Assignment 3: Data Exploration

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Fall 2023

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 Sakai.

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. Check your working directory, load necessary packages (tidyverse, lubridate), 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.

```
getwd() #check wd
```

[1] "/home/guest/module1/EDE_Fall2023"

```
#to install necessary packages if not already installed
#install.packages("tidyverse")
#install.packages("lubridate")

#load libraries
library("tidyverse")
library("lubridate")

#load data
Neonics <- read.csv("./Data/Raw/ECOTOX_Neonicotinoids_Insects_raw.csv",stringsAsFactors = TRUE)
Litter <- read.csv("./Data/Raw/NEON_NIWO_Litter_massdata_2018-08_raw.csv",stringsAsFactors = TRUE)</pre>
```

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: This information could be used to assess the chemical's effectiveness as an insecticide. It could also be used to see what species of insect it may harm when released into the environment.
- 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:The litter that falls to the ground could be very improtant to measure for environmental modeling. Specifically litter plays a key role in nutrient cycling and carbon sequestration because it allows nutrients and carbon to be transferred from plant life to the soil.
- 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. ground traps are sampled once per year 2. litter is collected in elevated 0.5m2 traps and ground traps of 3m by 0.5m 3. one ground trap and one elevated trap was deployed for every 400m2 area

Obtain basic summaries of your data (Neonics)

5. What are the dimensions of the dataset?

```
dim(Neonics) #get dimensions of Neonics
```

[1] 4623 30

Answer: The neonics dataset has 30 columns and 4623 rows

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) # sumarize 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: The most common effects that are studied are population and mortality. This might be specifically of interest because it is so relevant to environmental impact because if an insecticide causes changes in population it can cause a species to be threatened.

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: The sort() command can sort the output of the summary command...]

sort(summary(Neonics\$Species.Common.Name)) #sumarize the species data and sort it by number of apperance

##	Ant Family	Apple Maggot
##	9	9
##	Glasshouse Potato Wasp	Lacewing
##	10	10
##	Southern House Mosquito	Two Spotted Lady Beetle
##	10	10
##	Spotless Ladybird Beetle	Braconid Parasitoid
##	11	12
##	Common Thrip	Eastern Subterranean Termite
##	12	12
##	Jassid	Mite Order
##	12	12
##	Pea Aphid	Pond Wolf Spider
##	12	12
##	Armoured Scale Family	Diamondback Moth
##	13	13
##	Eulophid Wasp	Monarch Butterfly
##	13	13
##	Predatory Bug	Yellow Fever Mosquito
##	13	13
##	Corn Earworm	Green Peach Aphid
##	14	14
##	House Fly	Ox Beetle
##	14	14
##	Red Scale Parasite	Spined Soldier Bug
##	14	14
##	Western Flower Thrips	Hemlock Woolly Adelgid Lady Beetle
##	15	16
##	Hemlock Wooly Adelgid	Mite
##	16	16
##	Onion Thrip	Araneoid Spider Order
##	16	17
##	Bee Order	Egg Parasitoid
##	17	17
##	Insect Class	Moth And Butterfly Order
##	17	17
##	Oystershell Scale Parasitoid	Black-spotted Lady Beetle
##	17	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
##	Codling Moth	Flatheaded Appletree Borer
##	19	20
11 1 1	13	20

##	Horned Oak Gall Wasp	Leaf Beetle Family
##	20	20
##	Potato Leafhopper	Tooth-necked Fungus Beetle
##	20	20
##	Argentine Ant	Beetle
##	21	21
##	Mason Bee	Mosquito
##	22	22
## ##	Citrus Leafminer 23	Ladybird Beetle 23
##	Spider/Mite Class	Tobacco Flea Beetle
##	Spidel/Mite Class 24	24
##	Chalcid Wasp	Convergent Lady Beetle
##	25	25
##	Stingless Bee	Ground Beetle Family
##	25	27
##	Rove Beetle Family	Tobacco Aphid
##	27	27
##	Scarab Beetle	Spring Tiphia
##	29	29
##	Thrip Order	Ladybird Beetle Family
##	29	30
##	Parasitoid	Braconid Wasp
##	30	33
##	Cotton Aphid	Predatory Mite
##	33	33
## ##	Sweetpotato Whitefly 37	Aphid Family 38
##	Cabbage Looper	Buff-tailed Bumblebee
##	38	39
##	True Bug Order	Sevenspotted Lady Beetle
##	45	46
##	Beetle Order	Snout Beetle Family, Weevil
##	47	47
##	Erythrina Gall Wasp	Parasitoid Wasp
##	49	51
##	Colorado Potato Beetle	Parastic Wasp
##	57	58
##	Asian Citrus Psyllid	Minute Pirate Bug
##	60	62
##	European Dark Bee	Wireworm
##	66	69
##	Euonymus Scale 75	Asian Lady Beetle 76
## ##	Japanese Beetle	Italian Honeybee
##	94	113
##	Bumble Bee	Carniolan Honey Bee
##	140	152
##	Buff Tailed Bumblebee	Parasitic Wasp
##	183	285
##	Honey Bee	(Other)
##	667	670

