

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_DataWrangling.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.
6. Ensure that code in code chunks does not extend off the page in the PDF.

Set up your session

- 1a. Load the `tidyverse`, `lubridate`, and `here` packages into your session.
 - 1b. Check your working directory.
 - 1c. Read in all four raw data files associated with the EPA Air dataset, being sure to set string columns to be read in as 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. Apply the `glimpse()` function to reveal the dimensions, column names, and structure of each dataset.

```
#1a
# Loading tidyverse, lubridate, and here libraries
library(tidyverse)
library(lubridate)
library(here)

#1b
# check working directory. THIS NEEDS TO BE CHECKED DURING OH!!!
getwd()
```

```
## [1] "/Users/sarah/Documents/872_EDA/EDA_Spring2024"
```

```
#1c
# Calling the four EPA Datasets
EPA_03_NC2018 <- read.csv(
  "./Data/Raw/EPAair_03_NC2018_Raw.csv",
```

```

stringsAsFactors = TRUE
)

EPA_O3_NC2019 <- read.csv(
  "./Data/Raw/EPAair_O3_NC2019_Raw.csv",
  stringsAsFactors = TRUE
)

EPA_PM25_NC2018 <- read.csv(
  "./Data/Raw/EPAair_PM25_NC2018_Raw.csv",
  stringsAsFactors = TRUE
)

EPA_PM25_NC2019 <- read.csv(
  "./Data/Raw/EPAair_PM25_NC2019_Raw.csv",
  stringsAsFactors = TRUE
)

#2
# Used glimpse() function to see four EPA datasets
glimpse(
  EPA_O3_NC2018
)

```

```

## Rows: 9,737
## Columns: 20
## $ Date                <fct> 03/01/2018, 03/02/2018, 03/03/201~
## $ Source              <fct> AQS, AQS, AQS, AQS, AQS, AQS, AQS~
## $ Site.ID             <int> 370030005, 370030005, 370030005, ~
## $ POC                 <int> 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, ~
## $ Daily.Max.8.hour.Ozone.Concentration <dbl> 0.043, 0.046, 0.047, 0.049, 0.047~
## $ UNITS               <fct> ppm, ppm, ppm, ppm, ppm, ppm, ppm~
## $ DAILY_AQI_VALUE     <int> 40, 43, 44, 45, 44, 28, 33, 41, 4~
## $ Site.Name           <fct> Taylorsville Liledoun, Taylorsvil~
## $ DAILY_OBS_COUNT     <int> 17, 17, 17, 17, 17, 17, 17, 17, 1~
## $ PERCENT_COMPLETE    <dbl> 100, 100, 100, 100, 100, 100, 100~
## $ AQS_PARAMETER_CODE  <int> 44201, 44201, 44201, 44201, 44201~
## $ AQS_PARAMETER_DESC  <fct> Ozone, Ozone, Ozone, Ozone, Ozone~
## $ CBSA_CODE           <int> 25860, 25860, 25860, 25860, 25860~
## $ CBSA_NAME           <fct> "Hickory-Lenoir-Morganton, NC", "~
## $ STATE_CODE          <int> 37, 37, 37, 37, 37, 37, 37, 37, 3~
## $ STATE               <fct> North Carolina, North Carolina, N~
## $ COUNTY_CODE         <int> 3, 3, 3, 3, 3, 3, 3, 3, 3, 3, ~
## $ COUNTY              <fct> Alexander, Alexander, Alexander, ~
## $ SITE_LATITUDE       <dbl> 35.9138, 35.9138, 35.9138, 35.913~
## $ SITE_LONGITUDE      <dbl> -81.191, -81.191, -81.191, -81.19~

```

```

glimpse(
  EPA_O3_NC2019
)

```

```

## Rows: 10,592
## Columns: 20
## $ Date                <fct> 01/01/2019, 01/02/2019, 01/03/201~

