# 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 <>.

## 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)

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 ecotoxicologoy of neonicotinoids on insects? Feel free to do a brief internet search if you feel you need more background information.
  - Answer: This dataset examining the ecotoxicology of neonicotinoids on insects can help inform future ecotoxicology research by allowing the EPA to better understand the effects on insecticides on insects and how varying levels of exposure to neoincotinoids can affect the lifespan of an insect. This can then help scientists better understand the impacts of particular insecticides on insects and how particular types of neonicotinoids may be more or less potent.
- 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 may be interested in studying litter and woody debris to better understand their impact on ecosystem health as well as the cycling of carbon and other nutrients across the terrestrial sites.

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 at terrestrial NEON sites with woody vegetation in tower plots. The sampling occurs in a different number of plots depending on characteristics of the vegetation. Information on sampling methods *Trap placements within the plots are randomized when the sites contain over 50% aerial cover of woody vegetation over 2 meters in height.* Trap placements within plots are targeted when the sites contain less than 50% aerial cover of woody vegetation. \*The sampling frequency of sites varies based on the type of site. Ground traps are sampled once per year whereas elevated traps are sampled more frequently (approximately once every two weeks).

## Obtain basic summaries of your data (Neonics)

5. What are the dimensions of the dataset?

dim(Neonics)

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

#4623 rows by 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?

summary(Neonics\$Effect) #Shows which effects are the most prevalent

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

Answer: The most common effects that are studied are population and mortality. I believe the effects may be specifically of interest because these effects demonstrate how the neonics may affect the life expectancy of species being studied and how the neonics may place a role in insect mortality.

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)

##	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 Asian Citrus Psyllid	62 Parastic Wasp
##	ASIAN CITIUS ESYTTIC	rarastic wasp 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	December 2015
## ##	Cotton Aphid 33	Predatory Mite 33
##	Ladybird Beetle Family	Parasitoid
##	30	30
##	Scarab Beetle	Spring Tiphia
##	29	29
##	Thrip Order	Ground Beetle Family
##	29	27
##	Rove Beetle Family	Tobacco Aphid
##	27	27
##	Chalcid Wasp	Convergent Lady Beetle
##	25	25
## ##	Stingless Bee 25	Spider/Mite Class 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 Cadling Math	20
##	Codling Moth	Black-spotted Lady Beetle
## ##	19 Calico Scale	18 Fairyfly Parasitoid
##	18	rallylly ralasitold 18
##	Lady Beetle	Minute Parasitic Wasps
	Baay Decore	

##	18	18
##	Mirid Bug	Mulberry Pyralid
##	18	18
##	Silkworm	Vedalia Beetle
##	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 Mit a	16
## ##	Mite 16	Onion Thrip 16
##	Western Flower Thrips	Corn Earworm
##	western Flower inrips	torn Earworm
##	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	Dond Holf Chidan
##	Pea Aphid 12	Pond Wolf Spider 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 six most commonly studied species in the dataset are the following: 1) Honey Bee, 2) Parasitic Wasp 3) Buff tailed bumblebee 4) Carniolan Honey Bee 5)Bumble Bee 6)Italian Honeybee. With the exception of the parasitic wasp, all of these species are bees. However, all six of these species are pollinators which demonstrate that pollinators are of interest moreso than other species. Since pollinators are crucial in plant fertilization, the EPA may be more interested in studying the effects of neonics on these species in order to understand the chemicals potential impact on plant health.

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"

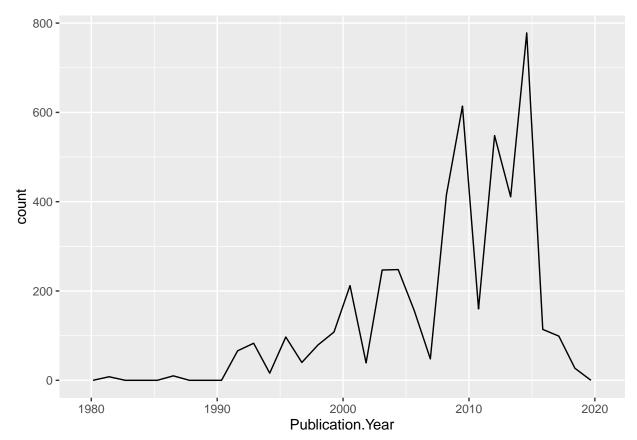
Answer: The class of Conc.1... Author is a factor. This is because some of the concentrations are not numeric (ex. NR which may be not reported) or are numeric values with a " $\sim$ " sign which indicate that they are approximations.

