Report

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Introduction

In this assignment, we use additional county level data from FEMA to produce maps with more features. The purpose of this assignment is to show our ability of using R to produce documents, presentations, and shiny apps.

Data Collection

Data 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. This is raw, unedited data from FEMA's National Emergency Management Information System (NEMIS).

Load Data

We download raw data PublicAssistanceFundedProjectsDetails.csv (https://www.fema.gov/openfema-data-page/public-assistance-funded-projects-details-v1) from the website of FEMA and load into R. Here are the first several rows and columns of the data:

```
incident Type\ pw Number\ application Title
##
     disasterNumber
                         declarationDate
               1239 1998-08-26 04:00:00 Severe Storm(s)
## 1
                                                                        Not Provided
## 2
               1239 1998-08-26 04:00:00 Severe Storm(s)
                                                                 51
                                                                        Not Provided
               1239 1998-08-26 04:00:00 Severe Storm(s)
                                                                 43
                                                                        Not Provided
## 3
## 4
               1239 1998-08-26 04:00:00 Severe Storm(s)
                                                                 2
                                                                                  (L)
## 5
               1239 1998-08-26 04:00:00 Severe Storm(s)
                                                                 47
                                                                        Not Provided
```

There are 79192 rows and 22 columns.

Obtain Hurricane Data from 2009 to 2018

After selecting hurricane data from 2009 to 2018 and deleting the missing data, we get 79171 observations. Here are the first several rows and columns of the data:

```
##
     disasterNumber
                         declarationDate incidentType pwNumber applicationTitle
## 1
               1866 2009-12-22 05:00:00
                                             Hurricane
                                                              1
                                                                     DIW-097-02F
               1866 2009-12-22 05:00:00
                                             Hurricane
                                                              2
## 2
                                                                     DIW-097-01F
## 3
               1866 2009-12-22 05:00:00
                                             Hurricane
                                                              3
                                                                     DIW-097-03F
## 4
                                                              4
               1866 2009-12-22 05:00:00
                                            Hurricane
                                                                     DIW-097-04F
## 5
               1866 2009-12-22 05:00:00
                                            Hurricane
                                                              5
                                                                     DIW-097-01B
```

EDA

In this part, we focus on column **projectAmount** – the estimated total cost of the Public Assistance grant project in dollars, without administrative costs. This amount is based on the damage survey. We calculate the total project amount for each state, each year, different project sizes and different damage categories and display it using barplot.

Total Project Amount for Each State

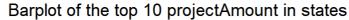
Table

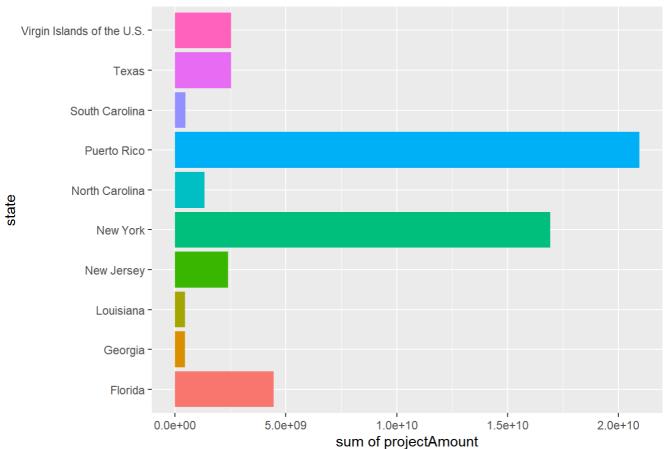
We calculate the total project amount for each state. Only 10 states with the largest number of project amount are shown. The whole table is available in Appendix.

state	sum
Puerto Rico	20936310576
New York	16927478331
Florida	4463746795
Texas	2540619000
Virgin Islands of the U.S	. 2523344651
New Jersey	2390999589
North Carolina	1325808430
South Carolina	486954380
Georgia	453679553
Louisiana	453214650

