

# 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.**

```
setwd("~/Desktop/Duke/Data Analytics/Environmental_Data_Analytics_2022")
getwd()
```

```
## [1] "/Users/devindomeyer/Desktop/Duke/Data Analytics/Environmental_Data_Analytics_2022"
```

```
#install.packages("tidyverse")
library(tidyverse)
```

```
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: Since neonicotinoids are a class of insecticides and are used widely, it would be beneficial to know how widespread its effects are. Although used in agriculture to control harmful insects, neonicotinoids are also effective at killing pollinators like bees which are already experiencing massive die-offs in the United States. There is also controversy surrounding exactly how dangerous neonicotinoids are to bees, with field-studies yielding different results than laboratory research.

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 are shown to increase risk of forest fires, especially in the drier more arid regions of the western United States. It is common practice to preemptively burn or remove the woody debris and litter to ensure incidental fires have less fodder.

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: \* One litter trap pair (1 elevated trap & 1 ground trap) is deployed for every 400 m<sup>2</sup> plot area. \* Ground traps are sampled once per year. \* Trap pair placement within plots may be either targeted or randomized, depending on the vegetation.

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

```
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 studies are 1. Population and 2. Mortality, both with an order of magnitude more observations than the other effects. Across studies these are likely measured more because they give the best indication of how effective the insecticide is at killing insects, and potentially how dangerous it is to non-target species.

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          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: 1. Honey Bee, Parasitic Wasp, Buff Tailed Bumblebee, Carniolan Honey Bee, Bumble

Bee, Italian Honeybee. These species are all important pollinator species and very important in agriculture. Parasitic Wasps not only pollinate, they are also a common and effective pest controller. It would be important to know how neonicotinoids would impact these species.

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 in the dataset. It isn't numeric because many of the observations have a backward slash after the number. I'm sure there is an important reason for this but I'm not sure why.

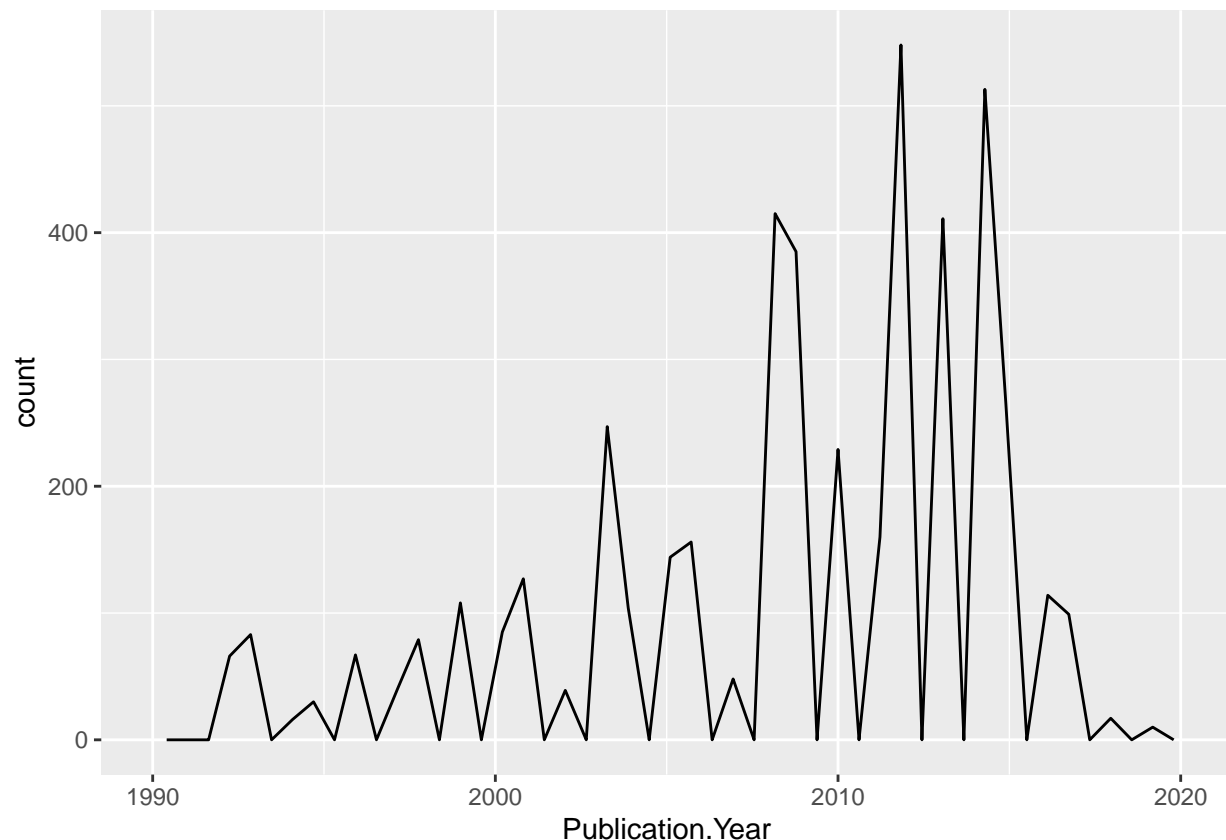
## Explore your data graphically (Neonics)

