

Assignment 4: Data Wrangling

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OVERVIEW

This exercise accompanies the lessons in Environmental Data Analytics on Data Wrangling

Directions

1. Rename this file `<FirstLast>_A04_DataExploration.Rmd` (replacing `<FirstLast>` with your first and last name).
2. Change “Student Name” on line 3 (above) with your name.
3. Work through the steps, **creating code and output** that fulfill each instruction.
4. Be sure to **answer the questions** in this assignment document.
5. When you have completed the assignment, **Knit** the text and code into a single PDF file.

The completed exercise is due on Friday, Oct7th @ 5:00pm.

Set up your session

1. Check your working directory, load the `tidyverse` and `lubridate` packages, and upload all four raw data files associated with the EPA Air dataset, being sure to set string columns to be read in a factors. See the README file for the EPA air datasets for more information (especially if you have not worked with air quality data previously).
2. Explore the dimensions, column names, and structure of the datasets.

```
# 1
```

```
getwd()
```

```
## [1] "C:/Users/leor9/OneDrive/Leonardo/MIDP Courses Fall 2022/R Class/EDA-Fall2022/Assignments"
```

```
setwd("C:/Users/leor9/OneDrive/Leonardo/MIDP Courses Fall 2022/R Class/EDA-Fall2022")
```

```
library(tidyverse)
```

```
library(lubridate)
```

```
EPAair_03_NC2018 <- read.csv("../Data/Raw/EPAair_03_NC2018_raw.csv",  
  stringsAsFactors = TRUE)
```

```
EPAair_03_NC2019 <- read.csv("../Data/Raw/EPAair_03_NC2019_raw.csv",  
  stringsAsFactors = TRUE)
```

```
EPAair_PM25_NC2018 <- read.csv("./Data/Raw/EPAair_PM25_NC2018_raw.csv",
  stringsAsFactors = TRUE)
```

```
EPAair_PM25_NC2019 <- read.csv("./Data/Raw/EPAair_PM25_NC2019_raw.csv",
  stringsAsFactors = TRUE)
```

```
# 2
```

```
# Dimensions, column names, and structure of  
# EPAair_03_NC2018 dataset
```

```
dim(EPAair_03_NC2018)
```

```
## [1] 9737 20
```

```
colnames(EPAair_03_NC2018)
```

```
## [1] "Date"  
## [2] "Source"  
## [3] "Site.ID"  
## [4] "POC"  
## [5] "Daily.Max.8.hour.Ozone.Concentration"  
## [6] "UNITS"  
## [7] "DAILY_AQI_VALUE"  
## [8] "Site.Name"  
## [9] "DAILY_OBS_COUNT"  
## [10] "PERCENT_COMPLETE"  
## [11] "AQS_PARAMETER_CODE"  
## [12] "AQS_PARAMETER_DESC"  
## [13] "CBSA_CODE"  
## [14] "CBSA_NAME"  
## [15] "STATE_CODE"  
## [16] "STATE"  
## [17] "COUNTY_CODE"  
## [18] "COUNTY"  
## [19] "SITE_LATITUDE"  
## [20] "SITE_LONGITUDE"
```

```
str(EPAair_03_NC2018)
```

```
## 'data.frame': 9737 obs. of 20 variables:  
## $ Date : Factor w/ 364 levels "01/01/2018","01/02/2018",...: 60 61 62  
## $ Source : Factor w/ 1 level "AQS": 1 1 1 1 1 1 1 1 1 1 ...  
## $ Site.ID : int 370030005 370030005 370030005 370030005 370030005 370030005  
## $ POC : int 1 1 1 1 1 1 1 1 1 1 ...  
## $ Daily.Max.8.hour.Ozone.Concentration: num 0.043 0.046 0.047 0.049 0.047 0.03 0.036 0.044 0.049 0  
## $ UNITS : Factor w/ 1 level "ppm": 1 1 1 1 1 1 1 1 1 1 ...  
## $ DAILY_AQI_VALUE : int 40 43 44 45 44 28 33 41 45 40 ...  
## $ Site.Name : Factor w/ 40 levels "", "Beaufort",...: 35 35 35 35 35 35 35  
## $ DAILY_OBS_COUNT : int 17 17 17 17 17 17 17 17 17 17 ...  
## $ PERCENT_COMPLETE : num 100 100 100 100 100 100 100 100 100 100 ...  
## $ AQS_PARAMETER_CODE : int 44201 44201 44201 44201 44201 44201 44201 44201 44201 4
```

```
## $ AQS_PARAMETER_DESC      : Factor w/ 1 level "Ozone": 1 1 1 1 1 1 1 1 1 1 ...
## $ CBSA_CODE                : int   25860 25860 25860 25860 25860 25860 25860 25860 25860 25860 ...
## $ CBSA_NAME                : Factor w/ 17 levels "","Asheville, NC",...: 9 9 9 9 9 9 9 9 9 9 ...
## $ 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 1 1 ...
## $ COUNTY_CODE              : int     3 3 3 3 3 3 3 3 3 3 ...
## $ COUNTY                   : Factor w/ 32 levels "Alexander","Avery",...: 1 1 1 1 1 1 1 1 1 1 ...
## $ SITE_LATITUDE            : num   35.9 35.9 35.9 35.9 35.9 ...
## $ SITE_LONGITUDE           : num  -81.2 -81.2 -81.2 -81.2 -81.2 ...
```

```
# Dimensions, column names, and structure of
# EPAair_03_NC2019 dataset
```

```
dim(EPAair_03_NC2019)
```

```
## [1] 10592    20
```

```
colnames(EPAair_03_NC2019)
```

```
## [1] "Date"
## [2] "Source"
## [3] "Site.ID"
## [4] "POC"
## [5] "Daily.Max.8.hour.Ozone.Concentration"
## [6] "UNITS"
## [7] "DAILY_AQI_VALUE"
## [8] "Site.Name"
## [9] "DAILY_OBS_COUNT"
## [10] "PERCENT_COMPLETE"
## [11] "AQS_PARAMETER_CODE"
## [12] "AQS_PARAMETER_DESC"
## [13] "CBSA_CODE"
## [14] "CBSA_NAME"
## [15] "STATE_CODE"
## [16] "STATE"
## [17] "COUNTY_CODE"
## [18] "COUNTY"
## [19] "SITE_LATITUDE"
## [20] "SITE_LONGITUDE"
```

