

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.

```
library(tidyverse)
getwd()
```

```
## [1] "/Users/ruanleyi/Documents/Year2/872 Environmental Data Analytics/Environmental_Data_Analytics_2018-08-08"

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: They are absorbed by plants and can be present in pollen and nectar, making them toxic to bees. There was uncertainty about the impact these insecticides were having on bees.

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: Plant debris, including woody debris and litter, is an essential but frequently overlooked component of carbon (C) storage in forest ecosystems.

- 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: * The litterfall and fine woody debris sampling data products provide mass data for plant functional groups from individual sampling bouts. * Spatial Sampling Design * Temporal Sampling Design

Obtain basic summaries of your data (Neonics)

- What are the dimensions of the dataset?

```
dim(Neonics)
```

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

- 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
```

Answer: Population is the most common effect that are studied

- 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: Honey Bee, Parasitic Wasp, Buff Tailed Bumblebee, Carniolan Honey Bee, Bumble Bee, Italian Honeybee. They are all in the bees category. They are of interest over other insects because bees are greatly impacted by the intesticides.

8. Concentrations are always a numeric value. What is the class of Conc.1..Author. in the dataset, and why is it not numeric?

```
summary(as.factor(Neonics$Conc.1..Author.))
```

```
##      0.37/      10/      NR/      NR      1      1023      0.40/      2/
##      208      127      108      94      82      80      69      63
##      10      0.053/      100      50/      0.5/      0.03      0.05/      0.45
##      62      59      56      51      45      44      43      43
##      0.1/      0.45/      1.0/      2.27/      50      0.125      500/      0.5
##      42      40      40      40      36      33      33      32
##      0.048/      0.15/      1/      48      25.0/      12/      0.027      2.4
##      30      30      30      30      28      27      26      26
##      0.2/      0.56/      100/      3      0.01/      1000/      3/      0.336
##      25      24      23      23      22      22      22      21
##      1.5/      0.05      1.5      2.60/      20.0/      6      6.80/      62.5/
##      21      20      20      20      20      20      20      20
##      0.005      0.4/      0.18/      0.3/      1000      40      0.00355/      0.1
##      18      18      17      17      17      17      16      16
##      0.4      150/      300      80/      0.053      0.24      0.28      125/
##      16      16      16      16      15      15      15      15
##      9      0.0001      0.0004/      0.084/      0.15      0.6      12.5/      144.0/
```

