

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” 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 <>.

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() #changed global options to current working directory so that it referred to main folder instead  
  
library(tidyverse)  
library(lubridate)  
Neonics <- read.csv("./Data/Raw/ECOTOX_Neonicotinoids_Insects_raw.csv")  
Litter <- 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: Neonicotinoids can inadvertently harm beneficial insects, such as bees and other pollinators. Furthermore, other non-target organisms, including humans, can be experiencing environmental risks associated with the pesticides. As many are water-soluble, chemical compounds

can be absorbed and processed by plants and crops and can be dispersed throughout the environment. Repeated first-hand exposure without proper PPE can cause negative side-effects to humans, in addition to impacting fetal development and causing birth defects.

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 can serve as indicators to the health of the forest. The amount of organic matter can impact the quality of habitat for organisms within the forest, as well as forage material. Composition of the litter and debris can also relay changing forest dynamics, such as if other tree species are dominating forests and altering the environment. Furthermore, the amount of debris, especially leaf litter, can also serve as a proxy for changes in climate and signify if seasons are changing at unusual rates. Woody debris, especially large woody debris, is important for sediment retention and stream channel morphology.

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 fine woody debris was collected from individual sampling bouts and collected from elevated and ground traps. * Mass data was recorded to the closest 0.01 g and sorted based on functional group. * Location of tower plots are selected randomly based on its relative location to primary and secondary airsheds, and depending on vegetative cover and height, plot size would vary.

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 “Effects” column, determine the most common effects that are studied. Why might these effects specifically be of interest?

```
summary(as.factor(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
```

```
#included as.factor because effect is a character
```

Answer: This information will indicate what effects the neonicotinoids have on the targeted species group. The most common effects including mortality and population growth.

- 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(as.factor(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	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 6 most common species are honey bees, parasitic wasps, buff tailed bumblebees, Carniolan honey bee, bumble bee, and Italian honey bee. The majority of these species are important pollinators, therefore it would be determinental if they suffered effects from the pesticide. However, parasitic wasps would be a crucial targeted species to reduce pollinator mortality.

