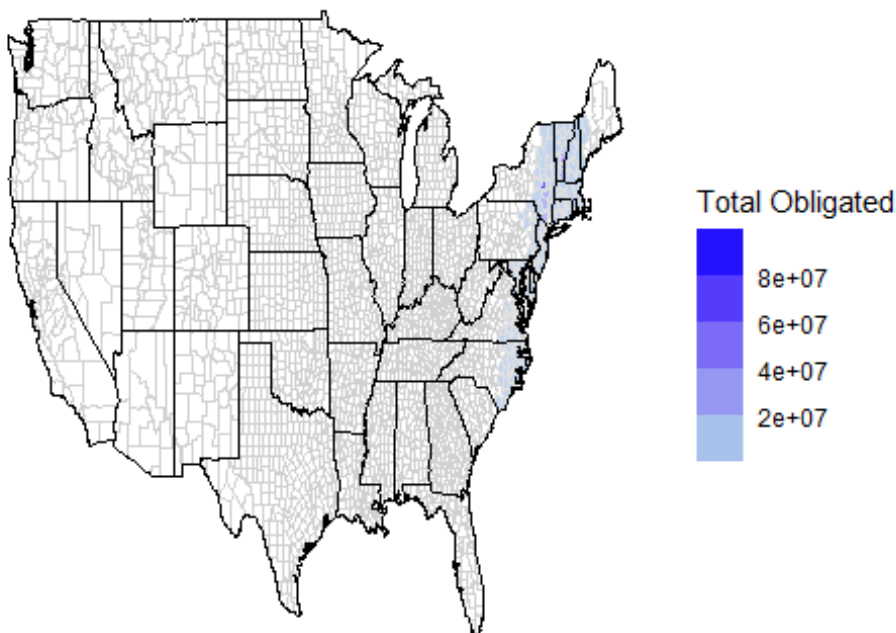
```

ggplot() +
  geom_polygon(data=AllCounty,
    aes(x=long, y=lat, group=group),
    color="gray", fill="white", size = .1 ) +
  geom_polygon(data = h,
    aes(x = long, y = lat, group = group, fill = `totalObligated`),
    color = "lightgrey", size = 0.2, alpha = 1.6,)+
  scale_fill_steps(low="lightblue",high="blue",name='Total Obligated',guide = "coloursteps")+
  geom_polygon(data=MainStates,
    aes(x=long, y=lat, group=group),
    color="black", fill="white", size = .5, alpha = .3)+
  ggtitle(paste(myyear,mytype,"total obligated: $",s))+
  theme(plot.title=element_text(hjust=0.5),
    panel.background=element_blank(),
    panel.border=element_blank(),
    axis.title=element_blank(),
    axis.text=element_blank(),
    axis.ticks=element_blank())
}

country_obligate_map(2011,"Hurricane")