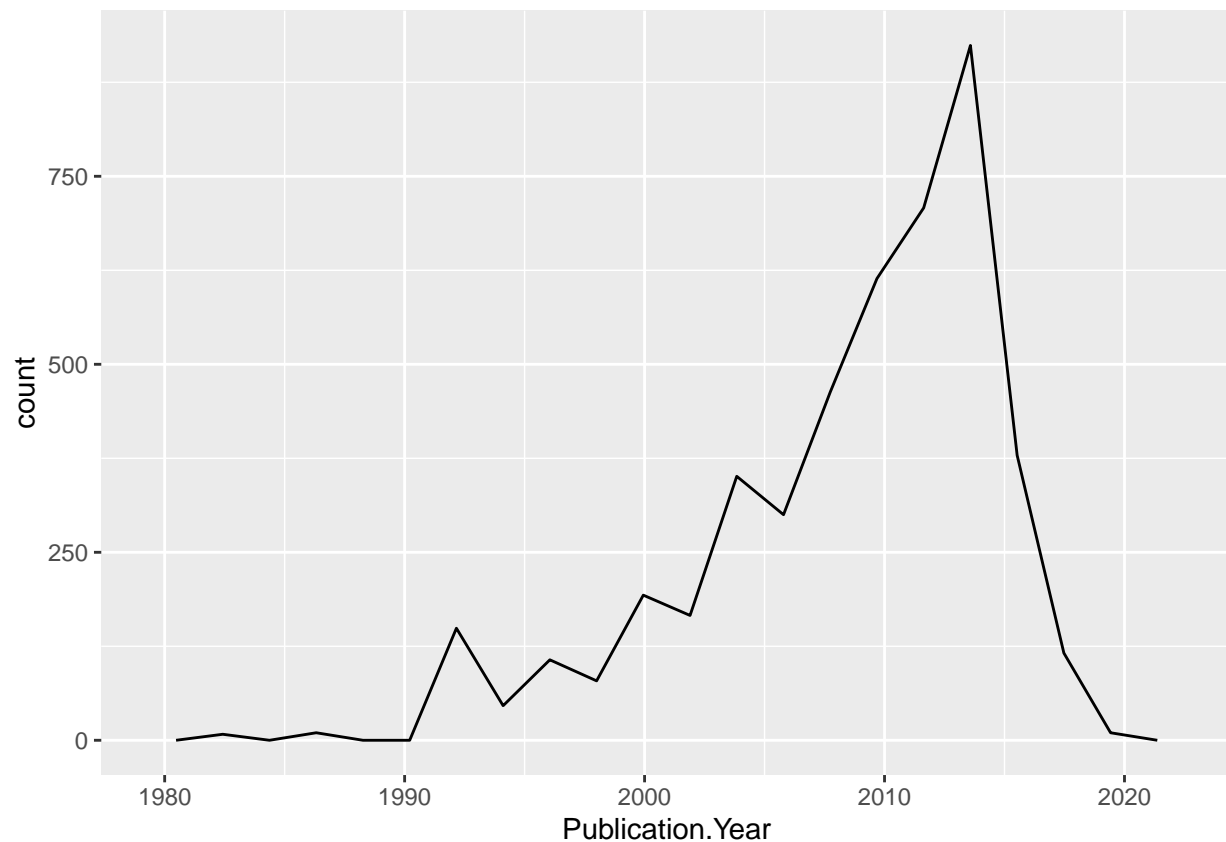
```
##      15      14      14      14      14      14      14      14
##    350/    40.0/    48/    56    84/    0.17/    125    14
##      14      14      14      14      14      13      13      13
##      16      17    0.047/    0.25/    0.28/    1.28/    1.81/    112
##      13      13      12      12      12      12      12      12
##     150     2.5/     25    60/    75/    0.02/    0.025/    0.29
##      12      12      