5: Part 1 - Data Visualization Basics

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Objectives

- 1. Perform simple data visualizations in the R package ggplot
- 2. Develop skills to adjust aesthetics and layers in graphs
- 3. Apply a decision tree framework for appropriate graphing methods

Opening discussion

Effective data visualization depends on purposeful choices about graph types. The ideal graph type depends on the type of data and the message the visualizer desires to communicate. The best visualizations are clear and simple. A good resource for data visualization is Data to Viz, which includes both a decision tree for visualization types and explanation pages for each type of data, including links to R resources to create them. Take a few minutes to explore this website.

Set Up

```
library(tidyverse); library(lubridate); library(here)
library(ggridges)
here()
```

[1] "/home/guest/EDE_Fall2024"

```
PeterPaul.chem.nutrients <-
    read.csv(here("Data/Processed_KEY/NTL-LTER_Lake_Chemistry_Nutrients_PeterPaul_Processed.csv"), string
PeterPaul.chem.nutrients.gathered <-
    read.csv(here("Data/Processed_KEY/NTL-LTER_Lake_Nutrients_PeterPaulGathered_Processed.csv"), stringsA
EPAair <- read.csv(here("Data/Processed_KEY/EPAair_03_PM25_NC1819_Processed.csv"), stringsAsFactors = T
EPAair$Date <- ymd(EPAair$Date)
PeterPaul.chem.nutrients$sampledate <- ymd(PeterPaul.chem.nutrients$sampledate)
PeterPaul.chem.nutrients.gathered$sampledate <- ymd(PeterPaul.chem.nutrients.gathered$sampledate)</pre>
```

ggplot

ggplot, called from the package ggplot2, is a graphing and image generation tool in R. This package is part of tidyverse. While base R has graphing capabilities, ggplot has the capacity for a wider range and more sophisticated options for graphing. ggplot has only a few rules:

- The first line of ggplot code always starts with ggplot()
- A data frame must be specified within the ggplot() function. Additional datasets can be specified in subsequent layers.
- Aesthetics must be specified, most commonly x and y variables but including others. Aesthetics can be specified in the ggplot() function or in subsequent layers.
- Additional layers must be specified to fill the plot.

Geoms

Here are some commonly used layers for plotting in ggplot:

```
geom_bar
geom_histogram
geom_freqpoly
geom_boxplot
geom_violin
geom_dotplot
geom_density_ridges
geom_point
geom_errorbar
geom_smooth
geom_line
geom_area
geom_abline (plus geom_hline and geom_vline)
```

Aesthetics

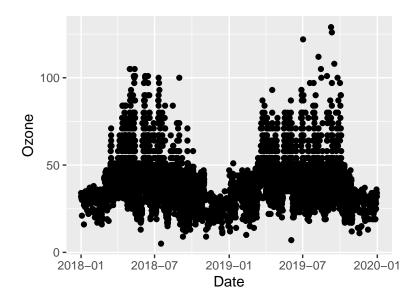
• geom_text

Here are some commonly used aesthetic types that can be manipulated in ggplot:

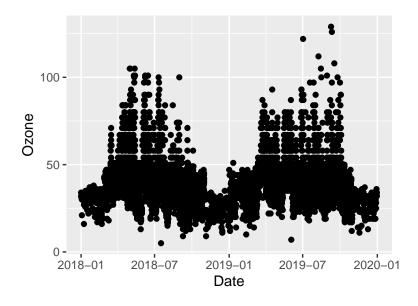
- color
- fill
- shape
- size
- transparency

Plotting continuous variables over time: Scatterplot and Line Plot

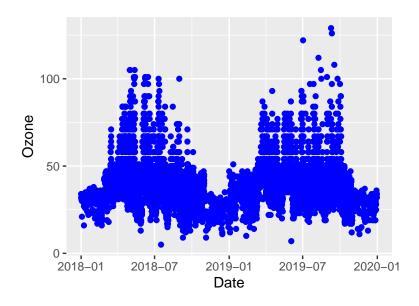
```
# Scatterplot
ggplot(EPAair, aes(x = Date, y = Ozone)) +
geom_point()
```



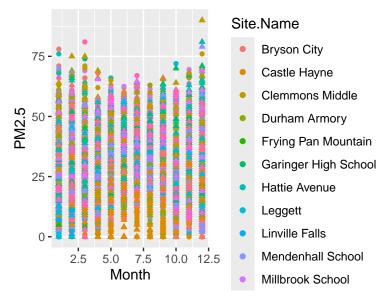
```
03plot <- ggplot(EPAair) +
  geom_point(aes(x = Date, y = Ozone))
print(03plot)</pre>
```



