Analysis of the effect of PM10 Concentration, Temperature, Geographic Location, Population, and distance to Electricity-Generation Combustion Points on the concentration of fine particulate matter (PM2.5) in North Carolina during the year 2018.

https://github.com/fr55/DataAnalytics_FinalProject

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Abstract

Experimental overview. This section should be no longer than 250 words.

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1   Research Question and Rationale

Nowadays air pollution is one of the most relevant health issues in the world. It refers to the contamination of the air by chemicals, biological materials, and other types of pollutants that are harmful to human health. To solve the problem of air pollution, it's necessary to understand the problem, what are the causes, and search for solutions based on the findings.

Particulate matter with a diameter of less than 2.5 micrometers is called PM2.5, and it is a extremely harmful air pollutant because it consists of particles with diameters that are less than or equal to 2.5 microns in size, which can get deeply into the lung, and ultimately impair lung function.

This study focus on trying to understand how PM2.5 concentration in North Carolina vary with temperature, PM10 concentration, zoning (piedmont, coastal, mountain), population, elevation, and distance to combustion points for electricity generation. This last variable was included because acording to the EPA combustion for electricity generation is the major point-source sector for PM2.5 in the USA (EPA, 2019).

The research question is: What are the effects of temperature, PM10 concentration, zoning (piedmont, coastal, mountain), population, elevation, distance to combustion points for electricity generation, in PM2.5 concentrations within North Carolina in the year 2018?

2 Dataset Information

For the analysis the following datasets were considered:

2.1 EPA PM2.5 Dataset

This dataset contains data from air quality monitoring of PM2.5 in North Carolina in 2018, and it was obtained using the Download Daily Data Tool in the United States Environmental Protection Agency (EPA) webpage https://www.epa.gov/outdoor-air-quality-data/download-daily-data where the options showed in Table 1 were selected:

Option	Selection
Pollutant	PM2.5
Year	2018
Geographic Area	North Carolina
Monitor Site	All Sites
Download	Download CSV (spreadsheet)

Table 1: Selections

The downloaded file was saved in the project folder path ./Data/Raw/ as EPAair_PM25_NC2018_raw.csv on 2019-03-31.

2.1.1 Data Content Information

The dataset contains daily mean PM2.5 concentration in ug/m3 in 2018. Data from 24 stations in 21 different counties of North Carolina with their location in NAD83 lat/long coordinates.

The dataset contains 19 columns, which are shown in Table 2. Column names without description are self-explanatory.

2.2 EPA PM10 Dataset

This dataset contains data from air quality monitoring of PM10 in North Carolina in 2018, and it was obtained using the Download Daily Data Tool in the United States Environmental Protection Agency (EPA) webpage https://www.epa.gov/outdoor-air-quality-data/download-daily-data where the options showed in Table 3 were selected:

The downloaded file was saved in the project folder path ./Data/Raw/ as EPAair_PM10_NC2018_raw.csv on 2019-03-31.

Column	Description
Date	mm/dd/YY
Source	AQS (Air Quality System)
Site ID	A unique number identifying the site.
POC	"Parameter Occurrence Code", distinguishes different instruments that measure the same parameter at the same site.
Daily Mean PM2.5 Concentration	
Units	Concentration Units
DAILY_AQI_VALUE	AQI = Air quality index
Site Name	
DAILY_OBS_COUNT	
PERCENT_COMPLETE	
AQS_PARAMETER_CODE	
AQS_PARAMETER_DESC	
CBSA_CODE	
CBSA_NAME	
STATE_CODE	
COUNTY CODE	A unique number identifying the County.
COUNTY	
SITE_LATITUDE	NAD83
SITE_LONGITUDE	NAD83

Table 2: Dataset content

2.2.1 Data Content Information

The dataset contains daily mean PM10 concentration in ug/m3 in 2018. Data from 9 stations in 8 different counties of North Carolina with their location in NAD83 lat/long coordinates.

The dataset contains 19 columns, which are shown in Table 4. Column names without description are self-explanatory.

2.3 NOAA Average Temperature Dataset

This dataset contains data from temperature monitoring in North Carolina in 2018, and it was obtained using the Data Search Tool in the National Center for Environmental Information of the National Oceanic and Atmospheric Administration (NOAA). Webpage https://www.ncdc.noaa.gov/cdo-web. Options showed in Table 5 were selected: XXXXXXArreglar

The downloaded file was saved in the project folder path ./Data/Raw/ as NOAA_TAVG_NC2018_raw.csv on 2019-03-28.

Option	Selection
Pollutant	PM10
Year	2018
Geographic Area	North Carolina
Monitor Site	All Sites
Download	Download CSV (spreadsheet)

Table 3: Selections

2.3.1 Data Content Information

The dataset contains daily mean air temperature in Farenheit in 2018. Data from 39 stations in North Carolina with their location in NAD83 lat/long coordinates. No county information.

The dataset contains 7 columns, which are shown in Table 6. Column names without description are self-explanatory.

2.4 US Census Bureau US counties shapefile

This dataset contains geographic and geometric information of all the counties of the US. The data is in NAD83 lat/long coordinates. De file was provided by John Fay in the Environmental Data Analytics (ENV 872L) course at Duke University, Spring 2019.

The files containing the information were saved in the project folder path ./Data/Spatial/ as cb_2017_us_county_20m on 2019-03-28.

2.4.1 Data Content Information

The dataset contains geographic and geometric information of all the counties of the US in NAD83 lat/long coordinates.

The dataset contains 10 columns, which are shown in Table 7. Column names without description are self-explanatory.

2.5 EPA combustion points for electricity generation in the US Dataset

This dataset contains facility-level locations for combustion points for electricity generation in the US, and it was obtained from the United States Environmental Protection Agency (EPA) webpage https://www3.epa.gov/air/emissions/where.htm. The Top PM2.5 emitting sectors link was selected.

The downloaded file was saved in the project folder path ./Data/Raw/ as EPA_ElecGenComb_US_raw.kml on 2019-03-31.

Column	Description
Date	mm/dd/YY
Source	AQS (Air Quality System)
Site ID	A unique number identifying the site.
POC	"Parameter Occurrence Code", distinguishes different instruments that measure the same parameter at the same site.
Daily Mean PM10 Concentration	
Units	Concentration Units
DAILY_AQI_VALUE	AQI = Air quality index
Site Name	
DAILY_OBS_COUNT	
PERCENT_COMPLETE	
AQS_PARAMETER_CODE	
AQS_PARAMETER_DESC	
CBSA_CODE	
CBSA_NAME	
STATE_CODE	A unique number identifying the County.
COUNTY CODE	A unique number identifying the County.
COUNTY	
SITE_LATITUDE	NAD83
SITE_LONGITUDE	NAD83

Table 4: Dataset content

2.5.1 Data Content Information

The dataset is a kml file that contains combustion points for electricity generation in the US. The data is in WGS84 lat/long coordinates.

All data sets, variable, and files are named according to the following naming convention: *databasename_datatype_details_stage.format*, where:

- databasename refers to the database from where the data originated

- datatype is a description of data

- details are additional descriptive details, particularly important for processed data

- stage refers to the stage in data management pipelines (e.g., raw, cleaned, temp or processed)

2.6 Analized data structure

With these datasets an exploratory data analysis was done and for the study. The datasets were wrangled and a file called PM2.5_Full_Elev_utm.shp was created, which has the data

Option	Selection
Pollutant	PM10
Year	2018
Geographic Area	North Carolina
Monitor Site	All Sites
Download	Download CSV (spreadsheet)

Table 5: Selections

Column	Description
STATION	A unique code identifying the site.
NAME	Station Name
Site ID	A unique number identifying the site.
LATITUDE	NAD83
LONGITUDE	NAD83
DATE	dd/mm/YY
TAVG	Daily Average Temperature in °F

Table 6: Dataset content

structure shown in Table 8.

Column	Description
STATEFP	A unique number identifying the State.
COUNTYFP	County Federal Information Processing Standards (FIPS) Code
COUNTYNS	Provides the American National Standards Institute (ANSI) code for the county or equivalent entity, as used by GNIS.
AFFGEOID	AFF Summary Level Code
GEOID	NAD83
NAME	County Name
LSAD	Legal/statistical area description
ALAND	County Land Area in square meters
AWATER	County Water Area in square meters
Geometry	Geometry and geographic information

Table 7: Dataset content

Variable	Units	N.Elements	Range	Source.File
Date	YY-mm-dd	343	From 2018-01-01 to 2018-12-09	EPAair_PM25_NC2018_raw.csv
Site_ID	-	24	-	EPAair_PM25_NC2018_raw.csv
COUNTY	-	21	-	EPAair_PM25_NC2018_raw.csv
Population	People	21	From 5,507 to 1,034,290	https://en.wikipedia.org/
Zone	-	3	Coastal, Piedmont, and Mountains	NC County Maps
PM2_5	ug/m3	6499	From -2.5 to 34.2	EPAair_PM25_NC2018_raw.csv
PM10	ug/m3	926	From 0 to 35	EPAair_PM10_NC2018_raw.csv
TAVG	Farenheit	4011	From 11 to 87	NOAA_TAVG_NC2018_raw.csv
Emiss_Dist	meters	24	From 813.5 to 81800.9	Self made
Elevation	meters	24	From 0.04 to 1418.8	Package elevatr

Table 8: Summary of data structure

3 Exploratory Data Analysis and Wrangling

3.1 EPA PM2.5 and PM10 Datasets

Uploading PM2.5 and PM10 2018 raw data files associated with EPA Air dataset and format date column.

```r
EPA_AQPM25_NC2018_raw <- read.csv("./Data/Raw/EPAair_PM25_NC2018_raw.csv")
EPA_AQPM10_NC2018_raw <- read.csv("./Data/Raw/EPAair_PM10_NC2018_raw.csv")

#Formatting Dates
EPA_AQPM25_NC2018_raw$Date <- as.Date(EPA_AQPM25_NC2018_raw$Date, format = "%m/%d/%Y")
EPA_AQPM10_NC2018_raw$Date <- as.Date(EPA_AQPM10_NC2018_raw$Date, format = "%m/%d/%Y")
```

