10: Data Visualization

Environmental Data Analytics | Sarah Ko notes Spring 2019

LESSON OBJECTIVES

1. Perform advanced edits on ggplot objects to follow best practices for data visualization

SET UP YOUR DATA ANALYSIS SESSION

```
getwd()
## [1] "C:/Users/Sarah/Documents/Duke/Year 2/Spring 2019/Data Analytics/Environmental_Data_Analytics"
library(tidyverse)

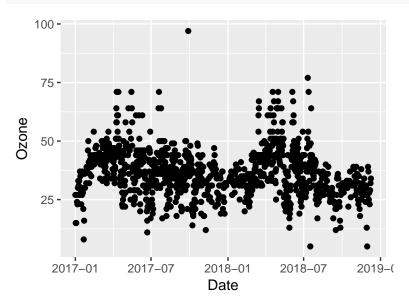
PeterPaul.chem.nutrients <- read.csv("./Data/Processed/NTL-LTER_Lake_Chemistry_Nutrients_PeterPaul_Proc
PeterPaul.nutrients.gathered <- read.csv("./Data/Processed/NTL-LTER_Lake_Nutrients_PeterPaulGathered_Pr
EPAair <- read.csv("./Data/Processed/EPAair_O3PM25_3sites1718_processed.csv")

EPAair$Date <- as.Date(EPAair$Date, format = "%Y-%m-%d")
PeterPaul.chem.nutrients$sampledate <- as.Date(PeterPaul.chem.nutrients$sampledate, format = "%Y-%m-%d")</pre>
```

Themes

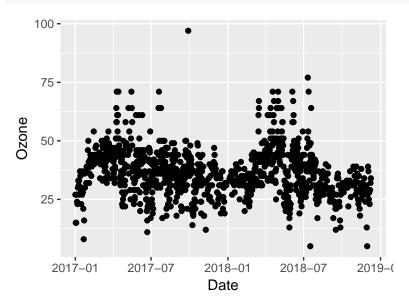
Often, we will want to change multiple visual aspects of a plot. Ggplot comes with pre-built themes that will adjust components of plots if you call that theme.

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

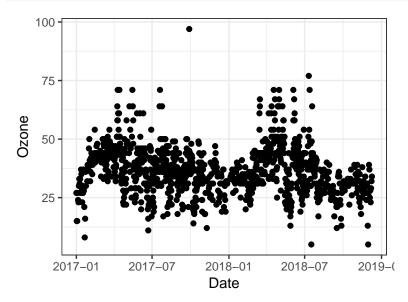


```
O3plot1 <- ggplot(EPAair) +
geom_point(aes(x = Date, y = Ozone)) +</pre>
```

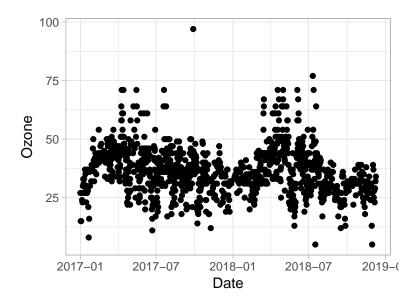
```
theme_gray()
print(03plot1)
```



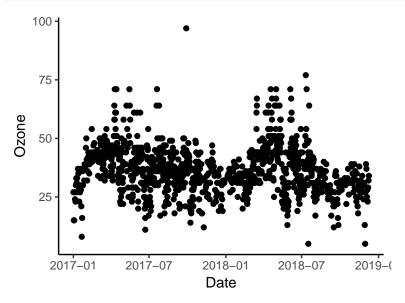
```
03plot2 <- ggplot(EPAair) +
  geom_point(aes(x = Date, y = Ozone)) +
  theme_bw()
print(03plot2)</pre>
```



```
O3plot3 <- ggplot(EPAair) +
  geom_point(aes(x = Date, y = Ozone)) +
  theme_light()
print(O3plot3)</pre>
```



```
03plot4 <- ggplot(EPAair) +
  geom_point(aes(x = Date, y = Ozone)) +
  theme_classic()
print(03plot4)</pre>
```



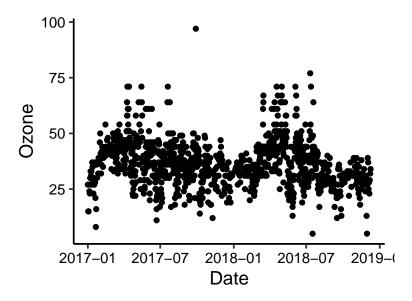
Notice that some aspects of your graph have not been adjusted, including:

- text size
- axis label colors
- legend position and justification

If you would like to set a common theme across all plots in your analysis session, you may define a theme and call up that theme for each graph. This eliminates the need to add multiple lines of code in each plot.

