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) #getting libraries
library(lubridate)
library(here)

Neonics <- read.csv ( #openning and naming files
file = here('./Data/Raw/ECOTOX_Neonicotinoids_Insects_raw.csv'),
stringsAsFactors = TRUE
)

Litter <- read.csv( #opening and naming</pre>
```

```
file= here('./Data/Raw/NEON_NIWO_Litter_massdata_2018-08_raw.csv'),
    stringsAsFactors = TRUE
)
```

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: They are an insecticide that binds to the nerves of insects, causing them of overstimulate and die. They are applied as a "drench" meaning 95% of what is applied ends up in the soil and eventually the water supply. It has been shown to have a negative effect on bees, who provide pollination and a vital part of plant life, as well as to vulnerable human population.

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: Litter and woody debris is a vital step in natural wildfire cycles. The monitoring and maintenance of woody debris can help us regulate controlled burns and also is important in understanding when natural burns are most needed and most productive to forest health.

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. Litter is collected from high and low traps
2.In this protocol, litter is defined as material that is dropped from the forest canopy and has a
butt end diameter <2cm and a length <50 cm

3. Locations of tower plots are selected randomly within the 90% flux footprint of the primary and secondary airsheds

Obtain basic summaries of your data (Neonics)

5. What are the dimensions of the dataset?

```
dim(Neonics) #4623 observations of 30 variables
```

```
## [1] 4623 30
```

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...]

```
Neonics_summary<- summary(Neonics$Effect)
Neonics_HightoLow <-sort(Neonics_summary, decreasing=TRUE)
print(Neonics_HightoLow) #fancy way</pre>
```

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

sort(summary(Neonics\$Effect), decreasing= TRUE) #less fancy way; is there a

##	Population	Mortality	Behavior	Feeding behavior
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#best practices for how to do this?

Answer: Most common effect is population, which is vital when understanding the systemic effects an insecticide could have on a species or collection of species. How much does it kill is an important question.

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...]

```
NeonicsSortedSpecies<- (sort(
    summary(Neonics$Species.Common.Name, maxsum=7) #maxsum=7 instead of 6
    , decreasing= TRUE #because otherwise "other" as the most common crowds out
    ) #number 6
)
print (NeonicsSortedSpecies)</pre>
```

(Other) Honey Bee Parasitic Wasp

```
## 3083 667 285

## Buff Tailed Bumblebee Carniolan Honey Bee Bumble Bee

## 183 152 140

## Italian Honeybee

## 113
```

Answer: Honey Bee, Parasitic Wasp, Buff Tailed Bumblebee, Carniolan Honey Bee, Bumble Bee, and Italian Honeybee are most commmon species because they are all pollinators. If these are killed by the insectcide it will negatively effect the whole ecoystem, meaning they are vital to study

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...]

```
class(Neonics$Conc.1..Author.)
```

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

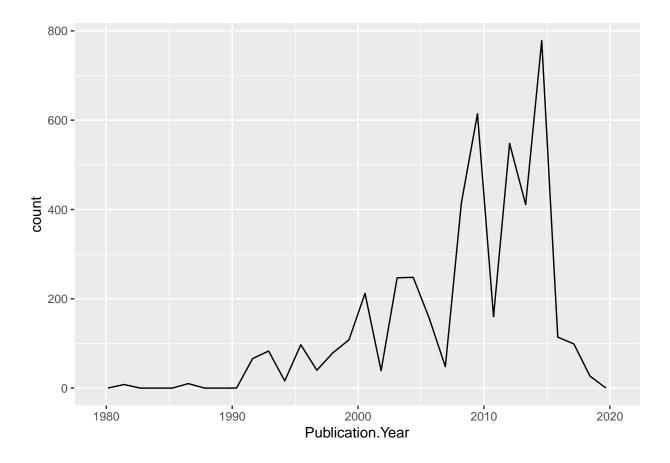
Answer: It is a factor. It is not numeric because mix of quantitive and qualitivate

Explore your data graphically (Neonics)

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

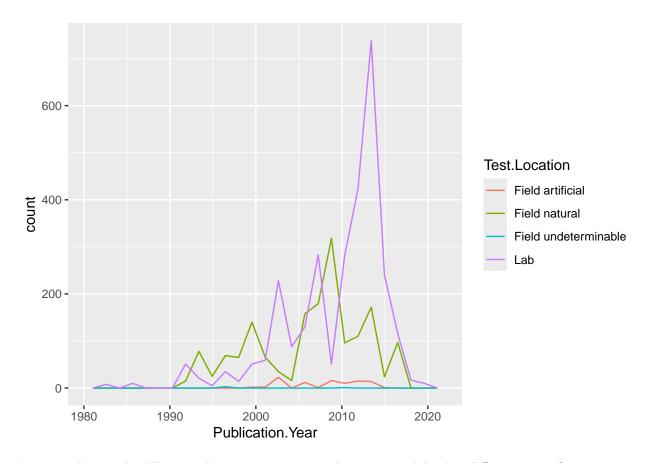
```
Neonics_Frequency<-ggplot(Neonics) +
  geom_freqpoly (aes(x=Publication.Year)
)
print(Neonics_Frequency)</pre>
```

'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.