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

```
ggplot(Neonics) +  
  geom_freqpoly(aes(x = Publication.Year, color = Publication.Year), bins = 50) +  
  scale_x_continuous(limits = c(1990, 2020)) +  
  theme(legend.position = "top")
```

```
## Warning: Removed 18 rows containing non-finite values (stat_bin).
```

```
## Warning: Removed 3 row(s) containing missing values (geom_path).
```

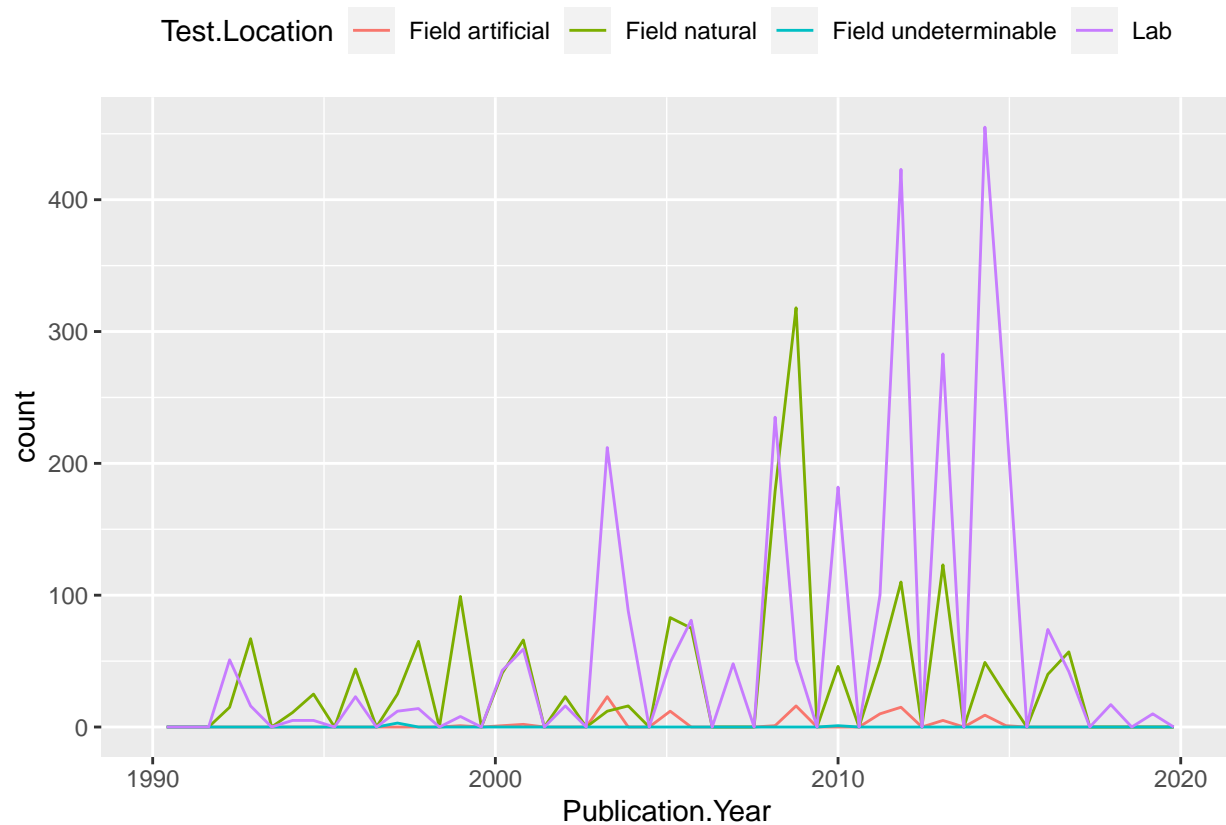


10. Reproduce the same graph but now add a color aesthetic so that different Test.Location are displayed as different colors.

