

Assignment 3: Data Exploration

Sena McCrory

OVERVIEW

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

Directions

1. Change “Student Name” on line 3 (above) with your name.
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., “Salk_A03_DataExploration.Rmd”) prior to submission.

The completed exercise is due on Tuesday, January 28 at 1:00 pm.

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.

```
getwd()

## [1] "C:/Users/senam/Box Sync/My Documents/MEM classes/Duke Spring 2020/DataAnalytics/Environmental_D

library(tidyverse)

neonic.data <- read.csv("./Data/Raw/ECOTOX_Neonicotinoids_Insects_raw.csv")

litter.data <- read.csv("./Data/Raw/NEON_NIWO_Litter_massdata_2018-08_raw.csv")
```

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: neonics are a widely used category of pesticides used in agricultural production. They have been implicated in the collapse of bee and other important/beneficial insect populations.

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 in forests can help us to understand the productivity of the forest and information about carbon and other nutrient cycles including fluxes, decomposition rates, and other biogeochemical measurements. Timing of debris amounts can also tell us about the phenology of leaf senescence and leaf drop in fall.

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:

- masses are reported for separate functional groups - e.g. leaves, twigs, seeds, flowers, etc
- the sampling design is spatially distributed with a pair (one elevated and one ground litter trap) per 400 sq m of the study plot (there are 20 plots), so there may be varying numbers of traps for different plots
- temporal sampling is irregular - with more frequent sampling during autumn and gaps during winter or dormant seasons

Obtain basic summaries of your data (Neonics)

5. What are the dimensions of the dataset?

```
dim(neonic.data)
```

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

6. Using the `summary` function, determine the most common effects that are studied. Why might these effects specifically be of interest?

```
summary(neonic.data$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: most common effects studied were “mortality” and “population” likely because these are the easiest and quickest to test. These effects are of interest because the researchers want to determine whether exposure to neonics could be responsible for decreased insect populations.

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(neonic.data$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	62
##	Asian Citrus Psyllid	Parastic Wasp
##	60	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	33
##	Cotton Aphid	Predatory Mite
##	33	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	Spider/Mite Class
##	25	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	20
##	Codling Moth	Black-spotted Lady Beetle
##	19	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
##	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 Woolly 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 6 most reported species are all hymenoptera (bees and wasps) - bees are ecologically important for pollination and therefore food production and parasitic wasps are often beneficial insects for agricultural crops because they control populations of unwanted pest insects

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(neonic.data$Conc.1..Author.)
```

