# 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 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.

```
#Loading tidyverse and lubridate
library(tidyverse)
library(lubridate)

#Uploading ECOTOX neonicotinoids dataset and Niwot Ridge NEON dataset
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)
```

#### 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 neonicotinoids on insects because, similar to our concern about human health effects related to the use of insecticides for treating crops, insecticides have the potential to affect other insects aside from the target "pest(s)." For example, the use of these insecticides could have undesired negative impacts on local pollinators that are depended upon to aid in the reproduction of the crops being grown!

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 that falls to the ground in forests because this litter/debris buildup can become hazardous fuel in the event of a forest fire. Conversely, data on litter and woody debris can be used as indicators for aboveground net primary productivity and biomass levels for the forest ecosystem being studied.

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 using both elevated and ground traps and then categorized into one of eight predefined functional groups, where 1. Litter is defined as material that has fallen from the forest canopy and has a butt end diamater less than 2 cm and length less than 50 cm that has been collected in 0.5 mx 0.5 m elevated PVC traps. 2. Woody debris is defined as material that has fallen from the forest canopy and has a butt end diameter less than 2 cm and length greater than 50 cm that has been collected in ground traps. 3. Sampling occurs at terrestrial NEON sites that feature woody vegetation that is over 2 meters in height with trap pairs (one elevated, one ground) being placed either randomly or in targeted areas depending on vegetation coverage within 400 mx 400 m plot areas.

#### Obtain basic summaries of your data (Neonics)

5. What are the dimensions of the dataset?

#Having R return the number of rows and the number of columns for the Neonics dataset. #The dimensions of this dataset are 4,632 rows x 30 columns. 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?

# #Extracting data from the "Effect" column of the Neonics dataset 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	

Answer: The most common effects that are studied appear to include population, mortality, and feeding behavior, ranked by descending order of their number of observations. These effects might be of particular interest because impacts they are all interrelated: detrimental impacts from the use of neonicotinoids on crops that these insects pollinate/consume will have a ripple effect in killing off individuals which in turn affects the entire population and increases mortality rates and, over time, will eventually result in a change in feeding behavior (for example, a bee colony is forced to identify a new food source and relocate their hive away from the cropland).

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...]

##	(Other)	Honey Bee
##	670	667
##	Parasitic Wasp	Buff Tailed Bumblebee
##	285	183
##	Carniolan Honey Bee	Bumble Bee
##	152	140
##	Italian Honeybee	Japanese Beetle
##	113	94
##	Asian Lady Beetle	Euonymus Scale
##	76	75
##	Wireworm	European Dark Bee
##	69	66
##	Minute Pirate Bug	Asian Citrus Psyllid
##	62	60
##	Parastic Wasp	Colorado Potato Beetle
##	58	57
##	Parasitoid Wasp	Erythrina Gall Wasp
##	51	49
##	Beetle Order	Snout Beetle Family, Weevil
##	47	47
##	Sevenspotted Lady Beetle	True Bug Order
##	46	45

##	Buff-tailed Bumblebee	Aphid Family
##	39	38
## ##	Cabbage Looper 38	Sweetpotato Whitefly 37
##	Braconid Wasp	Cotton Aphid
##	33	33
##	Predatory Mite	Ladybird Beetle Family
##	33	30
##	Parasitoid	Scarab Beetle
##	30	29
##	Spring Tiphia	Thrip Order
##	29	29
##	Ground Beetle Family	Rove Beetle Family
## ##	Tobacca Aphid	Chalcid Hage
##	Tobacco Aphid 27	Chalcid Wasp 25
##	Convergent Lady Beetle	Stingless Bee
##	25	25
##	Spider/Mite Class	Tobacco Flea Beetle
##	24	24
##	Citrus Leafminer	Ladybird Beetle
##	23	23
##	Mason Bee	Mosquito
##	22	22
## ##	Argentine Ant 21	Beetle 21
## ##	Flatheaded Appletree Borer	Horned Oak Gall Wasp
##	20	20
##	Leaf Beetle Family	Potato Leafhopper
##	20	20
##	Tooth-necked Fungus Beetle	Codling Moth
##	20	19
##	Black-spotted Lady Beetle	Calico Scale
##	18	18
##	Fairyfly Parasitoid	Lady Beetle
## ##	18 Minute Parasitic Wasps	18 Mirid Bug
##	18	18
##	Mulberry Pyralid	Silkworm
##	18	18
##	Vedalia Beetle	Araneoid Spider Order
##	18	17
##	Bee Order	Egg Parasitoid
##	17	17
##	Insect Class	Moth And Butterfly Order
##	17	17
## ##	Uystershell Scale Parasitoid 17	Hemlock Woolly Adelgid Lady Beetle 16
## ##	Hemlock Wooly Adelgid	Mite
##	nemiock wooly adeigid	16
##	Onion Thrip	Western Flower Thrips
##	16	15
##	Corn Earworm	Green Peach Aphid
##	14	14