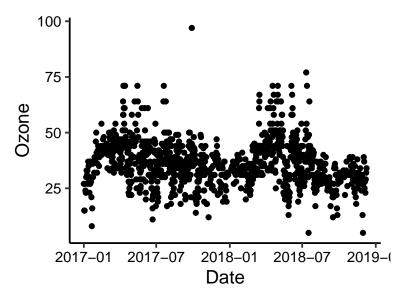
```
# options: call the theme in each plot or set the theme at the start.

03plot5 <- ggplot(EPAair) +
  geom_point(aes(x = Date, y = Ozone)) +
  mytheme
print(03plot5)</pre>
```



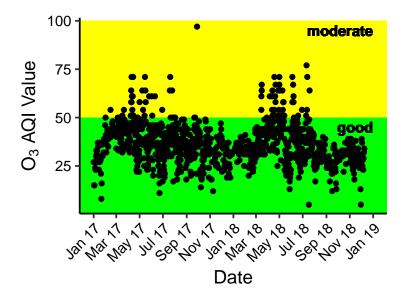
```
theme_set(mytheme)

03plot6 <- ggplot(EPAair) +
  geom_point(aes(x = Date, y = Ozone))
print(03plot6)</pre>
```



Adjusting multiple components of your plots

While the theme allows us to set multiple aspects of plots, ggplot allows us to adjust other parts of plots outside of the theme.



Color palettes

There are several color palettes that are designed to be more effective than palettes in base R. These include Viridis (https://cran.r-project.org/web/packages/viridis/vignettes/intro-to-viridis.html) and Color Brewer (http://colorbrewer2.org/). A few rules for choosing colors:

- Consider if your plot needs to be viewed in black and white. If so, choose a sequential palette with varying color intensity.
- Choose a palette that is color-blind friendly
- Maximize contrast (e.g., no pale colors on a white background)
- Diverging color palettes should be used for diverging values (e.g., warm-to-cool works well for values on a scale encompassing negative and positive values)

Perception is key! Choose palettes that are visually pleasing and will communicate what you are hoping your audience to perceive. Hint: base R palettes are not ideal.

```
#install.packages("viridis")
#install.packages("RColorBrewer")
```

```
#install.packages("colormap")
library(viridis)

## Loading required package: viridisLite
library(RColorBrewer)
library(colormap)

scales::show_col(colormap(colormap = colormaps$viridis, nshades = 16))
```

#440154ff	#461868ff	#472d7bff	#404284ff
#39558bff	#31668dff	#2a768eff	#24888dff
#23978aff	#26a784ff	#37b578ff	#55c467ff
#79d051ff	#a3da37ff	#cee12cff	#fde725ff

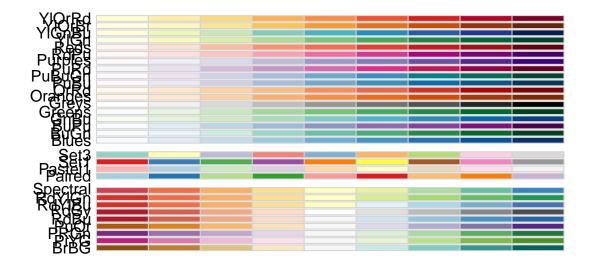
scales::show_col(colormap(colormap = colormaps\$inferno, nshades = 16))

	#100628ff	#210c49ff	#3f0e5eff
#5b116dff	#761b6bff	#902567ff	#ad315bff
#c43f4dff	#da513aff	#ea6827ff	#f6850fff
#f9a319ff	#f9c32eff	#fae063ff	#fcffa4ff

scales::show_col(colormap(colormap = colormaps\$magma, nshades = 16))