```
Neonics_Frequency<- ggplot(Neonics) +
  geom_freqpoly (aes(x=Publication.Year, color= Test.Location), bins=25
)
print(Neonics_Frequency)</pre>
```



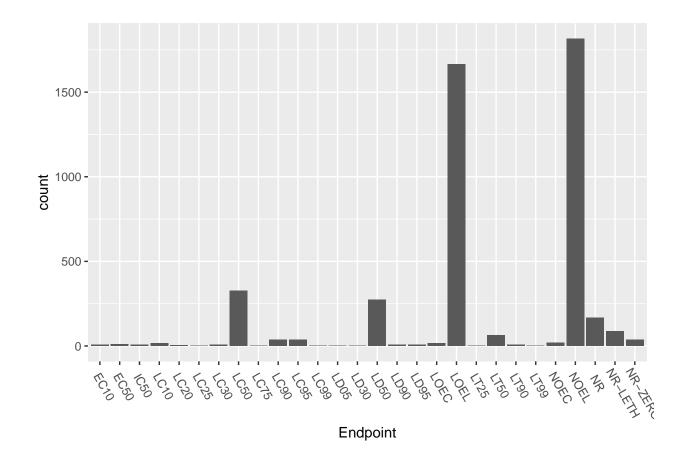
Interpret this graph. What are the most common test locations, and do they differ over time?

Answer: The most common test location recently is Lab, however there are peroids (\sim 1992-2002, 2008-2010) were "field natural" was the most common

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...]

```
Neonics_Bar <- ggplot(Neonics, aes(x=Endpoint)) +
   geom_bar() +
   theme(axis.text.x = element_text(angle = -60, vjust = .5, hjust=.2))
print(Neonics_Bar)</pre>
```



Answer:The two most common endpoints are NOEL, defined as No-observable-effect-level: highest dose (concentration) producing effects not significantly different from responses of controls according to author's reported statistical test (NOEAL/NOEC) and LOEL, defined as Lowest-observable-effect-level: lowest dose (concentration) producing effects that were significantly different (as reported by authors) from responses of controls (LOEAL/LOEC).

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.

```
#Litter$collectDate<-as.Date(Litter$collectDate, format="%Y-%m-%d") #base R OR
library(lubridate)

Litter$collectDate<- ymd(Litter$collectDate)
unique(Litter$collectDate) #only two dates, 2018-08-02 and 2018-08-30</pre>
```

[1] "2018-08-02" "2018-08-30"

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
```

summary(Litter\$plotID)