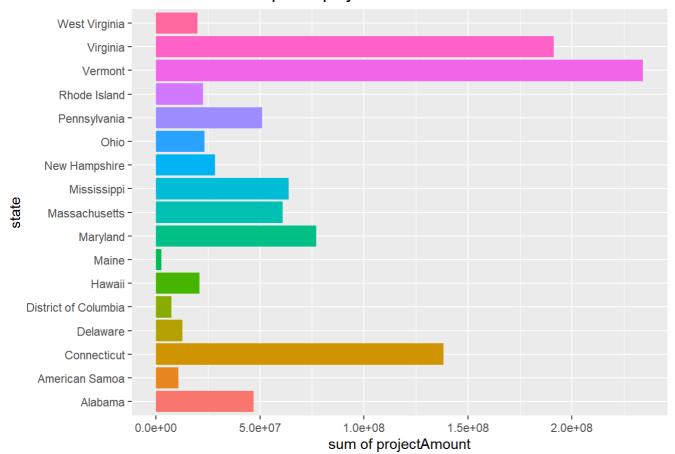
Barplot

Because the top 10 projectAmounts are too large, they are ploted in one graph. And the others are in another plot.





Barplot of projectAmount for other states



Total Project Amount for Each Year

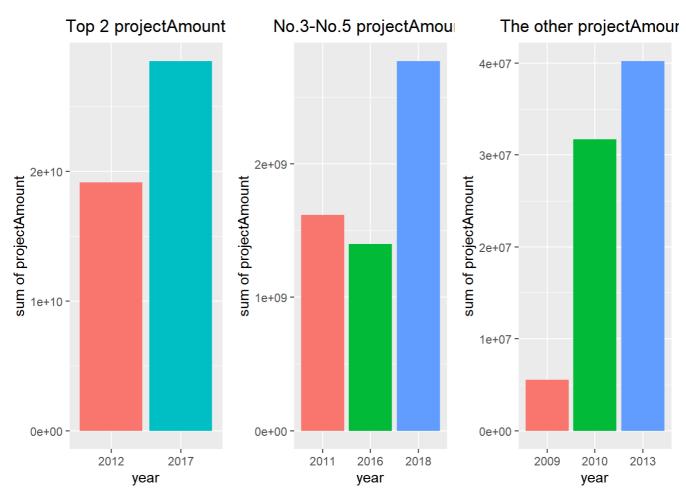
Table

We calculate the total project amount for each year.

beginyear	sum
20172	28502284412
20121	19141051210
2018	2772562070
2011	1619491113
2016	1401518239
2013	40234873
2010	31720861
2009	5547017

Barplot

Because the top 2 projectAmounts are too large, they are plotted in one graph. And the No.3-No.5 projectAmounts are in another plot for the same reason. The third plot contains the rest of projectAmounts.



Total Project Amount for Different Project Size

Table

We calculate the total project amount for different project sizes.

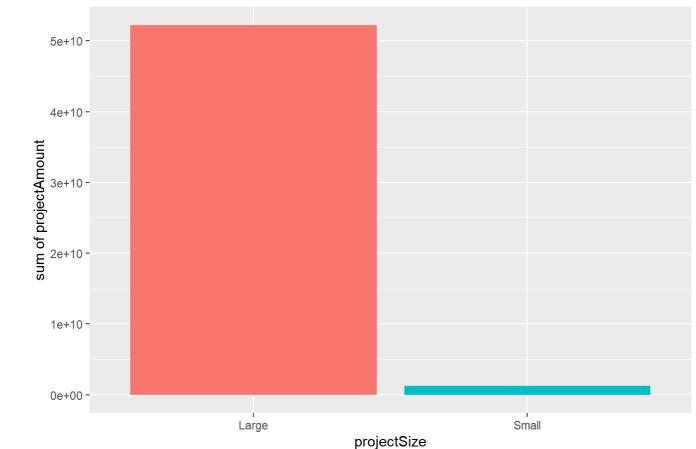
projectS	ize sum
Large	52242109567
Small	1272300227