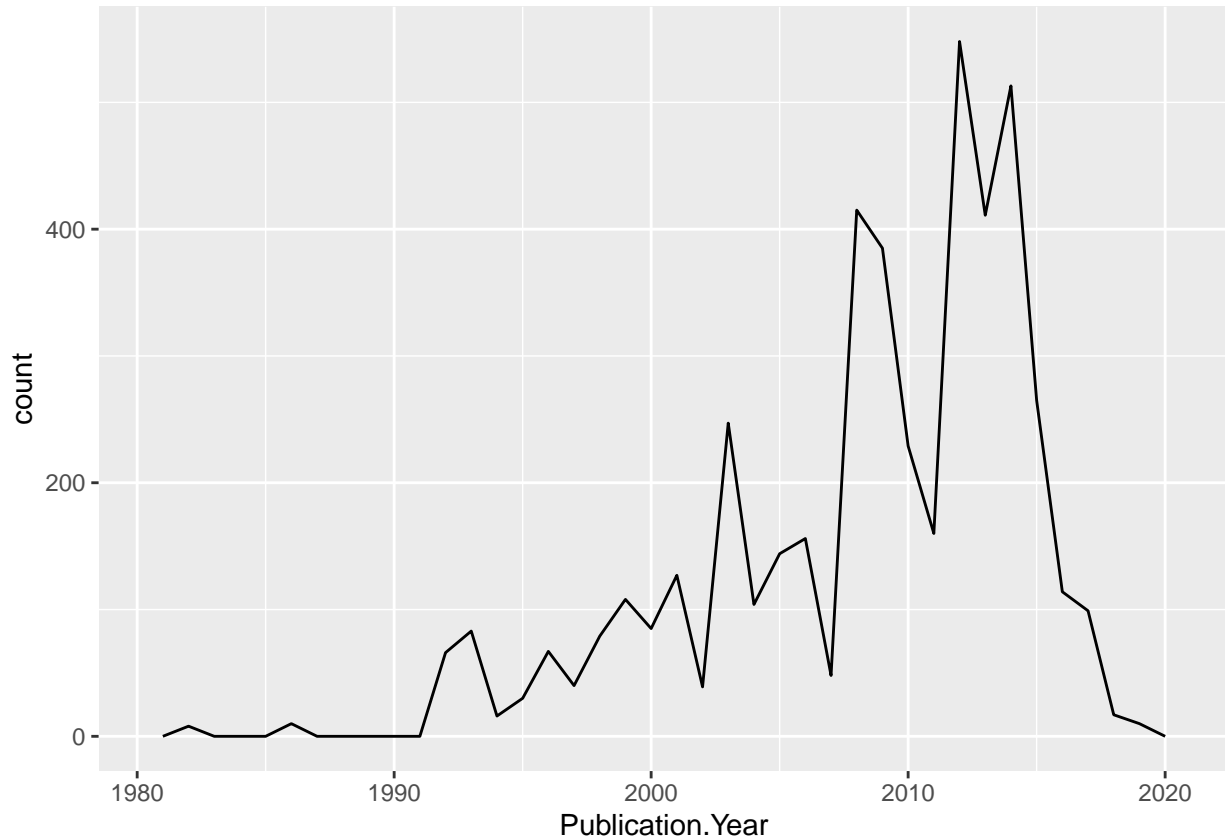
```
## [1] "factor"
```

Answer: some of the values include other symbols like ~ and / and so they cannot be interpreted as numeric values by R

Explore your data graphically (Neonics)

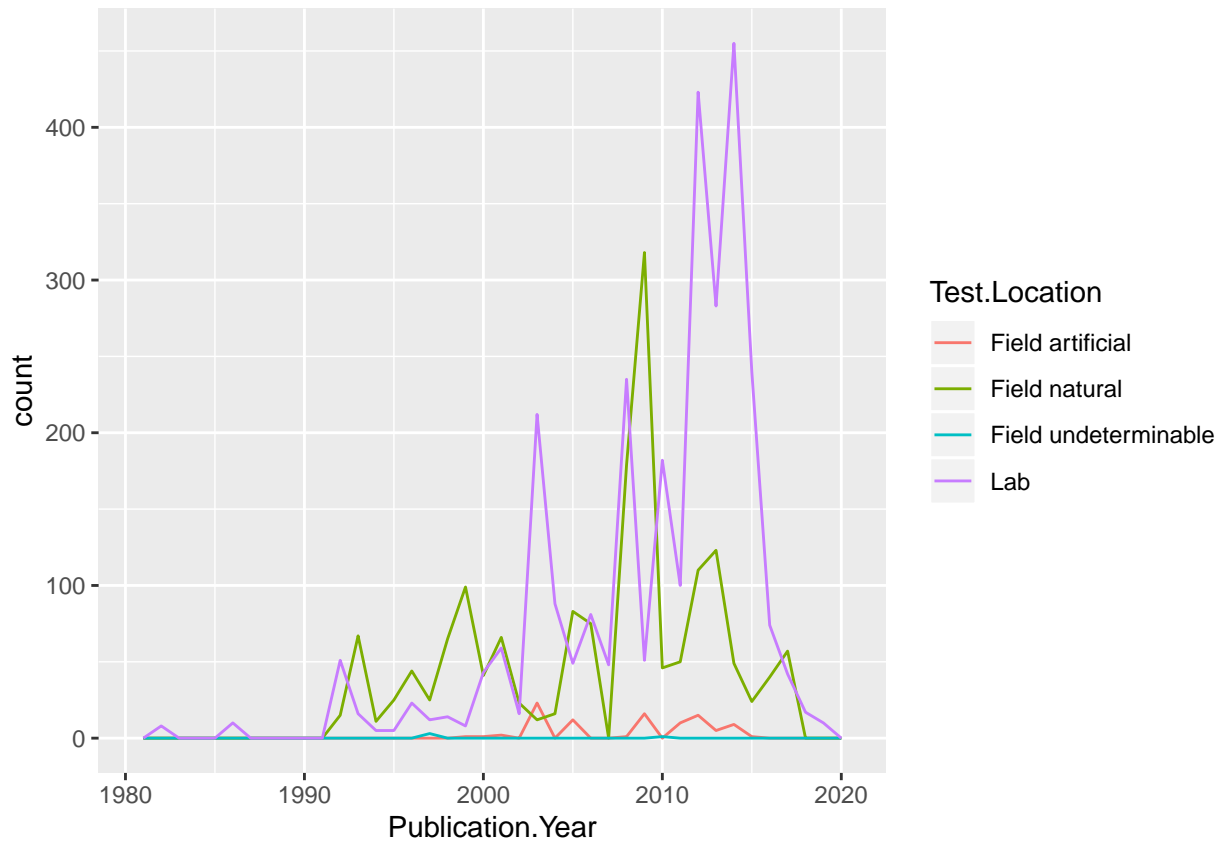
9. Using `geom_freqpoly`, generate a plot of the number of studies conducted by publication year.

```
ggplot(neonic.data) +  
  geom_freqpoly(aes(x=Publication.Year), binwidth = 1)
```