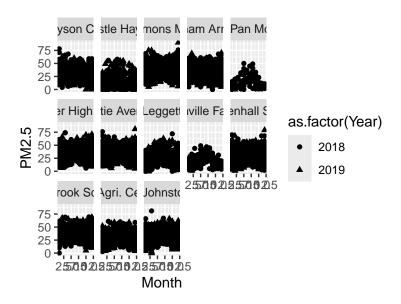
```
# Fix this code
03plot2 <- ggplot(EPAair) +
    #geom_point(aes(x = Date, y = Ozone, color = "blue"))
    geom_point(aes(x = Date, y = Ozone), color = "blue")
print(03plot2)</pre>
```

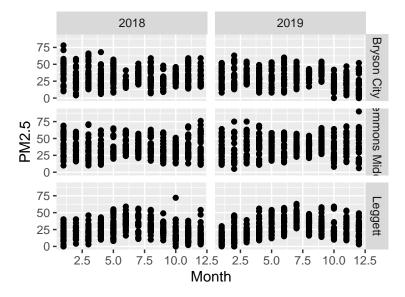


```
# Add additional variables
# How could you automatically assign a marker color to a variable?
PMplot <-
    ggplot(EPAair, aes(x = Month, y = PM2.5, shape = as.factor(Year), color = Site.Name)) +
    geom_point()
print(PMplot)</pre>
```

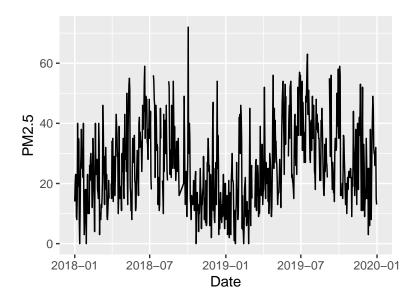


```
# Separate plot with facets
PMplot.faceted <-
    ggplot(EPAair, aes(x = Month, y = PM2.5, shape = as.factor(Year))) +
   geom_point() +
   facet_wrap(vars(Site.Name), nrow = 3)
print(PMplot.faceted)</pre>
```



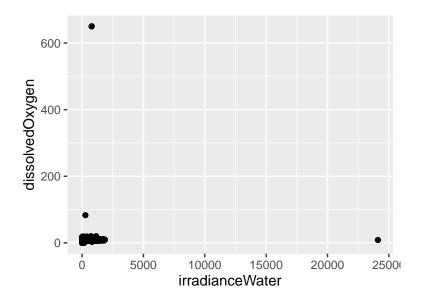


```
# Plot true time series with geom_line
PMplot.line <-
    ggplot(subset(EPAair, Site.Name == "Leggett"),
        aes(x = Date, y = PM2.5)) +
    geom_line()
print(PMplot.line)</pre>
```



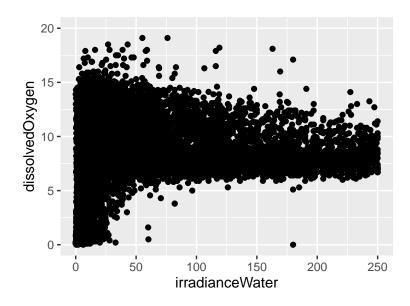
Plotting the relationship between two continuous variables: Scatterplot

```
# Scatterplot
lightvsD0 <-
ggplot(PeterPaul.chem.nutrients, aes(x = irradianceWater, y = dissolvedOxygen)) +
geom_point()
print(lightvsD0)</pre>
```

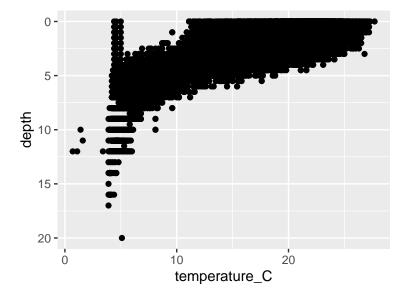


```
# Adjust axes
lightvsDOfixed <-
ggplot(PeterPaul.chem.nutrients, aes(x = irradianceWater, y = dissolvedOxygen)) +
geom_point() +
xlim(0, 250) +</pre>
```

```
ylim(0, 20)
print(lightvsDOfixed)
```

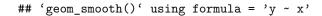


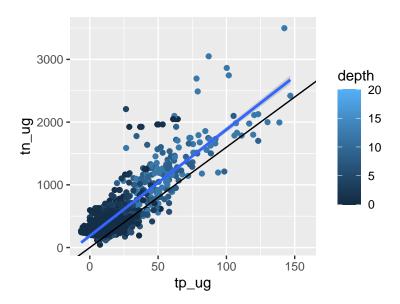
```
# Depth in the fields of limnology and oceanography is on a reverse scale
tempvsdepth <-
    ggplot(PeterPaul.chem.nutrients, aes(x = temperature_C, y = depth)) +
    #ggplot(PeterPaul.chem.nutrients, aes(x = temperature_C, y = depth, color = daynum)) +
    geom_point() +
    scale_y_reverse()
print(tempvsdepth)</pre>
```



```
NvsP <-
ggplot(PeterPaul.chem.nutrients, aes(x = tp_ug, y = tn_ug, color = depth)) +
geom_point() +
geom_smooth(method = lm) +</pre>
```

```
geom_abline(aes(slope = 16, intercept = 0))
print(NvsP)
```