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.)
```

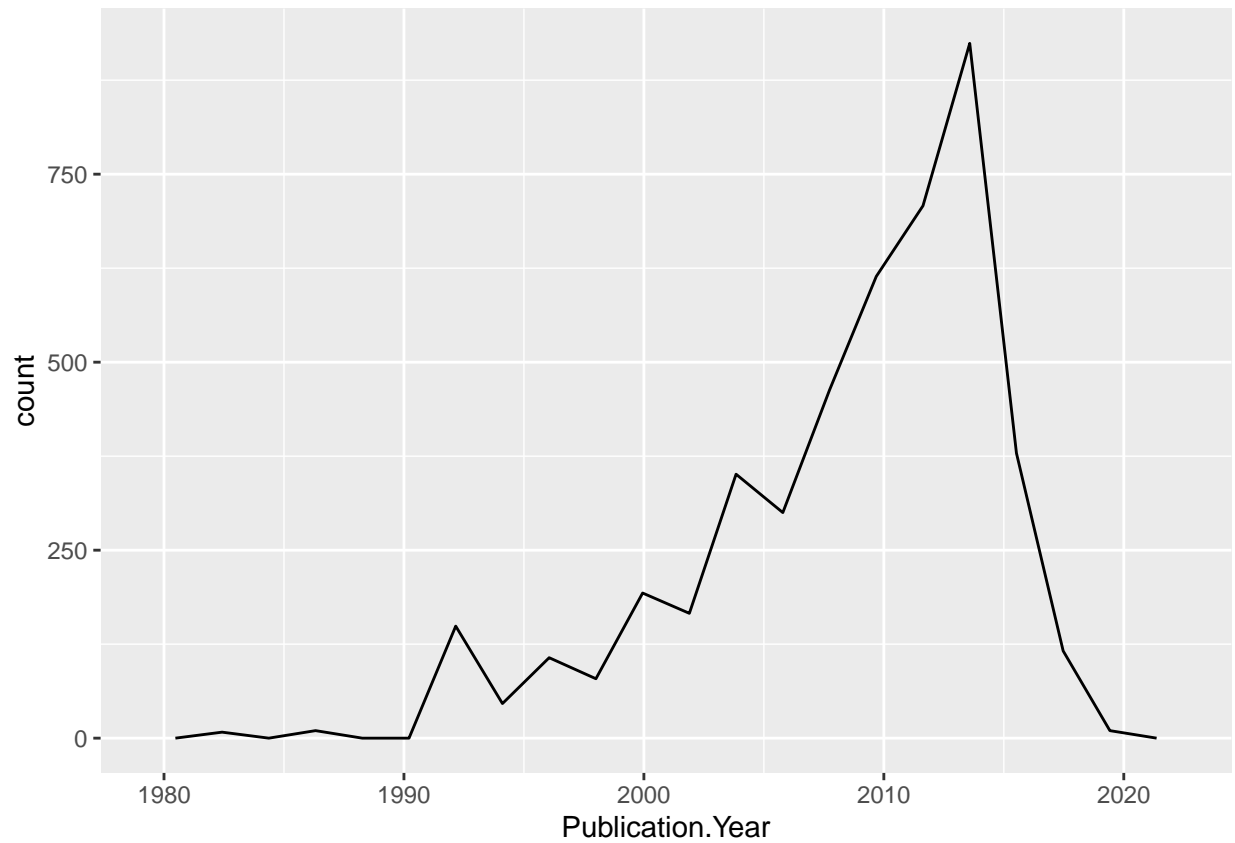
```
## [1] "character"
```

Answer: There are non-numerical values within the class of `Conc.1..Author.` Since these are not all numbers it makes the value character

Explore your data graphically (Neonics)

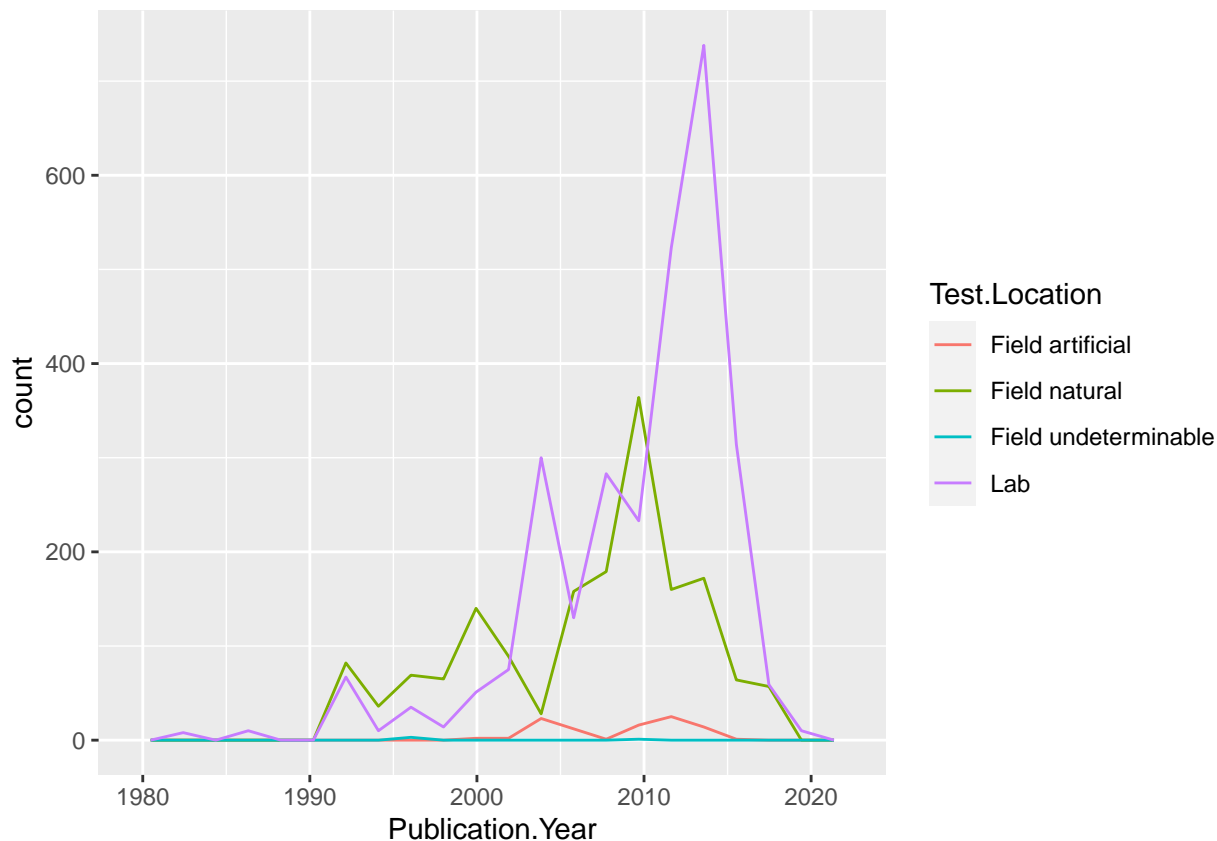
9. Using `geom_freqpoly`, generate a plot of the number of studies conducted by publication year.

