

Assignment 3: Data Exploration

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OVERVIEW

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

Directions

1. Rename this file `<FirstLast>_A03_DataExploration.Rmd` (replacing `<FirstLast>` with your first and last name).
2. Change “Student Name” on line 3 (above) with your name.
3. Work through the steps, **creating code and output** that fulfill each instruction.
4. Be sure to **answer the questions** in this assignment document.
5. When you have completed the assignment, **Knit** the text and code into a single PDF file.
6. After Knitting, submit the completed exercise (PDF file) to the dropbox in Sakai.

The completed exercise is due on Sept 30th.

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 include the subcommand to read strings in as factors.

```
getwd()
```

```
## [1] "C:/Users/Caroline/Desktop/EDA-Fall2022"
```

```
setwd("C:/Users/Caroline/Desktop/EDA-Fall2022")
library("tidyverse")
```

```
## -- Attaching packages ----- tidyverse 1.3.2 --
## v ggplot2 3.3.6      v purrr  0.3.4
## v tibble  3.1.8      v dplyr  1.0.10
## v tidyr   1.2.1      v stringr 1.4.1
## v readr   2.1.2      v forcats 0.5.2
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()    masks stats::lag()
```

```
library("lubridate")

##
## Attaching package: 'lubridate'
##
## The following objects are masked from 'package:base':
##
##     date, intersect, setdiff, union

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)

# This chunk of code loads necessary packages, downloads and renames the
# desired csv files, and commands R to read the strings in as factors. I also
# had to set my working directory.
```

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: Ecotoxicology of neonicotinoids on insects is important for environmentalist and conservationist because the neonicotinoids can have a negative impact on insects life cycles. It can result in death, alter behavior and/or reduce reproduction. This can result in population decline. Additionally, insects exposed to neonicotinoids can impact the food chain and the health of the animals that consume insects (and the animals that consume those animals and so on).

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: Woody debris and litter are important parts of forest ecosystems because they play a role in carbon budgets and nutrient cycling in forest ecosystems. Both can impact the fuel load and how fires might function inside of an ecosystem. They also impact the underlying soil and can alter nitrogen availability, which can influence flora composition.

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: 1. Sampling is executed at terrestrial NEON sites that contain woody vegetation >2m tall 2. Sampling occurs only in tower plots, which are selected randomly within the 90% flux footprint of the primary and secondary airsheds 3. In forested sites, the sample plots are 20 40m x 40m plots. In sites with low-statured vegetation, sampling is targeted to 4 40m x 40m plots and 2 20m x 20m plots.

Obtain basic summaries of your data (Neonics)

5. What are the dimensions of the dataset?

```
dim(Neonics)
```

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

```
# this code asks for the dimensions of the Neonics dataset. There are 4623  
# observations and 30 variables.
```

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
```

```
# this code shows the summary of the effects column in the Neonics dataset.
```

Answer: Population is the most common effect studied. This is probably because it gives a sense of the overall health of the population in terms of abundance and population trends. The second most common effect studied is Mortality, which is related to population. Additionally, these two effect categories are probably relatively “easy” to study.

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

```

this chunk of code shows the summary of species in the Neonics dataset.

Answer: The six most common species are the Honey Bee, the Parasitic Wasp, the Buff Tailed Bumblebee, the Carniolan Honey Bee, the Bumble Bee, and the Italian Honeybee. Many of these species are closely related (bees), and the Parasitic Wasp is considered to be a “beneficial” insect similar to many bee species. They are likely of interest to this dataset because of the important role they play in the ecosystem as pollinators.

