

Assignment 4: Data Wrangling

Monisha Eadala

OVERVIEW

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

Directions

1. Change “Student Name” on line 3 (above) with your name.
2. Work through the steps, **creating code and output** that fulfill each instruction.
3. Be sure to **answer the questions** in this assignment document.
4. When you have completed the assignment, **Knit** the text and code into a single PDF file.
5. After Knitting, submit the completed exercise (PDF file) to the dropbox in Sakai. Add your last name into the file name (e.g., “Salk_A04_DataWrangling.Rmd”) prior to submission.

The completed exercise is due on Tuesday, February 4 at 1:00 pm.

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. 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() # Checks the working directory

## [1] "/Users/monishaeadala/Environmental_Data_Analytics_2020"

library(tidyverse) # Loads 'tidyverse'
library(lubridate) # Loads `lubridate`

EPA.Air.O3.2018 <- read.csv("./Data/Raw/EPAair_O3_NC2018_raw.csv") # Imports
the file
EPA.Air.O3.2019 <- read.csv("./Data/Raw/EPAair_O3_NC2019_raw.csv") # Imports
the file
EPA.Air.PM25.2018 <- read.csv("./Data/Raw/EPAair_PM25_NC2018_raw.csv") #
Imports the file
EPA.Air.PM25.2019 <- read.csv("./Data/Raw/EPAair_PM25_NC2019_raw.csv") #
Imports the file
```

#2

```
dim(EPA.Air.03.2018) # Checks the dimensions
```

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

```
colnames(EPA.Air.03.2018) # Checks the column names
```

```
## [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] "AQ5_PARAMETER_CODE"
## [12] "AQ5_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(EPA.Air.03.2018) # Checks the structure of the dataset
```

```
## 'data.frame': 9737 obs. of 20 variables:
## $ Date : Factor w/ 364 levels
"01/01/2018","01/02/2018",...: 60 61 62 63 64 65 66 67 68 69 ...
## $ Source : Factor w/ 1 level "AQ5": 1 1 1 1
1 1 1 1 1 1 ...
## $ Site.ID : int 370030005 370030005
370030005 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 0.043 ...
## $ 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 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 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 ...
## $ 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 ...
```

```
dim(EPA.Air.03.2019) # Checks the dimensions
```

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

```
colnames(EPA.Air.03.2019) # Checks the column names
```

```
## [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(EPA.Air.03.2019) # Checks the structure of the dataset
```

```
## 'data.frame':    10592 obs. of  20 variables:
## $ Date              : Factor w/ 365 levels
"01/01/2019", "01/02/2019", ...: 1 2 3 4 5 6 7 8 9 10 ...
```

```

## $ 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 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 0.03 ...
## $ 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 ...

```

```

dim(EPA.Air.PM25.2018) # Checks the dimensions

```

```

## [1] 8983 20

```

```

colnames(EPA.Air.PM25.2018) # Checks the column names

```

```

## [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(EPA.Air.PM25.2018) # Checks the structure of the dataset

## 'data.frame':      8983 obs. of  20 variables:
## $ Date                : Factor w/ 365 levels
"01/01/2018","01/02/2018",...: 2 5 8 11 14 17 20 23 26 29 ...
## $ 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 ...
## $ 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 ...

dim(EPA.Air.PM25.2019) # Checks the dimensions

## [1] 8581    20

colnames(EPA.Air.PM25.2019) # Checks the column names

## [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(EPA.Air.PM25.2019) # Checks the structure of the dataset

## 'data.frame':      8581 obs. of  20 variables:
## $ Date                : Factor w/ 365 levels
"01/01/2019","01/02/2019",...: 3 6 9 12 15 18 21 24 27 30 ...
## $ 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

```
class(EPA.Air.O3.2018$Date) # Checks the class prior to the changing
## [1] "factor"

EPA.Air.O3.2018$Date <- as.Date(EPA.Air.O3.2018$Date, format = "%m/%d/%Y") #
Changes date to date in EPA.Air.O3.2018
class(EPA.Air.O3.2018$Date) # Checks the class after changing the Date to
date
## [1] "Date"

class(EPA.Air.O3.2019$Date) # Checks the class prior to the changing
## [1] "factor"

EPA.Air.O3.2019$Date <- as.Date(EPA.Air.O3.2019$Date, format = "%m/%d/%Y") #
Changes date to date in EPA.Air.O3.2019
class(EPA.Air.O3.2019$Date) # Checks the class after changing the Date to
date
## [1] "Date"

class(EPA.Air.PM25.2018$Date) # Checks the class prior to the changing
## [1] "factor"

EPA.Air.PM25.2018$Date <- as.Date(EPA.Air.PM25.2018$Date, format =
"%m/%d/%Y") # Changes date to date in EPA.Air.PM25.2018
class(EPA.Air.PM25.2018$Date) # Checks the class after changing the Date to
date
## [1] "Date"

class(EPA.Air.PM25.2019$Date) # Checks the class prior to the changing
## [1] "factor"

EPA.Air.PM25.2019$Date <- as.Date(EPA.Air.PM25.2019$Date, format =
"%m/%d/%Y") # Changes date to date in EPA.Air.PM25.2019
class(EPA.Air.PM25.2019$Date) # Checks the class after changing the Date to
date
## [1] "Date"
```