```
ggplot(Neonics) +
  geom_freqpoly(aes(x = Publication.Year, color = Test.Location), bins = 50) +
  scale_x_continuous(limits = c(1990, 2020)) +
  theme(legend.position = "top")
```

```
## Warning: Removed 18 rows containing non-finite values (stat_bin).
```

```
## Warning: Removed 12 row(s) containing missing values (geom_path).
```

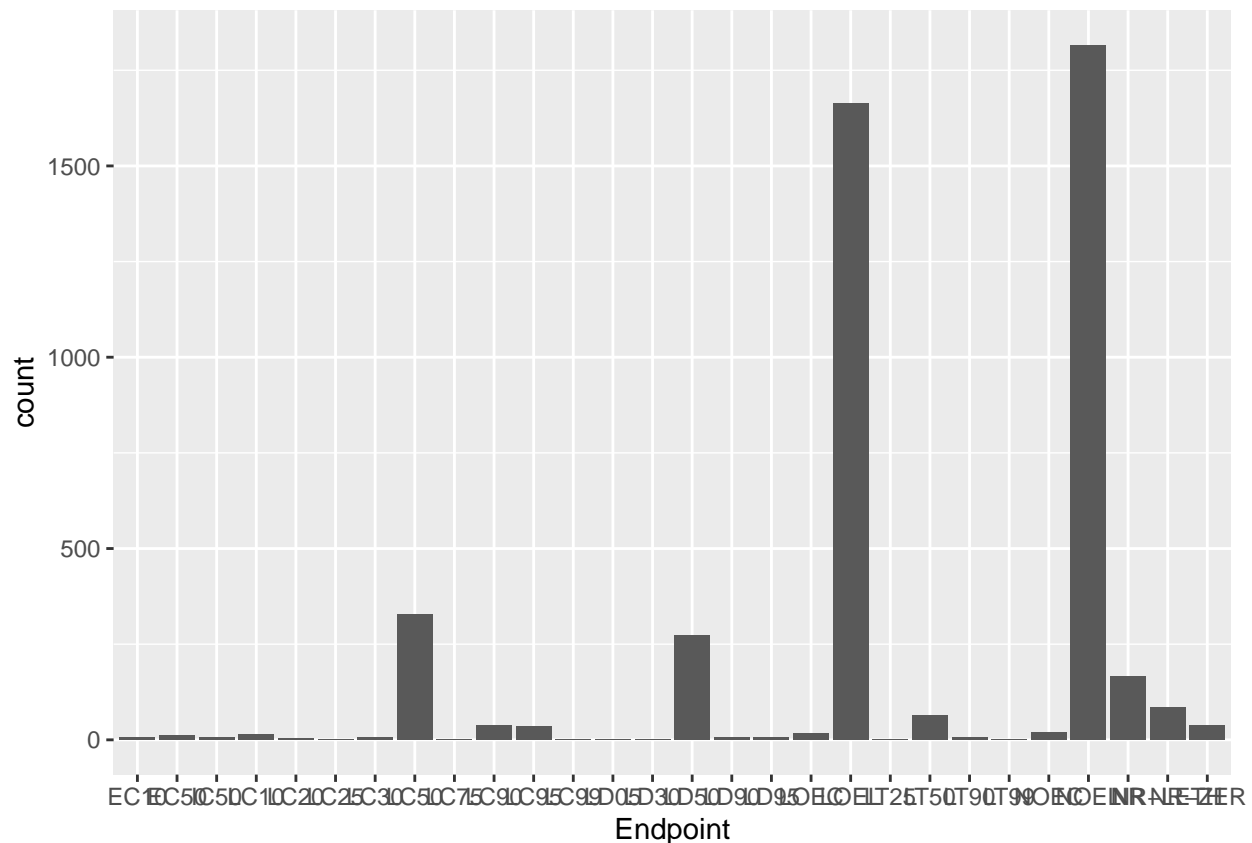


Interpret this graph. What are the most common test locations, and do they differ over time?

