Assignment 3: Data Exploration

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OVERVIEW

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

Directions

- 1. Change "Student Name, Section #" on line 3 (above) with your name and section number.
- 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., "FirstLast_A03_DataExploration.Rmd") prior to submission.

The completed exercise is due on <1/31/22>.

Set up your R session

1. Check your working directory, load necessary packages (tidyverse), 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 add the stringsAsFactors = TRUE parameter to the function when reading in the CSV files.

```
library(tidyverse)
library(ggplot2)
library(dplyr)
library(readr)
getwd()
```

[1] "/Users/rorymccollum/Desktop/Rdata/Environmental_Data_Analytics_2022"

```
Neonics <- read.csv('/Users/rorymccollum/Desktop/Rdata/Environmental_Data_Analytics_2022/Data/Raw/ECOTO_
View(Neonics)
```

Litter <- read.csv('/Users/rorymccollum/Desktop/Rdata/Environmental_Data_Analytics_2022/Data/Raw/NEON_N View(Litter)

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 ecotoxicologoy of neonicotinoids on insects? Feel free to do a brief internet search if you feel you need more background information.

Answer: It's interesting to note what effects are seen in insects because some effects might also occur in humans.

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: Forest litter and woody debris are important factors in 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: Litter and woody debris are sampled in tower plots. There is different litter trap placement (randomized or targeted) depending on vegetation Plots must be specific distances from man made structures * Streams must not intersect plots.

Obtain basic summaries of your data (Neonics)

5. What are the dimensions of the dataset?

dim(Neonics)

[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?

summary(Neonics\$Effect)

##	Accumulation	Avoidance	Behavior	Biochemistry
##	12	102	360	11
##	Cell(s)	Development	<pre>Enzyme(s)</pre>	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	

Answer: It looks like population and mortality are the two most studied effects. Population might indicate that there is a population level effect, which is good to know that it's widespread. Mortality is good to know because death is a pretty strong effect.

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.

summary(Neonics\$Species.Common.Name)

Parasitic Wasp	Honey Bee	##
285	667	##
Carniolan Honey Bee	Buff Tailed Bumblebee	##
152	183	##
Italian Honeybee	Bumble Bee	##
113	140	##
Asian Lady Beetle	Japanese 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 Erythrina Gall Wasp	51 Beetle Order
##	Erythrina Gair Wasp 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
##	30	30
## ##	Scarab Beetle 29	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 22	Argentine Ant 21
##	Beetle	Flatheaded Appletree Borer
##	21	20
##	Horned Oak Gall Wasp	Leaf Beetle Family
##	20	20
##	Potato Leafhopper	Tooth-necked Fungus Beetle
##	20	20
##	Codling Moth	Black-spotted Lady Beetle
##	19	18
##	Calico Scale	Fairyfly Parasitoid
##	18	18
##	Lady Beetle	Minute Parasitic Wasps
## ##	18 Mirid Bug	18 Mulberry Pyralid
##	mirid bug 18	Mulberry Pyralid 18
##	Silkworm	Vedalia Beetle
	SIIIWOIII	, caarra 200010

##	18	18
##	Araneoid Spider Order	Bee Order
##	17	17
##	Egg Parasitoid	Insect Class
##	17	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	14
##	Ox Beetle	Red Scale Parasite
##	14	14
##	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	Eastern Subterranean Termite
##	12	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
##	Two Spotted Lady Beetle	Ant Family
##	10	9
##	Apple Maggot	(Other)
##	9	670

Answer: The top six species are all bees (except for the one wasp). These are likely of particular interest as important pollinators.

Honey	Bee	P	arasitic Wasp		
		667			285
	Buff	Tailed Bumblebee		Carniolan Honey	Bee
		183			152
		Bumble Bee		Italian Hone	ybee
		140			113

8. Concentrations are always a numeric value. What is the class of Conc.1..Author. in the dataset, and why is it not numeric?

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

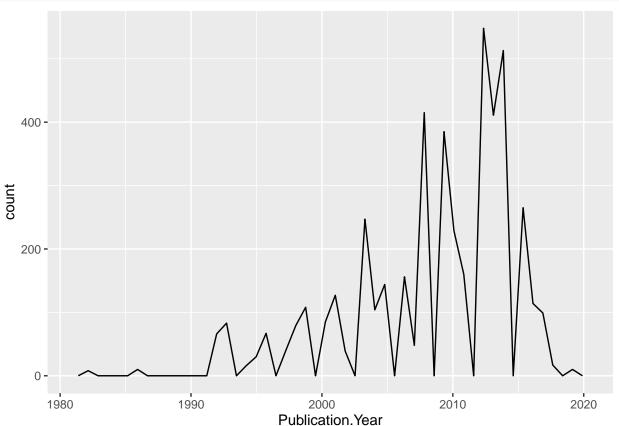
[1] "factor"

Answer: Conc.1..Author is a factor because it uses only a certain set of values.