Data exploration of the PM2.5 and PM10 2018 raw data files associated with EPA Air dataset.

```r
dim(EPA_AQPM25_NC2018_raw)
```

```
## [1] 9644   20
```

```r
dim(EPA_AQPM10_NC2018_raw)
```

```
## [1] 2905   20
```

```r
str(EPA_AQPM25_NC2018_raw)
```

```
## 'data.frame':    9644 obs. of  20 variables:
##  $ Date                      : Date, format: "2018-01-02" "2018-01-05" ...
##  $ Source                    : Factor w/ 2 levels "AirNow","AQS": 2 2 2 2 2 2 2 2
##  $ Site.ID                   : int  370110002 370110002 370110002 370110002 37011
##  $ POC                       : int  1 1 1 1 1 1 1 1 1 1 ...
##  $ Daily.Mean.PM2.5.Concentration: num  2.9 3.7 5.3 0.8 2.5 4.5 1.8 2.5 4.2 1.7 ...
##  $ UNITS                     : Factor w/ 1 level "ug/m3 LC": 1 1 1 1 1 1 1 1 1 1 1
##  $ DAILY_AQI_VALUE           : int  12 15 22 3 10 19 8 10 18 7 ...
##  $ Site.Name                 : Factor w/ 25 levels "","Blackstone",..: 15 15 15 1
##  $ DAILY_OBS_COUNT           : int  1 1 1 1 1 1 1 1 1 1 ...
##  $ PERCENT_COMPLETE          : num  100 100 100 100 100 100 100 100 100 100 ...
##  $ AQS_PARAMETER_CODE        : int  88502 88502 88502 88502 88502 88502 88502 885
##  $ AQS_PARAMETER_DESC        : Factor w/ 2 levels "Acceptable PM2.5 AQI & Speciat
##  $ CBSA_CODE                 : int  NA NA NA NA NA NA NA NA NA NA ...
##  $ CBSA_NAME                 : Factor w/ 14 levels "","Asheville, NC",..: 1 1 1 1
##  $ STATE_CODE                : int  37 37 37 37 37 37 37 37 37 37 ...
##  $ STATE                     : Factor w/ 1 level "North Carolina": 1 1 1 1 1 1 1 1
##  $ COUNTY_CODE               : int  11 11 11 11 11 11 11 11 11 11 ...
##  $ COUNTY                    : Factor w/ 21 levels "Avery","Buncombe",..: 1 1 1 1
##  $ SITE_LATITUDE             : num  36 36 36 36 36 ...
##  $ SITE_LONGITUDE            : num  -81.9 -81.9 -81.9 -81.9 -81.9 ...
```

```r
str(EPA_AQPM10_NC2018_raw)
```

```
## 'data.frame':    2905 obs. of  20 variables:
##  $ Date                      : Date, format: "2018-01-01" "2018-01-02" ...
##  $ Source                    : Factor w/ 1 level "AQS": 1 1 1 1 1 1 1 1 1 1 1 ...
##  $ Site.ID                   : int  370510009 370510009 370510009 370510009 370510
##  $ POC                       : int  3 3 3 3 3 3 3 3 3 3 ...
##  $ Daily.Mean.PM10.Concentration: int  3 9 15 12 12 10 9 15 22 15 ...
##  $ UNITS                     : Factor w/ 1 level "ug/m3 SC": 1 1 1 1 1 1 1 1 1 1 1 .
##  $ DAILY_AQI_VALUE           : int  3 8 14 11 11 9 8 14 20 14 ...
##  $ Site.Name                 : Factor w/ 9 levels "Durham Armory",..: 9 9 9 9 9 9
##  $ DAILY_OBS_COUNT           : int  1 1 1 1 1 1 1 1 1 1 ...
##  $ PERCENT_COMPLETE          : num  100 100 100 100 100 100 100 100 100 100 ...
##  $ AQS_PARAMETER_CODE        : int  81102 81102 81102 81102 81102 81102 81102 8110
##  $ AQS_PARAMETER_DESC        : Factor w/ 1 level "PM10 Total 0-10um STP": 1 1 1 1
```

```
##   $ CBSA_CODE                 : int  22180 22180 22180 22180 22180 22180 22180 2218
##   $ CBSA_NAME                 : Factor w/ 8 levels "","Charlotte-Concord-Gastonia,
##   $ STATE_CODE                : int  37 37 37 37 37 37 37 37 37 37 ...
##   $ STATE                     : Factor w/ 1 level "North Carolina": 1 1 1 1 1 1 1 1
##   $ COUNTY_CODE               : int  51 51 51 51 51 51 51 51 51 51 ...
##   $ COUNTY                    : Factor w/ 8 levels "Cumberland","Durham",..: 1 1 1
##   $ SITE_LATITUDE             : num  35 35 35 35 35 ...
##   $ SITE_LONGITUDE            : num  -79 -79 -79 -79 -79 ...
```

colnames(EPA_AQPM25_NC2018_raw)

```
##  [1] "Date"                      "Source"
##  [3] "Site.ID"                   "POC"
##  [5] "Daily.Mean.PM2.5.Concentration" "UNITS"
##  [7] "DAILY_AQI_VALUE"           "Site.Name"
##  [9] "DAILY_OBS_COUNT"           "PERCENT_COMPLETE"
## [11] "AQS_PARAMETER_CODE"        "AQS_PARAMETER_DESC"
## [13] "CBSA_CODE"                 "CBSA_NAME"
## [15] "STATE_CODE"                "STATE"
## [17] "COUNTY_CODE"               "COUNTY"
## [19] "SITE_LATITUDE"             "SITE_LONGITUDE"
```

colnames(EPA_AQPM10_NC2018_raw)

```
##  [1] "Date"                      "Source"
##  [3] "Site.ID"                   "POC"
##  [5] "Daily.Mean.PM10.Concentration" "UNITS"
##  [7] "DAILY_AQI_VALUE"           "Site.Name"
##  [9] "DAILY_OBS_COUNT"           "PERCENT_COMPLETE"
## [11] "AQS_PARAMETER_CODE"        "AQS_PARAMETER_DESC"
## [13] "CBSA_CODE"                 "CBSA_NAME"
## [15] "STATE_CODE"                "STATE"
## [17] "COUNTY_CODE"               "COUNTY"
## [19] "SITE_LATITUDE"             "SITE_LONGITUDE"
```

summary(EPA_AQPM25_NC2018_raw)

```
##       Date                 Source        Site.ID              POC
##   Min.   :2018-01-01   AirNow: 873   Min.   :370110002   Min.   :1.000
##   1st Qu.:2018-04-04   AQS   :8771   1st Qu.:370650099   1st Qu.:3.000
##   Median :2018-06-27                 Median :371190041   Median :3.000
##   Mean   :2018-06-30                 Mean   :371023866   Mean   :2.948
##   3rd Qu.:2018-09-30                 3rd Qu.:371230001   3rd Qu.:3.000
##   Max.   :2018-12-31                 Max.   :371830021   Max.   :5.000
##
##   Daily.Mean.PM2.5.Concentration      UNITS        DAILY_AQI_VALUE
##   Min.   :-2.80                   ug/m3 LC:9644   Min.   : 0.00
```

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```
## 1st Qu.: 5.00                                     1st Qu.:21.00
## Median : 7.20                                     Median :30.00
## Mean   : 7.61                                     Mean   :31.22
## 3rd Qu.: 9.80                                     3rd Qu.:41.00
## Max.   :34.20                                     Max.   :97.00
##
##                          Site.Name    DAILY_OBS_COUNT PERCENT_COMPLETE
## Garinger High School         : 732    Min.   :1       Min.   :100
## Millbrook School             : 722    1st Qu.:1       1st Qu.:100
## Remount                      : 668    Median :1       Median :100
## Montclaire Elementary School: 648     Mean   :1       Mean   :100
## Hattie Avenue                : 510    3rd Qu.:1       3rd Qu.:100
## Board Of Ed. Bldg.           : 478    Max.   :1       Max.   :100
## (Other)                      :5886
## AQS_PARAMETER_CODE                               AQS_PARAMETER_DESC
## Min.   :88101     Acceptable PM2.5 AQI & Speciation Mass:2008
## 1st Qu.:88101       PM2.5 - Local Conditions            :7636
## Median :88101
## Mean   :88184
## 3rd Qu.:88101
## Max.   :88502
##
##    CBSA_CODE                              CBSA_NAME       STATE_CODE
## Min.   :11700    Charlotte-Concord-Gastonia, NC-SC:2048  Min.   :37
## 1st Qu.:16740    Raleigh, NC                      :1418  1st Qu.:37
## Median :24780    Winston-Salem, NC                :1323  Median :37
## Mean   :29881                                     :1165  Mean   :37
## 3rd Qu.:40580    Asheville, NC                    : 532  3rd Qu.:37
## Max.   :49180    Durham-Chapel Hill, NC           : 469  Max.   :37
## NA's   :1165    (Other)                           :2689
##             STATE       COUNTY_CODE          COUNTY     SITE_LATITUDE
## North Carolina:9644   Min.   : 11.0   Mecklenburg:2048  Min.   :34.36
##                       1st Qu.: 65.0   Wake       :1069  1st Qu.:35.24
##                       Median :119.0   Forsyth    : 876  Median :35.64
##                       Mean   :102.4   Buncombe   : 478  Mean   :35.58
##                       3rd Qu.:123.0   Durham     : 469  3rd Qu.:35.91
##                       Max.   :183.0   Pitt       : 461  Max.   :36.11
##                                       (Other)    :4243
## SITE_LONGITUDE
## Min.   :-83.44
## 1st Qu.:-80.87
## Median :-80.23
## Mean   :-80.03
## 3rd Qu.:-78.82
## Max.   :-76.21
```