Ox Beetle	House Fly	##
14	14	##
Spined Soldier Bug	Red Scale Parasite	##
	14	
14	<del></del>	##
Diamondback Moth	Armoured Scale Family	##
13	13	##
Monarch Butterfly	Eulophid Wasp	##
13	13	##
Yellow Fever Mosquito	Predatory Bug	##
13	13	##
Common Thrip	Braconid Parasitoid	##
12	12	##
Jassid	Eastern Subterranean Termite	##
12	12	##
Pea Aphid	Mite Order	##
12	12	##
Spotless Ladybird Beetle	Pond Wolf Spider	##
11	12	##
Lacewing	Glasshouse Potato Wasp	##
-	-	
10	10	##
Two Spotted Lady Beetle	Southern House Mosquito	##
10	10	##
Apple Maggot	Ant Family	##
9	9	##

Answer: The six most commonly studied insect species from the Neonics dataset, based on their common names, include the honey bee, parasitic wasp, buff tailed bumblebee, Carniolan honey bee, bumble bee, Italian honeybee, and Japanese beetle. However, it is worth noting that the "Other" category had the greatest number of observations (670 whereas the honey bee had the second-highest at 667)! All of these species are pollinators - I learned that beetles are also considered pollinators, being called "mess and soil pollinators" by the Xerces Society for Invertebrate Conservation based on their messy eating habits - collecting and transferring pollen between the plants that they feed on. Thus, these insects, in comparison to others within the dataset, are more vulnerable to the application of insecticides on crops since they are directly reliant on them as a primary food source.

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?

```
#Determining the class of "Conc.1..Author" column in the Neonics dataset class(Neonics$Conc.1..Author.)
```

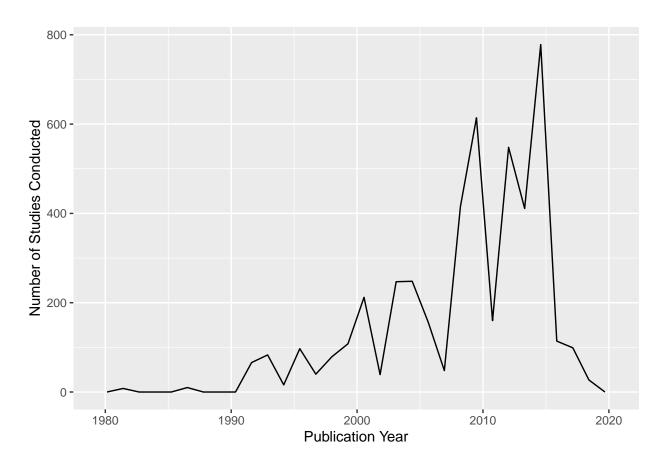
#### ## [1] "factor"

Answer: Despite concentrations always being documented as a numeric value, R reports the class type for "Conc.1..Author" as a "Factor" because we specified with the "stringsAsFactors = TRUE" command when loading the Neonics dataset that we wanted R to read-in factors, which it automatically assigns to any vector that does not feature integer values.

### Explore your data graphically (Neonics)

9. Using geom\_freqpoly, generate a plot of the number of studies conducted by publication year.

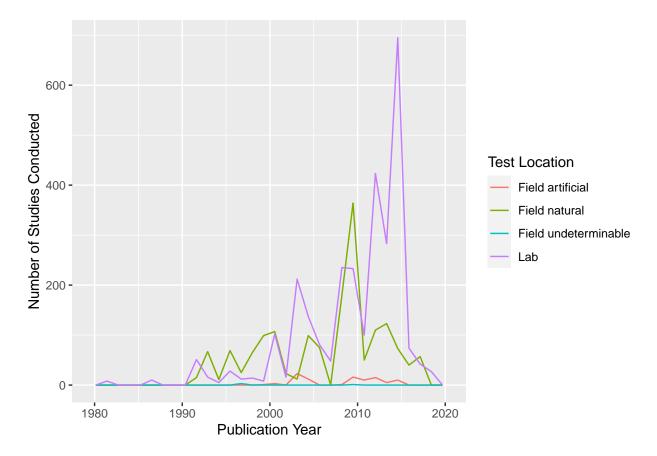
```
#Constructing a frequency line graph from Neonics to display studies conducted by publication year
ggplot(Neonics) +
  geom_freqpoly(aes(x = Publication.Year), bins = 30) +
  labs(x = "Publication Year", y = "Number of Studies Conducted")
```