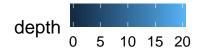
	#0f0926ff	#1e1046ff	#3b1165ff
#55147cff	#701e7fff	#8a2880ff	#a7317cff
#c13d75ff	#db4a69ff	#ec6163ff	#f88061ff
#fc9d6fff	#febc83ff	#fddc9fff	#fcfdbfff

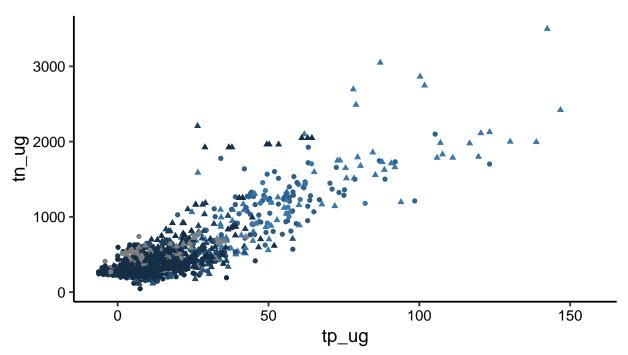
display.brewer.all(n = 9)



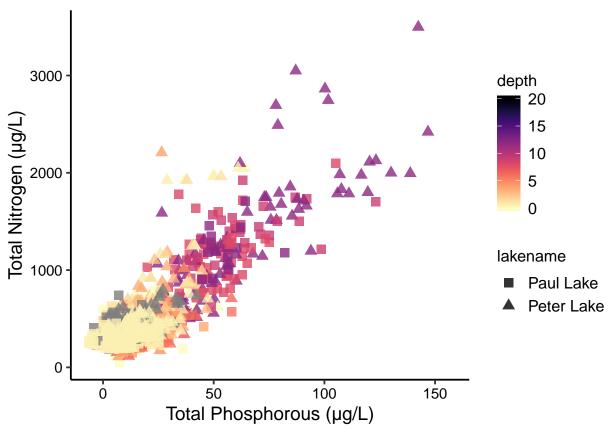
```
NvsP <-
    ggplot(PeterPaul.chem.nutrients, aes(x = tp_ug, y = tn_ug, color = depth, shape = lakename)) +
    geom_point()
print(NvsP)</pre>
```

lakename • Paul Lake ▲ Peter Lake





```
# let's first make the plot look better.
# change your axis labels to reflect TN and TP in micrograms per liter.
# change your legend labels
NvsP2 <-
 ggplot(PeterPaul.chem.nutrients, aes(x = tp_ug, y = tn_ug, color = depth, shape = lakename)) +
 geom_point(alpha = 0.8, size = 3) +
 xlab("Total Phosphorous (\U003BCg/L)") + #can also use paste function
 ylab("Total Nitrogen (\U003BCg/L)") +
  # change your legend labels here
 scale_shape_manual(values = c(15, 17)) +
  scale_color_distiller(palette = "Blues", direction = 1) + # use scale_color_brewer for discrete varia
  scale_color_viridis(option = "magma", direction = -1) +
  theme(legend.position = "right",
        legend.text = element_text(size = 12), legend.title = element_text(size = 12))
## Scale for 'colour' is already present. Adding another scale for
## 'colour', which will replace the existing scale.
print(NvsP2)
```



```
# can change the shape to something you can define both the fill and border of

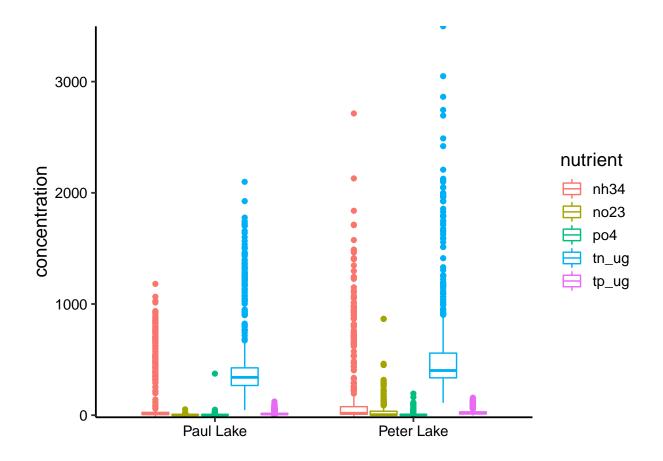
# change your y axis label to list concentration in micrograms per liter

# remove your x axis label

# change labels for nutrients in the legend

Nutrientplot <-
ggplot(PeterPaul.nutrients.gathered, aes(x = lakename, y = concentration, color = nutrient)) +
geom_boxplot() +

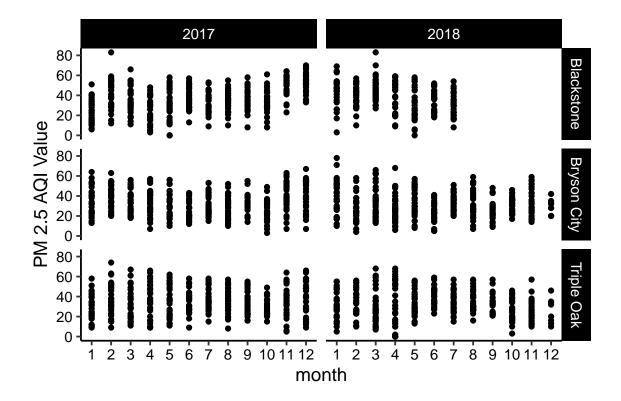
# place your additional edits here
xlab(NULL) +
scale_y_continuous(expand = c(0, 0)) +
#scale_color_brewer(palette = "YlGnBu") +
#scale_color_manual(values = c("#7fcdbb", "#41b6c4", "#1d91c0", "#225ea8", "#0c2c84")) +
#scale_color_viridis(discrete = TRUE) +
theme(legend.position = "right")
print(Nutrientplot)</pre>
```