```

```
## $ Source <fct> AirNow, AirNow, AirNow, AirNow, A~
## $ Site.ID <int> 370030005, 370030005, 370030005, ~
## $ POC <int> 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, ~
## $ Daily.Max.8.hour.Ozone.Concentration <dbl> 0.029, 0.018, 0.016, 0.022, 0.037~
## $ UNITS <fct> ppm, ppm, ppm, ppm, ppm, ppm, ppm~
## $ DAILY_AQI_VALUE <int> 27, 17, 15, 20, 34, 34, 27, 35, 3~
## $ Site.Name <fct> Taylorsville Liledoun, Taylorsvil~
## $ DAILY_OBS_COUNT <int> 24, 24, 24, 24, 24, 24, 24, 24, 2~
## $ PERCENT_COMPLETE <dbl> 100, 100, 100, 100, 100, 100, 100~
## $ AQS_PARAMETER_CODE <int> 44201, 44201, 44201, 44201, 44201~
## $ AQS_PARAMETER_DESC <fct> Ozone, Ozone, Ozone, Ozone, Ozone~
## $ CBSA_CODE <int> 25860, 25860, 25860, 25860, 25860~
## $ CBSA_NAME <fct> "Hickory-Lenoir-Morganton, NC", "~
## $ STATE_CODE <int> 37, 37, 37, 37, 37, 37, 37, 37, 3~
## $ STATE <fct> North Carolina, North Carolina, N~
## $ COUNTY_CODE <int> 3, 3, 3, 3, 3, 3, 3, 3, 3, ~
## $ COUNTY <fct> Alexander, Alexander, Alexander, ~
## $ SITE_LATITUDE <dbl> 35.9138, 35.9138, 35.9138, 35.913~
## $ SITE_LONGITUDE <dbl> -81.191, -81.191, -81.191, -81.19~
```

```
glimpse(
  EPA_PM25_NC2018
)
```

```
## Rows: 8,983
## Columns: 20
## $ Date <fct> 01/02/2018, 01/05/2018, 01/08/2018, 01/~
## $ Source <fct> AQS, AQS, AQS, AQS, AQS, AQS, AQS, AQS,~
## $ Site.ID <int> 370110002, 370110002, 370110002, 370110~
## $ POC <int> 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, ~
## $ Daily.Mean.PM2.5.Concentration <dbl> 2.9, 3.7, 5.3, 0.8, 2.5, 4.5, 1.8, 2.5,~
## $ UNITS <fct> ug/m3 LC, ug/m3 LC, ug/m3 LC, ug/m3 LC,~
## $ DAILY_AQI_VALUE <int> 12, 15, 22, 3, 10, 19, 8, 10, 18, 7, 24~
## $ Site.Name <fct> Linville Falls, Linville Falls, Linvill~
## $ DAILY_OBS_COUNT <int> 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, ~
## $ PERCENT_COMPLETE <dbl> 100, 100, 100, 100, 100, 100, 100, 100,~
## $ AQS_PARAMETER_CODE <int> 88502, 88502, 88502, 88502, 88502, 8850~
## $ AQS_PARAMETER_DESC <fct> Acceptable PM2.5 AQI & Speciation Mass,~
## $ CBSA_CODE <int> NA, NA, NA, NA, NA, NA, NA, NA, NA, NA,~
## $ CBSA_NAME <fct> "", "", "", "", "", "", "", "", "", "",~
## $ STATE_CODE <int> 37, 37, 37, 37, 37, 37, 37, 37, 37, 37,~
## $ STATE <fct> North Carolina, North Carolina, North C~
## $ COUNTY_CODE <int> 11, 11, 11, 11, 11, 11, 11, 11, 11, 11,~
## $ COUNTY <fct> Avery, Avery, Avery, Avery, Avery, Aver~
## $ SITE_LATITUDE <dbl> 35.97235, 35.97235, 35.97235, 35.97235,~
## $ SITE_LONGITUDE <dbl> -81.93307, -81.93307, -81.93307, -81.93~
```

```
glimpse(
  EPA_PM25_NC2019
)
```

```
## Rows: 8,581
## Columns: 20
```

```
## $ Date          <fct> 01/03/2019, 01/06/2019, 01/09/2019, 01/~
## $ Source        <fct> AQS, AQS, AQS, AQS, AQS, AQS, AQS, AQS,~
## $ Site.ID       <int> 370110002, 370110002, 370110002, 370110~
## $ POC           <int> 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, ~
## $ Daily.Mean.PM2.5.Concentration <dbl> 1.6, 1.0, 1.3, 6.3, 2.6, 1.2, 1.5, 1.5,~
## $ UNITS         <fct> ug/m3 LC, ug/m3 LC, ug/m3 LC, ug/m3 LC,~
## $ DAILY_AQI_VALUE <int> 7, 4, 5, 26, 11, 5, 6, 6, 15, 7, 14, 20~
## $ Site.Name     <fct> Linville Falls, Linville Falls, Linvill~
## $ DAILY_OBS_COUNT <int> 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, ~
## $ PERCENT_COMPLETE <dbl> 100, 100, 100, 100, 100, 100, 100, 100,~
## $ AQS_PARAMETER_CODE <int> 88502, 88502, 88502, 88502, 88502, 8850~
## $ AQS_PARAMETER_DESC <fct> Acceptable PM2.5 AQI & Speciation Mass,~
## $ CBSA_CODE      <int> NA, NA, NA, NA, NA, NA, NA, NA, NA, NA,~
## $ CBSA_NAME      <fct> "", "", "", "", "", "", "", "", "", "",~
## $ STATE_CODE     <int> 37, 37, 37, 37, 37, 37, 37, 37, 37, 37,~
## $ STATE          <fct> North Carolina, North Carolina, North C~
## $ COUNTY_CODE    <int> 11, 11, 11, 11, 11, 11, 11, 11, 11, 11,~
## $ COUNTY         <fct> Avery, Avery, Avery, Avery, Avery, Aver~
## $ SITE_LATITUDE  <dbl> 35.97235, 35.97235, 35.97235, 35.97235,~
## $ SITE_LONGITUDE <dbl> -81.93307, -81.93307, -81.93307, -81.93~
```