```
str(EPAair_03_NC2019)
```

```
## 'data.frame':    10592 obs. of  20 variables:
## $ Date                : Factor w/ 365 levels "01/01/2019","01/02/2019",...: 1 2 3 4 ...
## $ Source               : Factor w/ 2 levels "AirNow","AQS": 1 1 1 1 1 1 1 1 1 1 ...
## $ Site.ID              : int   370030005 370030005 370030005 370030005 370030005 370030005 ...
## $ POC                  : int    1 1 1 1 1 1 1 1 1 1 ...
## $ Daily.Max.8.hour.Ozone.Concentration: num   0.029 0.018 0.016 0.022 0.037 0.037 0.029 0.038 0.038 ...
## $ UNITS                : Factor w/ 1 level "ppm": 1 1 1 1 1 1 1 1 1 1 ...
## $ DAILY_AQI_VALUE       : int    27 17 15 20 34 34 27 35 35 28 ...
## $ Site.Name            : Factor w/ 38 levels "","Beaufort",...: 33 33 33 33 33 33 33 33 33 33 ...
## $ DAILY_OBS_COUNT       : int    24 24 24 24 24 24 24 24 24 24 ...
```

```
## $ PERCENT_COMPLETE      : num  100 100 100 100 100 100 100 100 100 100 ...
## $ AQS_PARAMETER_CODE    : int   44201 44201 44201 44201 44201 44201 44201 44201 44201 44201 ...
## $ AQS_PARAMETER_DESC    : Factor w/ 1 level "Ozone": 1 1 1 1 1 1 1 1 1 1 ...
## $ CBSA_CODE             : int   25860 25860 25860 25860 25860 25860 25860 25860 25860 25860 ...
## $ CBSA_NAME             : Factor w/ 15 levels "","Asheville, NC",...: 8 8 8 8 8 8 8 8 8 8 ...
## $ 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 1 1 ...
## $ COUNTY_CODE           : int     3 3 3 3 3 3 3 3 3 3 ...
## $ COUNTY                : Factor w/ 30 levels "Alexander","Avery",...: 1 1 1 1 1 1 1 1 1 1 ...
## $ SITE_LATITUDE         : num   35.9 35.9 35.9 35.9 35.9 ...
## $ SITE_LONGITUDE        : num  -81.2 -81.2 -81.2 -81.2 -81.2 ...
```

```
# Dimensions, column names, and structure of
# EPAair_PM25_NC2018 dataset
```

```
dim(EPAair_PM25_NC2018)
```

```
## [1] 8983    20
```

```
colnames(EPAair_PM25_NC2018)
```

```
## [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"
```

```
str(EPAair_PM25_NC2018)
```

```
## 'data.frame':    8983 obs. of  20 variables:
## $ Date              : Factor w/ 365 levels "01/01/2018","01/02/2018",...: 2 5 8 11 14 17 ...
## $ Source            : Factor w/ 1 level "AQS": 1 1 1 1 1 1 1 1 1 1 ...
## $ Site.ID           : int   370110002 370110002 370110002 370110002 370110002 370110002 370110002 370110002 370110002 370110002 ...
## $ 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 ...
## $ DAILY_AQI_VALUE    : int   12 15 22 3 10 19 8 10 18 7 ...
## $ Site.Name          : Factor w/ 25 levels "","Blackstone",...: 15 15 15 15 15 15 15 15 15 15 ...
## $ 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 88502 88502 88502 ...
## $ AQS_PARAMETER_DESC : Factor w/ 2 levels "Acceptable PM2.5 AQI & Speciation Mass",...: 1 1 1 1 1 1 1 1 1 1 ...
## $ 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 1 1 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 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 1 1 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 ...
```

```
# Dimensions, column names, and structure of
# EPAair_PM25_NC2019 dataset
```

```
dim(EPAair_PM25_NC2019)
```

```
## [1] 8581 20
```

```
colnames(EPAair_PM25_NC2019)
```

```
## [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"
```

```
str(EPAair_PM25_NC2019)
```

```
## 'data.frame': 8581 obs. of 20 variables:
## $ Date : Factor w/ 365 levels "01/01/2019","01/02/2019",...: 3 6 9 12 15 18 ...
## $ Source : Factor w/ 2 levels "AirNow","AQS": 2 2 2 2 2 2 2 2 2 2 ...
## $ Site.ID : int 370110002 370110002 370110002 370110002 370110002 370110002 370110002 370110002 370110002 370110002 ...
## $ POC : int 1 1 1 1 1 1 1 1 1 1 ...
## $ Daily.Mean.PM2.5.Concentration: num 1.6 1 1.3 6.3 2.6 1.2 1.5 1.5 3.7 1.6 ...
## $ UNITS : Factor w/ 1 level "ug/m3 LC": 1 1 1 1 1 1 1 1 1 1 ...
## $ DAILY_AQI_VALUE : int 7 4 5 26 11 5 6 6 15 7 ...
## $ Site.Name : Factor w/ 25 levels "", "Board Of Ed. Bldg.",...: 14 14 14 14 14 14 14 14 14 14 ...
## $ 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 88502 88502 88502 ...
## $ AQS_PARAMETER_DESC : Factor w/ 2 levels "Acceptable PM2.5 AQI & Speciation Mass",...: 1 1 1 1 1 1 1 1 1 1 ...
## $ 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 1 1 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 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 1 1 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 ...
```