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

```
ggplot(neonic.data) +  
  geom_freqpoly(aes(x=Publication.Year, color = Test.Location), binwidth = 1)
```

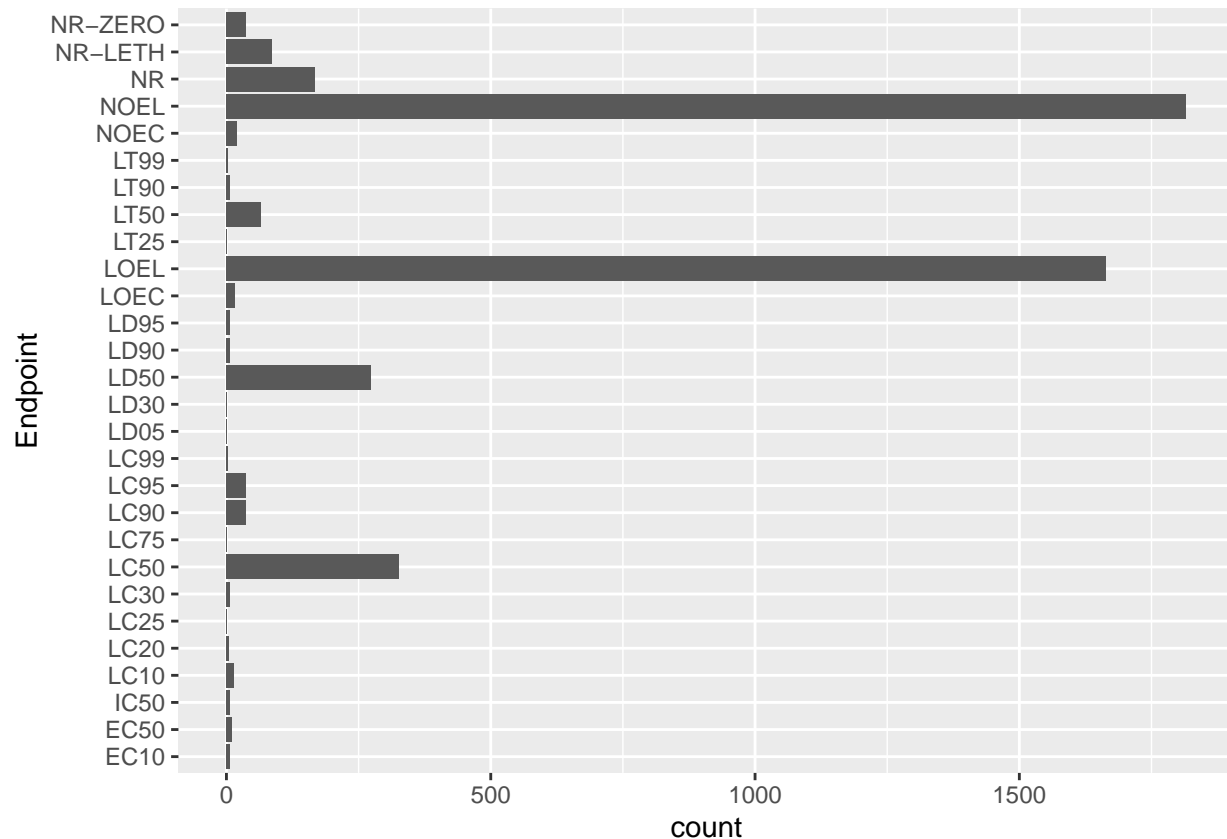


Interpret this graph. What are the most common test locations, and do they differ over time?