## Explore your data graphically (Neonics)

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

```
library(ggplot2)
ggplot(Neonics) + geom_freqpoly(aes(x=Publication.Year))
```

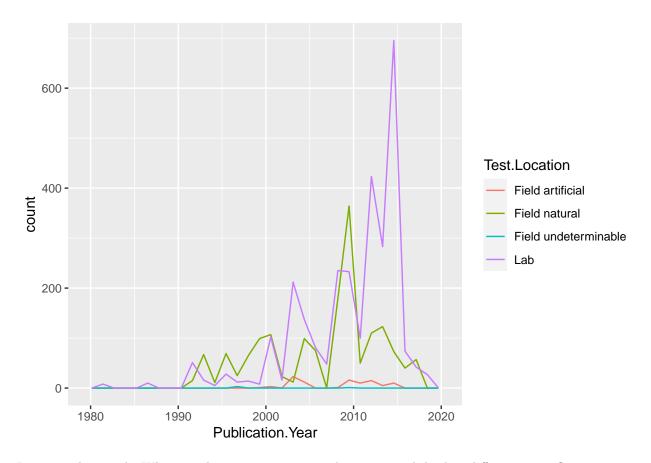
## `stat\_bin()` using `bins = 30`. Pick better value with `binwidth`.



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))
```

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

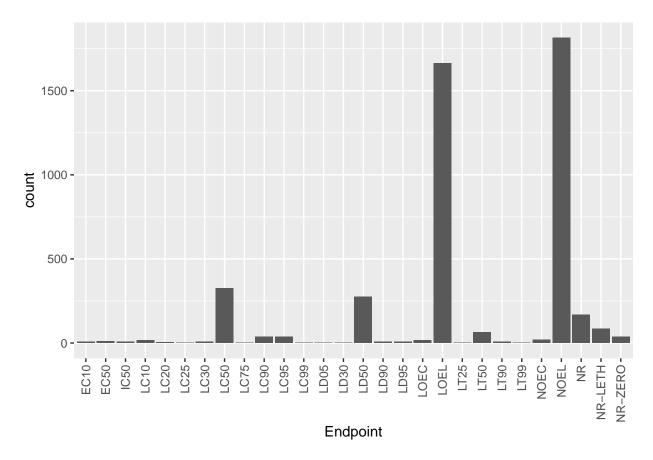


Interpret this graph. What are the most common test locations, and do they differ over time?

Answer: By examining the graph, it is evident that the most common test locations from 1990 to present day are a lab and a field-natural. The field-natural is the most prevalent test location from the early 1990s to early 2000s while the lab surpasses the field-natural in frequency starting in the early 2000s. Based on this graph, the labs' usage increases rapidly and peaks between 2010-2015 and begins to sharply decline starting in 2015. Prior to the peak in lab testing locations, the field-natural experiences a peak followed by a decrease.

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), na.rm=TRUE) +geom\_bar(aes(x=Endpoint)) +theme(axis.text.x = element\_text(angle=90, vj



Answer: The most common endpoint coordinates are NOEL and LOEL. NOEL corresponds to no-observable-effect-level and LOEL corresponds to the lowest-observable-effect-level. Therefore, the more commonly observed endpoints in this dataset correspond to the highest dose where there is no effect on the control group and the lowest dose where there is an effect on the group control. These endpoints can help one better understand the threshold at which a neonics has an effect on insects.

## 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) #collectDate is a factor

## [1] "factor"
library(lubridate)

##
## Attaching package: 'lubridate'

## The following objects are masked from 'package:base':

##
## date, intersect, setdiff, union
Litter$collectDate<-as.Date(Litter$collectDate) #collectDate is now a date
unique(Litter$collectDate)</pre>
```