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

```
#Reproducing the previous frequency line graph but adding display of test locations in different colors
ggplot(Neonics) +
  geom_freqpoly(aes(x = Publication.Year, color = Test.Location)) +
  labs(x = "Publication Year", y = "Number of Studies Conducted") +
  labs(color = "Test Location")
```

## 'stat\_bin()' using 'bins = 30'. Pick better value with 'binwidth'.



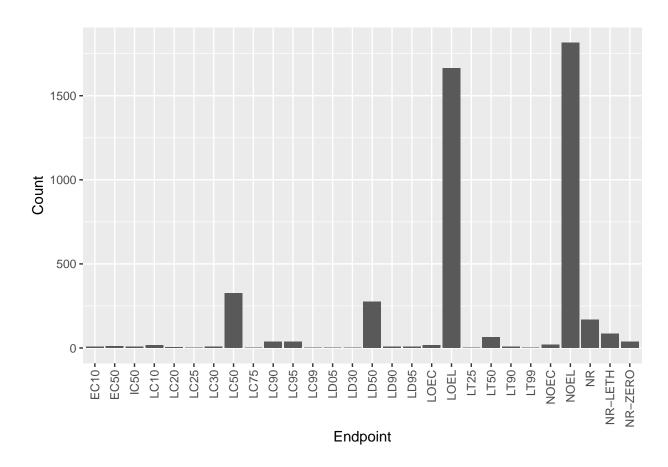
Interpret this graph. What are the most common test locations, and do they differ over time?

Answer: The most common test locations are "lab" and "field natural" and yes, they vary over time. "Lab" begins as the most common test location, then is overtaken by "field natural" between 1990 and 2000. After the early 2000s, "Lab" takes the lead again until 2010, when "field natural" reaches its peak number of studies (close to 400). However, after 2010, close to 2015, lab reaches its own peak, being used as the location setting for nearly 700 studies.

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...]

```
#Creation of a bar graph for Neonics Endpoint counts
ggplot(Neonics) +
  geom_bar(aes(x = Endpoint)) +
  theme(axis.text.x = element_text(angle = 90, vjust = 0.5, hjust=1)) +
  labs(x = "Endpoint", y = "Count")
```



#Checking findings from my bar graph summary(Neonics\$Endpoint)											
##	EC10	EC50	IC50	LC10	LC20	LC25	LC30	LC50	LC75	LC90	
##	6	11	6	15	5	1	6	327	1	37	
##	LC95	LC99	LD05	LD30	LD50	LD90	LD95	LOEC	LOEL	LT25	
##	36	2	1	1	274	6	7	17	1664	1	
##	LT50	LT90	LT99	NOEC	NOEL	NR N	IR-LETH	NR-ZERO			

Answer: NOEL and LOEL are the two most common endpoints from neonicotinoid toxicity testing. NOEL is defined as the no-observable-effect-level, or the highest concentration that produced effects that were not significantly different from those under the control responses, whereas LOEL is the lowest-observable-effect-level, or the lowest dose that produced significantly different effects from control responses.

#### 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.

```
#Determining the class of collectDate from the Litter dataset.
#It is not recognized by R as a date (it is a "factor")!
class(Litter$collectDate)
```

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

```
#Changing collectDate so that it can become a date class
collectDate <- ymd(Litter$collectDate)

#Confirming that collectDate is now recognized as a "date"!
class(collectDate)</pre>
```

## [1] "Date"

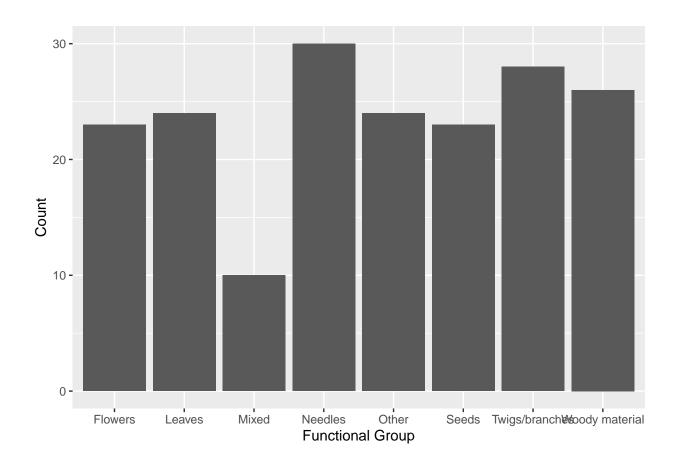
13. Using the unique function, determine how many plots were sampled at Niwot Ridge. How is the information obtained from unique different from that obtained from summary?

```
#Use of unique function to inspect the sample plots at Niwot Ridge
unique(Litter$plotID)
   [1] NIWO 061 NIWO 064 NIWO 067 NIWO 040 NIWO 041 NIWO 063 NIWO 047 NIWO 051
  [9] NIWO_058 NIWO_046 NIWO_062 NIWO_057
## 12 Levels: NIWO_040 NIWO_041 NIWO_046 NIWO_047 NIWO_051 NIWO_057 ... NIWO_067
#Comparing the use of summary to inspect plot IDs
summary(Litter$plotID)
## NIWO 040 NIWO 041 NIWO 046 NIWO 047 NIWO 051 NIWO 057 NIWO 058 NIWO 061
                                             14
                  19
                           18
                                    15
                                                        8
                                                                16
## NIWO_062 NIWO_063 NIWO_064 NIWO_067
##
         14
                  14
                           16
                                    17
```

