Assignment 3: Data Exploration

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OVERVIEW

This exercise accompanies the lessons in Environmental Data Analytics on Data Exploration.

Directions

- 1. Rename this file <FirstLast>_A03_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. Assign a useful name to each code chunk and include ample comments with your code.
- 5. Be sure to **answer the questions** in this assignment document.
- 6. When you have completed the assignment, **Knit** the text and code into a single PDF file.
- 7. After Knitting, submit the completed exercise (PDF file) to the dropbox in Canvas.

TIP: If your code extends past the page when knit, tidy your code by manually inserting line breaks.

TIP: If your code fails to knit, check that no install.packages() or View() commands exist in your code.

Set up your R session

1. Load necessary packages (tidyverse, lubridate, here), check your current working directory and upload two datasets: the ECOTOX neonicotinoid dataset (ECOTOX_Neonicotinoids_Insects_raw.csv) and the Niwot Ridge NEON dataset for litter and woody debris (NEON_NIWO_Litter_massdata_2018-08_raw.csv). Name these datasets "Neonics" and "Litter", respectively. Be sure to include the subcommand to read strings in as factors.

library(tidyverse) #Installing the relevant packages

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr
              1.1.4
                        v readr
                                    2.1.5
## v forcats
              1.0.0
                        v stringr
                                    1.5.1
                        v tibble
## v ggplot2
              3.5.1
                                    3.2.1
## v lubridate 1.9.3
                        v tidyr
                                    1.3.1
## v purrr
              1.0.2
## -- Conflicts ----- tidyverse conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                    masks stats::lag()
## i Use the conflicted package (<a href="http://conflicted.r-lib.org/">http://conflicted.r-lib.org/</a>) to force all conflicts to become error
```

```
library(lubridate)
library(here)
## here() starts at /home/guest/EDE_Fall2024
getwd() #Making sure my directory still connects to Git
## [1] "/home/guest/EDE_Fall2024"
here()
## [1] "/home/guest/EDE Fall2024"
Neonics <- here('Data/Raw/ECOTOX Neonicotinoids Insects raw.csv')
print(Neonics)
## [1] "/home/guest/EDE_Fall2024/Data/Raw/ECOTOX_Neonicotinoids_Insects_raw.csv"
Neonics <- read.csv(</pre>
 file = here('Data/Raw/ECOTOX_Neonicotinoids_Insects_raw.csv'),
  stringsAsFactors = T
Litter <- here('Data/Raw/NEON NIWO Litter massdata 2018-08 raw.csv')
print(Litter)
## [1] "/home/guest/EDE_Fall2024/Data/Raw/NEON_NIWO_Litter_massdata_2018-08_raw.csv"
Litter <- read.csv(</pre>
  file = here('Data/Raw/NEON_NIWO_Litter_massdata_2018-08_raw.csv'),
  stringsAsFactors = T
) #Loaded necessary packages and added the required datasets
```

Learn about your system

2. The neonicotinoid dataset was collected from the Environmental Protection Agency's ECOTOX Knowledgebase, a database for ecotoxicology research. Neonicotinoids are a class of insecticides used widely in agriculture. The dataset that has been pulled includes all studies published on insects. Why might we be interested in the ecotoxicology of neonicotinoids on insects? Feel free to do a brief internet search if you feel you need more background information.

Answer: We might be interested in the ecotoxicology of neonicotoids based on how they effect pollinating insects. If insects that aid in crop reproduction die, there could be less yield of food.

3. The Niwot Ridge litter and woody debris dataset was collected from the National Ecological Observatory Network, which collectively includes 81 aquatic and terrestrial sites across 20 ecoclimatic domains. 32 of these sites sample forest litter and woody debris, and we will focus on the Niwot Ridge long-term ecological research (LTER) station in Colorado. Why might we be interested in studying litter and woody debris that falls to the ground in forests? Feel free to do a brief internet search if you feel you need more background information.

Answer: We might be interested in studying litter and woody debris based on how it plays a very important role in carbon storage and nutrient cycling.

4. How is litter and woody debris sampled as part of the NEON network? Read the NEON_Litterfall_UserGuide.pdf document to learn more. List three pieces of salient information about the sampling methods here:

Answer: 1. Trap placement is both targeted and randomized depending on vegetation cover and height. 2. 1-30 m² nested subplots utilized in 400 or 1600 m² plots. 3. All data taken at sites where there is evidence of woody vegetation being at least 2m tall.

Obtain basic summaries of your data (Neonics)

5. What are the dimensions of the dataset?