```
ggplot(Neonics) +
  geom_freqpoly(aes(x = Publication.Year), bins=20 )
```



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=20)
```

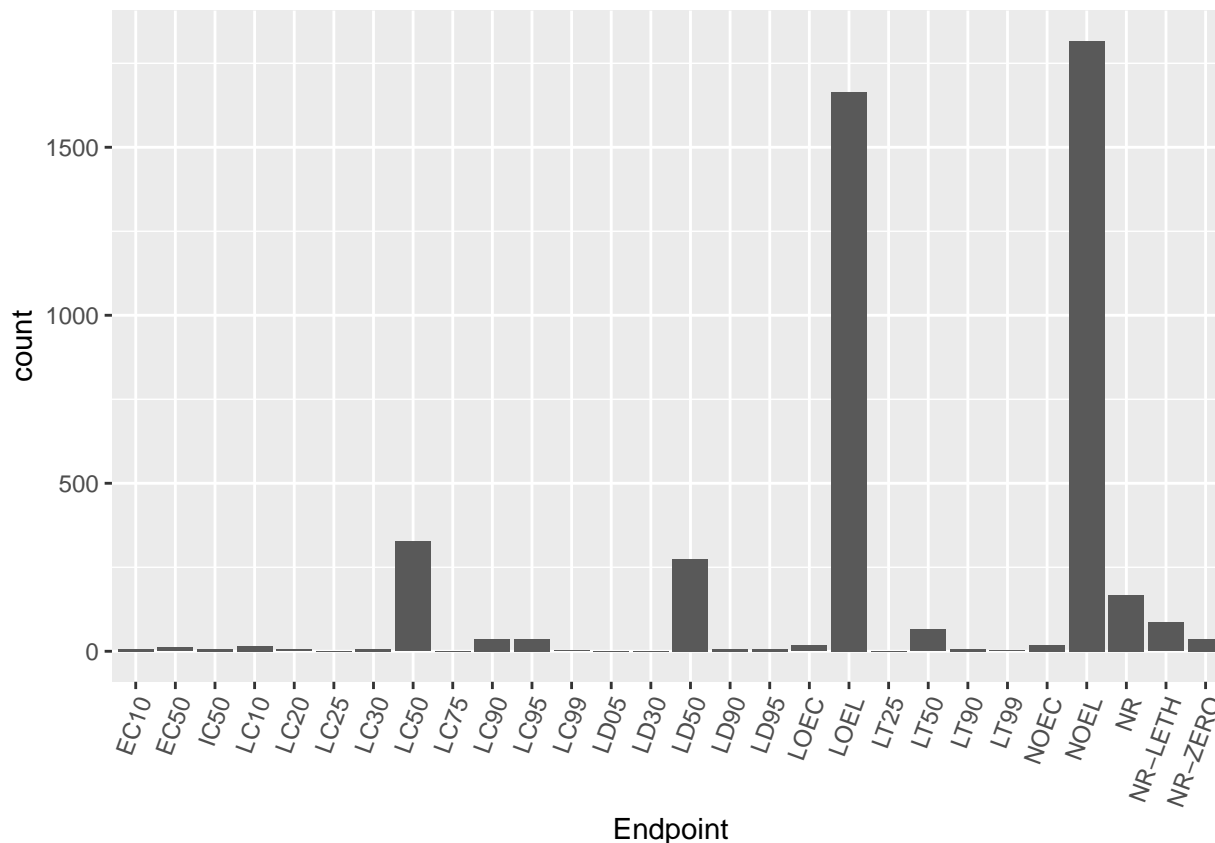


Interpret this graph. What are the most common test locations, and do they differ over time?

Answer: The most common test locations are in the lab and in the natural field settings. Lab settings increased after 2000, and sharply increased in 2010. Natural field settings increased at a more gradual rate between 1990 and 2000 with some drops, but experienced a significant drop around 2005. Subsequently they increased in 2010 but then decreased again.