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"
```

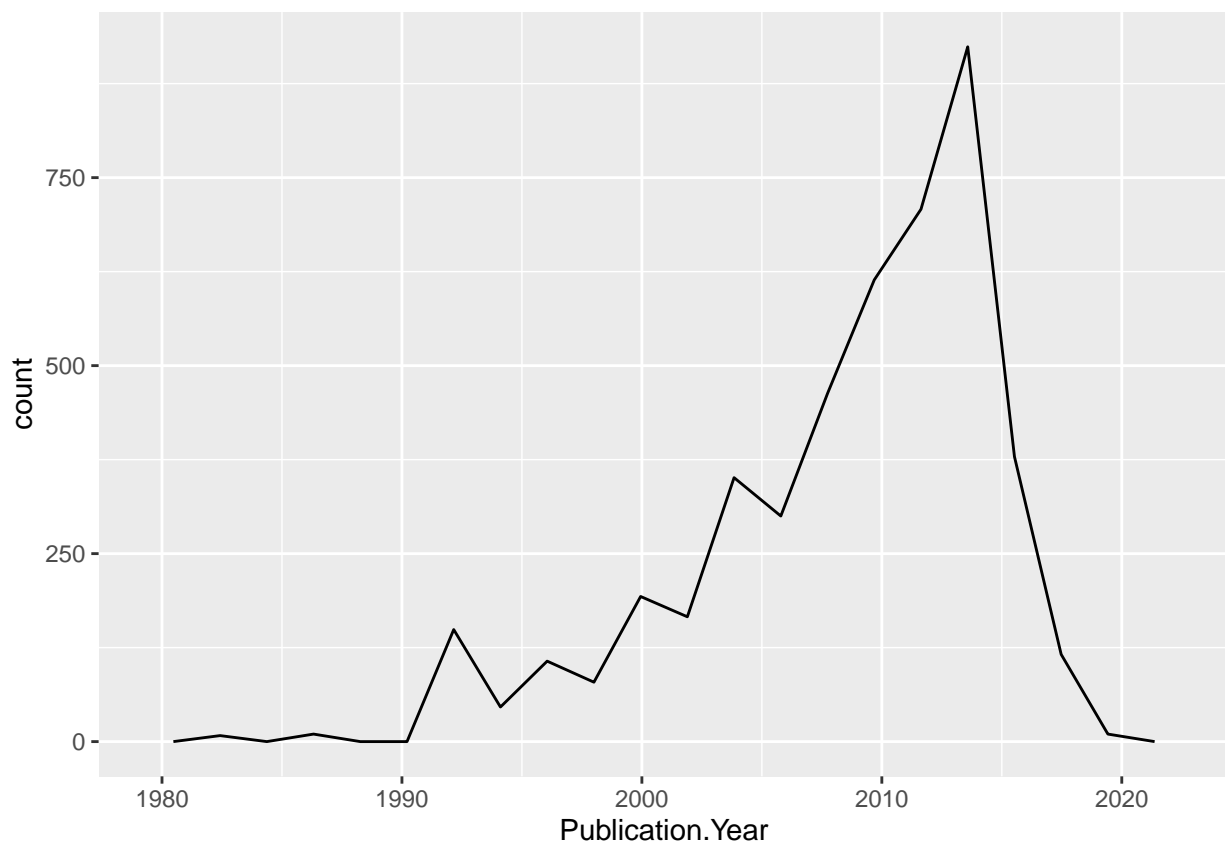
```
# this code is asking for the class associated with the Conc.1..Author column
# in the Neonics dataset.
```

Answer: The class is a factor, likely because the column has approximations, less than and greater than symbols in the column's