```
dim(Neonics)
```

[1] 4623 30

#Gave me the number of rows in columns listed in the console pane (4623 rows, 30 columns).

6. Using the summary function on the "Effect" column, determine the most common effects that are studied. Why might these effects specifically be of interest? [Tip: The sort() command is useful for listing the values in order of magnitude...]

summary(Neonics\$Effect)

##	Accumulation	Avoidance	Behavior	Biochemistry
##	12	102	360	11
##	Cell(s)	Development	Enzyme(s)	Feeding behavior
##	9	136	62	255
##	Genetics	Growth	Histology	Hormone(s)
##	82	38	5	1
##	Immunological	Intoxication	Morphology	Mortality
##	16	12	22	1493
##	Physiology	Population	Reproduction	
##	7	1803	197	

```
Common_Effects <- sort(summary(Neonics$Effect))
Common_Effects #Used sorted data to pick two most common effects
```

##	Hormone(s)	Histology	Physiology	Cell(s)
##	1	5	7	9
##	Biochemistry	Accumulation	Intoxication	Immunological
##	11	12	12	16
##	Morphology	Growth	Enzyme(s)	Genetics
##	22	38	62	82
##	Avoidance	Development	Reproduction	Feeding behavior
##	102	136	197	255
##	Behavior	Mortality	Population	
##	360	1493	1803	

Answer: Population and mortality may be of the highest interest based on how they represent the efficacy of the insecticides being used, as lower population and higher mortality represent an effective insecticide.

7. Using the summary function, determine the six most commonly studied species in the dataset (common name). What do these species have in common, and why might they be of interest over other insects? Feel free to do a brief internet search for more information if needed. [TIP: Explore the help on the summary() function, in particular the maxsum argument...]

summary (Neonics \$Species.Common.Name) #Gets me a list of all species freq.

##	Honey Bee	Parasitic Wasp
##	667	285
##	Buff Tailed Bumblebee	Carniolan Honey Bee
##	183	152
##	Bumble Bee	Italian Honeybee
##	140	113
##	Japanese Beetle	Asian Lady Beetle
##	94	76
##	Euonymus Scale	Wireworm
##	75	69
##	European Dark Bee	Minute Pirate Bug
##	66	62
##	Asian Citrus Psyllid	Parastic Wasp
##	60	58
##	Colorado Potato Beetle	Parasitoid Wasp
##	57	51
##	Erythrina Gall Wasp	Beetle Order
##	49	47
##	Snout Beetle Family, Weevil	Sevenspotted Lady Beetle
##	47	46
##	True Bug Order	Buff-tailed Bumblebee
##	45	39
##	Aphid Family	Cabbage Looper
##	38	38
##	Sweetpotato Whitefly	Braconid Wasp
##	37	33
##	Cotton Aphid	Predatory Mite
##	33	33
##	Ladybird Beetle Family	Parasitoid
##	adybiid beetie ramily	30
##		
##	Scarab Beetle	Spring Tiphia 29
##	Thrip Order	Ground Beetle Family
##	29	27
##	Rove Beetle Family	Tobacco Aphid
##	27	27
##	Chalcid Wasp	Convergent Lady Beetle
##	25	25
##	Stingless Bee	Spider/Mite Class
##	25	24
##	Tobacco Flea Beetle	Citrus Leafminer
##	24	23

##	Ladybird Beetle	Mason Bee
##	23	22
##	Mosquito	Argentine Ant
##	22	21
##	Beetle	Flatheaded Appletree Borer
##	21	20
##	Horned Oak Gall Wasp	Leaf Beetle Family
##	20	20
##	Potato Leafhopper	Tooth-necked Fungus Beetle
##	20 Codling Moth	20
##	Codling Moth 19	Black-spotted Lady Beetle 18
##	Calico Scale	Fairyfly Parasitoid
##	18	18
##	Lady Beetle	Minute Parasitic Wasps
##	18	18
##	Mirid Bug	Mulberry Pyralid
##	18	18
##	Silkworm	Vedalia Beetle
##	18	18
##	Araneoid Spider Order	Bee Order
##	17	17
##	Egg Parasitoid 17	Insect Class 17
##	Moth And Butterfly Order	Oystershell Scale Parasitoid
##	17	17
	Hemlock Woolly Adelgid Lady Beetle	Hemlock Wooly Adelgid
##	16	16
##	Mite	Onion Thrip
##	16	16
##	Western Flower Thrips	Corn Earworm
##	15	14
##	Green Peach Aphid	House Fly
##	14 Ox Beetle	14 Red Scale Parasite
##	OX Beetle	ned Scale Falasite
##	Spined Soldier Bug	Armoured Scale Family
##	14	13
##	Diamondback Moth	Eulophid Wasp
##	13	13
##	Monarch Butterfly	Predatory Bug
##	13	13
##	Yellow Fever Mosquito	Braconid Parasitoid
##	13	12
##	Common Thrip 12	Eastern Subterranean Termite 12
##	Jassid	Mite Order
##	12	12
##	Pea Aphid	Pond Wolf Spider
##	12	12
##	Spotless Ladybird Beetle	Glasshouse Potato Wasp
##	11	10
##	Lacewing	Southern House Mosquito
##	10	10