Answer: The number of studies has increased substantially over time. Field studies were more common at first but then lab studies took over in mid 2000s, field studies again had a resurgence but then were quickly replaced with lab studies after 2010. Other test locations remained pretty uncommon

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(neonic.data)+
  geom_bar(aes(x=Endpoint))+
  coord_flip()
```



Answer: most common endpoints reported are NOEL (No observed effect level) and LOEL (lowest observed effect level).

In a study design using multiple different concentrations, the LOEL is defined as the lowest concentration at which a statistically significant effect (deviation from the control) is seen. And the NOEL is the highest concentration at which there is no statistically significant difference from the control.

Explore your data (Litter)

- 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.data$collectDate)
```

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

```
litter.data$collectDate <- as.Date(litter.data$collectDate, format = "%Y-%m-%d")
class(litter.data$collectDate)
```

```
## [1] "Date"
```

```
unique(litter.data$collectDate)
```

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

- 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.data$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
```

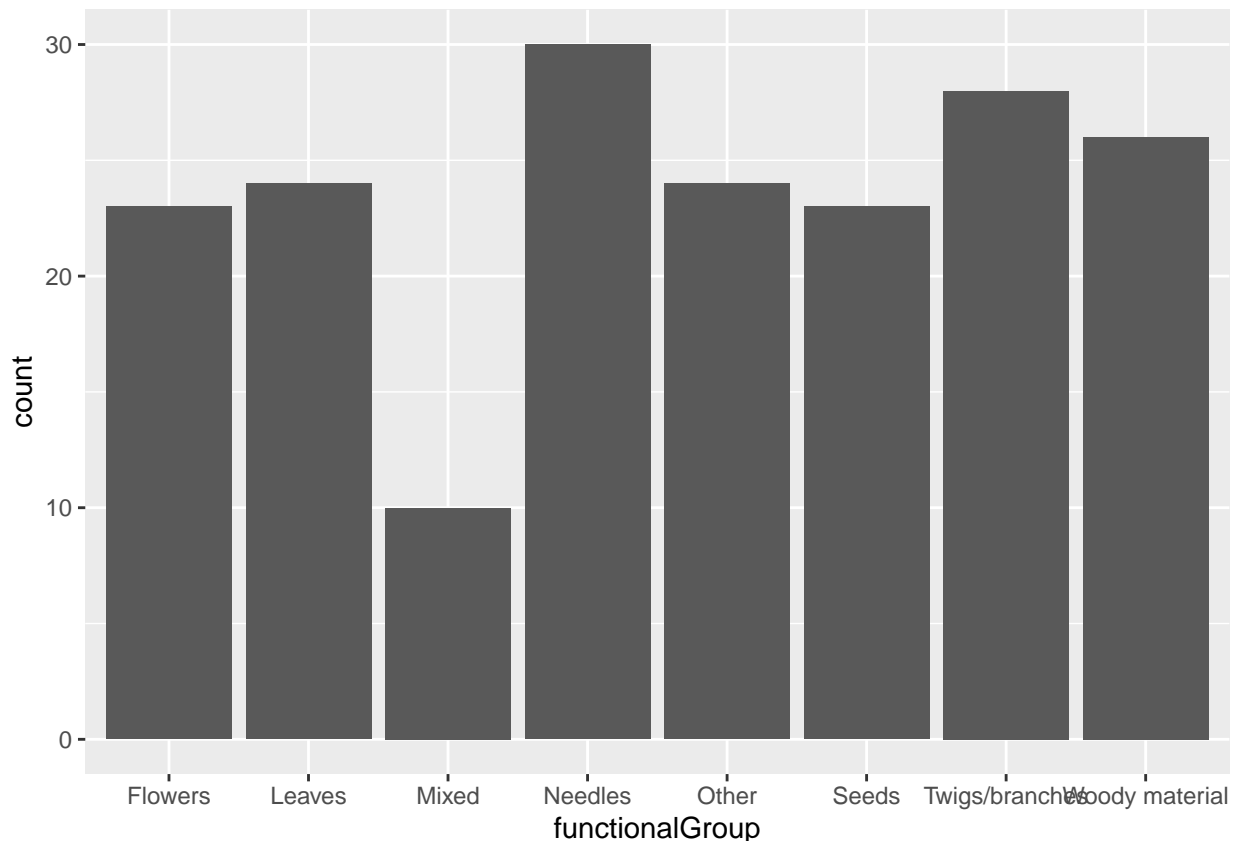
```
summary(litter.data$plotID)
```