```
##
```

```r
summary(EPA_AQPM10_NC2018_raw)
```

```
##      Date              Source      Site.ID                POC
##  Min.   :2018-01-01  AQS:2905   Min.   :370510009   Min.   :1.000
##  1st Qu.:2018-04-07             1st Qu.:370670022   1st Qu.:3.000
##  Median :2018-07-02             Median :371170001   Median :3.000
##  Mean   :2018-07-03             Mean   :371072712   Mean   :3.172
##  3rd Qu.:2018-10-04             3rd Qu.:371190042   3rd Qu.:4.000
##  Max.   :2018-12-31             Max.   :371830014   Max.   :5.000
##
##  Daily.Mean.PM10.Concentration       UNITS      DAILY_AQI_VALUE
##  Min.   : 0.00                    ug/m3 SC:2905   Min.   : 0.00
##  1st Qu.:10.00                                    1st Qu.: 9.00
##  Median :13.00                                    Median :12.00
##  Mean   :13.72                                    Mean   :12.67
##  3rd Qu.:17.00                                    3rd Qu.:16.00
##  Max.   :64.00                                    Max.   :55.00
##
##                         Site.Name   DAILY_OBS_COUNT PERCENT_COMPLETE
##  Millbrook School           :588   Min.   :1        Min.   :100
##  Garinger High School       :351   1st Qu.:1        1st Qu.:100
##  Montclaire Elementary School:344  Median :1        Median :100
##  Hattie Avenue              :342   Mean   :1        Mean   :100
##  Durham Armory              :335   3rd Qu.:1        3rd Qu.:100
##  William Owen School        :321   Max.   :1        Max.   :100
##  (Other)                    :624
##  AQS_PARAMETER_CODE          AQS_PARAMETER_DESC   CBSA_CODE
##  Min.   :81102      PM10 Total 0-10um STP:2905   Min.   :16740
##  1st Qu.:81102                                   1st Qu.:16740
##  Median :81102                                   Median :22180
##  Mean   :81102                                   Mean   :28310
##  3rd Qu.:81102                                   3rd Qu.:39580
##  Max.   :81102                                   Max.   :49180
##                                                  NA's   :247
##                            CBSA_NAME    STATE_CODE
##  Charlotte-Concord-Gastonia, NC-SC:695   Min.   :37
##  Raleigh, NC                      :588   1st Qu.:37
##  Winston-Salem, NC                :342   Median :37
##  Durham-Chapel Hill, NC           :335   Mean   :37
##  Fayetteville, NC                 :321   3rd Qu.:37
##  Greensboro-High Point, NC        :320   Max.   :37
##  (Other)                          :304
##        STATE       COUNTY_CODE          COUNTY     SITE_LATITUDE
```
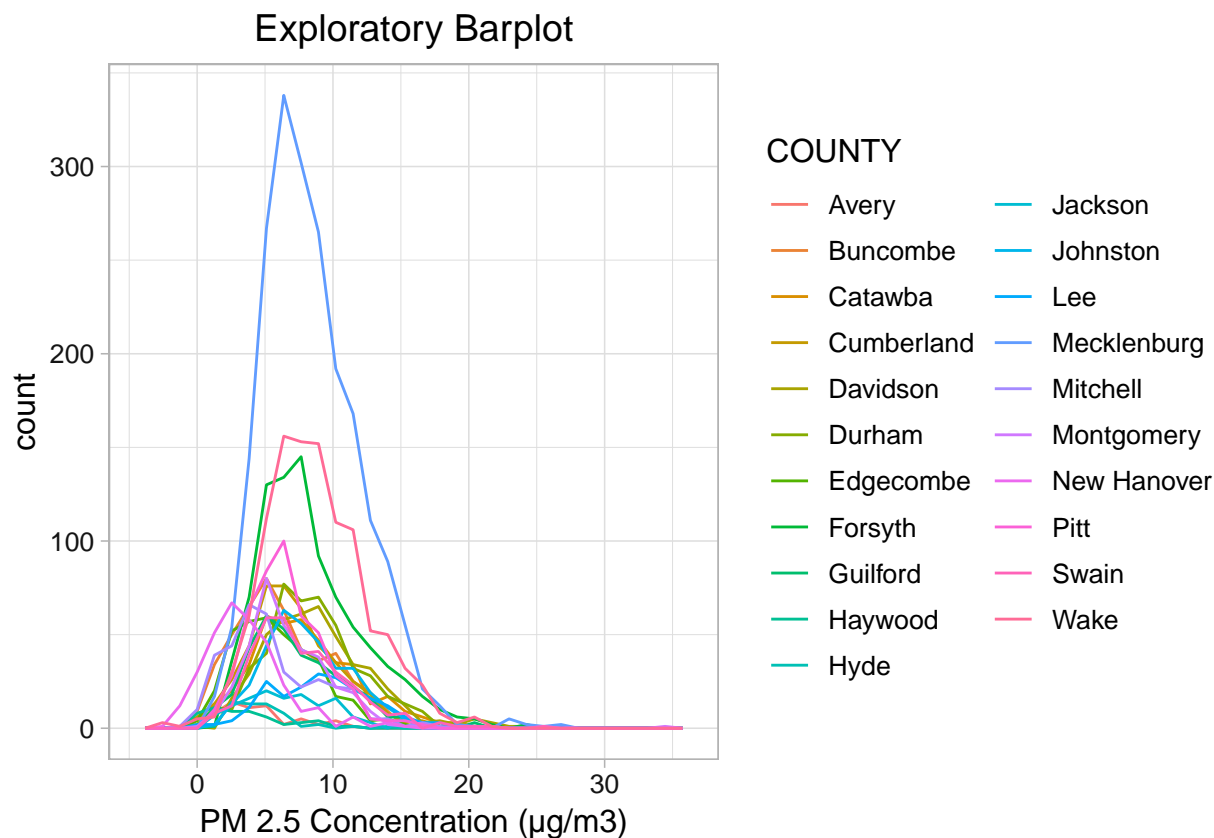
Figure 1: PM2.5 NC 2018 frequency polygon.