values. The computer reads these symbols and classifies the column as a factor because it has both numeric and non-numeric values.

Explore your data graphically (Neonics)

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

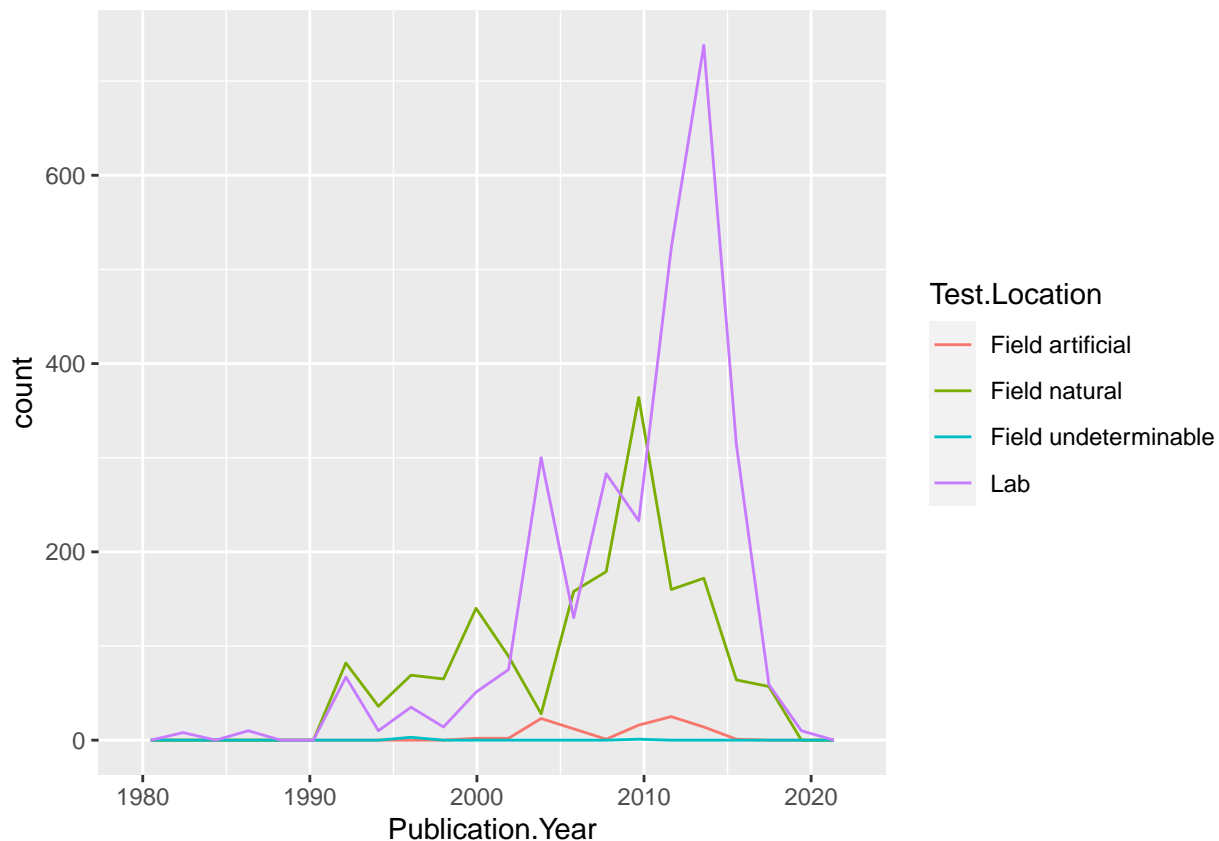
```
ggplot(Neonics) + geom_freqpoly(aes(x = Publication.Year), bins = 20)
```