Wrangle individual datasets to create processed files.

3. Change the Date columns to be date objects.
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
# Used as.Date() to turn Date Columns to date objects
EPA_03_NC2018$Date <- as.Date(EPA_03_NC2018$Date, format = "%m/%d/%Y")
EPA_03_NC2019$Date <- as.Date(EPA_03_NC2019$Date, format = "%m/%d/%Y")
EPA_PM25_NC2018$Date <- as.Date(EPA_PM25_NC2018$Date, format = "%m/%d/%Y")
EPA_PM25_NC2019$Date <- as.Date(EPA_PM25_NC2019$Date, format = "%m/%d/%Y")

#4
#Creating new datasets based on the specified columns asked for
EPA_03_NC2018_df <- select(EPA_03_NC2018,
                           Date,
                           DAILY_AQI_VALUE,
                           Site.Name,
                           AQS_PARAMETER_DESC,
                           COUNTY:SITE_LONGITUDE
                           )

EPA_03_NC2019_df <- select(EPA_03_NC2019,
                           Date,
                           DAILY_AQI_VALUE,
```

```

        Site.Name,
        AQS_PARAMETER_DESC,
        COUNTY:SITE_LONGITUDE
    )

EPA_PM25_NC2018_df <- select(EPA_PM25_NC2018,
    Date,
    DAILY_AQI_VALUE,
    Site.Name,
    AQS_PARAMETER_DESC,
    COUNTY:SITE_LONGITUDE
)

EPA_PM25_NC2019_df <- select(EPA_PM25_NC2019,
    Date,
    DAILY_AQI_VALUE,
    Site.Name,
    AQS_PARAMETER_DESC,
    COUNTY:SITE_LONGITUDE
)

#5
# Used pipes to mutate the AQS_PARAMETER_DESC column by changing the descriptions, manually, to PM2.5
# Used the same process for each dataset
EPA_PM25_NC2018_df <- EPA_PM25_NC2018_df %>%
    mutate(AQS_PARAMETER_DESC = fct_recode(AQS_PARAMETER_DESC,
        "PM2.5" = "PM2.5 - Local Conditions",
        "PM2.5" = "Acceptable PM2.5 AQI & Speciation Mass"
    ))

EPA_PM25_NC2019_df <- EPA_PM25_NC2019_df %>%
    mutate(AQS_PARAMETER_DESC = fct_recode(AQS_PARAMETER_DESC,
        "PM2.5" = "PM2.5 - Local Conditions",
        "PM2.5" = "Acceptable PM2.5 AQI & Speciation Mass"
    ))