Explore your data graphically (Neonics)

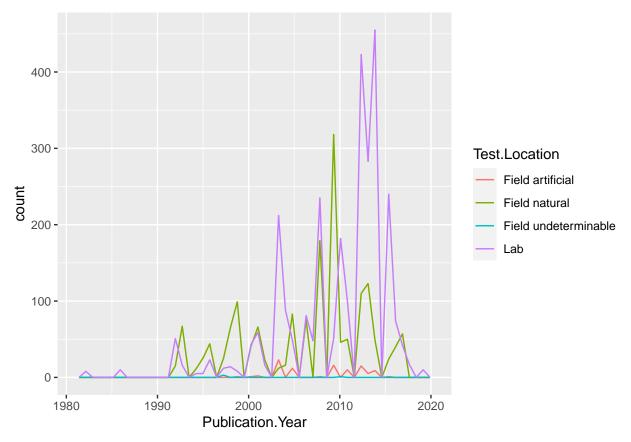
9. Using geom_freqpoly, generate a plot of the number of studies conducted by publication year.

```
ggplot(Neonics)+
  geom_freqpoly(aes(x=Publication.Year), bins=50)
```



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

```
ggplot(Neonics)+
geom_freqpoly(aes(x=Publication.Year, color=Test.Location), bins=50)
```

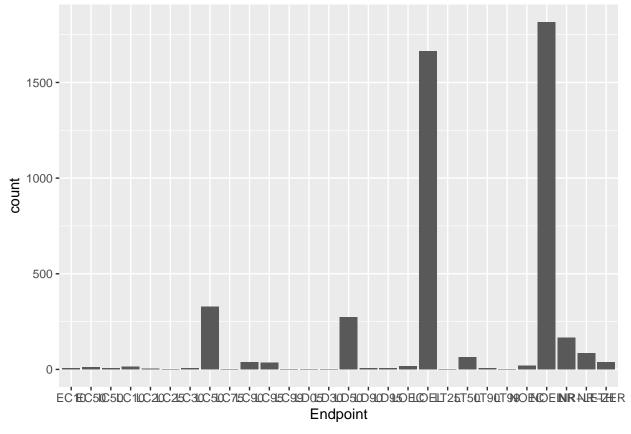


Interpret this graph. What are the most common test locations, and do they differ over time?

Answer: The most common test location is the lab, and it has slowly been gaining traction over time. The second most common test location is "field natural" and it varied over time, the most popular time being just before 2010.

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.

```
ggplot(Neonics, aes(x=Endpoint)) +
  geom_bar()
```



Answer: The two most common endpoints and LOEL= Lowest observed effect level and NOEL= no observed effect level. These describe the lowest dose of a toxin where you see a statistically significant effect and the highest does of a toxin where you see no effect, respectively.

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)
unique(Litter$collectDate)</pre>
```

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

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

unique(Litter\$namedLocation)

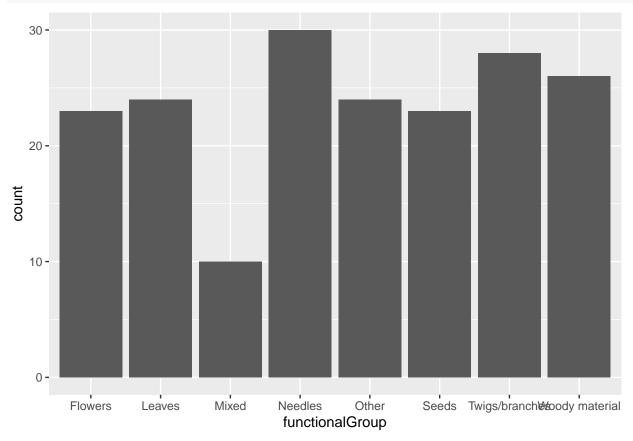
```
## [1] NIWO_061.basePlot.ltr NIWO_064.basePlot.ltr NIWO_067.basePlot.ltr
## [4] NIWO_040.basePlot.ltr NIWO_041.basePlot.ltr NIWO_063.basePlot.ltr
## [7] NIWO_047.basePlot.ltr NIWO_051.basePlot.ltr NIWO_058.basePlot.ltr
## [10] NIWO_046.basePlot.ltr NIWO_062.basePlot.ltr NIWO_057.basePlot.ltr
## 12 Levels: NIWO_040.basePlot.ltr ... NIWO_067.basePlot.ltr
```

summary(Litter\$namedLocation)

Answer: There were 12 different plot locations. Summary shows the name of the plot with the nuber of times it was sampled underneath while unique shows the names of the plots in rows and at the end tells you the number of levels.

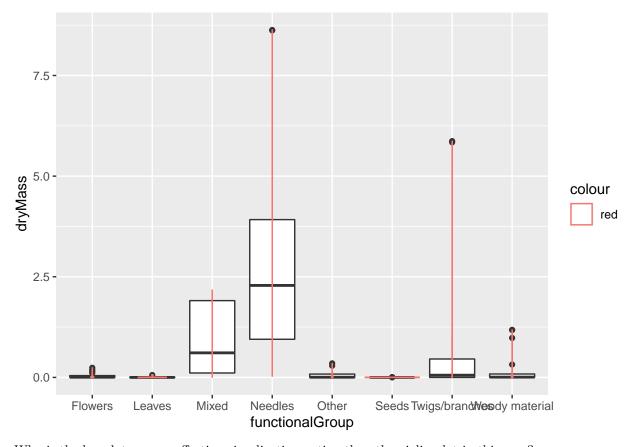
14. Create a bar graph of functionalGroup counts. This shows you what type of litter is collected at the Niwot Ridge sites. Notice that litter types are fairly equally distributed across the Niwot Ridge sites.

```
ggplot(Litter, aes(x=functionalGroup)) +
  geom_bar()
```



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

```
ggplot(Litter) +
  geom_boxplot(aes(x=functionalGroup, y=dryMass)) +
  geom_violin(aes(x=functionalGroup, y=dryMass, color="red"))
```



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

Answer: A boxplot tells you more information than the violin plot. Vioin only shows a line up to the uppermost bound for each group, while boxplots show you where the data is concentrated.

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

Answer: Needles tend to have the highest biomass, followed by mixed.