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) +
  geom_bar(aes (Endpoint)) +
  theme(axis.text.x = element_text(angle=70, hjust=1))
```



Answer: The two most common Endpoints are LOEL and NOEL. LOEL, lowest-observable-effect-level, is a terrestrial endpoint. It reports the lowest dose (concentration) that produces effects that were significantly different from control responses. NOEL, also a terrestrial endpoint, is no observed effects residue. This reports the highest residue concentration that produces effects, which is not significantly different from control responses.

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$collectDate) #collectDate originally a character
```

```
## [1] "character"
```

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

```
class(Litter$collectDate)
```

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



```
unique(Litter$collectDate)
```

```
## [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$plotID)
```

```
## [1] "NIWO_061" "NIWO_064" "NIWO_067" "NIWO_040" "NIWO_041" "NIWO_063"  
## [7] "NIWO_047" "NIWO_051" "NIWO_058" "NIWO_046" "NIWO_062" "NIWO_057"
```

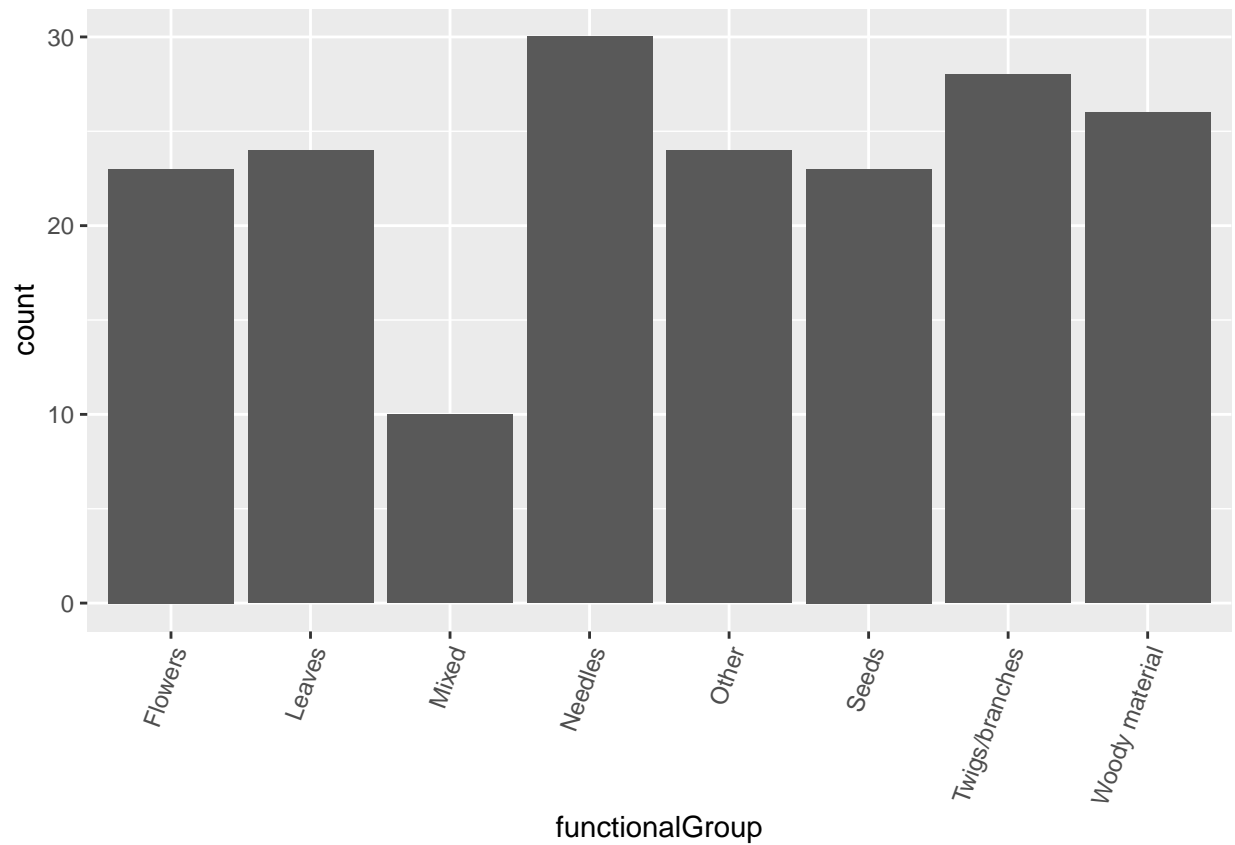
```
summary(Litter$plotID)
```

```
##      Length      Class      Mode  
##      188 character character
```

Answer: The `unique` function presents the unique values within the field. For `plotID`, there are 12 different types of plots, where samples correspond to one of those types. This differs from the `summary` function, where it reports the total number of observations within the field. This reports that there are 188 observations.