```
##   North Carolina:2905     Min.   : 51.0    Mecklenburg:695     Min.   :35.04
##                           1st Qu.: 67.0    Wake       :588     1st Qu.:35.24
##                           Median :117.0    Forsyth    :342     Median :35.86
##                           Mean   :107.3    Durham     :335     Mean   :35.67
##                           3rd Qu.:119.0    Cumberland :321     3rd Qu.:36.03
##                           Max.   :183.0    Guilford   :320     Max.   :36.11
##                                            (Other)    :304
##   SITE_LONGITUDE
##   Min.   :-80.87
##   1st Qu.:-80.23
##   Median :-78.95
##   Mean   :-79.36
##   3rd Qu.:-78.57
##   Max.   :-76.91
##
```

Visual data exploration of the PM2.5 2018 raw data file in Figure 1, Figure 2, and Figure 3.

Visual data exploration of the PM10 2018 raw data file in Figure 4.
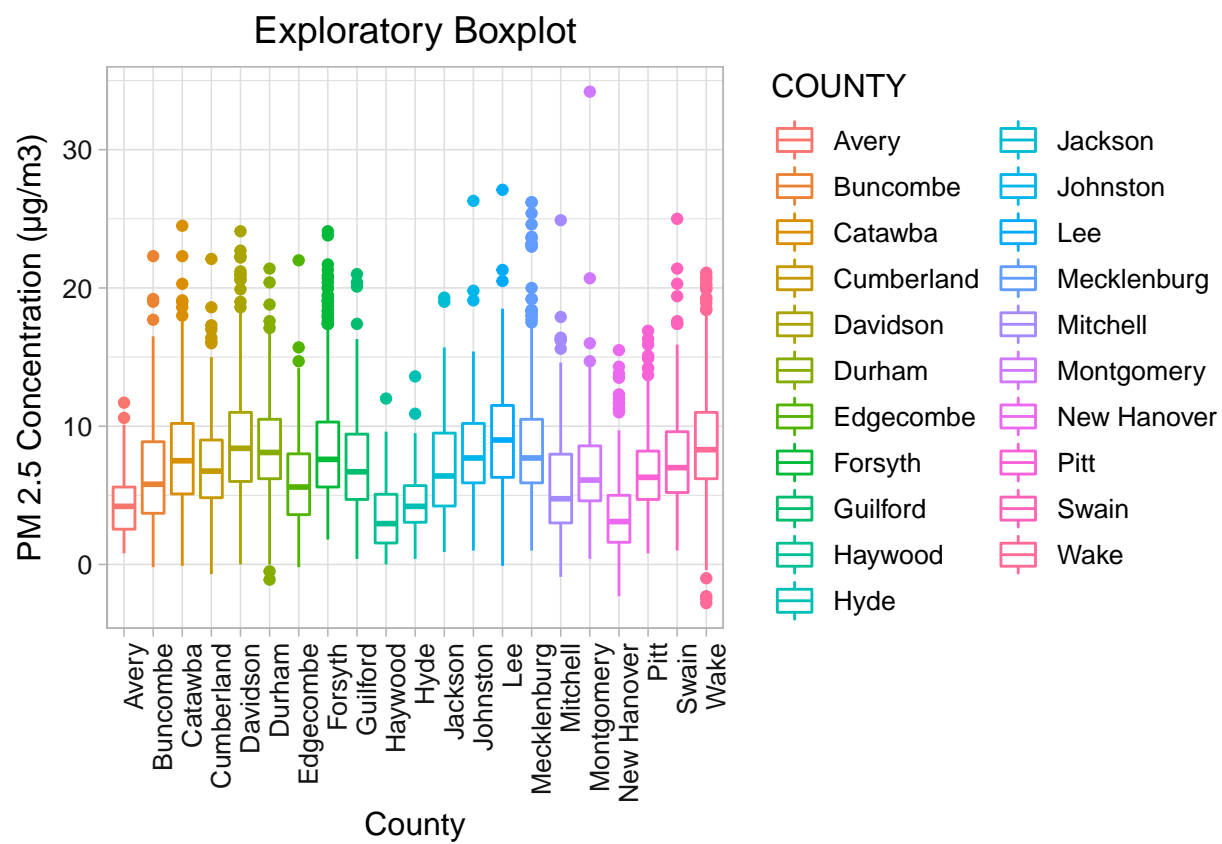
Figure 2: PM2.5 NC 2018 boxplot.
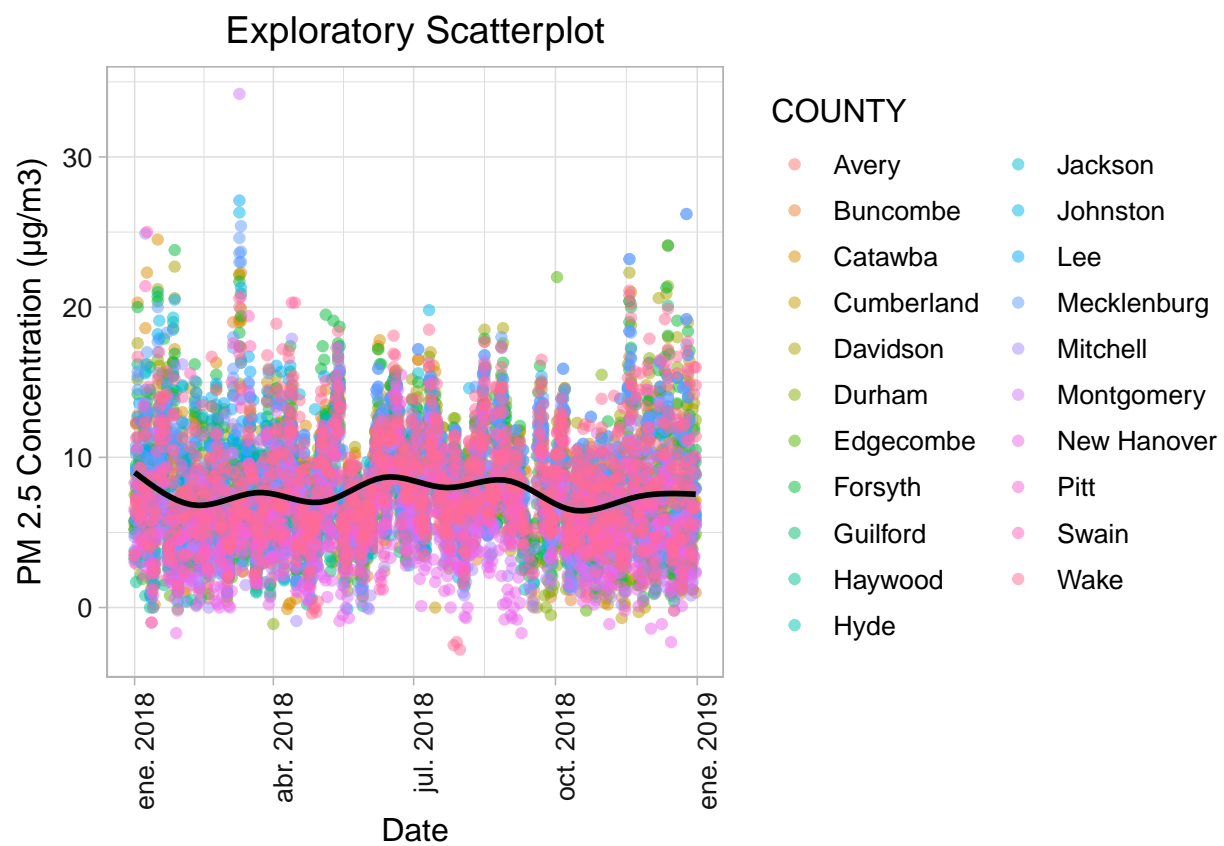
Figure 3: PM2.5 NC 2018 scatterplot.
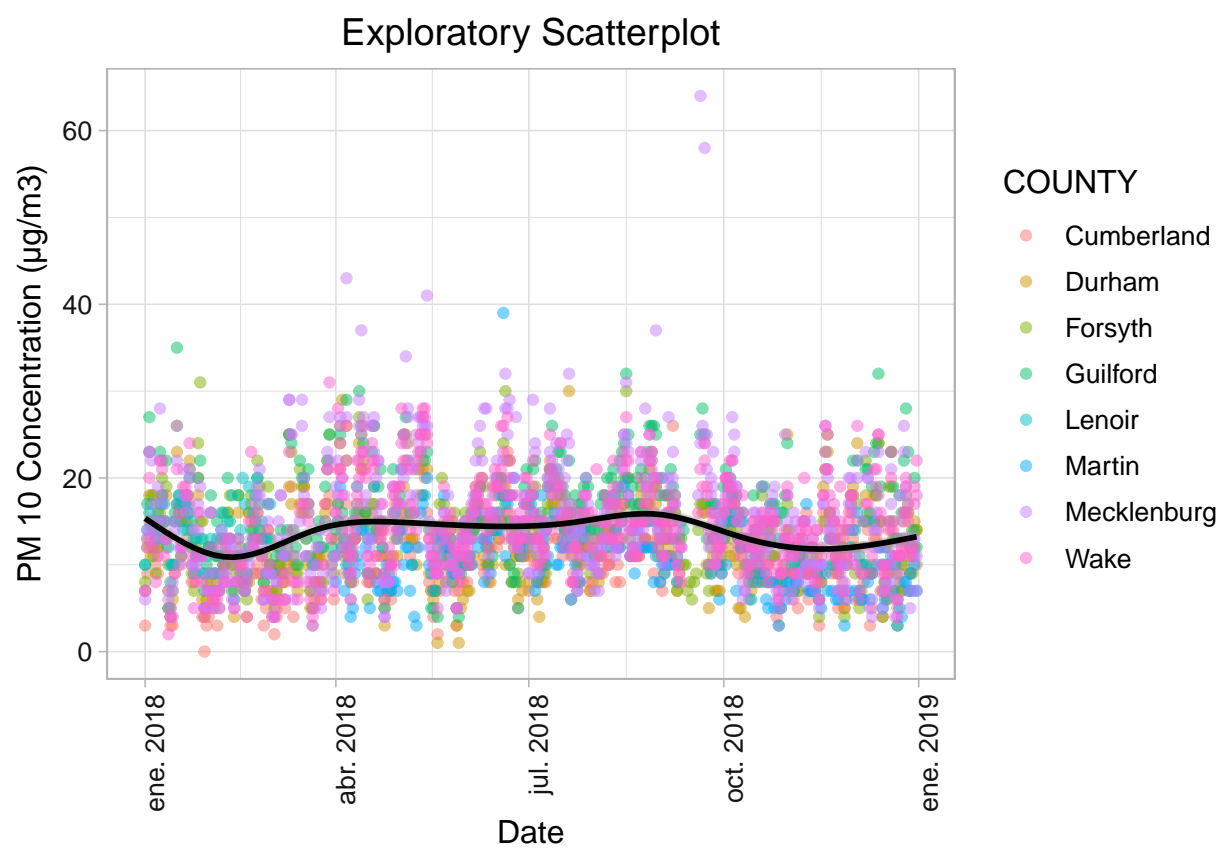
Figure 4: PM10 NC 2018 scatterplot.

Data wrangling of the PM2.5 and PM10 2018 raw data files.

```r
#Selecting Columns
EPA_AQ_PM25_NC2018_Temp <- select(EPA_AQPM25_NC2018_raw, Date, Site.ID,
                                  Daily.Mean.PM2.5.Concentration, AQS_PARAMETER_DESC,
                                  COUNTY:SITE_LONGITUDE)


#Changing column name
colnames(EPA_AQ_PM25_NC2018_Temp)[colnames(EPA_AQ_PM25_NC2018_Temp)
                                  =="Daily.Mean.PM2.5.Concentration"]<-"Daily.Mean.Concent

#Selecting Columns
EPA_AQ_PM10_NC2018_Temp <- select(EPA_AQPM10_NC2018_raw, Date, Site.ID,
                                  Daily.Mean.PM10.Concentration, AQS_PARAMETER_DESC,
                                  COUNTY:SITE_LONGITUDE)


#Changing column name
colnames(EPA_AQ_PM10_NC2018_Temp)[colnames(EPA_AQ_PM10_NC2018_Temp)
                                  =="Daily.Mean.PM10.Concentration"]<-"Daily.Mean.Concentr

#Create AQS_PARAMETER_DESC Column with Contaminant description.
EPA_AQ_PM25_NC2018_Temp$AQS_PARAMETER_DESC <- "PM2.5"
EPA_AQ_PM10_NC2018_Temp$AQS_PARAMETER_DESC <- "PM10"


#Eliminates duplicate dates
EPA_AQ_PM25_NC2018_Cleaned <- EPA_AQ_PM25_NC2018_Temp [!duplicated(EPA_AQ_PM25_NC2018_Te


EPA_AQ_PM10_NC2018_Cleaned <- EPA_AQ_PM10_NC2018_Temp [!duplicated(EPA_AQ_PM10_NC2018_Te


# Combine the data.
EPA_AQ_PM2.5PM10_NC2018_Cleaned <- rbind(EPA_AQ_PM25_NC2018_Cleaned, EPA_AQ_PM10_NC2018_

#Save the data in the processed folder
write.csv(EPA_AQ_PM2.5PM10_NC2018_Cleaned,
          "./Data/Processed/EPA_AQ_PM2.5PM10_NC2018_Cleaned.csv")


#Spread PM2.5 and PM10
EPA_AQ_PM2.5PM10_NC2018_Spread <-
  EPA_AQ_PM2.5PM10_NC2018_Cleaned %>%
  spread(AQS_PARAMETER_DESC, Daily.Mean.Concentration)


#Remove rows without PM2.5 data
EPA_AQ_PM2.5PM10_NC2018_Spread <- EPA_AQ_PM2.5PM10_NC2018_Spread[!is.na(EPA_AQ_PM2.5PM10

#Convert the dataset to a spatially enabled "sf" data frame
```
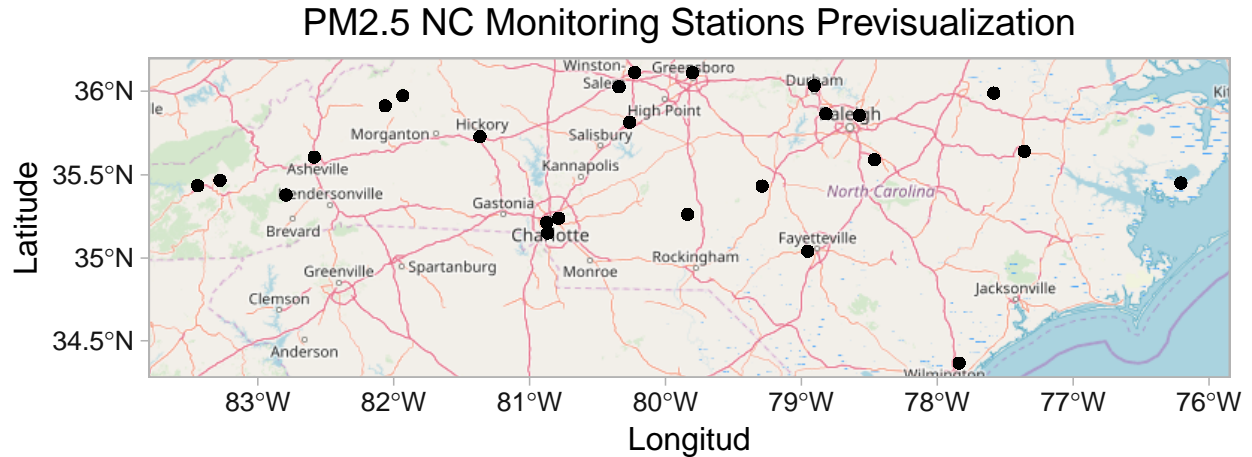
Figure 5: PM2.5 NC Monitoring Stations Previsualization.