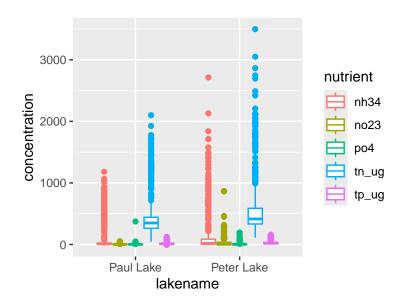


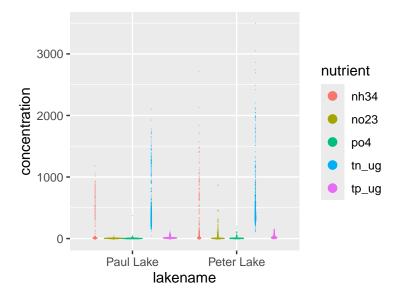
Plotting continuous vs. categorical variables

A traditional way to display summary statistics of continuous variables is a bar plot with error bars. Let's explore why this might not be the most effective way to display this type of data. Navigate to the Caveats page on Data to Viz (https://www.data-to-viz.com/caveats.html) and find the page that explores barplots and error bars.

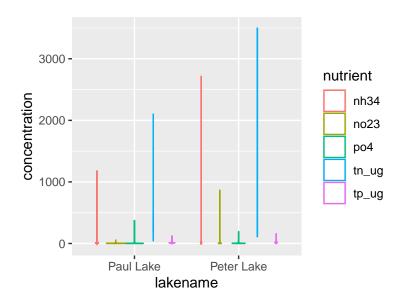
What might be more effective ways to display the information? Navigate to the boxplots page in the Caveats section to explore further.

```
# Box and whiskers plot
Nutrientplot3 <-
    ggplot(PeterPaul.chem.nutrients.gathered, aes(x = lakename, y = concentration)) +
    geom_boxplot(aes(color = nutrient)) # Why didn't we use "fill"? keeping contour black but changing fi
print(Nutrientplot3)</pre>
```



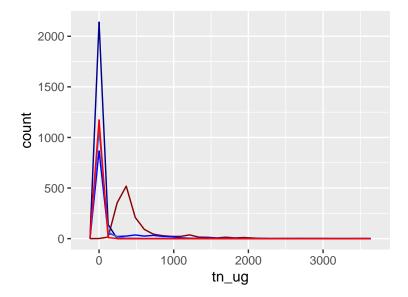


```
# Violin plot
Nutrientplot5 <-
    ggplot(PeterPaul.chem.nutrients.gathered, aes(x = lakename, y = concentration)) +
    geom_violin(aes(color = nutrient)) #
print(Nutrientplot5)</pre>
```



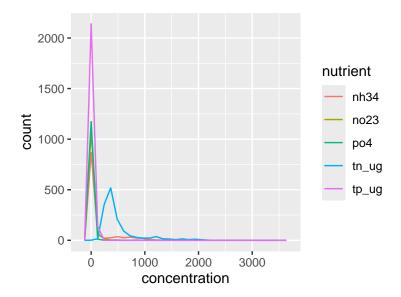
```
# Frequency polygons
# Using a tidy dataset
Nutrientplot6 <-
ggplot(PeterPaul.chem.nutrients) +
geom_freqpoly(aes(x = tn_ug), color = "darkred") +
geom_freqpoly(aes(x = tp_ug), color = "darkblue") +
geom_freqpoly(aes(x = nh34), color = "blue") +
geom_freqpoly(aes(x = no23), color = "royalblue") +
geom_freqpoly(aes(x = po4), color = "red")
print(Nutrientplot6)</pre>
```

```
## 'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.
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## 'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.
## 'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.
```



```
# Using a gathered dataset
Nutrientplot7 <-
    ggplot(PeterPaul.chem.nutrients.gathered) +
    geom_freqpoly(aes(x = concentration, color = nutrient))
print(Nutrientplot7)</pre>
```

'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.



```
# Frequency polygons have the risk of becoming spaghetti plots.
# See <https://www.data-to-viz.com/caveat/spaghetti.html> for more info.

# Ridgeline plot
Nutrientplot6 <-
ggplot(PeterPaul.chem.nutrients.gathered, aes(y = nutrient, x = concentration)) +
geom_density_ridges(aes(fill = lakename), alpha = 0.5) #
print(Nutrientplot6)</pre>
```

Picking joint bandwidth of 10.9