Answer: the top 6 most commonly studdied species are the western honey bee (Apis mellifera),

the buff tailed bumblebee (Bombus terrestris), the parasitic wasp, the carniolan Honey Bee (Apis mellifera ssp. carnica), the common eastern bumble bee (Bombus impatiens), and the italian honey bee (Apis mellifera ssp. ligustica). All of these species are kinds of bees except for the parasitic wasp which is closely related. These species are likely most studied because bees are important polinators and bees have many populations that are under threat.

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?

class(Neonics\$Conc.1..Author.) #check class of `Conc.1..Author.`

[1] "factor"

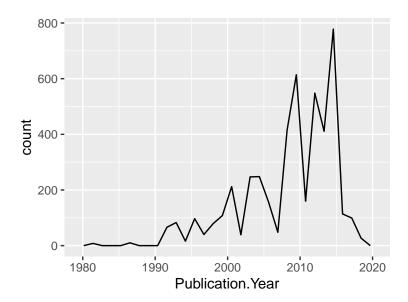
sort	(summary)	(Neonics\$C	Conc.1A	uthor.))	#take a	look at wl	nat option	ns there o	are for	the	input	of	this	fie
##	0.02/	0.025/	0.29	37.5/	4/	5	0.047/	0.25/						
##	11	11	11	11	11	11	12							
##	0.28/	1.28/	1.81/	112	150	2.5/	25							
##	12	1.207	1.017	12	12	12	12	•						
##	75/	0.17/	125	14	16	17	0.0001							
##	12	13	13	13	13	13	14							
##	0.084/	0.15	0.6	12.5/			40.0/							
##	14	14	14	14	144.07	14	14							
##	56	84/	0.053	0.24	0.28	125/		0.00355/						
##	14	14	15	15	15	15	15	•						
##	0.1	0.4	150/	300	80/	0.18/	0.3/							
##	16	16	16	16	16	17	17	17						
##	40	0.005	0.4/	0.05	1.5	2.60/	20.0/	6						
##	17	18	18	20	20	2.007	20.07							
##	6.80/	62.5/	0.336	1.5/	0.01/		3/							
##	20	20	21	21	22	22	22							
##	3	0.56/	0.2/	0.027	2.4	12/	25.0/							
##	23	24	25	26	26	27	28							
##	0.15/	1/	48	0.5	0.125	500/	50							
##	30	30	30	32	33	33	36	-						
##	1.0/	2.27/	0.1/	0.05/	0.45	0.03	0.5/							
##	40	40	42	43	43	44	45	51						
##	100	0.053/	10	2/	0.40/	1023	1	NR						
##	56	59	62	63	69	80	82							
##	NR/	10/	0.37/	(Other)	30	30	02	01						
##	108	127	208	1817										
ππ	100	121	200	1011										

Answer: The class of Conc.1..Author. is factor. This is due to there being a few entries that are not numeric including the NR and also the / that is at the end of the numbered entries that prevents it from being read as a number. It is instead read as a string which is taken in as a factor.

Explore your data graphically (Neonics)

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

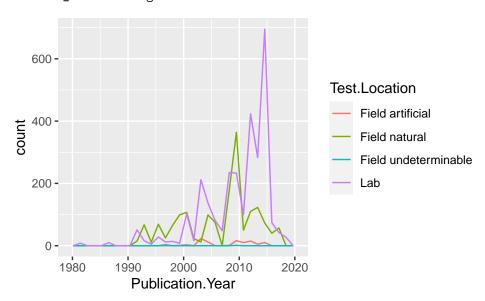
```
ggplot(Neonics, aes(x = Publication.Year)) + geom_freqpoly() #plot time series of number of publication
## `stat bin()` using `bins = 30`. Pick better value with `binwidth`.
```



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

```
ggplot(Neonics)+
  geom_freqpoly(aes(x = Publication.Year, color = Test.Location)) #plot time series of number of public
```

`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: The most common test locations are natural fields and labs. Field tests reached their peak popularity shortly before 2010. Lab tests Reached their peak popularity in 2015.

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...]