Answer: The most common test locations are Lab and Field Natural. Overtime, the relative proportion of Lab to Field Natural locations increases.

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, aes(x = Endpoint)) +
  geom_bar()
```



Answer: NOEL and LOEL are the most common endpoints. NOEL means “no observable effect level”, in which the highest dose produces effects not significantly different from responses of controls. LOEL means “lowest observable effect level”, in which the lowest dose produces effects that were significantly different from the responses of controls.

## 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)
```

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

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

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

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

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

Answer: The class of collectDate is “factor.” On August 2 and August 30 litter was sampled in August 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`?

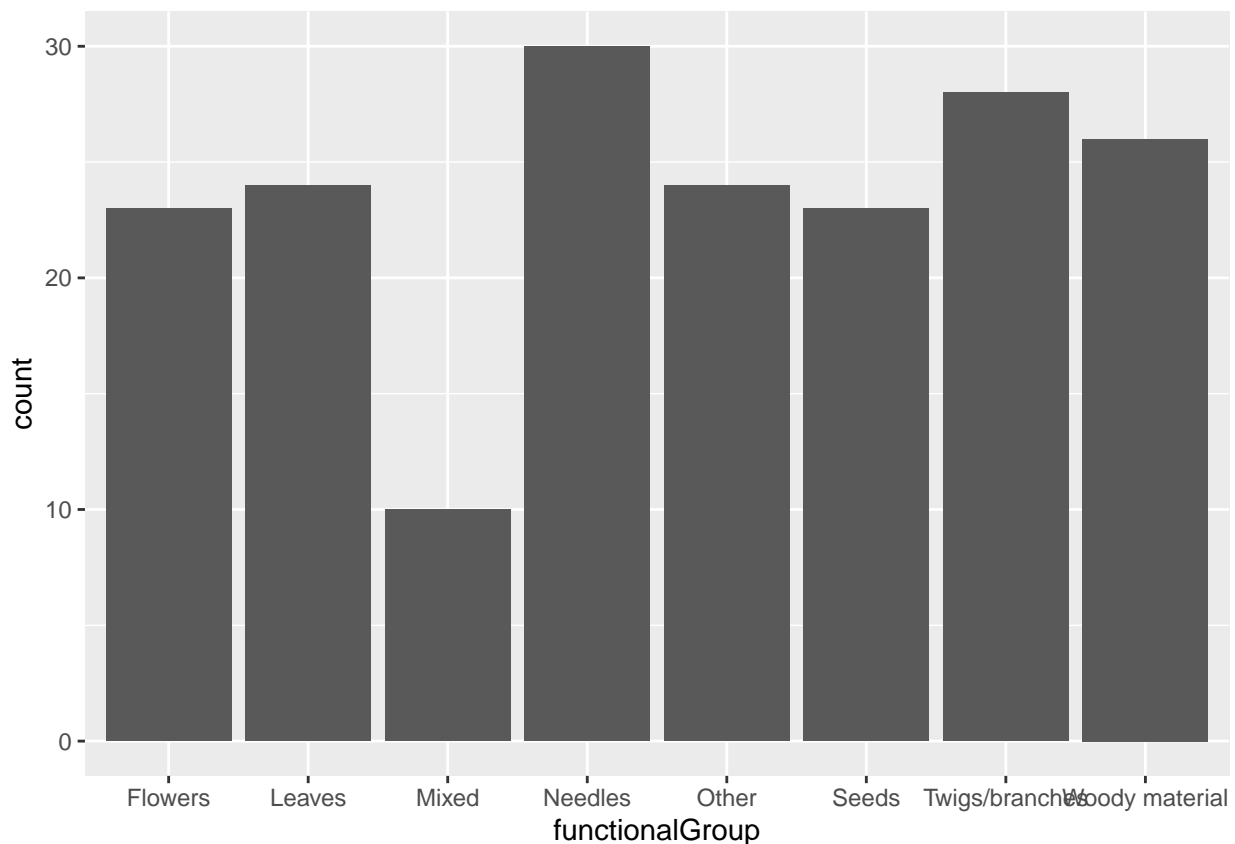
```
unique(Litter$plotID)
```

```
## [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
```