```
##
                  (Other)
                                        Honey Bee
                                                          Parasitic Wasp
##
                     3083
                                              667
                                                                      285
## Buff Tailed Bumblebee
                             Carniolan Honey Bee
                                                              Bumble Bee
##
                                                                      140
##
        Italian Honeybee
##
```

Answer: The six most commonly studied species are Honey Bee, Parasitic Wasp, Buff Tailed Bumblebee, Carniolan Honey Bee, Bumble Bee, and Italian Honeybee. These species all appear to bee in the bee/wasp family (striped with stingers). They of much importance due to how vital they are in ecosystem services such as pollination and even making honey for three of them (we do not want them to die).

8. Concentrations are always a numeric value. What is the class of Conc.1..Author. column in the dataset, and why is it not numeric? [Tip: Viewing the dataframe may be helpful...]

```
Conc_Class <- class(Neonics$Conc.1..Author.) #Reads off the class
Conc_Class
```

```
## [1] "factor"
```

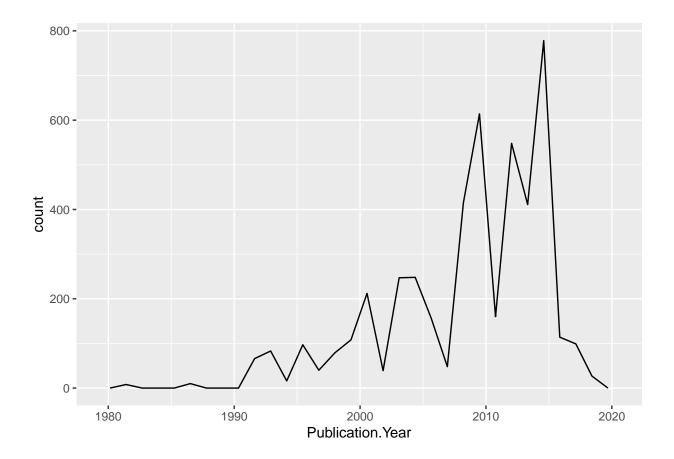
Answer: The class of "Conc.1..Author." is Factor. It likely is not numeric because of How it is a measurement (or categorical value) and not meant to be used in operations.

Explore your data graphically (Neonics)

9. Using geom_freqpoly, generate a plot of the number of studies conducted by publication year.

```
help("geom_freqpoly")
ggplot(Neonics, aes(x = Publication.Year)) +
  geom_freqpoly() #Made graph based on dataset, set one column as x variable
```

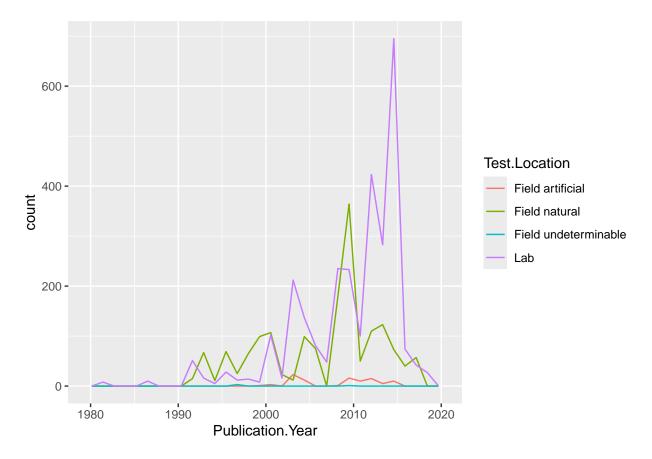
'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.



10. Reproduce the same graph but now add a color aesthetic so that different Test.Location are displayed as different colors.