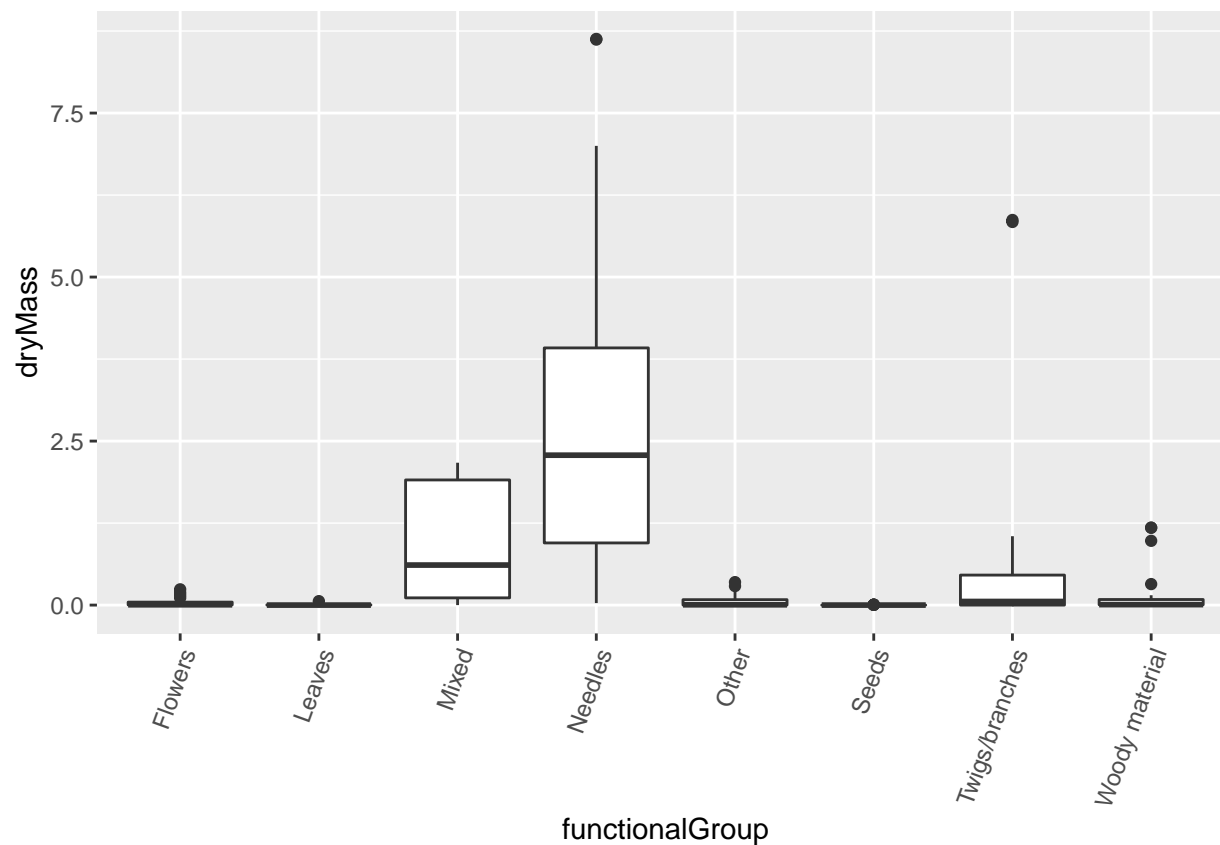
14. Create a bar graph of `functionalGroup` counts. This shows you what type of litter is collected at the Niwot Ridge sites. Notice that litter types are fairly equally distributed across the Niwot Ridge sites.

```
ggplot(Litter) +  
  geom_bar(aes (functionalGroup)) +  
  theme(axis.text.x = element_text(angle=70, hjust=1))
```



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

```
ggplot(Litter) +  
  geom_boxplot(aes(x = functionalGroup, y = dryMass))+  
  theme(axis.text.x = element_text(angle=70, hjust=1))
```