Answer: Whereas "summary" returns the number of observations associated with each plot ID, "unique" simply extracts unique elements from the Litter dataframe, returning all unique elements/rows for plot IDs by their name.

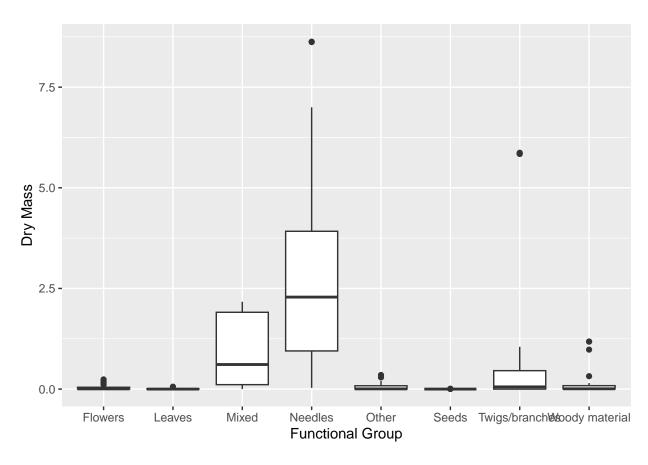
14. Create a bar graph of functional Group counts. This shows you what type of litter is collected at the Niwot Ridge sites. Notice that litter types are fairly equally distributed across the Niwot Ridge sites.

```
#Constructing a bar graph to visualize the count per functional group from Litter
ggplot(Litter) +
  geom_bar(aes(x = functionalGroup)) +
  labs(x = "Functional Group", y = "Count")
```

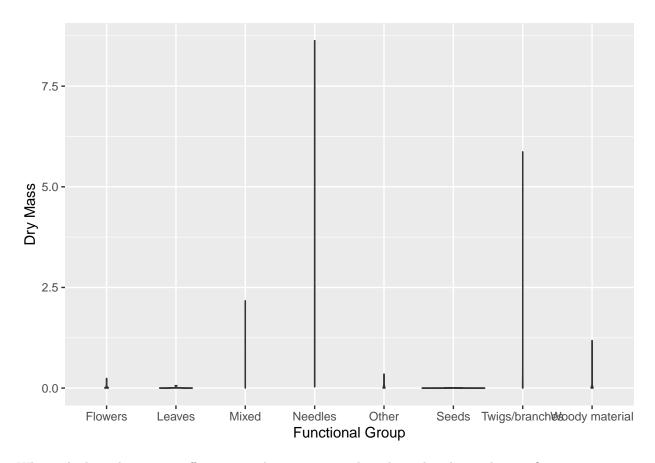


15. Using geom\_boxplot and geom\_violin, create a boxplot and a violin plot of dryMass by functional-Group.

```
#Constructing a boxplot to see the range of Litter dry mass values as they pertain to functional groups ggplot(Litter) + geom_boxplot(aes(x = functionalGroup, y = dryMass)) + labs(x = "FunctionalGroup", y = "Dry Mass")
```



```
#Constructing a violin plot for comparison of the same values
ggplot(Litter) +
  geom_violin(aes(x = functionalGroup, y = dryMass)) +
  labs(x = "Functional Group", y = "Dry Mass")
```



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

Answer: The boxplot is a more effective visualization option than the violin plot in this case because it allows us to effectively see the interquartile range and median values for "mixed" and "needles" functional groups (the two groups featuring the largest IQR and median values) whereas the violin plot is able to show the range of values for these two functional groups but not their median values or the density distribution of each functional group's values. The violin plot is including outliers within its depicted ranges, drawing attention to "twigs/branches" as the functional group with the second largest range of values to needles, leaving no knowledge about the narrow IQR that it has relative to the "mixed" functional group, as revealed by the box plot.

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

Answer: Needles and mixed litter tend to have the highest dry biomass at these sites based on the reported median dry mass value depicted on the box plot.