```
ggplot(Neonics, aes(x = Publication.Year, color = Test.Location)) +#Added color
geom_freqpoly()
```

'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.



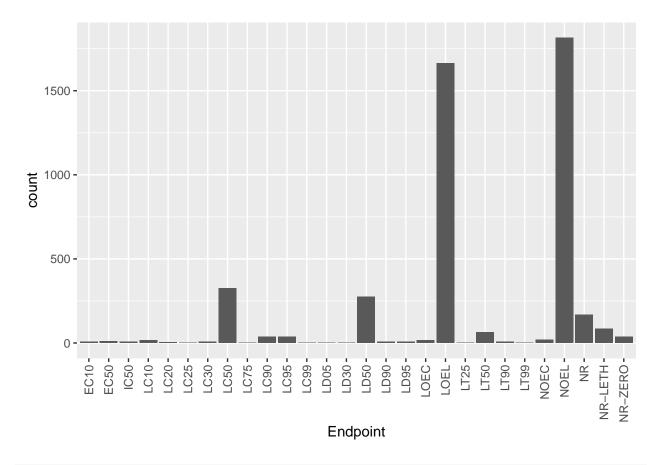
Interpret this graph. What are the most common test locations, and do they differ over time?

Answer: Lab and Field natural appear to be the most common test locations. Field natural is the most common in most of the 1990's and a few years around 2010, while Lab is the most common for most of the rest of the years.

11. Create a bar graph of Endpoint counts. What are the two most common end points, and how are they defined? Consult the ECOTOX_CodeAppendix for more information.

[TIP: Add theme(axis.text.x = element_text(angle = 90, vjust = 0.5, hjust=1)) to the end of your plot command to rotate and align the X-axis labels...]

```
help("geom_bar")
ggplot(Neonics, aes(x = Endpoint)) +
  geom_bar() +
  theme(axis.text.x = element_text(angle = 90, vjust = 0.5, hjust=1))
```



#Defined x as the endpoint column, used the provided code to rotate labels

Answer: NOEL and LOEL are the most common endpoints. LOEL is defined as "Lowest-observable-effect-level", which refers to a low dose producing significant effects. NOEL is defined as "No-observable-effect-level", which refers to the highest dose producing no significant effects.

Explore your data (Litter)

12. Determine the class of collectDate. Is it a date? If not, change to a date and confirm the new class of the variable. Using the unique function, determine which dates litter was sampled in August 2018.

```
class(Litter$collectDate)

## [1] "factor"

Litter$collectDate <- as_date(Litter$collectDate)
class(Litter$collectDate) #Changed the class to date and found two unique values

## [1] "Date"

unique(Litter$collectDate)