```
PM2.5_PM10_sf <- st_as_sf(EPA_AQ_PM2.5PM10_NC2018_Spread,coords = c('SITE_LONGITUDE','SI

#Convert all to UTM Zone 17 (crs = 26917)
PM2.5_PM10_sf_utm <- st_transform(PM2.5_PM10_sf, c=26917)
```

In Figure 5 is presented the locations of the PM2.5 monitoring stations.

3.2   North Carolina Counties Zoning, Geographic information, and Population Data

Downloading the list of North Carolina Counties and Population from a Wikipedia URL.

```
#North Carolina Counties
url <- "https://en.wikipedia.org/wiki/List_of_counties_in_North_Carolina"
webpage <- read_html(url)

County_Name <- webpage %>% html_nodes("th:nth-child(1)") %>% html_text()
County_Population <- webpage %>% html_nodes("tr :nth-child(7)") %>% html_text()
```

```
#Remove unwanted info and characters
County_Info <- data_frame(County = County_Name[9:108])
County_Info$County <- str_replace(County_Info$County, " County", "")
County_Info$County <- str_replace(County_Info$County, "\n", "")

Population <- data_frame(Population=County_Population[2:101])

County_Info <- cbind(County_Info, Population)

County_Info$Population <- str_replace(County_Info$Population,",","")
County_Info$Population <- str_replace(County_Info$Population,",","")

County_Info$Population <- as.numeric(County_Info$Population)
```

Assigning the corresponding zone to each county. Info from: Rudersdorf, Amy. 2010. "NC County Maps." Government & Heritage Library, State Library of North Carolina.

```
#North Carolina Zones
County_Info$Zone<-ifelse(County_Info$County == 'Ashe'|County_Info$County =='Alleghany'|C
                         ifelse(County_Info$County == 'Surry'|County_Info$County =='Stok
                                ifelse(County_Info$County == 'Scotland'|County_Info$Coun
```

Data exploration of the County_Info dataframe.

```
dim(County_Info)
```

```
## [1] 100    3
```

```
str(County_Info)
```

```
## 'data.frame':    100 obs. of  3 variables:
##  $ County    : chr  "Alamance" "Alexander" "Alleghany" "Anson" ...
##  $ Population: num  157844 37159 10935 25531 26833 ...
##  $ Zone      : chr  "Piedmont" "Piedmont" "Mountains" "Piedmont" ...
```

```
colnames(County_Info)
```

```
## [1] "County"     "Population" "Zone"
```

```
summary(County_Info)
```

```
##     County            Population           Zone
##  Length:100        Min.   :   4090    Length:100
##  Class :character  1st Qu.:  25001    Class :character
##  Mode  :character  Median :  55311    Mode  :character
##                    Mean   : 100526
##                    3rd Qu.: 114764
##                    Max.   :1034290
```

```r
unique(County_Info$Zone)
```

```
## [1] "Piedmont"  "Mountains" "Coastal"
```

Adding the County Information to the PM2.5_PM10_sf_utm dataframe.

```r
PM2.5_PM10_Info_sf_utm <- PM2.5_PM10_sf_utm %>%
left_join(County_Info, by = c("COUNTY"="County"))
```

3.3   US Census Bureau US counties shapefile

Reading the USA county shapefile, sub-setting for NC.

```r
counties_sf<- st_read('./Data/Spatial/cb_2017_us_county_20m.shp') %>%
  filter(STATEFP == 37) #Filter for just NC Counties
```

```
## Reading layer `cb_2017_us_county_20m' from data source `C:\Users\Felipe\OneDrive - Du
## Simple feature collection with 3220 features and 9 fields
## geometry type:  MULTIPOLYGON
## dimension:      XY
## bbox:           xmin: -179.1743 ymin: 17.91377 xmax: 179.7739 ymax: 71.35256
## epsg (SRID):    4269
## proj4string:    +proj=longlat +datum=NAD83 +no_defs
```

```r
#CRS
st_crs(counties_sf) #crs=4269 = NAD83.
```

```
## Coordinate Reference System:
##   EPSG: 4269
##   proj4string: "+proj=longlat +datum=NAD83 +no_defs"
```

Converting the counties_sf to UTM Zone 17

```r
#Convert all to UTM Zone 17 (crs = 26917)
counties_sf_utm <- st_transform(counties_sf, c=26917)
```

Data exploration of the County_Info dataframe.

```r
dim(counties_sf_utm)
```

```
## [1] 100  10
```

```r
str(counties_sf_utm)
```

```
## Classes 'sf' and 'data.frame':   100 obs. of  10 variables:
##  $ STATEFP : Factor w/ 52 levels "01","02","04",..: 34 34 34 34 34 34 34 34 34 34 ...
##  $ COUNTYFP: Factor w/ 325 levels "001","003","005",..: 116 71 94 54 62 125 84 102 26
##  $ COUNTYNS: Factor w/ 3220 levels "00023901","00025441",..: 1672 1649 1659 1639 1644
```

```
##  $ AFFGEOID: Factor w/ 3220 levels "0500000US01001",..: 1981 1944 1964 1931 1937 1988
##  $ GEOID   : Factor w/ 3220 levels "01001","01003",..: 1981 1944 1964 1931 1937 1988
##  $ NAME    : Factor w/ 1910 levels "Abbeville","Acadia",..: 1765 1000 1358 721 797 18
##  $ LSAD    : Factor w/ 9 levels "00","03","04",..: 5 5 5 5 5 5 5 5 5 5 ...
##  $ ALAND   : num  6.54e+08 1.03e+09 1.69e+09 1.67e+09 1.01e+09 ...
##  $ AWATER  : num  42187365 5900300 8248766 30723331 3981006 ...
##  $ geometry:sfc_MULTIPOLYGON of length 100; first list element: List of 1
##   ..$ :List of 1
##   .. ..$ : num [1:10, 1:2] 724066 727615 739570 744478 741926 ...
##   ..- attr(*, "class")= chr  "XY" "MULTIPOLYGON" "sfg"
##  - attr(*, "sf_column")= chr "geometry"
##  - attr(*, "agr")= Factor w/ 3 levels "constant","aggregate",..: NA NA NA NA NA NA NA
##   ..- attr(*, "names")= chr  "STATEFP" "COUNTYFP" "COUNTYNS" "AFFGEOID" ...
```

```
colnames(counties_sf_utm)
```

```
## [1] "STATEFP"  "COUNTYFP" "COUNTYNS" "AFFGEOID" "GEOID"    "NAME"
## [7] "LSAD"     "ALAND"    "AWATER"   "geometry"
```

```
summary(counties_sf_utm)
```

```
##    STATEFP       COUNTYFP      COUNTYNS           AFFGEOID       GEOID
## 37     :100   001    : 1   01008531: 1   0500000US37001: 1   37001  : 1
## 01     :  0   003    : 1   01008532: 1   0500000US37003: 1   37003  : 1
## 02     :  0   005    : 1   01008533: 1   0500000US37005: 1   37005  : 1
## 04     :  0   007    : 1   01008534: 1   0500000US37007: 1   37007  : 1
## 05     :  0   009    : 1   01008535: 1   0500000US37009: 1   37009  : 1
## 06     :  0   011    : 1   01008536: 1   0500000US37011: 1   37011  : 1
## (Other):  0   (Other):94   (Other) :94   (Other)       :94   (Other):94
##        NAME        LSAD         ALAND              AWATER
## Alamance : 1   06     :100   Min.   :4.472e+08   Min.   :4.453e+05
## Alexander: 1   00     :  0   1st Qu.:8.936e+08   1st Qu.:7.287e+06
## Alleghany: 1   03     :  0   Median :1.192e+09   Median :1.595e+07
## Anson    : 1   04     :  0   Mean   :1.259e+09   Mean   :1.347e+08
## Ashe     : 1   05     :  0   3rd Qu.:1.501e+09   3rd Qu.:3.831e+07
## Avery    : 1   12     :  0   Max.   :2.453e+09   Max.   :3.001e+09
## (Other)  :94   (Other):  0
##         geometry
## MULTIPOLYGON :100
## epsg:26917   :  0
## +proj=utm ...:  0
##
##
##
##
```

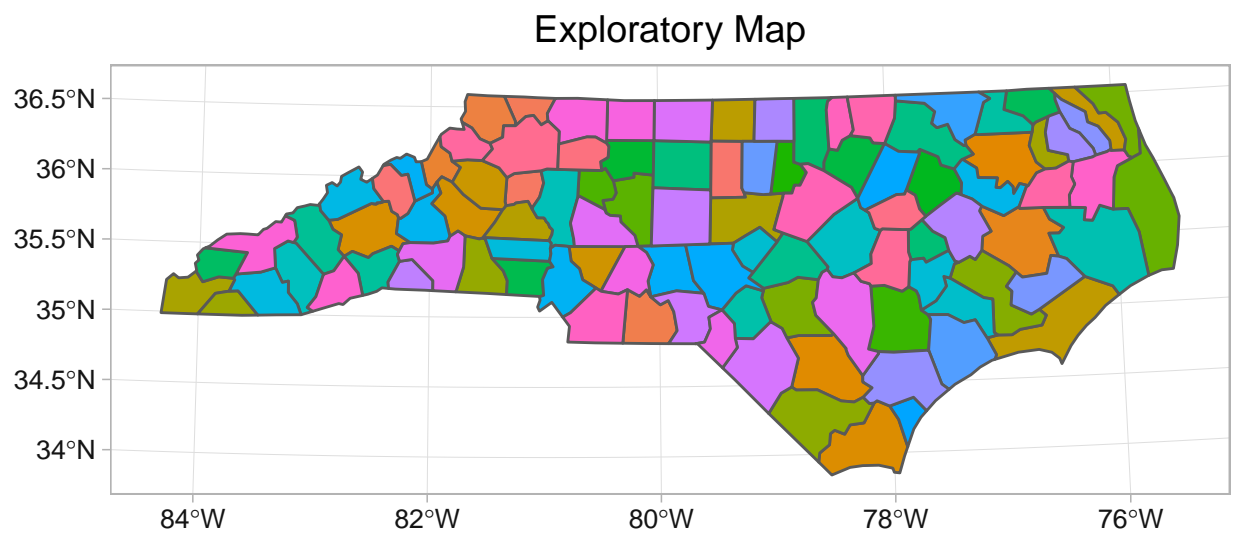Visual data exploration of the counties_sf_utm dataframe in Figure 6.

Figure 6: Counties exploratory map.

3.4 NOAA Average Temperature Dataset

Reading the 2018 North Carolina Air Temperature data.