12      12      12      11      11      11
##    37.5/     4/       5 (Other)
##      11      11      11    1817
```

Answer: Conc.1...Author is character variable. It is not numeric because concentrains are based on various levels.

Explore your data graphically (Neonics)

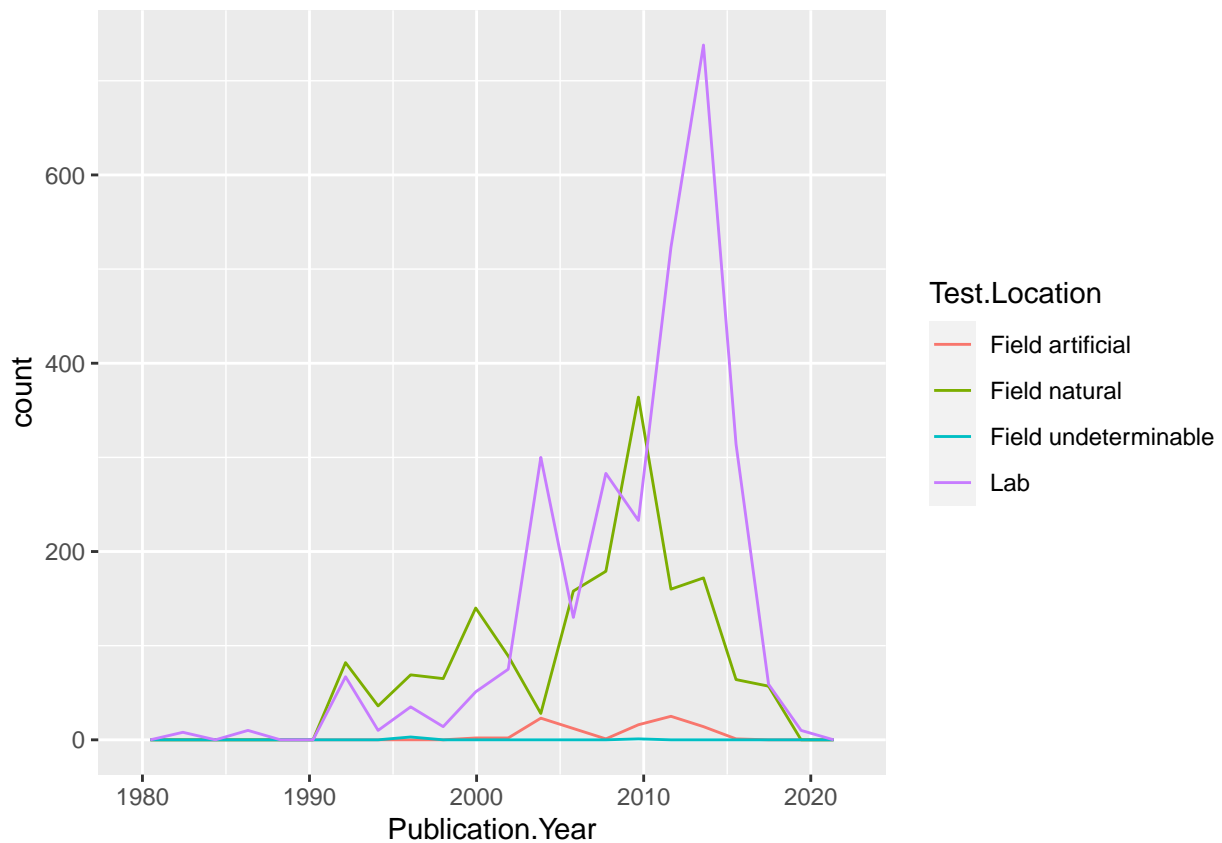
9. Using `geom_freqpoly`, generate a plot of the number of studies conducted by publication year.

```
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)
```