```

2011 Hurricane total obligated: \$ 31681662421.9



barplot

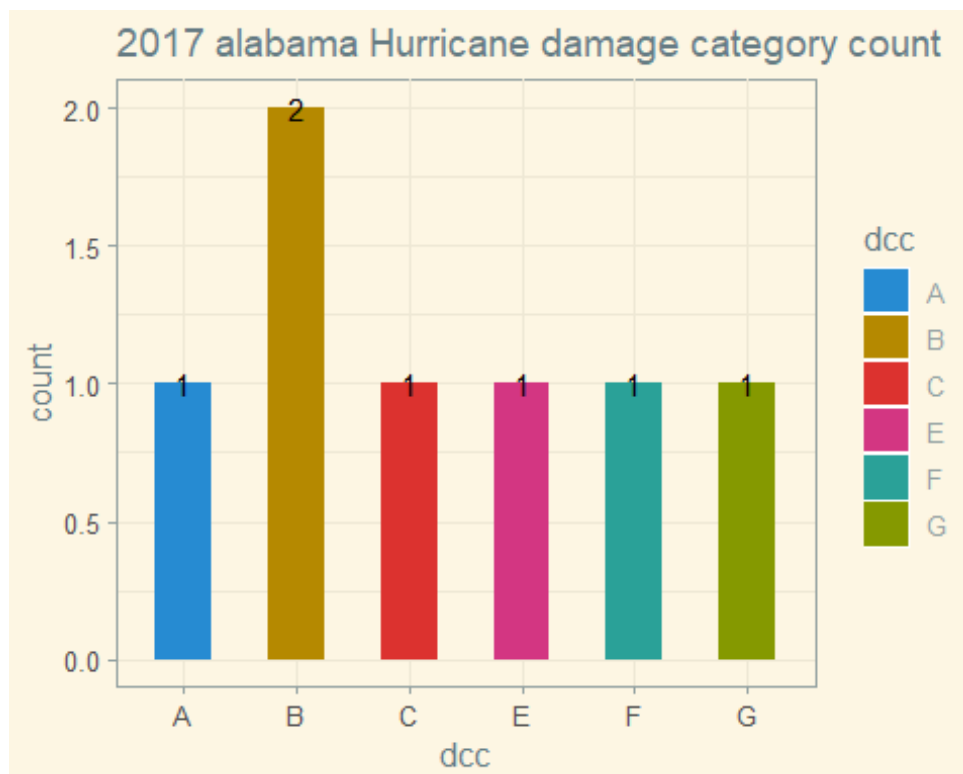
For this bar plot, it counts damage categories of one disaster in a specific year. There are two examples. The first bar plot counts the categories of Hurricane damage in Alabama, 2017. The second bar plot counts the categories of Hurricane damage in U.S, 2017.

```
library(ggthemes)

## Warning: package 'ggthemes' was built under R version 4.0.3

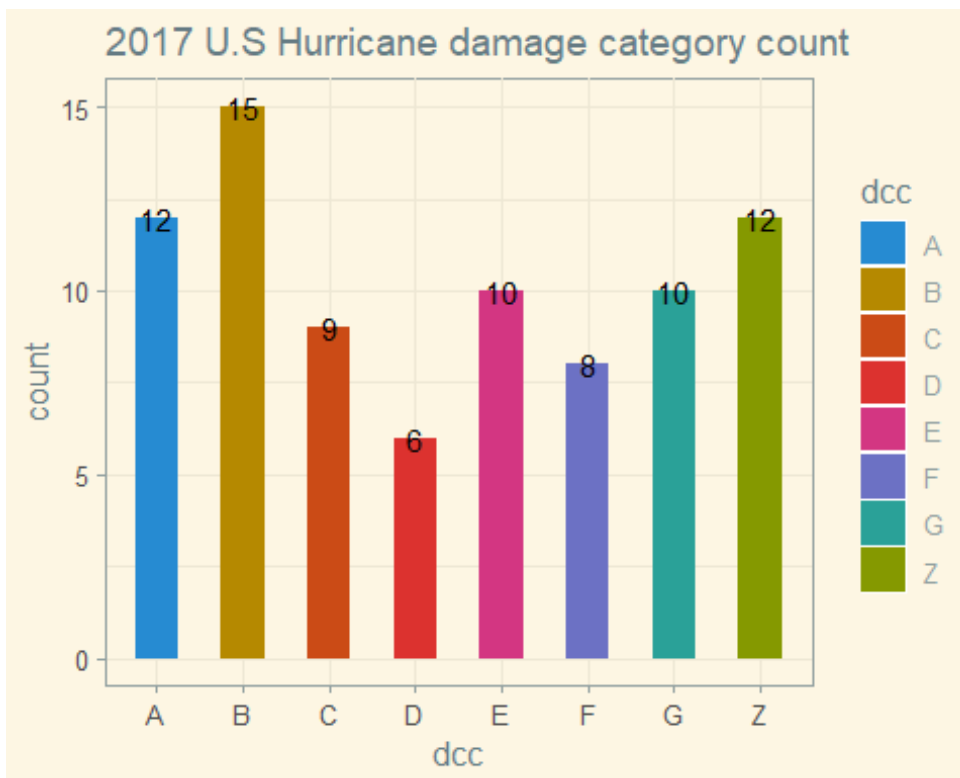
df_countywide <- read.csv("df_countywide.csv")
dcc_bar = function(myyear, mystate, mytype){
  hh= df_countywide %>% filter(declarationYear==myyear, state==mystate, incidentType==mytype)%>%
    group_by(disasterNumber)%>%
    summarise(dcc, .groups="drop")%>%
    unique()

  ggplot(hh, aes(x=dcc, fill=dcc))+
    geom_bar(stat = "count", width=0.5)+
    geom_text(aes(label=as.character(..count..)), stat="count")+
    theme_solarized()+
    scale_fill_solarized()+
    ggtitle(paste(myyear, mystate, mytype, "damage category count"))
}
dcc_bar(2017, "alabama", "Hurricane")
```



```
country_dcc_bar = function(myyear,mytype){
  hh= df_countywide %>% filter(declarationYear==myyear,incidentType==mytype)%>%
    group_by(disasterNumber)%>%
    summarise(dcc,.groups="drop")%>%
    unique()

  ggplot(hh,aes(x=dcc,fill=dcc))+
    geom_bar(stat = "count",width=0.5)+
    geom_text(aes(label=as.character(..count..)),stat="count")+
    theme_solarized()+
    scale_fill_solarized()+
    ggtitle(paste(myyear,"U.S",mytype,"damage category count"))
}
country_dcc_bar(2017,"Hurricane")
```



Shiny app

After cleaning and mapping, we build an interactive web app by Shiny to present our data and plots more visually. Here is the link to our app:
<https://zc9714.shinyapps.io/Mapping2/>