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

## #Litter was sampled on August 2 and August 30 in 2018

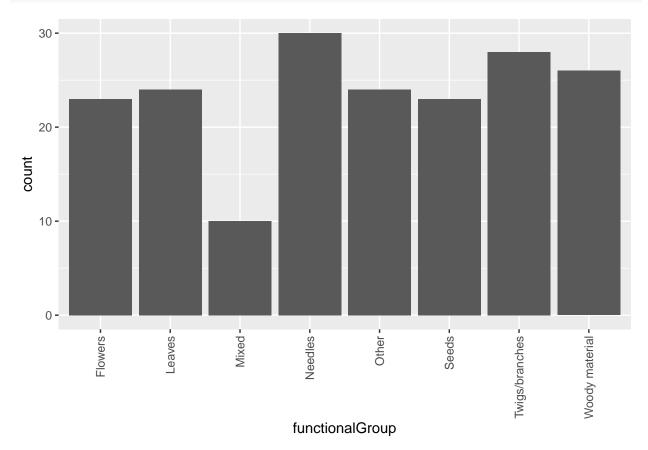
13. Using the unique function, determine how many plots were sampled at Niwot Ridge. How is the information obtained from unique different from that obtained from summary?

```
unique(Litter$plotID) #12 plots were sampled at Niwot Ridge
    [1] NIWO_061 NIWO_064 NIWO_067 NIWO_040 NIWO_041 NIWO_063 NIWO_047 NIWO_051
    [9] NIWO_058 NIWO_046 NIWO_062 NIWO_057
## 12 Levels: NIWO_040 NIWO_041 NIWO_046 NIWO_047 NIWO_051 NIWO_057 ... NIWO_067
summary(Litter$plotID)
## NIWO_040 NIWO_041 NIWO_046 NIWO_047 NIWO_051 NIWO_057 NIWO_058 NIWO_061
##
                  19
                           18
                                     15
                                                                 16
                                                                          17
## NIWO 062 NIWO 063 NIWO 064 NIWO 067
##
         14
                  14
```

Answer: By using the "unique" function, we can determine that 12 plots were sampled at Niwot Ridge. When obtaining data on the plots using the summary function, we obtain the number of observations at each unique plot rather than the number of unique plots being sampled in the dataset.

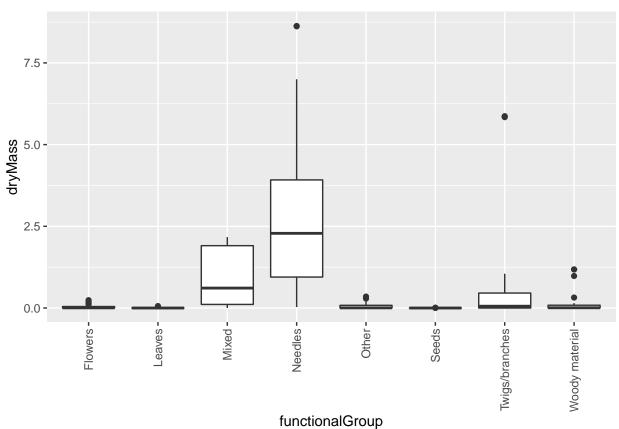
14. Create a bar graph of functionalGroup counts. This shows you what type of litter is collected at the Niwot Ridge sites. Notice that litter types are fairly equally distributed across the Niwot Ridge sites.

ggplot(Litter) + geom\_bar(aes(x=functionalGroup)) + theme(axis.text.x = element\_text(angle=90, vjust=0.

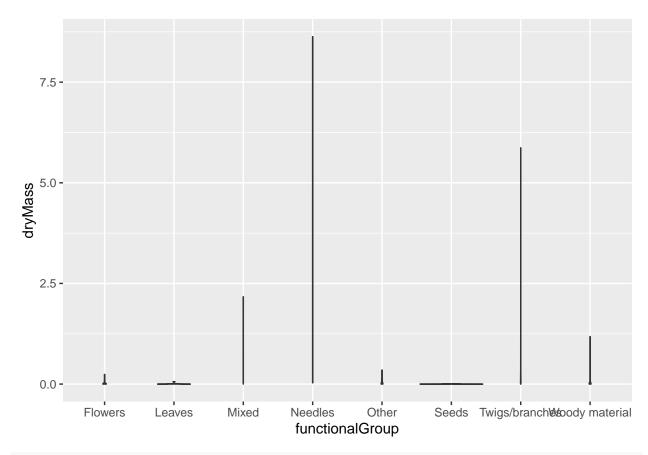


15. Using geom\_boxplot and geom\_violin, create a boxplot and a violin plot of dryMass by functional-Group.

```
ggplot(Litter) +
  geom_boxplot(aes(x=functionalGroup, y= dryMass)) + theme(axis.text.x = element_text(angle=90, vjust=0)
```



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



#Tried to visualize data with the violin plot. The violin does not show spread as easily

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

Answer: In this case, the boxplot is a more effective visualization option than the violin plot because it displays the spread of data and provides a visualization of the dryMass summary statistics. The boxplot displays the median, interquartile range, and outliers across each functional group and those statistics are not apparent in the violin 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 biomass at the sites in this study. While twigs and branches tend to have a lower biomass relative to needles and mixed litter, there is an outlier that is heavier.