Wrangle individual datasets to create processed files.

3. Change date to date

4. Select the following columns: Date, DAILY_AQI_VALUE, Site.Name, AQS_PARAMETER_DESC, COUNTY, SITE_LATITUDE, SITE_LONGITUDE
5. For the PM2.5 datasets, fill all cells in AQS_PARAMETER_DESC with "PM2.5" (all cells in this column should be identical).
6. Save all four processed datasets in the Processed folder. Use the same file names as the raw files but replace "raw" with "processed".

3

```
EPAair_03_NC2018$Date <- mdy(EPAair_03_NC2018$Date)
EPAair_03_NC2019$Date <- mdy(EPAair_03_NC2019$Date)
EPAair_PM25_NC2018$Date <- mdy(EPAair_PM25_NC2018$Date)
EPAair_PM25_NC2019$Date <- mdy(EPAair_PM25_NC2019$Date)
```

4

```
EPAair_03_NC2018 <- select(EPAair_03_NC2018, Date,
  DAILY_AQI_VALUE, Site.Name, AQS_PARAMETER_DESC,
  COUNTY, SITE_LATITUDE, SITE_LONGITUDE)

EPAair_03_NC2019 <- select(EPAair_03_NC2019, Date,
  DAILY_AQI_VALUE, Site.Name, AQS_PARAMETER_DESC,
  COUNTY, SITE_LATITUDE, SITE_LONGITUDE)

EPAair_PM25_NC2018 <- select(EPAair_PM25_NC2018,
  Date, DAILY_AQI_VALUE, Site.Name, AQS_PARAMETER_DESC,
  COUNTY, SITE_LATITUDE, SITE_LONGITUDE)

EPAair_PM25_NC2019 <- select(EPAair_PM25_NC2019,
  Date, DAILY_AQI_VALUE, Site.Name, AQS_PARAMETER_DESC,
  COUNTY, SITE_LATITUDE, SITE_LONGITUDE)
```

5

```
EPAair_PM25_NC2018$AQS_PARAMETER_DESC = "PM2.5"
EPAair_PM25_NC2019$AQS_PARAMETER_DESC = "PM2.5"
```

6

```
setwd("C:/Users/leor9/OneDrive/Leonardo/MIDP Courses Fall 2022/R Class/EDA-Fall2022")

write.csv(EPAair_03_NC2018, row.names = FALSE,
  file = "./Data/Processed/EPAair_03_NC2018_Processed.csv")

write.csv(EPAair_03_NC2019, row.names = FALSE,
  file = "./Data/Processed/EPAair_03_NC2019_Processed.csv")

write.csv(EPAair_PM25_NC2018, row.names = FALSE,
  file = "./Data/Processed/EPAair_PM25_NC2018_Processed.csv")

write.csv(EPAair_PM25_NC2019, row.names = FALSE,
  file = "./Data/Processed/EPAair_PM25_NC2019_Processed.csv")
```