## [1] "2018-08-02" "2018-08-30"</pre>
```

13. Using the unique function, determine how many different plots were sampled at Niwot Ridge. How is the information obtained from unique different from that obtained from summary?

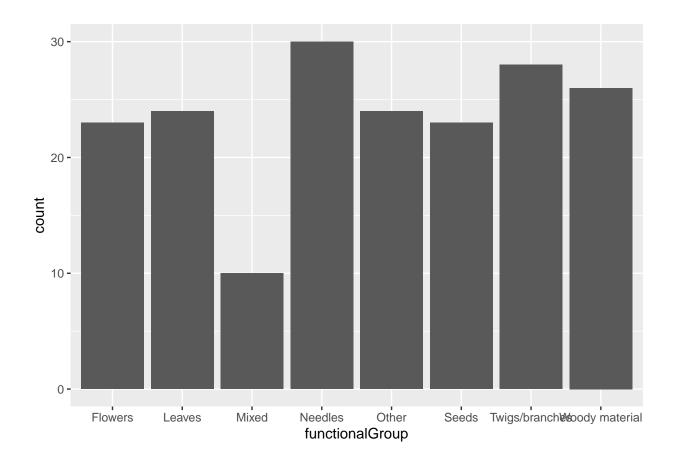
```
unique(Litter$plotID)
   [1] NIWO 061 NIWO 064 NIWO 067 NIWO 040 NIWO 041 NIWO 063 NIWO 047 NIWO 051
   [9] NIWO_058 NIWO_046 NIWO_062 NIWO_057
## 12 Levels: NIWO_040 NIWO_041 NIWO_046 NIWO_047 NIWO_051 NIWO_057 ... NIWO_067
length(unique(Litter$plotID)) #Counted number of unique values using length()
## [1] 12
summary(Litter$plotID)
## NIWO_040 NIWO_041 NIWO_046 NIWO_047 NIWO_051 NIWO_057 NIWO_058 NIWO_061
                  19
                                                                16
         20
                           18
## NIWO_062 NIWO_063 NIWO_064 NIWO_067
         14
                  14
                           16
length(summary(Litter$plotID)) #Counted number of unique values using length()
```

Answer: 12 different plots were sampled at Niwot Ridge. The information obtained from Unique includes the same factors as Summary, but Summary also includes the frequency of each factor in the dataset.

[1] 12

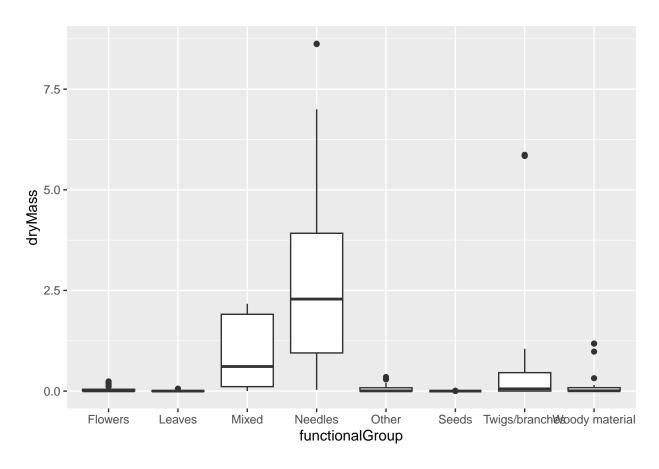
14. Create a bar graph of functional Group counts. This shows you what type of litter is collected at the Niwot Ridge sites. Notice that litter types are fairly equally distributed across the Niwot Ridge sites.

```
ggplot(Litter, aes(x = functionalGroup)) +
geom_bar() #Counts appear relatively even with exception to "Mixed".
```

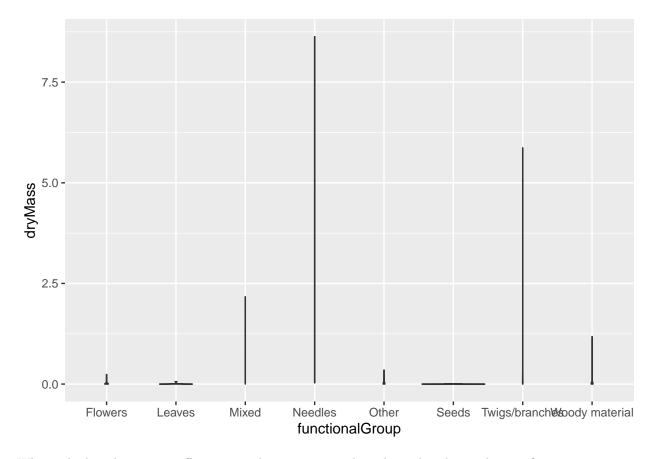


15. Using geom_boxplot and geom_violin, create a boxplot and a violin plot of dryMass by functional-Group.

```
ggplot(Litter, aes(x = functionalGroup, y = dryMass)) +
geom_boxplot()
```



```
ggplot(Litter, aes(x = functionalGroup, y = dryMass)) +
geom_violin() #Made both graphs and compared shapes
```



Why is the boxplot a more effective visualization option than the violin plot in this case?

Answer: The boxplot is a more effective visualization than the violin plot in this case because of no particular dryMass measurement having a high density/frequency.

What type(s) of litter tend to have the highest biomass at these sites?

Answer: It appears that needles tend to have the highest biomass, with mixed coming in second (based on averages).