Answer: 12 plots were sampled at Niwot Ridge. This information is different from `summary` because it can be defined to smaller subsets of the data frame.

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.

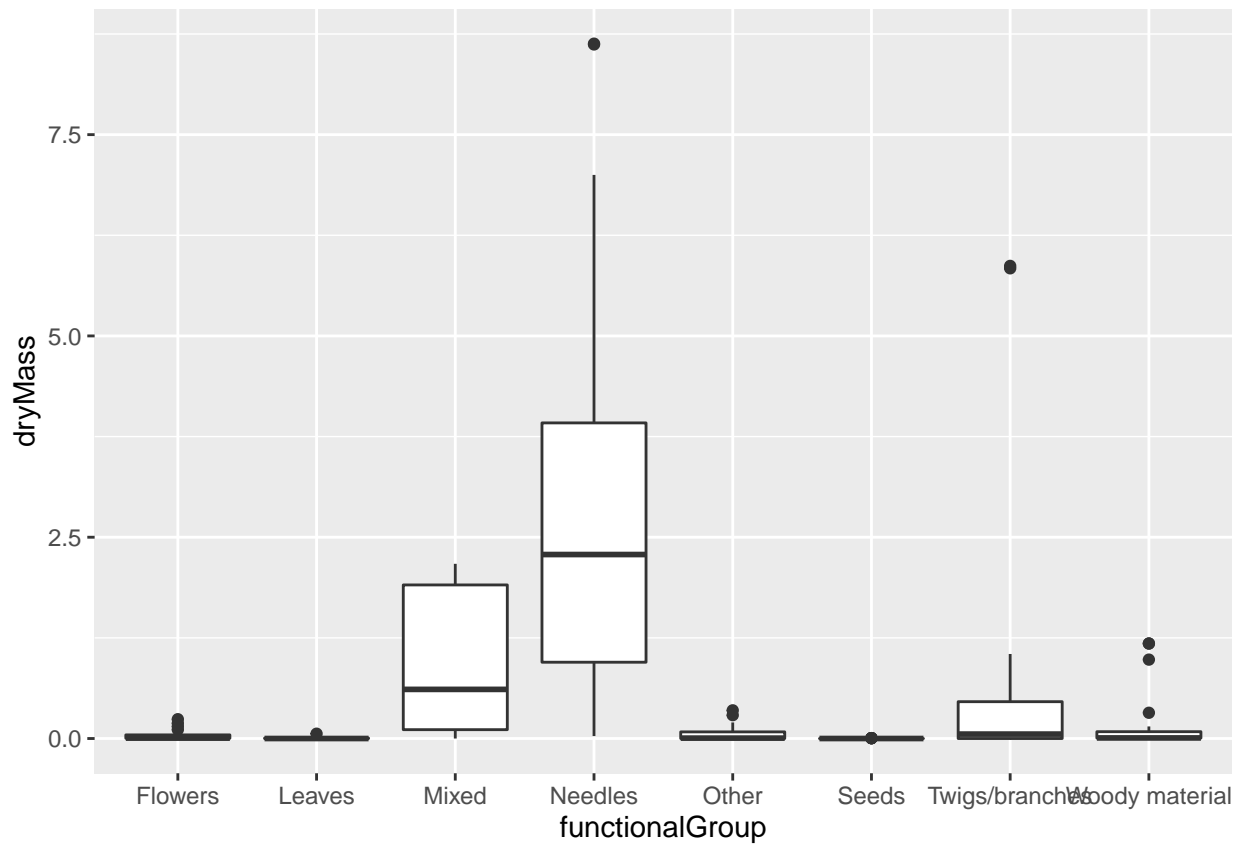
```
ggplot(Litter, aes(x = functionalGroup)) +  
  geom_bar()
```