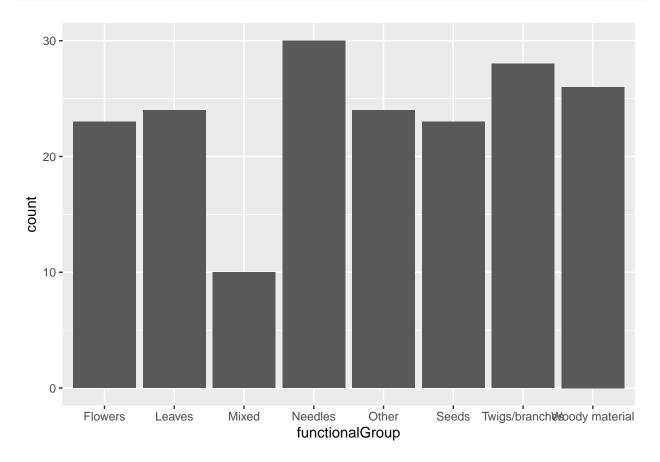
```
## NIWO_040 NIWO_041 NIWO_046 NIWO_047 NIWO_051 NIWO_057 NIWO_058 NIWO_061 ## 20 19 18 15 14 8 16 17 ## NIWO_062 NIWO_063 NIWO_064 NIWO_067 ## 14 14 16 17
```

Answer: With unique() it will tell me the number of unique classes that I ask for, but summary of that class will list each one out (if there's not too many I can count) but also it will list the number of observations within each class

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.

```
Litter_functionalGroup <- ggplot(
  data=Litter, aes(x=functionalGroup)) +
  geom_bar()

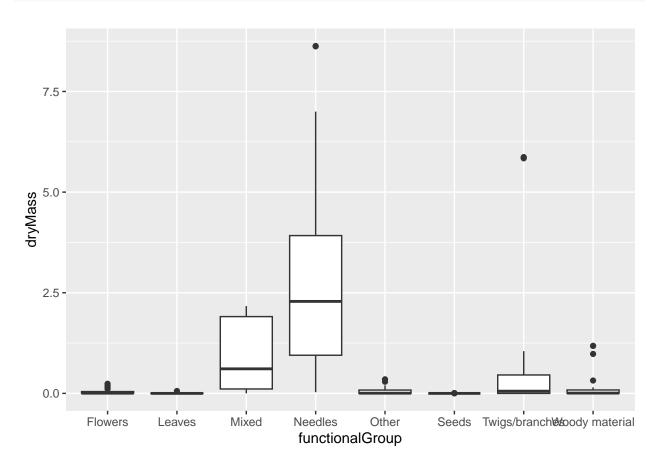
print(Litter_functionalGroup)</pre>
```



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

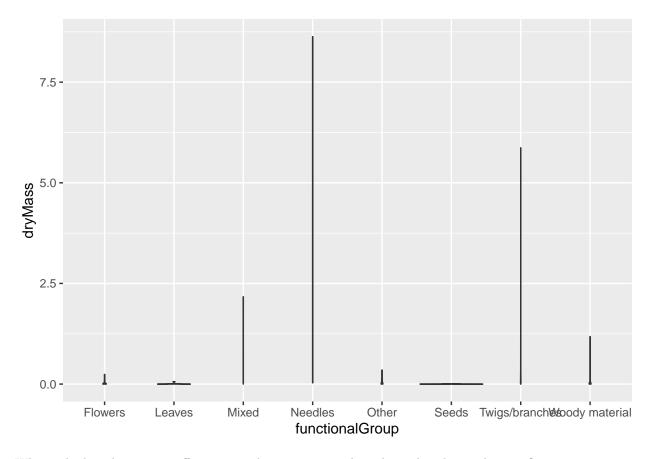
```
Litter_dryMassBox <- ggplot(
  data=Litter, aes(x=functionalGroup, y= dryMass)) +
  geom_boxplot()

print(Litter_dryMassBox)</pre>
```



```
Litter_dryMassFiddle <- ggplot(
  data=Litter, aes(x=functionalGroup, y= dryMass)) +
  geom_violin()

print(Litter_dryMassFiddle)</pre>
```



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

Answer: The box plot is more effective because it more effectively conveys the range, in this case using y axis to try to convey the size of each range instead of the x. The violin plot uses the x axis instead of to convey the size of each individual part of the range, and in this case the granularity doesn't create a ton of clarity.

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

Answer: Needles by a long shot.