Assignment 4: Joins and Visualization

WenLi

2020/10/3

Problem 1

This problem will involve the nycflights13 dataset (including tables airlines, airports, planes and weather), which we saw in class. Start by installing and importing the dataset to your chosen platform. We will first use joins to search and manipulate the dataset, then we will produce a flightpath visualization.

a. Filter the dataset (using a left join) to display the tail number, year, month, day, hour, origin, and humidity for all flights heading to Tampa International Airport (TPA) on the afternoon of November 1, 2013.

```
## # A tibble: 7 x 7
##
     tailnum year month
                            day hour origin humid
##
             <int> <int> <dbl> <chr>
     <chr>
## 1 N580JB
              2013
                      11
                              1
                                   14 JFK
                                               63.1
## 2 N337NB
              2013
                              1
                                   14 LGA
                                              56.5
                       11
## 3 N567UA
              2013
                       11
                              1
                                   15 EWR
                                               52.8
                                   14 JFK
## 4 N515MQ
              2013
                       11
                              1
                                              63.1
## 5 N779JB
              2013
                       11
                              1
                                   15 EWR
                                              52.8
## 6 N561JB
              2013
                                   16 LGA
                                               50.6
                       11
                              1
## 7 N974DL
              2013
                                   18 JFK
                                              74.8
```

- b. What is the difference between the following two joins? anti_join(flights, airports, by = c("dest" = "faa")) anti_join(airports, flights, by = c("faa" = "dest"))
 - anti_join(flights, airports, by = c("dest" = "faa")): this operation will drop from table flights all observations that have a match with the condition ("dest" = "faa") in table airports. the results is a subset of table flights.
 - anti_join(airports, flights, by = c("faa" = "dest")): this operation will drop from table airports all observations that have a match with the condition ("dest" = "faa") in table flights the results is a subset of table airports

c. Select the origin and destination airports and their latitude and longitude for all fights in the dataset (using one or more inner joins).

```
Result = MyFlights %>%
  select (origin, dest) %>%
  inner_join(select(airports, faa, origin_lat=lat, origin_lon=lon), by = c("origin" = "faa")) %>%
  inner_join(select(airports, faa, dest_lat=lat, dest_lon=lon), by = c("dest" = "faa"))
as_tibble(Result)
## # A tibble: 329,174 x 6
##
      origin dest origin_lat origin_lon dest_lat dest_lon
##
      <chr>
             <chr>
                         <dbl>
                                    <dbl>
                                             dbl>
                                                       <dbl>
   1 EWR
             IAH
                         40.7
                                    -74.2
                                              30.0
                                                      -95.3
##
    2 LGA
                         40.8
                                    -73.9
                                              30.0
                                                      -95.3
##
             IAH
##
    3 JFK
             MIA
                         40.6
                                    -73.8
                                              25.8
                                                      -80.3
##
   4 LGA
                         40.8
                                    -73.9
                                              33.6
                                                      -84.4
             ATL
   5 EWR
                                    -74.2
                                              42.0
                                                      -87.9
##
             ORD
                         40.7
   6 EWR
             FLL
                         40.7
                                    -74.2
                                              26.1
                                                       -80.2
##
##
   7 LGA
             IAD
                         40.8
                                    -73.9
                                              38.9
                                                      -77.5
##
  8 JFK
             MCO
                         40.6
                                    -73.8
                                              28.4
                                                      -81.3
## 9 LGA
             ORD
                         40.8
                                    -73.9
                                              42.0
                                                      -87.9
## 10 JFK
             PBI
                         40.6
                                    -73.8
                                              26.7
                                                       -80.1
## # ... with 329,164 more rows
```

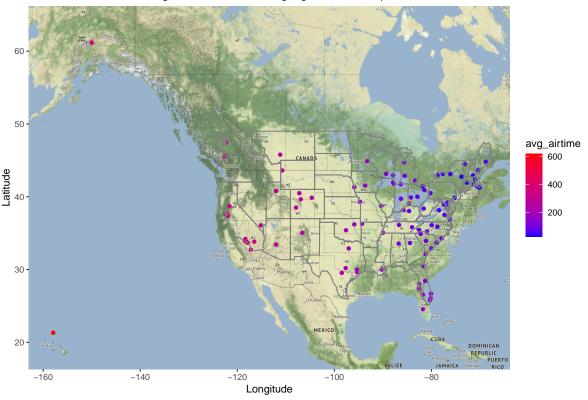
d. Use groupby and count to get the number of flights to each unique origin/destination combination.

```
Result = MyFlights %>%
         group_by (origin, dest) %>%
         dplyr::count (origin, dest)
as_tibble (Result)
## # A tibble: 224 x 3
##
      origin dest
##
      <chr> <chr> <int>
##
   1 EWR
             ALB
                      439
##
    2 EWR
             ANC
                        8
##
    3 EWR
             ATL
                     5022
##
   4 EWR
             AUS
                      968
##
   5 EWR
             AVL
                      265
##
    6 EWR
             BDL
                      443
##
    7 EWR
             BNA
                     2336
##
   8 EWR
             BOS
                     5327
##
  9 EWR
             BQN
                      297
## 10 EWR
                      931
             BTV
## # ... with 214 more rows
```