#6
# Saving the four datasets into the processed folder
write.csv(
    EPA_O3_NC2018_df,
    row.names = FALSE,
    file = "./Data/Processed/EPAair_O3_NC2018_Processed.csv"
)

write.csv(
    EPA_O3_NC2019_df,
    row.names = FALSE,
    file = "./Data/Processed/EPAair_O3_NC2019_Processed.csv"
)

write.csv(
    EPA_PM25_NC2018_df,
    row.names = FALSE,
    file = "./Data/Processed/EPAair_PM25_NC2018_Processed.csv"
)

```

```
)

write.csv(
  EPA_PM25_NC2019_df,
  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 only 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 - but it will include sites with missing site information, which you don’t want...)
 - Some sites have multiple measurements per day. Use the split-apply-combine strategy to generate daily means: group by date, site name, 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_NC1819_Processed.csv”

```
#7
# Used the rbind() function to combine the four EPA datasets together
EPA_O3_PM25_NC1819 <- rbind(EPA_O3_NC2018_df,
                           EPA_O3_NC2019_df,
                           EPA_PM25_NC2018_df,
                           EPA_PM25_NC2019_df
                           )

#8
EPA_O3_PM25_NC1819_processed <- EPA_O3_PM25_NC1819 %>%
  filter(Site.Name == "Linville Falls" | Site.Name == "Durham Armory" | #Used Filter() of common sites
         Site.Name == "Leggett" | Site.Name == "Hattie Avenue" |
         Site.Name == "Clemmons Middle" | Site.Name == "Mendenhall School" |
         Site.Name == "Frying Pan Mountain" | Site.Name == "West Johnston Co." |
         Site.Name == "Garinger High School" | Site.Name == "Castle Hayne" |
         Site.Name == "Pitt Agri. Center" | Site.Name == "Bryson City"
         | Site.Name == "Millbrook School") %>%
  group_by(Date, #beginning of split-apply-combine strategy)
```

```

      Site.Name,
      AQS_PARAMETER_DESC,
      COUNTY) %>%
summarise(meanAQI = mean(DAILY_AQI_VALUE),
          meanlatitude = mean(SITE_LATITUDE),
          meanlongitude = mean(SITE_LONGITUDE)) %>%
mutate(Month = month(Date), Year = year(Date)) #Creating the Month and Year columns with mutate()

```

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

```

#9
# Separated PM2.5 and Ozone to become one row per day per location
EPA_03_PM25_NC1819_processed.spread <- pivot_wider(EPA_03_PM25_NC1819_processed,
                                                    names_from = AQS_PARAMETER_DESC,
                                                    values_from = meanAQI
                                                    )

#10
# 8976 rows and 9 columns
dim(EPA_03_PM25_NC1819_processed.spread)

```

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

```

#11
write.csv(
  EPA_03_PM25_NC1819_processed.spread,
  row.names = FALSE,
  file = "./Data/Processed/EPAair_03_PM25_NC1819_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 mean **ozone** values are not available (use the function `drop_na` in your pipe). It's ok to have missing mean PM2.5 values in this result.

13. Call up the dimensions of the summary dataset.

```

#12
EPA_03_PM25_NC1819.summaries <-
  EPA_03_PM25_NC1819_processed.spread %>%
  group_by(Site.Name, Month, Year) %>% # grouping by site, month, and year
  summarise(mean.O3.AQI = mean(PM2.5),
            mean.PM25.AQI = mean(Ozone)) %>% #creating the mean for ozone and PM2.5
  filter(!is.na(mean.O3.AQI)) # removing where mean ozone is not available

```

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

```
#13
# Dimensions are 211 rows by 5 columns
dim(EPA_03_PM25_NC1819.summaries)
```

```
## [1] 211 5
```

14. Why did we use the function `drop_na` rather than `na.omit`? Hint: replace `drop_na` with `na.omit` in part 12 and observe what happens with the dimensions of the summary data frame.

Answer: We used the `drop_na` function rather than `na.omit` function because it also deleted rows where cells in `mean.PM25.AQI` were `na`. This decreased the amount of rows from 211 to 101.