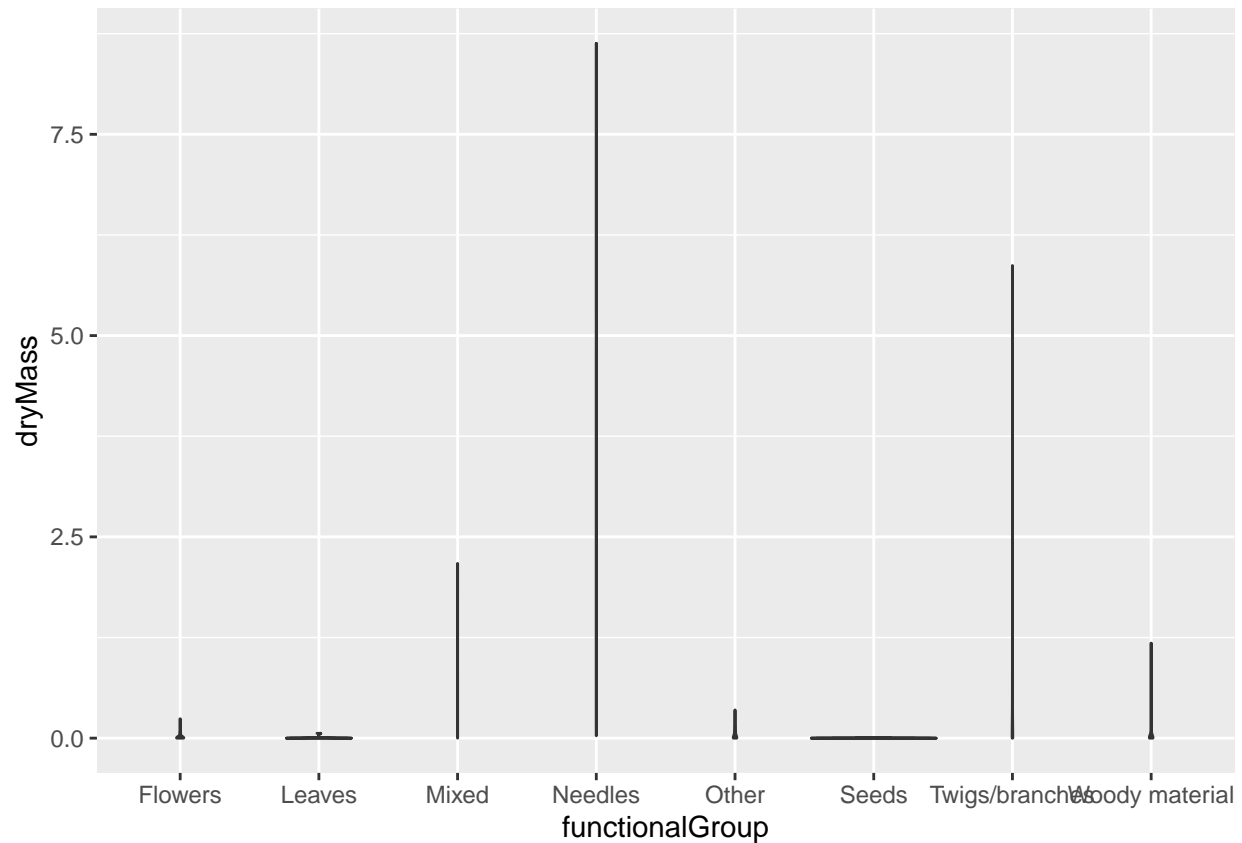


```
ggplot(Litter) +
  geom_violin(aes(x = functionalGroup, y = dryMass),
    draw_quantiles = c(0.25, 0.5, 0.75))
```

```
## Warning in regularize.values(x, y, ties, missing(ties), na.rm = na.rm):
## collapsing to unique 'x' values
```

```
## Warning in regularize.values(x, y, ties, missing(ties), na.rm = na.rm):
## collapsing to unique 'x' values
```

```
## Warning in regularize.values(x, y, ties, missing(ties), na.rm = na.rm):
## collapsing to unique 'x' values
```



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

Answer: The boxplot is better at displaying the spread of the data for each functional group. Even though the range for some of the groups was very small, you are also able to visualize outliers. However, the violin plot is not an adequate visualization. Since the values in each functional group are similar and the distribution density is small, the plot is not capable of displaying the probability distribution within the plot since the values are too similar.

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

Answer: Mixed litter and needle litter have the highest biomass at the sample sites.