e. Produce a map that colors each destination airport by the average air time of its incoming flights. Here is a code snippet to draw a map of all flight destinations, which you can use as a starting point. You may need to install the maps packages if you have not already. Adjust the title, axis labels and aesthetics to make this visualization as clear as possible.

```
# Get average time and join the tables
AvgairTime = select(flights, dest, air_time) %>%
```

Average air time of incoming flights for the airports



Problem 2

You may recall on the lecture on Friday, Sep 25 (when we had Dr. Ofer Amram as a guest speaker), the warm-up question that day was to type in the city and state (or city and country) where you grew up in. The result of that warm-up question is summarized in a 2 file that is posted under Lecture/Sep 25. The task you have for this problem is to visualize that list on a world map, indicating in some way the cities. You could use the same mark to denote the cities. Try to make your visualization as nice looking as possible.

```
register_google(key = APIKey)
# Get the coordinates of all locations
```

```
= read_rtf("warmUpQuestion-Sep25.RTF")
GupLocDf = data.frame(Location=GupLoc, stringsAsFactors=FALSE)
GupLocMap = geocode(location=GupLoc, output="more", source="google")
GupLocMap = cbind(GupLocDf, GupLocMap) %>%
            select (Location, Longtitude=lon, Latitude=lat)
head (GupLocMap)
##
                 Location Longtitude Latitude
## 1
       Seoul, South Korea 126.97797 37.56654
## 2
      Olympia, Washington -122.90070 47.03787
## 3
           Lucknow ,India
                            80.94617 26.84669
## 4
             Richland, WA -119.27520 46.28042
## 5 Chandpur, Bangladesh
                            90.66307 23.23210
       Lahore, Pakistan.
                            74.35875 31.52037
## 6
# Get map box according to the coordinates
MapBox = c(min(GupLocMap$Longtitude)-20, min(GupLocMap$Latitude)-20,
           max(GupLocMap$Longtitude)+20, max(GupLocMap$Latitude)+20)
Map = get_map(location=MapBox, source = "stamen", maptype = "toner", zoom = 3)
#Draw the map
ggmap(Map) +
  coord_fixed(ratio = 1.5) +
  geom_point(data=GupLocMap, aes(x=Longtitude, y=Latitude), color="red") +
  labs(title="Locations of growing up", x="Longitude", y="Latitude") +
  theme(plot.title = element_text(hjust = 0.5))
```

Locations of growing up



Problem 3

Create a word cloud for an interesting (relatively short, say a couple pages) document of your own choice. Examples of suitable documents include: summary of a recent project you are working or have worked on; your own recent Statement of Purpose or Research Statement or some other similar document.

```
# Load texts
Texts = readLines("PCAIntroduction.txt")
Docs = Corpus(VectorSource(Texts))
# Clean the data
Docs = Docs %>%
  tm_map(removeNumbers) %>%
  tm_map(removePunctuation) %>%
  tm_map(stripWhitespace) %>%
  tm_map(content_transformer(tolower)) %>%
  tm_map(removeWords, stopwords("english"))
# Create a document-term-matrix
DocWords = Docs %>%
  TermDocumentMatrix() %>%
  as.matrix() %>%
 rowSums() %>%
  sort(decreasing=TRUE)
Df = data.frame(word = names(DocWords), freq=DocWords)
head (Df)
##
                              word freq
## analysis
                          analysis
                                     24
## data
                              data
                                     18
## pca
                               pca
                                     14
## dependence
                        dependence
                                     11
## interprocedural interprocedural
                                     10
# Generate the word cloud
layout(matrix(c(1, 2), nrow=2), heights=c(1, 4))
par(mar=rep(0, 4))
plot.new()
text(x=0.5, y=0.5, "Introduction of PCA: a new program analysis tool")
wordcloud(words = Df$word, freq = Df$freq, min.freq = 1,
          max.words=200, random.order=FALSE, rot.per=0.35, colors=brewer.pal(4, "Dark2"))
```