#4

```
EPA.Air.O3.2018.select <- select(EPA.Air.O3.2018, Date, DAILY_AQI_VALUE,
Site.Name, AQS_PARAMETER_DESC, COUNTY, SITE_LATITUDE, SITE_LONGITUDE) #
Selects Date, DAILY_AQI_VALUE, Site.Name, AQS_PARAMETER_DESC, COUNTY,
SITE_LATITUDE, SITE_LONGITUDE from EPA.Air.O3.2018

EPA.Air.O3.2019.select <- select(EPA.Air.O3.2019, Date, DAILY_AQI_VALUE,
Site.Name, AQS_PARAMETER_DESC, COUNTY, SITE_LATITUDE, SITE_LONGITUDE) #
```

```
Selects Date, DAILY_AQI_VALUE, Site.Name, AQS_PARAMETER_DESC, COUNTY,
SITE_LATITUDE, SITE_LONGITUDE from EPA.Air.03.2019
```

```
EPA.Air.PM25.2018.select <- select(EPA.Air.PM25.2018, Date, DAILY_AQI_VALUE,
Site.Name, AQS_PARAMETER_DESC, COUNTY, SITE_LATITUDE, SITE_LONGITUDE) #
Selects Date, DAILY_AQI_VALUE, Site.Name, AQS_PARAMETER_DESC, COUNTY,
SITE_LATITUDE, SITE_LONGITUDE from EPA.Air.PM25.2018
```

```
EPA.Air.PM25.2019.select <- select(EPA.Air.PM25.2019, Date, DAILY_AQI_VALUE,
Site.Name, AQS_PARAMETER_DESC, COUNTY, SITE_LATITUDE, SITE_LONGITUDE) #
Selects Date, DAILY_AQI_VALUE, Site.Name, AQS_PARAMETER_DESC, COUNTY,
SITE_LATITUDE, SITE_LONGITUDE from EPA.Air.PM25.2018
```

```
#5
```

```
EPA.Air.PM25.2018.select <- mutate(EPA.Air.PM25.2018.select,
AQS_PARAMETER_DESC = "PM2.5") # Fills all cells in AQS_PARAMETER_DESC with
"PM2.5" in EPA.Air.PM25.2018.select
```

```
EPA.Air.PM25.2019.select <- mutate(EPA.Air.PM25.2019.select,
AQS_PARAMETER_DESC = "PM2.5") # Fills all cells in AQS_PARAMETER_DESC with
"PM2.5" in EPA.Air.PM25.2019.select
```

```
#6
```

```
write.csv(EPA.Air.03.2018.select, row.names = FALSE, file =
"./Data/Processed/EPAair_03_NC2018_processed.csv") # Saved the
EPA.Air.03.2018.select dataset in the Processed folder by the name
"EPAair_03_NC2018_processed"
```

```
write.csv(EPA.Air.03.2019.select, row.names = FALSE, file =
"./Data/Processed/EPAair_03_NC2019_processed.csv") # Saved the
EPA.Air.03.2019.select dataset in the Processed folder by the name
"EPAair_03_NC2019_processed"
```

```
write.csv(EPA.Air.PM25.2018.select, row.names = FALSE, file =
"./Data/Processed/EPAair_PM25_NC2018_processed.csv") # Saved the
EPA.Air.PM25.2018.select dataset in the Processed folder by the name
"EPAair_PM25_NC2018_processed"
```

```
write.csv(EPA.Air.PM25.2019.select, row.names = FALSE, file =
"./Data/Processed/EPAair_PM25_NC2019_processed.csv") # Saved the
EPA.Air.PM25.2019.select dataset in the Processed folder by the name
"EPAair_PM25_NC2019_processed"
```

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

```
EPA.Air.combined <- rbind(EPA.Air.O3.2018.select, EPA.Air.O3.2019.select,
EPA.Air.PM25.2018.select, EPA.Air.PM25.2019.select) # Combines the four
datasets
```

#8

```
EPA.Air.combined.piped <-
  EPA.Air.combined %>%
    filter(Site.Name == "Linville Falls" | Site.Name == "Durham Armory" |
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") %>%
  dplyr::group_by(Date, Site.Name, AQS_PARAMETER_DESC, COUNTY) %>% # Groups
by date, site, aqs parameter, and county
  dplyr::summarise(meanAQI = mean(DAILY_AQI_VALUE),
                    meanLat = mean(SITE_LATITUDE),
                    meanLong = mean(SITE_LONGITUDE)) %>% # Takes the mean of
the AQI value, latitude, and longitude.
  mutate(Month = month(Date), Year = year(Date)) # makes new columns "Month"
and "Year"
dim(EPA.Air.combined.piped) # Checks the dimensions of the dataset; In this
case they are 14752 by 9
```

```
## [1] 14752      9
```

#9

```
EPA.Air.combined.piped.spread <- spread(EPA.Air.combined.piped,
AQS_PARAMETER_DESC, meanAQI) # Spreads dataset into separate columns 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
dim(EPA.Air.combined.piped.spread) # Checks the dimensions of the dataset; In
this case it is 8976 by 9

## [1] 8976    9

#11
write.csv(EPA.Air.combined.piped.spread, row.names = FALSE, file =
"./Data/Processed/EPAair_03_PM25_NC1718_Processed.csv") # Saves the
EPA.Air.combined.piped dataset in the Processed folder by the name
"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
EPA.Air.combined.piped.spread.summaries <-
  EPA.Air.combined.piped.spread %>%
  dplyr::group_by(Site.Name, Month, Year) %>% # Groups the data by site,
month, and year
  dplyr::summarise(meanO3 = mean(Ozone),
                    meanPM25 = mean(PM2.5)) %>% # Generates the mean AQI values for
ozone and PM2.5 for each group
#12b
  drop_na(Month, Year) # Drops the rows where a month and year are not
available

#13
dim(EPA.Air.combined.piped.spread.summaries) # Checks the dimensions of the
dataset; In this case it is 308 by 5

## [1] 308    5
```

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

Answer: 'na.omit' would omit all the NAs in the dataset, while in 'drop_na' we can select the specific columns for the rows that contains NAs we want to drop. Therefore, we use the 'drop_na' function since we want to remove instances where only the month and year are not available.