Combine datasets

7. Combine the four datasets with `rbind`. Make sure your column names are identical prior to running this code.
8. Wrangle your new dataset with a pipe function (`%>%`) so that it fills the following conditions:
 - Include all sites that the four data frames have in common: “Linville Falls”, “Durham Armory”, “Leggett”, “Hattie Avenue”, “Clemmons Middle”, “Mendenhall School”, “Frying Pan Mountain”, “West Johnston Co.”, “Garinger High School”, “Castle Hayne”, “Pitt Agri. Center”, “Bryson City”, “Millbrook School” (the function `intersect` can figure out common factor levels)
 - Some sites have multiple measurements per day. Use the split-apply-combine strategy to generate daily means: group by date, site, aqs parameter, and county. Take the mean of the AQI value, latitude, and longitude.
 - Add columns for “Month” and “Year” by parsing your “Date” column (hint: `lubridate` package)
 - Hint: the dimensions of this dataset should be 14,752 x 9.
9. Spread your datasets such that AQI values for ozone and PM2.5 are in separate columns. Each location on a specific date should now occupy only one row.
10. Call up the dimensions of your new tidy dataset.
11. Save your processed dataset with the following file name: “EPAair_O3_PM25_NC1718_Processed.csv”

7

```
EPAair_NC1819 <- rbind(EPAair_O3_NC2018, EPAair_O3_NC2019,  
  EPAair_PM25_NC2018, EPAair_PM25_NC2019)
```

8

```
EPAair_NC1819_filtered <- filter(EPAair_NC1819,  
  Site.Name %in% c("Linville Falls", "Durham Armory",  
    "Leggett", "Hattie Avenue", "Clemmons Middle",  
    "Mendenhall School", "Frying Pan Mountain",  
    "West Johnston Co.", "Garinger High School",  
    "Castle Hayne", "Pitt Agri. Center", "Bryson City",  
    "Millbrook School")) %>%  
  group_by(Date, Site.Name, AQS_PARAMETER_DESC,  
    COUNTY) %>%  
  summarise(mean_DAILY_AQI_VALUE = mean(DAILY_AQI_VALUE),  
    mean_latitude = mean(SITE_LATITUDE), mean_longitude = mean(SITE_LONGITUDE)) %>%  
  mutate(month = month(Date), year = year(Date))
```

```
## 'summarise()' has grouped output by 'Date', 'Site.Name', 'AQS_PARAMETER_DESC'.  
## You can override using the '.groups' argument.
```

9

```
EPAair_NC1819_filtered_spread <- pivot_wider(EPAair_NC1819_filtered,  
  names_from = AQS_PARAMETER_DESC, values_from = mean_DAILY_AQI_VALUE)
```

10

```
dim(EPAair_NC1819_filtered_spread)
```

```
## [1] 8976    9
```

```
# 11

setwd("C:/Users/leor9/OneDrive/Leonardo/MIDP Courses Fall 2022/R Class/EDA-Fall2022")

write.csv(EPAair_NC1819_filtered_spread, row.names = FALSE,
          file = "./Data/Processed/EPAair_03_PM25_NC1718_Processed.csv")
```

Generate summary tables

12. Use the split-apply-combine strategy to generate a summary data frame. Data should be grouped by site, month, and year. Generate the mean AQI values for ozone and PM2.5 for each group. Then, add a pipe to remove instances where a month and year are not available (use the function `drop_na` in your pipe).
13. Call up the dimensions of the summary dataset.

```
# 12a

EPAair_NC1819_filtered_spread_summary <- EPAair_NC1819_filtered_spread %>%
  group_by(Site.Name, month, year) %>%
  summarise(mean_AQI_Value_Ozone = mean(Ozone),
            mean_AQI_Value_PM2.5 = mean(PM2.5)) %>%
  drop_na(mean_AQI_Value_Ozone, mean_AQI_Value_PM2.5)
```

```
## 'summarise()' has grouped output by 'Site.Name', 'month'. You can override
## using the '.groups' argument.
```

```
# 12b

# 13

dim(EPAair_NC1819_filtered_spread_summary)
```

```
## [1] 101    5
```

14. Why did we use the function `drop_na` rather than `na.omit`?

Answer: We use `drop_na` instead of `na.omit` because the first function eliminates the missing values of a specific column of a data frame, while `drop_na` remove rows with missing data in the columns specified inside the function.