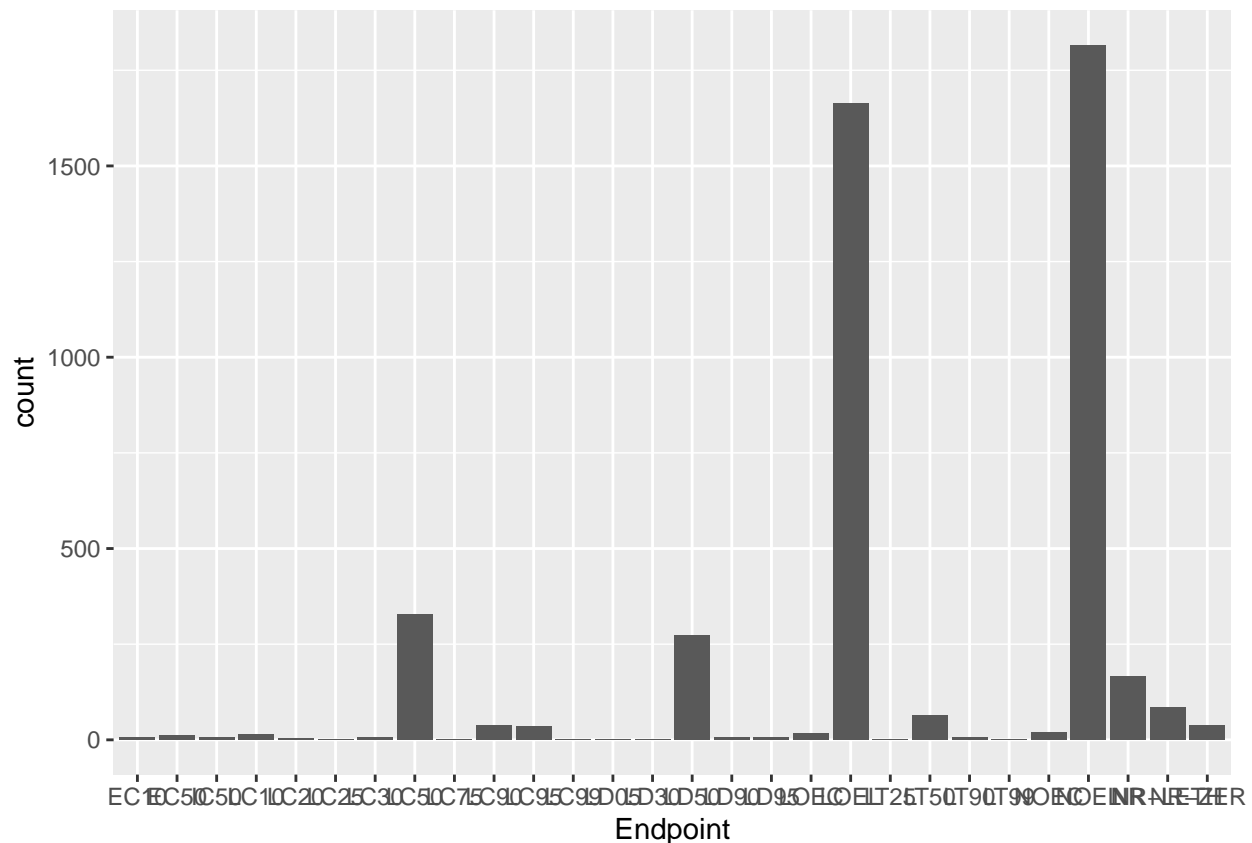


Interpret this graph. What are the most common test locations, and do they differ over time?

Answer: Lab is the most common test locations. They do differ over time, difference between number of tests conducted in Lab and other locations are getting larger.

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()
```



```
summary(as.factor(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 NR-LETH NR-ZERO
##       65         7         2        19     1816     167        86        37
```

Answer: LOEL and NOEL are two most common points. They are defined lowest-observable-effect-level, and no-observed-effects-residue.

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)
```

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

```
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"
```