```
## 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: 12 plots were sampled. “Unique” lists the different unique values in the column and the total number of levels while “summary” lists the values also with the number of times they occur

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.

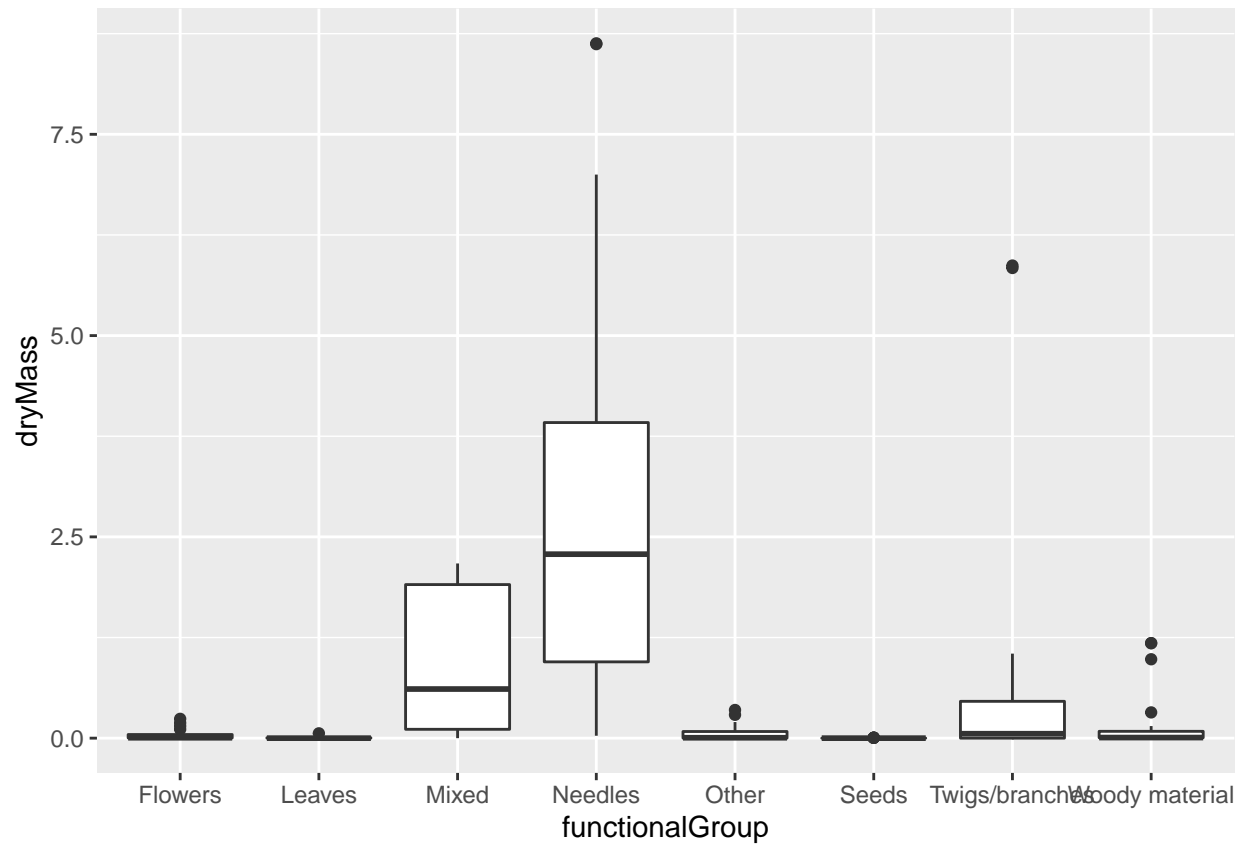
```
ggplot(litter.data, aes(x=functionalGroup)) +
  geom_bar()
```