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

1500-
500-
```

Endpoint

Answer: the two most common endpoints are LOEC (Lowest observable effect concentration) and NOCE (No observable effect concentration)

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) #check class of collect date

## [1] "factor"

Litter$collectDate <- as.Date(Litter$collectDate,format = "%Y-%m-%d") #update class to be date instead
Litter$collectDate # printo out new collect date data</pre>
```

```
[1] "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02"
##
##
     [6] "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02"
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    [11] "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02"
    [16] "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02"
##
    [21] "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02"
##
    [26] "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02"
##
##
    [31] "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02"
    [36] "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02"
##
    [41] "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02"
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##
    [46] "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02"
    [51] "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02"
    [56] "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02"
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    [61] "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02"
    [66] "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02"
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    [71] "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02"
##
##
    [76] "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02"
##
    [81] "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02"
```

[86] "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02" "2018-08-02" [91] "2018-08-02" "2018-08-30" "2018-08-30" "2018-08-30" "2018-08-30" "2018-08-30" "2018-08-30"

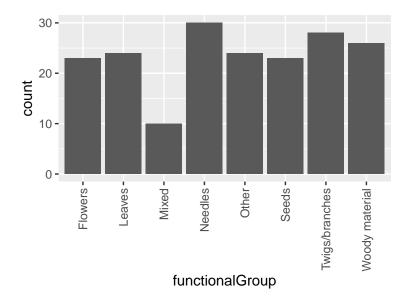
```
[96] "2018-08-30" "2018-08-30" "2018-08-30" "2018-08-30" "2018-08-30"
## [101] "2018-08-30" "2018-08-30" "2018-08-30" "2018-08-30" "2018-08-30"
     [106] "2018-08-30" "2018-08-30" "2018-08-30" "2018-08-30" "2018-08-30"
## [111] "2018-08-30" "2018-08-30" "2018-08-30" "2018-08-30" "2018-08-30"
## [116] "2018-08-30" "2018-08-30" "2018-08-30" "2018-08-30" "2018-08-30"
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## [151] "2018-08-30" "2018-08-30" "2018-08-30" "2018-08-30" "2018-08-30"
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## [186] "2018-08-30" "2018-08-30" "2018-08-30"
class(Litter$collectDate) # check that class has been updated
## [1] "Date"
unique(Litter$collectDate) # find unique collection dates
## [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?
summary(Litter$namedLocation) #summarize locations
## NIWO_040.basePlot.ltr NIWO_041.basePlot.ltr NIWO_046.basePlot.ltr
##
                                        20
                                                                                 19
## NIWO_047.basePlot.ltr NIWO_051.basePlot.ltr NIWO_057.basePlot.ltr
##
                                        15
## NIWO_058.basePlot.ltr NIWO_061.basePlot.ltr NIWO_062.basePlot.ltr
                                                                                 17
## NIWO_063.basePlot.ltr NIWO_064.basePlot.ltr NIWO_067.basePlot.ltr
                                        14
                                                                                 16
                                                                                                                         17
unique(Litter$namedLocation) # find unique locations
```

```
## [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
```

Answer:12 plots were sampled at Niwot Ridge. Summary gives both the locations and how many records are at each location. Unique just gives a list of the different unique locations and a count of how many there are.

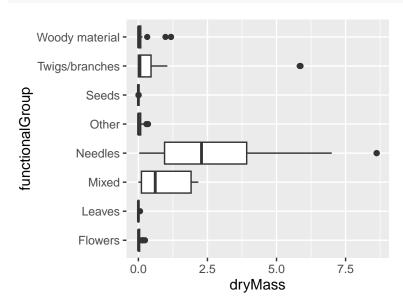
14. Create a bar graph of functionalGroup counts. This shows you what type of litter is collected at the Niwot Ridge sites. Notice that litter types are fairly equally distributed across the Niwot Ridge sites.

```
ggplot(Litter)+
  geom_bar(aes(x = functionalGroup)) + theme(axis.text.x = element_text(angle = 90, vjust = 0.5, hjust=
```

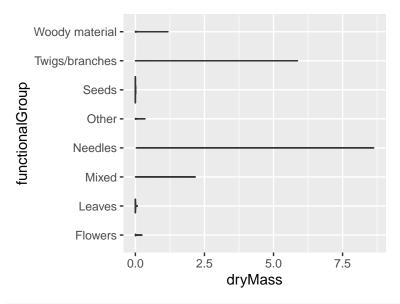


15. Using geom_boxplot and geom_violin, create a boxplot and a violin plot of dryMass by functional-Group.

ggplot(Litter) +
 geom_boxplot(aes(x = dryMass, y = functionalGroup)) # plot boxplot of dry mass by functional group



```
ggplot(Litter) +
  geom_violin(aes(x = dryMass, y = functionalGroup)) # plot boxplot of dry mass by functional group
```



I am not sure if it was supposed to have x and y axis but I think it looks best this way

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

Answer: the box plot is more effective because the violin plots don't show the difference between where the majority of the measurments are and the outliers. the boxplots make it clear where the outliers are

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

Answer: needles