```
#Read the 2018 Air Temperature data
NOAA_DTAVG_NC2018_raw <- read.csv("./Data/Raw/NOAA_TAVG_NC2018_raw.csv")
```

Data exploration of the NOAA_DTAVG_NC2018_raw dataframe.

```
dim(NOAA_DTAVG_NC2018_raw)
```

```
## [1] 283423      7
```

```
str(NOAA_DTAVG_NC2018_raw)
```

```
## 'data.frame':    283423 obs. of  7 variables:
##  $ STATION  : Factor w/ 1066 levels "US1NCAG0001",..: 217 217 217 217 217 217 217 217
##  $ NAME     : Factor w/ 1063 levels "ABERDEEN 0.7 NW, NC US",..: 716 716 716 716 716
##  $ LATITUDE : num  34.8 34.8 34.8 34.8 34.8 ...
##  $ LONGITUDE: num  -76.9 -76.9 -76.9 -76.9 -76.9 ...
##  $ ELEVATION: num  8.5 8.5 8.5 8.5 8.5 8.5 8.5 8.5 8.5 8.5 ...
##  $ DATE     : Factor w/ 365 levels "01/01/2018","01/02/2018",..: 37 133 145 205 265 2
##  $ TAVG     : int  NA NA NA NA NA NA NA NA NA NA ...
```

```
colnames(NOAA_DTAVG_NC2018_raw)
```

```
## [1] "STATION"   "NAME"      "LATITUDE"  "LONGITUDE" "ELEVATION" "DATE"
## [7] "TAVG"
```

```
summary(NOAA_DTAVG_NC2018_raw)
```

```
##         STATION                        NAME              LATITUDE
##  US1NCBC0005:   365    SPARTA 3.5 SSW, NC US     :   545   Min.   :33.88
##  US1NCBC0041:   365    HILLSBOROUGH 5.6 NNW, NC US:  502   1st Qu.:35.16
##  US1NCBK0004:   365    ADVANCE 0.2 ESE, NC US    :   365   Median :35.56
##  US1NCCH0004:   365    ALBEMARLE, NC US          :   365   Mean   :35.49
##  US1NCCS0002:   365    ARDEN 1.6 ENE, NC US      :   365   3rd Qu.:35.90
##  US1NCCY0003:   365    ASHEBORO 1.3 SSE, NC US   :   365   Max.   :36.56
##  (Other)    :281233    (Other)                   :280916
##    LONGITUDE          ELEVATION             DATE               TAVG
##  Min.   :-84.30    Min.   :    0.0   16/04/2018:   887   Min.   : 9.00
##  1st Qu.:-81.67    1st Qu.:   29.3   17/05/2018:   874   1st Qu.:47.00
##  Median :-79.16    Median :  150.9   20/03/2018:   871   Median :63.00
##  Mean   :-79.70    Mean   :  279.9   12/06/2018:   870   Mean   :60.19
##  3rd Qu.:-78.01    3rd Qu.:  389.5   30/05/2018:   869   3rd Qu.:75.00
##  Max.   :-75.46    Max.   : 1902.0   01/08/2018:   867   Max.   :87.00
##                                      (Other)   :278185   NA's   :269572
```

Data wrangling of the NOAA_DTAVG_NC2018_raw dataframe.

```r
#Remove stations without Temperature information
NOAA_DTAVG_NC2018_Complete <- na.omit(NOAA_DTAVG_NC2018_raw)

#Convert the dataset to a spatially enabled "sf" data frame
NOAA_DTAVG_NC2018_sf <- st_as_sf(NOAA_DTAVG_NC2018_Complete,coords = c('LONGITUDE','LATI

#Convert all to UTM Zone 17 (crs = 26917)
NOAA_DTAVG_NC2018_sf_utm <- st_transform(NOAA_DTAVG_NC2018_sf, c=26917)

#Formatting dates
NOAA_DTAVG_NC2018_sf_utm$DATE <- as.Date(NOAA_DTAVG_NC2018_sf_utm$DATE, format = "%d/%m/
```

The 2018 Air Temperature data does not have County information, so the location is used with the counties_sf_utm dataframe to locate the county of each station.

```r
#Adding the county and zone information to the Temperature dataframe

#Index of the matching feature
county_index <- st_nearest_feature(NOAA_DTAVG_NC2018_sf_utm, counties_sf_utm)

#Eliminates geo info
aux1 <- st_set_geometry(counties_sf_utm[county_index,"NAME"], value=NULL)

#adds the columns
NOAA_DTAVG_NC2018_sf_utm$COUNTY <- aux1$NAME

#Reordering
NOAA_DTAVG_NC2018_sf_utm <- NOAA_DTAVG_NC2018_sf_utm[,c(1,2,3,4,5,7,6)]
```

Visual data exploration of the 2018 North Carolina Air Temperature data in Figure 7, Figure 8, Figure 9, and Figure 10..

Next, the temperature of the nearest Temperature Station is added to each PM2.5 Station in the PM2.5_PM10_Info_sf_utm dataframe.

```r
#Create a Data frame with only the PM2.5 station info
PM2.5_Stations <- PM2.5_PM10_Info_sf_utm %>%
  select(Site.ID, geometry) %>%
  subset(!duplicated(Site.ID))

#Distances bewteen the PM2.5 stations and the Temperature Stations
Nearest <- st_nearest_feature(PM2.5_Stations, NOAA_DTAVG_NC2018_sf_utm)

a <- length(unique(PM2.5_Stations$Site.ID))