projectSizecount

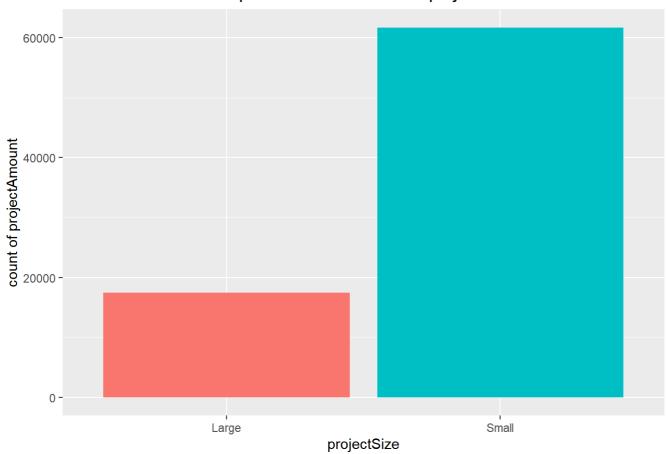
Large 17428 Small 61743

Barplot

Barplot of projectAmount for different projectSize



Barplot of count for different projectSize



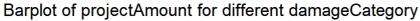
Total Project Amount for Different Damage Categories

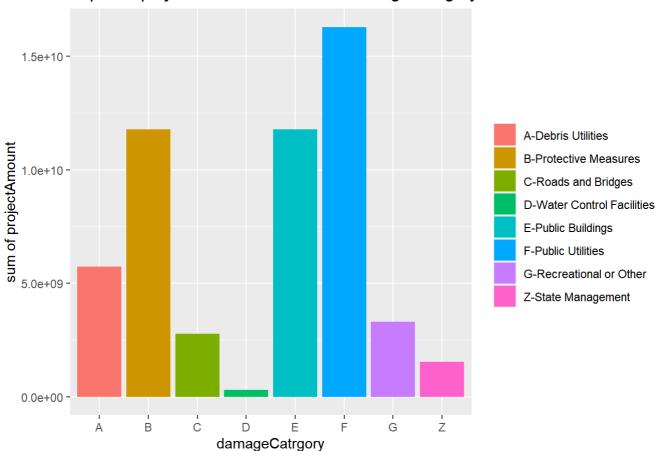
Table

We calculate the total project amount for different damage categories.

dcc	sum
F	16291752391
В	11784476587
Ε	11774539483
Α	5738378249
G	3309757035
С	2771202628
Z	1537575545
D	306727876

Barplot





Mapping with 'leaflet'

In this part, we draw maps to show the estimated total cost in states and counties.

Combine Fips Code with Maps Data

```
#combine the fips with maps data
##county
data (county. fips)
mapcounty=st as sf(map('county', plot=F, fill=T))
colnames (county. fips) [2] = colnames (mapcounty) [1]
mapcounty=left_join(mapcounty, county.fips, 'ID')
##state
data(state.fips)
mapstate=st as sf(map('state', plot = F, fill = T))
colnames (state. fips) [6] = colnames (mapstate) [1]
mapstate=left join(mapstate, state.fips, 'ID')
mapstate[20, 2:6] <-c (25, 22, 1, 1, "MA")
mapstate [21, 2:6] < -c(26, 23, 2, 3, "MI")
mapstate [31, 2:6] < -c (36, 33, 1, 2, "NY")
mapstate[32,2:6] <-c (37,34,3,5,"NC")
mapstate [45, 2:6] < -c (51, 49, 3, 5, "VA")
mapstate[46, 2:6] <-c (53, 50, 4, 9, "WA")
colnames(mapstate)[2] <- "stateNumberCode"</pre>
mapstate$stateNumberCode<- as.numeric(mapstate$stateNumberCode)</pre>
#combine data
#by state
hurricane_group<- group_by(PAFPD_0918, stateNumberCode)
h GroupByfips - summarise(hurricane group, sumcost = sum(projectAmount))
h GroupByfips - h GroupByfips[order(h GroupByfips$sumcost, decreasing=F),]
hcount_s <- right_join(mapstate, h_GroupByfips, "stateNumberCode")
#by county
hurricane group <- filter(PAFPD 0918, countyCode!=0)
hurricane_group <- group_by (hurricane_group, fips)
h_GroupByfips <- summarise(hurricane_group, sumcost = sum(projectAmount))
h_GroupByfips \( - h_GroupByfips[order(h_GroupByfips$sumcost, decreasing=F), ]
hcount_c <- right_join(mapcounty, h_GroupByfips, "fips")</pre>
hcount_c <- filter(hcount_c, is. na(ID) == F)
hcount_c2 <- filter(hcount_c, sumcost<200000000)
hcount c3 <- filter(hcount c, sumcost>200000000)
```