15. Using `geom_boxplot` and `geom_violin`, create a boxplot and a violin plot of `dryMass` by `functionalGroup`.

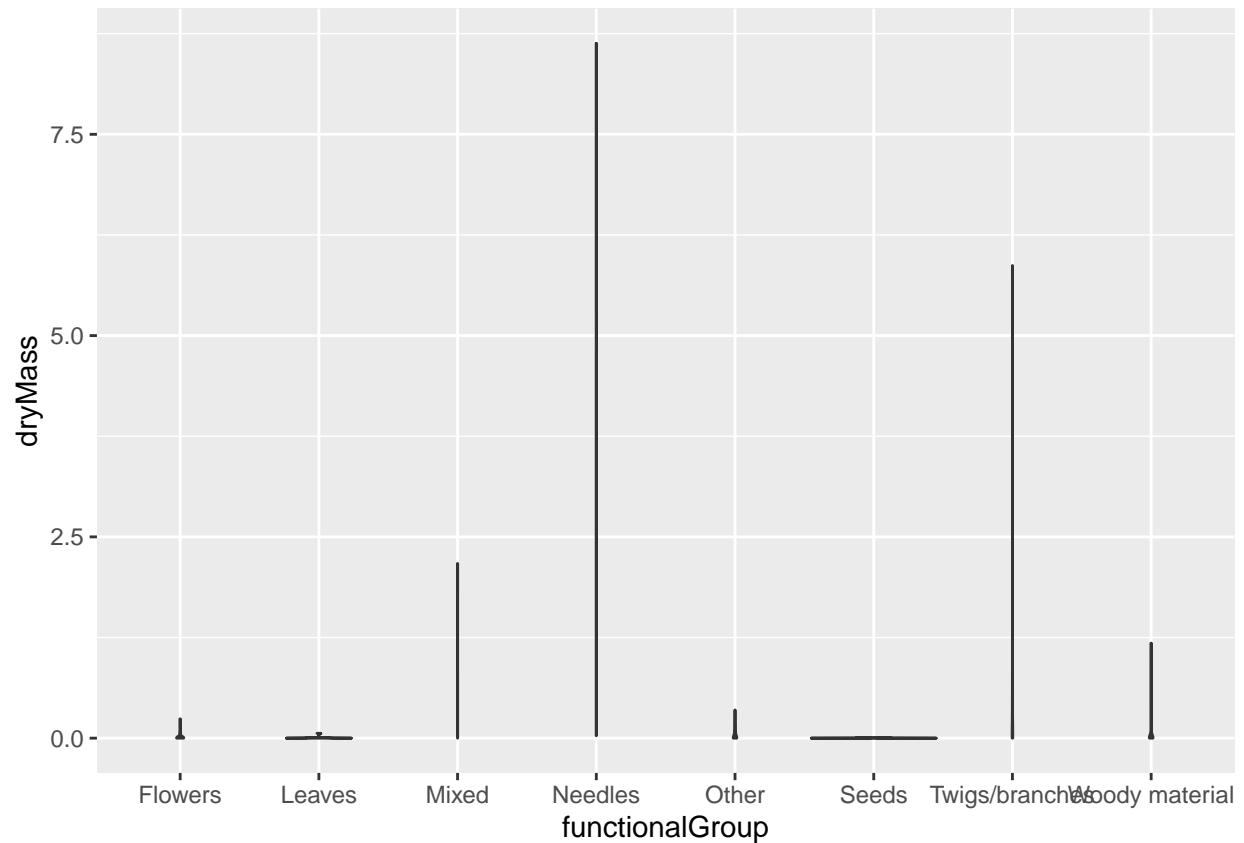
```
litter.plot <- ggplot(litter.data, aes(x=functionalGroup, y = dryMass))
```

```
litter.plot +
  geom_boxplot() #+
```

```
#scale_y_log10()

litter.plot +
  geom_violin() #+
```



```
#scale_y_log10()
```

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

Answer: the distributions of dryMass have a wide spread so the violin plot does not effectively show the shape of the distribution unless a log transformation is used

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

Answer: “Needles” and “Mixed” seem to have the highest dryMass of the different functional groups