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))
```

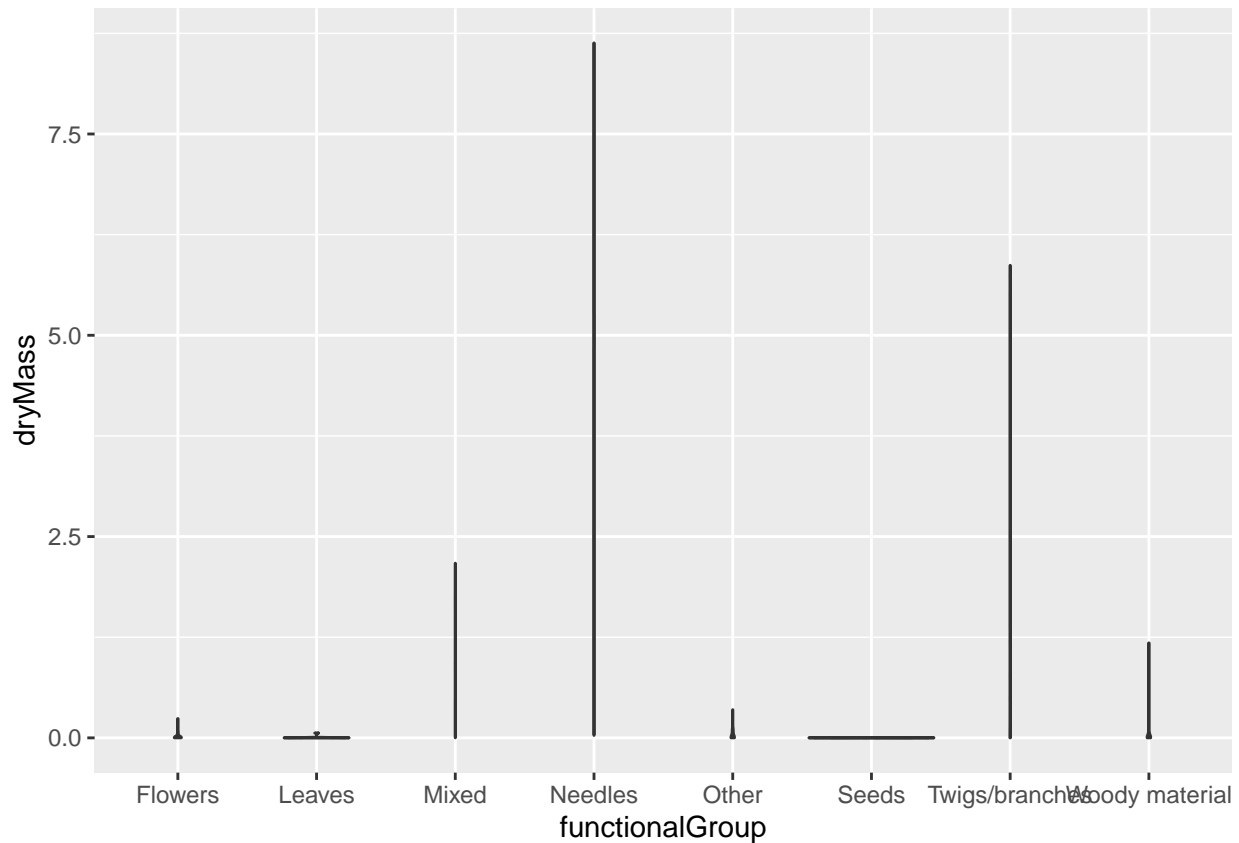


```
ggplot(Litter) +
  geom_violin(aes(x = functionalGroup, y = dryMass),
    draw_quantiles = c(0.25, 0.5, 0.75))
```

```
## 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: A boxplot is more effective in this case because across functional groups the data is fairly evenly distributed across drymass levels. There aren't any large data clusters that would demonstrate the violin effect.

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

Answer: Needles tend to have the highest biomass at these sites.