NOAA_DTAVG_NC2018_sf_utm$NAME <- as.character(NOAA_DTAVG_NC2018_sf_utm$NAME)
```

NC 2018 Temperature Exploratory Map, Mean Annual Temperature
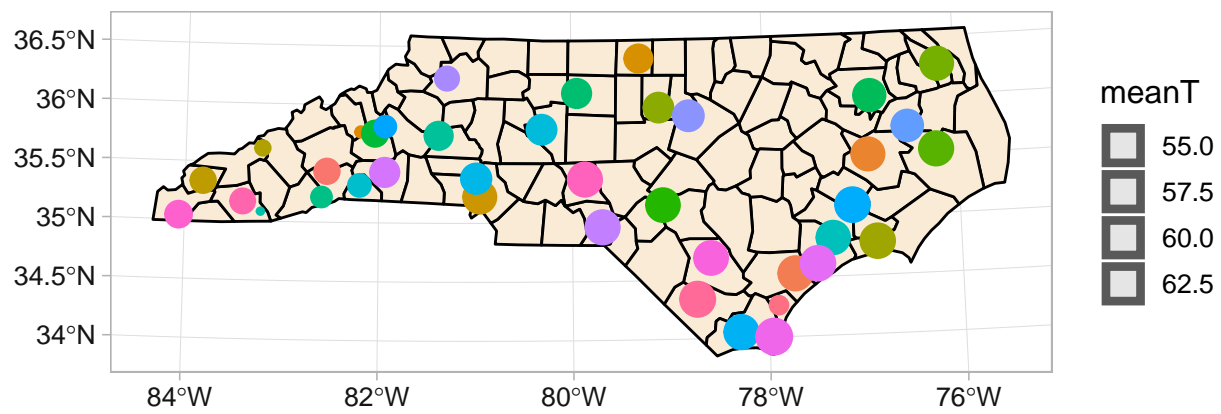
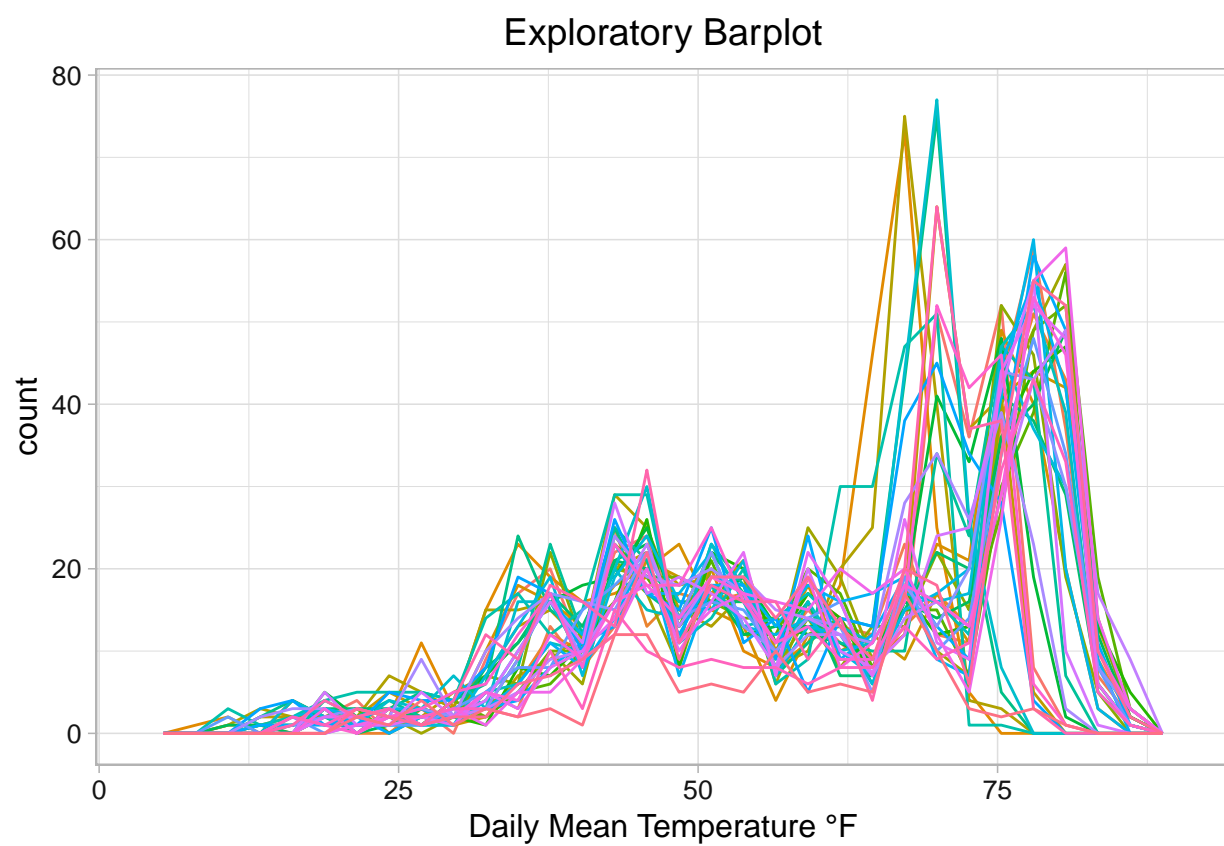Figure 7: Mean Annual Temperature exploratory map.

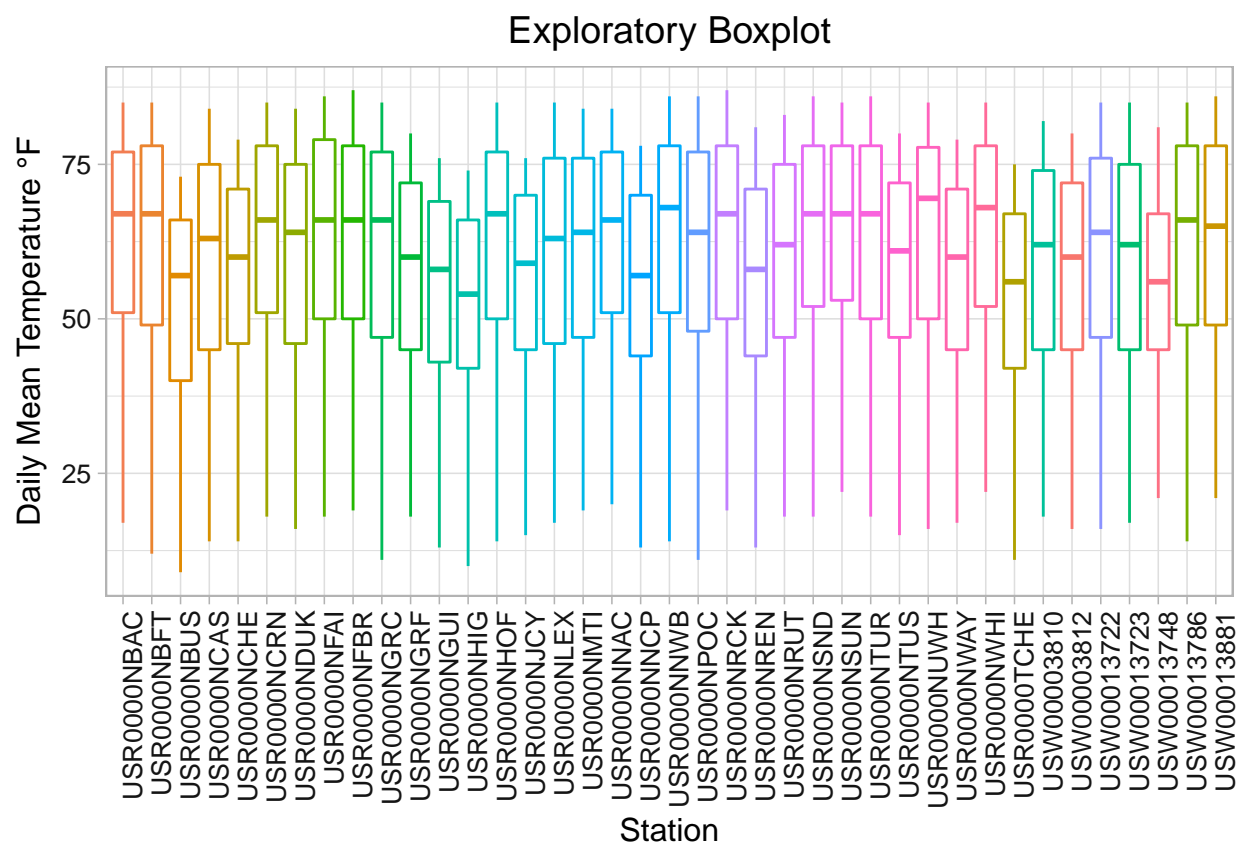Figure 8: Daily Mean Temperature NC 2018 frequency polygon.

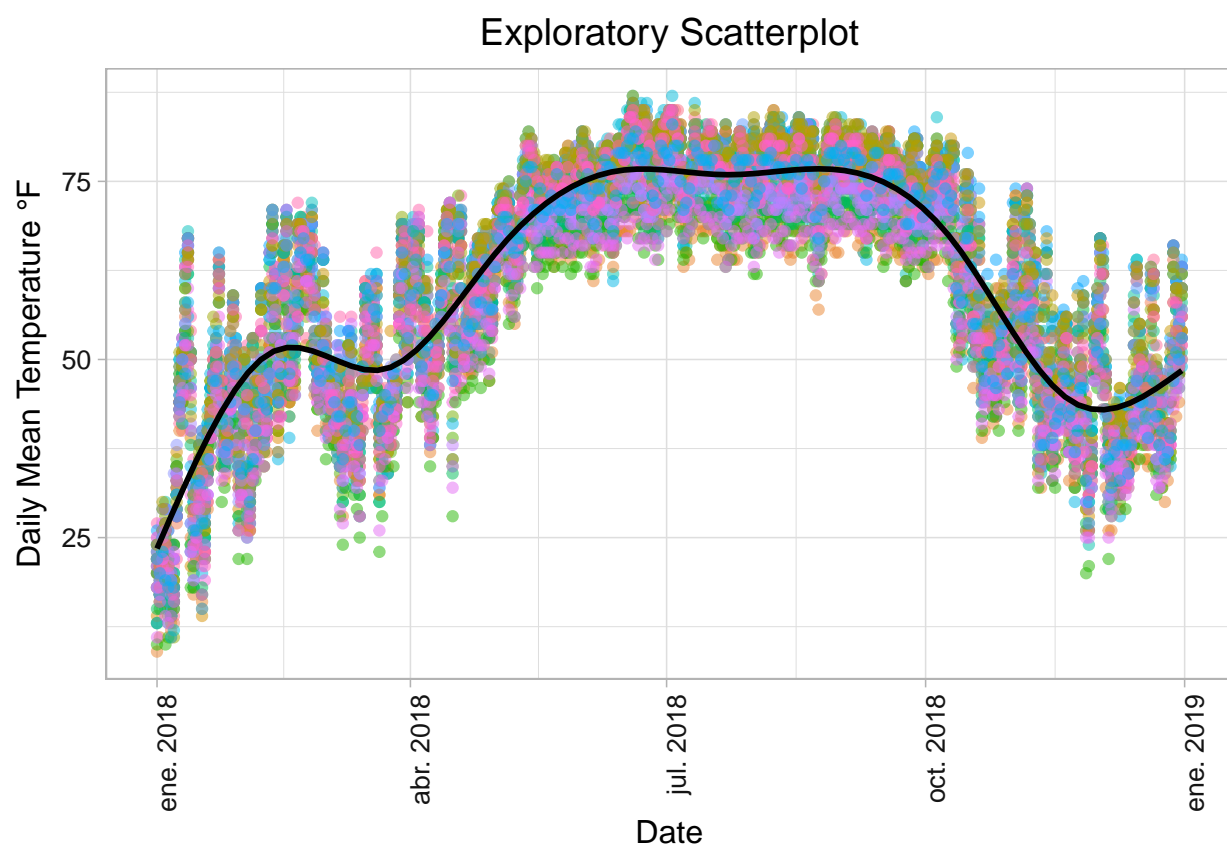Figure 9: Daily Mean Temperature NC 2018 boxplot.

Figure 10: Daily Mean Temperature NC 2018 scatterplot.

```r
#Assingning the nearest Temperature Station to each PM2.5 station.
for (i in 1:a){
  PM2.5_Stations$Temp_Est[i] <- NOAA_DTAVG_NC2018_sf_utm$NAME[Nearest[i]]
}

#Drop the geo data
aux2 <- st_set_geometry(PM2.5_Stations, value=NULL)

#Left_join the data
PM2.5_PM10_Temp_sf_utm <- PM2.5_PM10_Info_sf_utm %>%
left_join(aux2)

#Assingning the Temperature of the nearest Temperature Station to each PM2.5 station.

#Drops the geo data
aux3 <- st_set_geometry(NOAA_DTAVG_NC2018_sf_utm, value=NULL)

#Left_join the data
PM2.5_PM10_Temp_sf_utm <- PM2.5_PM10_Temp_sf_utm %>%
left_join(aux3, by = c("Temp_Est"="NAME", "Date"="DATE", "COUNTY")) %>%
  select(Date,Site.ID,COUNTY,Population,Zone,PM2.5,PM10,TAVG,geometry)
```

3.5 EPA combustion points for electricity generation in the US Dataset

Reading the Electricity Generation via Combustion data.