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"
```

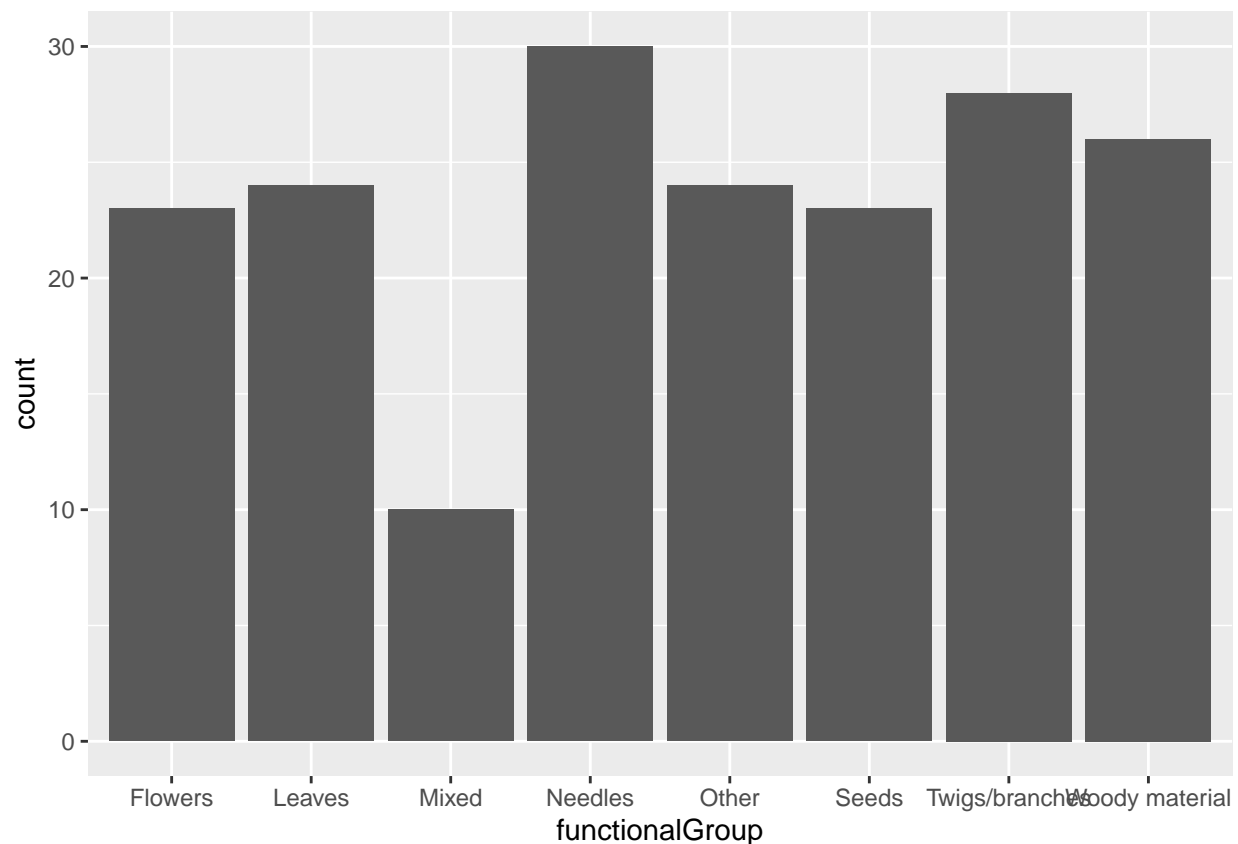
```
summary(Litter$plotID)
```

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

Answer: `unique` function gives the values that `plotID` can be; `summary` function returns the class, mode, and length of `plotID`, which is not what we need.

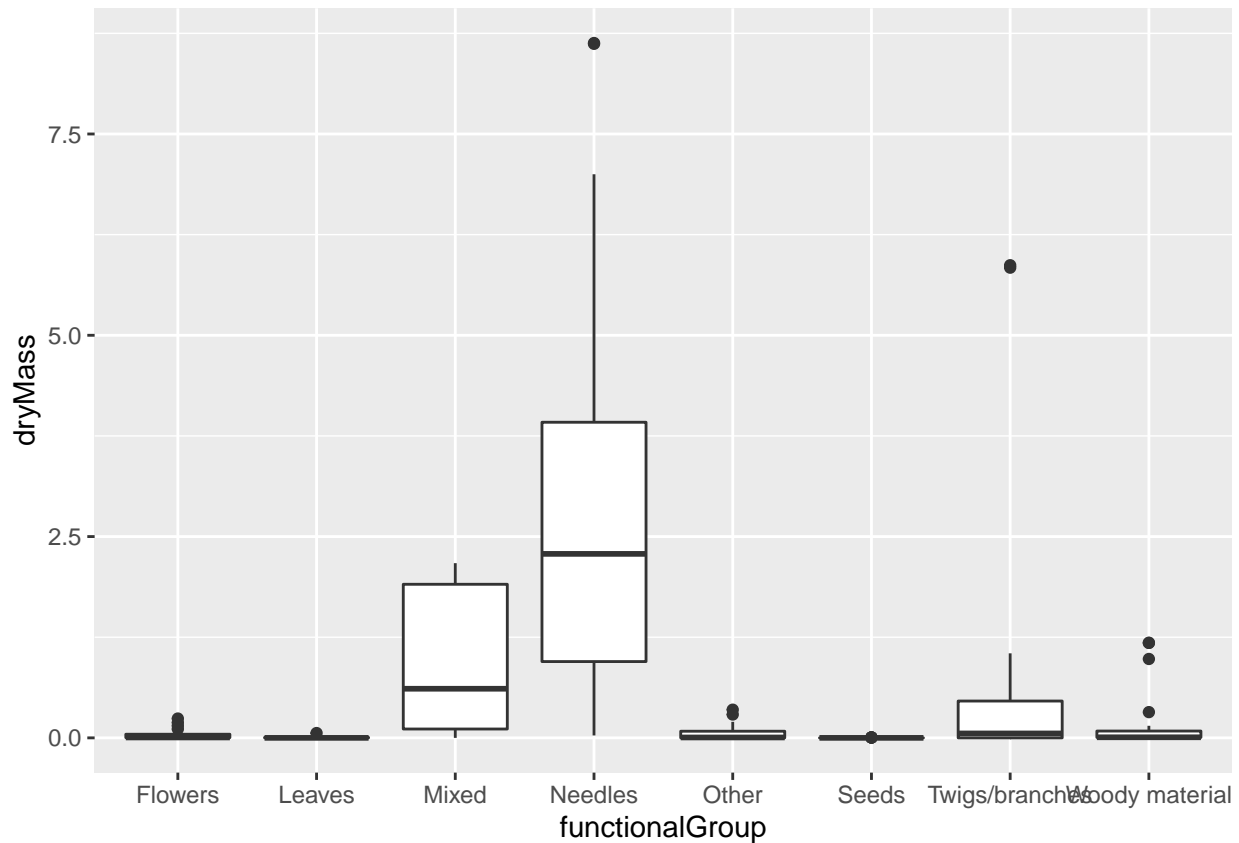
14. Create a bar graph of `functionalGroup` counts. This shows you what type of litter is collected at the Niwot Ridge sites. Notice that litter types are fairly equally distributed across the Niwot Ridge sites.