Maps of States

We draw the total cost of every state in map.





As shown in the picture, hurricanes mainly influence the east coast. New York has the largest estimated total cost in 2009-2018.

Maps of Counties

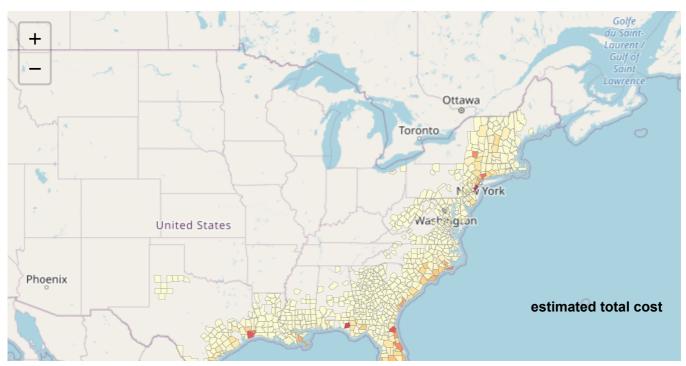
If we put the top 10 counties with the largest estimated total cost in the same graph with other counties, the differences among counties will be hard to find.

So we draw the top 10 counties alone and it appears that they are still hard to find in map because of their tiny area.

As a result, we simply list them as follows:

county	funding
7 new york, new york	9800666891
6 new york,nassau	1812897932
8 new york,queens	1181782058
10texas,harris	815788676
1 florida,bay	696292642
3 florida,miami-dade	382560652
4 new jersey, monmouth	317317875
5 new jersey,ocean	277788906
9 new york,suffolk	248392959
2 florida,broward	203482941

And the rest counties are drawn in this map:





Shiny

In shiny we not only display the 'projectAmount', but also show the number of hurricanes from 2009 to 2018.

- You can choose from the sidebar to see the mapping or graphs.
- Mapping are displayed in two scale:county level and state level. Buttons are at the topleft of mapping box.
- In Control box you can choose which period and which state to display.
- Some graphs also provide slide bar to choose period or certain year.

Click here go to shiny app (https://yuzhe.shinyapps.io/MA615-midtermmap/)

Appendix

Total Project Amount for Each State

##		state	sum
##		Puerto Rico	
##			16927478331
##	3	Florida	4463746795
##	4	Texas	2540619000
##	5	Virgin Islands of the U.S.	2523344651
##	6	New Jersey	2390999589
##	7	North Carolina	1325808430
##	8	South Carolina	486954380
##	9	Georgia	453679553
##	10	Louisiana	453214650
##	11	Vermont	234192622
##	12	Virginia	191292785
##	13	Connecticut	138181734
##	14	Maryland	77073165
##	15	Mississippi	63891963
##	16	Massachusetts	60956986
##	17	Pennsylvania	51092293
##	18	Alabama	46862086
##	19	New Hampshire	28459543
##	20	Ohio	23355813
##	21	Rhode Island	22589632
##	22	Hawaii	20814535
##	23	West Virginia	19854405
##	24	Delaware	12779360
##	25	American Samoa	10747350
##	26	District of Columbia	7448842
##	27	Maine	2660726