```r
EPA_US_CombEmissions <- st_read("./Data/Raw/EPA_ElecGenComb_US_raw.kml")
```

```
## Reading layer `Electricity Generation via Combustion' from data source `C:\Users\Feli
## Simple feature collection with 2042 features and 2 fields
## geometry type:  POINT
## dimension:      XYZ
## bbox:           xmin: -176.6593 ymin: 19.63283 xmax: -67.00325 ymax: 71.29221
## epsg (SRID):    4326
## proj4string:    +proj=longlat +datum=WGS84 +no_defs
```

Wrangling the data

```r
st_crs(EPA_US_CombEmissions) #crs=4326 = WGS 84
```

```
## Coordinate Reference System:
##   EPSG: 4326
##   proj4string: "+proj=longlat +datum=WGS84 +no_defs"
```
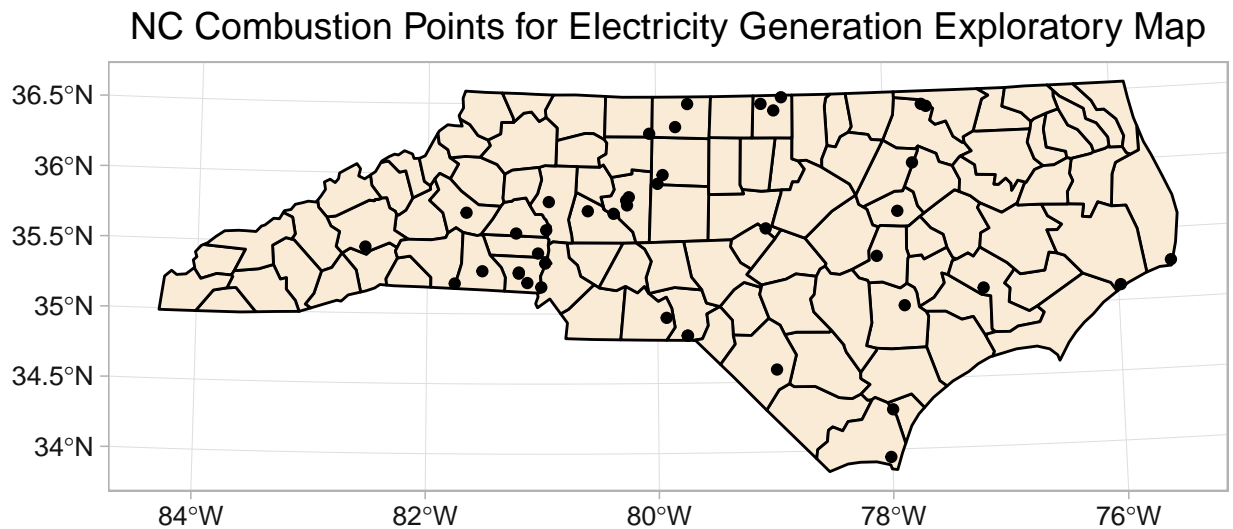
Figure 11: Combustion points for electricity generation in the North Carolina.

```
#Convert all to UTM Zone 17 (crs = 26917)
EPA_US_CombEmissions_utm <- st_transform(EPA_US_CombEmissions, c=26917)

#Clip the EPA_US_CombEmissions data set by the NC State boundary dataset

#First create a State_sf file
#Aggregate the data using group_by and summarize, just as you would a non-spatial data
state_sf_utm <- st_union(counties_sf_utm)

#Eliminate the emission points outside NC
EPA_NC_CombEmissions_utm <- st_intersection(EPA_US_CombEmissions_utm,state_sf_utm)
```

Visual data exploration of the EPA combustion points for electricity generation in the North Carolina in Figure 7, Figure 8, Figure 11, and Figure 10..

Now the distance between PM2.5 stations and Electricity Generation via Combustion points is determined and added to the PM2.5_PM10_Temp_sf_utm dataframe.

```
#Distances between PM2.5 stations and Electricity Generation via Combustion points
Distances <- st_distance(PM2.5_Stations, EPA_NC_CombEmissions_utm)
```

```r
a <- length(unique(PM2.5_Stations$Site.ID))

#Determining the minimum distance of each PM2.5 station to a combustion point in meter
for (i in 1:a){
  PM2.5_Stations$Emiss_Dist[i] <- min(Distances[i,])
}

#Filling the PM2.5_PM10_Temp_sf_utm file with the distances

#Drops the geo data
aux4 <- PM2.5_Stations %>%
  st_set_geometry(value=NULL) %>%
  select(Site.ID,Emiss_Dist)

#Left_join the data
PM2.5_Full_utm <- PM2.5_PM10_Temp_sf_utm %>%
left_join(aux4, by = c("Site.ID")) %>%
  select(Date,Site.ID,COUNTY,Population,Zone,PM2.5,PM10,TAVG,Emiss_Dist,geometry)
```

Finally, using the elevatr package, elevation information is added to the PM 2.5 stations in the PM2.5_Full_utm dataframe, creating the PM2.5_Full_Elev_utm, which is saved in the Project folder ./Data/Processed.

Elevations for the PM2.5 Stations

```r
prj_dd <- "+proj=utm +zone=17 +ellps=GRS80 +towgs84=0,0,0,0,0,0,0 +units=m +no_defs"
PM2.5_Full_Elev_utm <- get_elev_point(PM2.5_Full_utm, prj = prj_dd, src = "epqs")

st_write(PM2.5_Full_Elev_utm,
         "./Data/Processed/PM2.5_Full_Elev_utm.shp", driver = "ESRI Shapefile")

PM2.5_Full_Elev_utm <- st_read('./Data/Processed/PM2.5_Full_Elev_utm.shp')
```

```
## Reading layer `PM2.5_Full_Elev_utm' from data source `C:\Users\Felipe\OneDrive - Duke
## Simple feature collection with 6499 features and 11 fields
## geometry type:  POINT
## dimension:      XY
## bbox:           xmin: 278314.3 ymin: 3807066 xmax: 935107.5 ymax: 3996722
## epsg (SRID):    NA
## proj4string:    +proj=utm +zone=17 +ellps=GRS80 +units=m +no_defs
```

3.6   Additional previsualization of the data

The next chunk is to start looking at the data

```r
#Spatially
ggplot() +
  annotation_map_tile(zoom = 7) +
  geom_sf(data = counties_sf_utm_simple, color = 'Black', aes(fill=Zone), alpha=0.7) +
  geom_sf(data = NOAA_DTAVG_NC2018_sf_utm, color = 'Blue') +
  geom_sf(data = EPA_NC_CombEmissions_utm, color = 'Black') +
  scale_fill_brewer(palette = "Blues", name = "Zoning")   +
  xlab(expression("Longitud")) +
  ylab(expression("Latitude")) +
  ggtitle("Data Visualization")


#Scatterplot
ggplot(NOAA_DTAVG_NC2018_sf_utm, aes(x=DATE, y=TAVG)) +
  geom_point(aes(color = ELEVATION)) +
  xlab(expression("Date")) +
  ylab(expression("Average Daily Temp °F")) +
  #labs(color = 'Station') +
  scale_color_gradient(low="cyan", high="blue4", name = "Elevation (m)")   +
  ggtitle("Average Temperature in North Carolina 2018")


ggplot(NOAA_DTAVG_NC2018_sf_utm, aes(x=DATE, y=TAVG)) +
  geom_point(aes(color = Zone)) +
  xlab(expression("Date")) +
  ylab(expression("Average Daily Temp °F")) +
  #labs(color = 'Station') +
  scale_color_brewer(palette = "Reds", name = "Zone")   +
  ggtitle("Average Temperature in North Carolina 2018")


PM2.5_Dist_Plot <- ggplot(PM2.5_PM10_sf_utm, aes(x=Date, y=PM2.5)) +
  geom_point(aes(color = Emiss_Dist)) +
  geom_smooth(color ="black") +
  xlab(expression("Date")) +
  ylab(expression("PM 2.5 Concentration (\U003BCg/m3)")) +
  scale_color_gradient(low="cyan", high="blue4", name = "Dist. to combustion point (m)")
  ggtitle("2018 Daily PM2.5 concentration, North Carolina") +
  geom_hline(yintercept=12, linetype="dashed", color = "black", size = 1)


PM2.5_Zone_Plot <- ggplot(PM2.5_PM10_sf_utm, aes(x=Date, y=PM2.5)) +
  geom_point(aes(color = Zone)) +
  geom_smooth(color ="black") +
  xlab(expression("Date")) +
  ylab(expression("PM 2.5 Concentration (\U003BCg/m3)")) +
  scale_fill_manual("red","blue","yellow")   +
  ggtitle("2018 Daily PM2.5 concentration, North Carolina") +
```

```r
  geom_hline(yintercept=12, linetype="dashed", color = "black", size = 1)

PM10_Dist_Plot <- ggplot(PM2.5_PM10_sf_utm, aes(x=Date, y=PM10)) +
  geom_point(aes(color = Emiss_Dist)) +
  geom_smooth(color ="black") +
  xlab(expression("Date")) +
  ylab(expression("PM 10 Concentration (\U003BCg/m3)")) +
  scale_color_gradient(low="cyan", high="blue4", name = "Dist. to combustion point (m)")
  ggtitle("2018 Daily PM10 concentration, North Carolina")

grid.arrange(PM2.5_Dist_Plot, PM10_Dist_Plot, PM2.5_Zone_Plot, nrow = 3)

ggplot(PM2.5_PM10_Temp_sf_utm_elev, aes(x=Date, y=PM2.5)) +
  geom_point(aes(color = elevation)) +
  geom_smooth(color ="black") +
  xlab(expression("Date")) +
  ylab(expression("PM 10 Concentration (\U003BCg/m3)")) +
  scale_color_gradient(low="cyan", high="blue4", name = "Elev. (m)")  +
  ggtitle("2018 Daily PM10 concentration, North Carolina")

ggplot(PM2.5_PM10_Temp_sf_utm_elev, aes(x=Date, y=PM2.5)) +
  geom_point(aes(color = TAVG)) +
  geom_smooth(color ="black") +
  xlab(expression("Date")) +
  ylab(expression("PM 10 Concentration (\U003BCg/m3)")) +
  scale_color_gradient(low="cyan", high="blue4", name = "Temp (°F)")  +
  ggtitle("2018 Daily PM10 concentration, North Carolina")




#PM 2.5 Regulatory Standard. Based on a yearly average value set at 12 micrograms per

#Plot counties with a new variable (zone)
ggplot() +
  geom_sf(data = counties_sf_utm_simple, aes(fill=Zone))




summary(PM2.5_Stations$Emiss_Dist)
```

4   Analysis

In 2012, the United States Environmental Protection Agency (USEPA) established two complementary primary regulatory standards for PM2.5. The first is based on a yearly average value and is set at 12 micrograms per cubic meter, ug/m3,

First statistical test should look at the standard in each station.

FRA: This model structure should look familiar, with a typical linear model structure and dataframe defined. The addition here is that we have defined Week as a random variable. Essentially, we are interested not in the specific effects of each week but in the variability among weeks, so we have defined it as a random effect (essentially coming from a larger distribution of seasonal variability).

5   Summary and Conclusions