```
# This code asks ggplot to chose the Neonics dataset, and then create a
# frequency polygon of the publication year. It also sets the bins to 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)
```



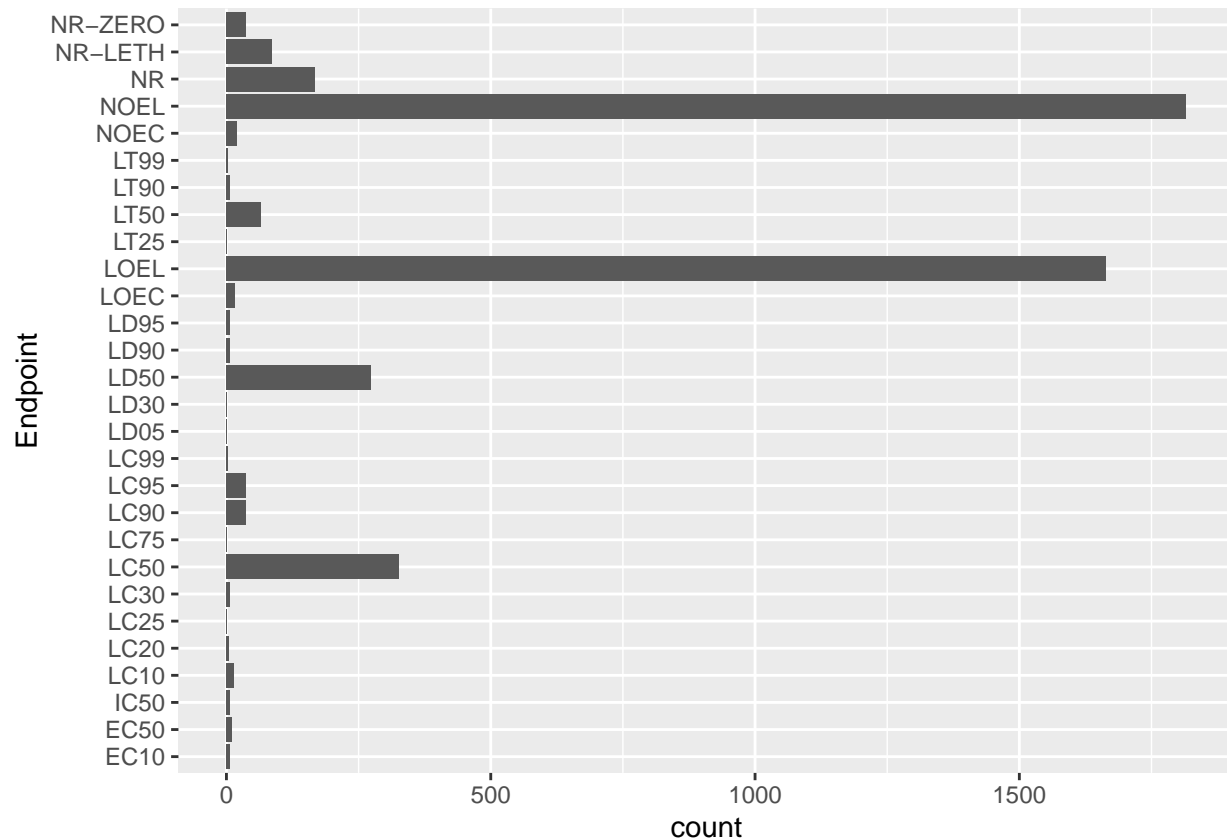
```
# This code asks ggplot to chose the Neonics dataset, create a frequency
# polygon of the publication years, and create different plots for each Test
# Location. It also sets the bins to 20.
```

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

Answer: The most common test location appears to be lab, with a significant number of tests being run in 2014. Field natural appears to be the 2nd most common test location. Field Artificial is far less common, and Field undeterminable is the least common.

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(x = Endpoint)) + coord_flip()
```