```
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))
```

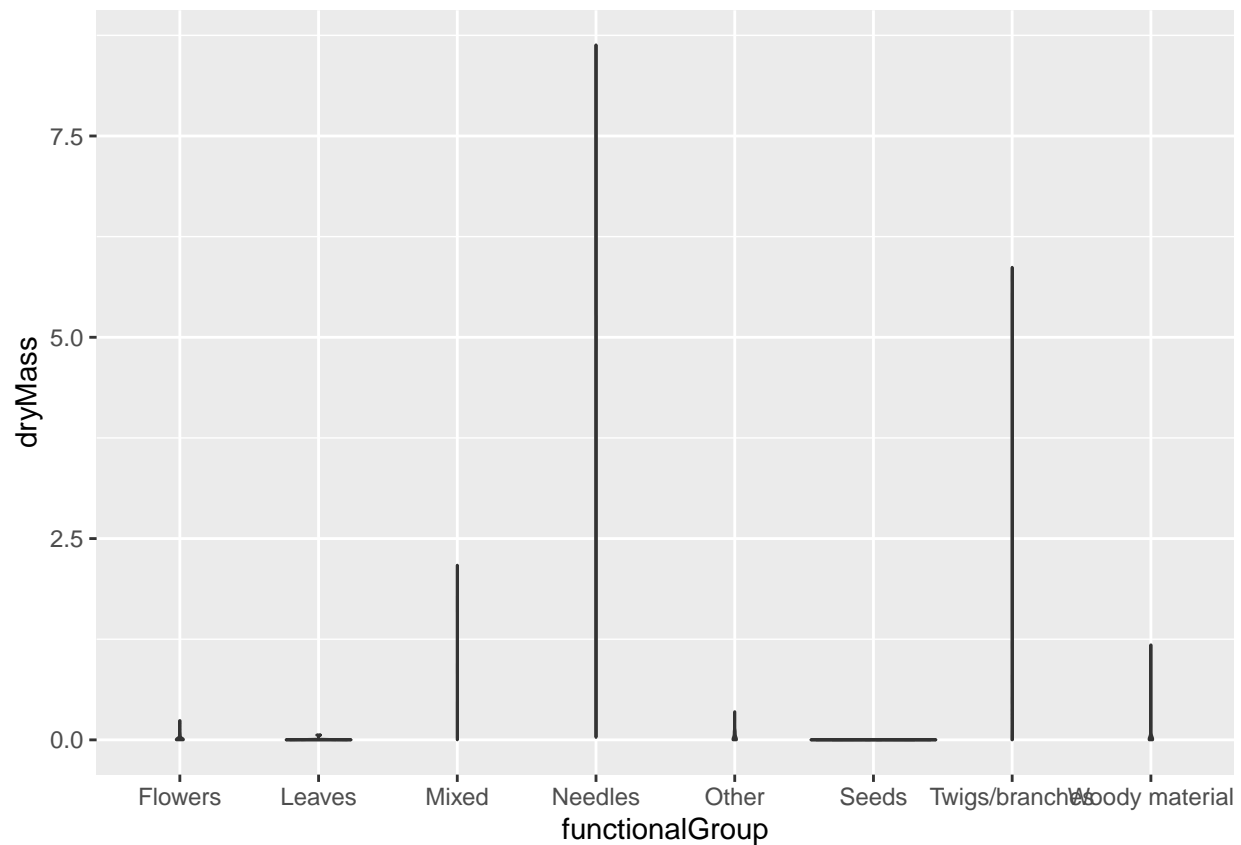



```
ggplot(Litter) +
  geom_violin(aes(x = functionalGroup, y = dryMass),
    draw_quantiles = c(0.25, 0.5, 0.75))
```

```
## 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: because violin chart cannot reflect how the drymass data is distributed on each functional group.

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

Answer: Needles