*# This code asked ggplot to create a bar graph from the Neonics dataset of the
endpoints I flipped the axis so that the end point categories can be read*

Answer: The two most common end points are NOEL (No observable effect level: highest dose (concentration)- the test produced effects that were not significantly different from responses of controls) and LOEL (Lowest observable effect level: lowest dose(concentration)- the test produced effects that were significantly different from responses of controls)

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] "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"
```

```
# This code checks the class of the collectDate column in the Litter dataset.  
# It then informs R that this column is a date, and the format of the date  
# inside that column. It then asks R what the unique values are in the  
# collectDate column.
```

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
```

```
# This code asks for the unique values in the plotID column of the Litter  
# dataset
```

Answer: There are 12 plots sampled in this data set. The information obtained from `unique` is different than `summary` because the `unique` command will tell you how many different values are present (in this case, 12), whereas `summary` would have told how many samples were taken at each plot.

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.

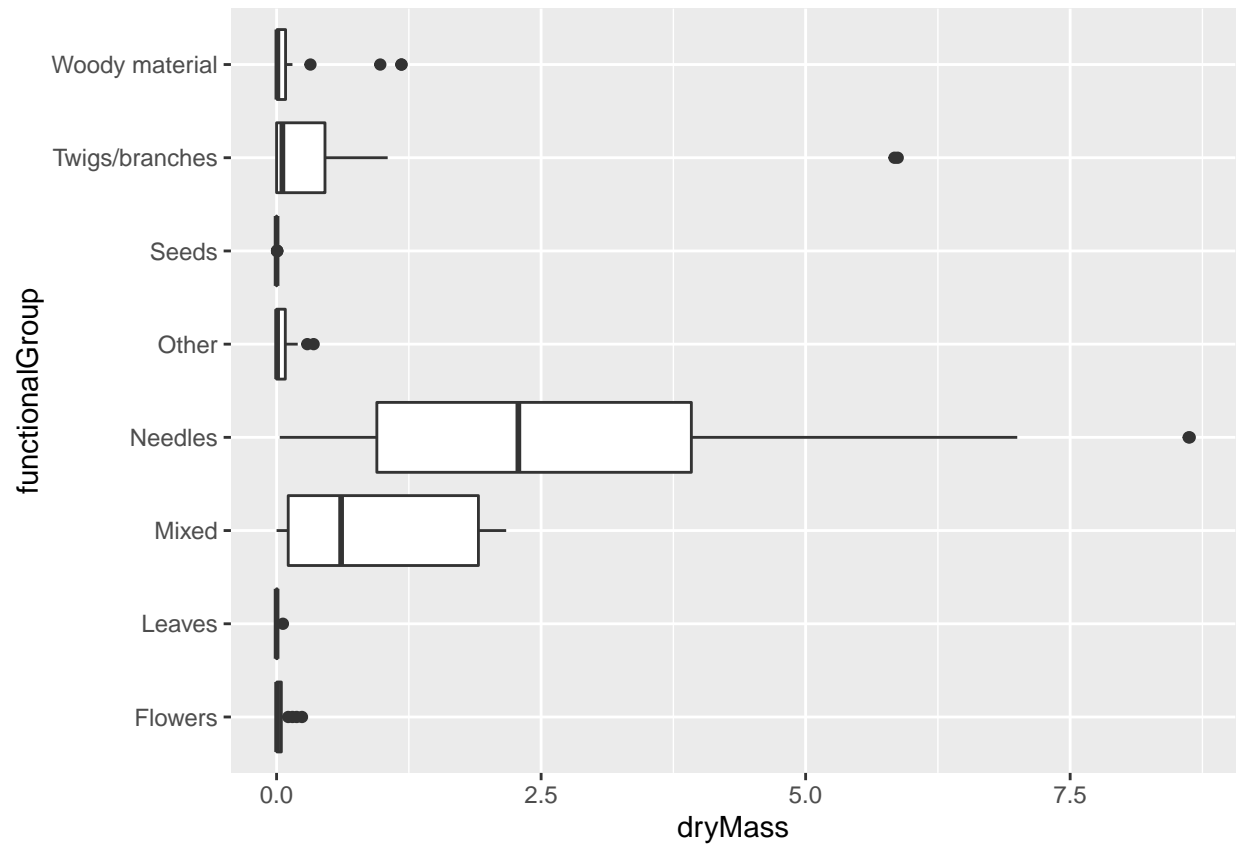
```
ggplot(Litter) + geom_bar(aes(x = functionalGroup))
```



```
# this code asks ggplot to make a bar plot of the functionalGroup column from  
# the Litter dataset.
```

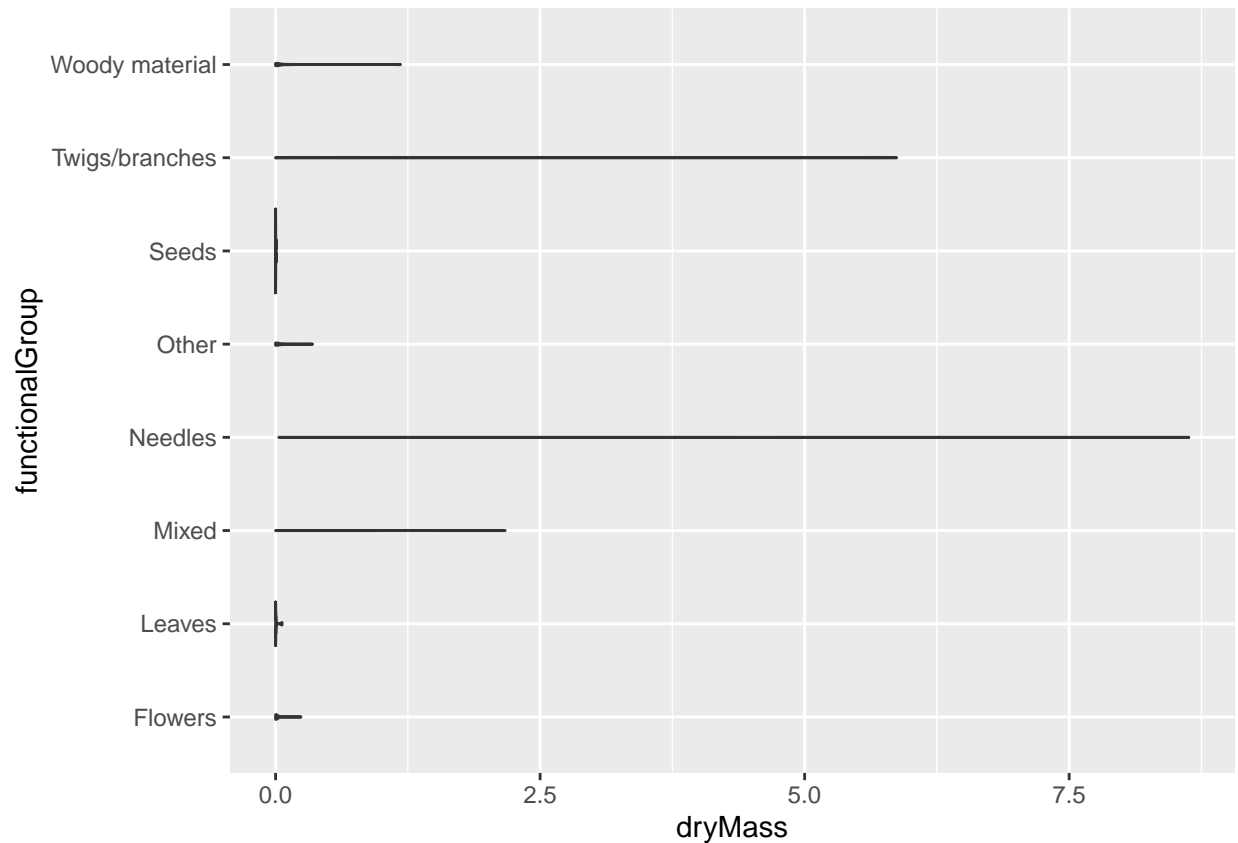
15. Using `geom_boxplot` and `geom_violin`, create a boxplot and a violin plot of `dryMass` by `functionalGroup`.

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



*# this code asks ggplot to make a boxplots of the dryMass column and to sort
the boxplots by functionalGroup*

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



```
# this code asks ggplot to make a violin plot of the dryMass column and to sort
# the violin plots by functionalGroup
```

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

Answer: Violin plots are used to show the distributions of numeric data using density curves. The width corresponds to the approximate frequency of that points in each region. In this case, the violin plots are extremely thin; our data is not robust enough to create meaningful density curves. If we had a larger data set, we might be able to extrapolate more from violin plots. Box Plots show the minimum and maximum values (the length of the line), outliers (as dots), the median (as a line inside the box) and the IQR (the length of the box). For our data set, this is a much more helpful visualization.

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