Adjusting facets

```
PMplot.faceted <-
    ggplot(EPAair, aes(x = month, y = PM2.5)) +
    geom_point() +
    facet_grid(Site.Name ~ year) +
    scale_x_continuous(breaks = c(1:12)) +
    theme(strip.background = element_rect(fill = "black"), strip.text = element_text(color = "white")) +
    ylab(expression("PM 2.5 AQI Value"))
print(PMplot.faceted)</pre>
```

Warning: Removed 52 rows containing missing values (geom_point).



Multiple plots on a page

In situations where facets don't fill our needs to place multiple plots on a page, we can use the package <code>gridExtra</code> to arrange plots. The <code>grid.arrange</code> function is extremely flexible in its ability to arrange plots in specific configurations. A useful guide can be found here: https://cran.r-project.org/web/packages/egg/vignettes/Ecosystem.html.

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

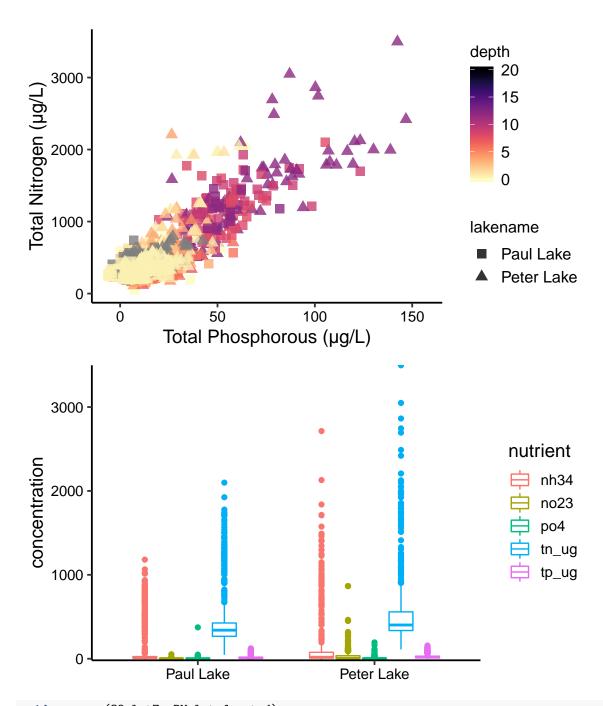
## Warning: package 'gridExtra' was built under R version 3.5.2

##
## Attaching package: 'gridExtra'

## The following object is masked from 'package:dplyr':

##
## combine
grid.arrange(NvsP2, Nutrientplot)
```

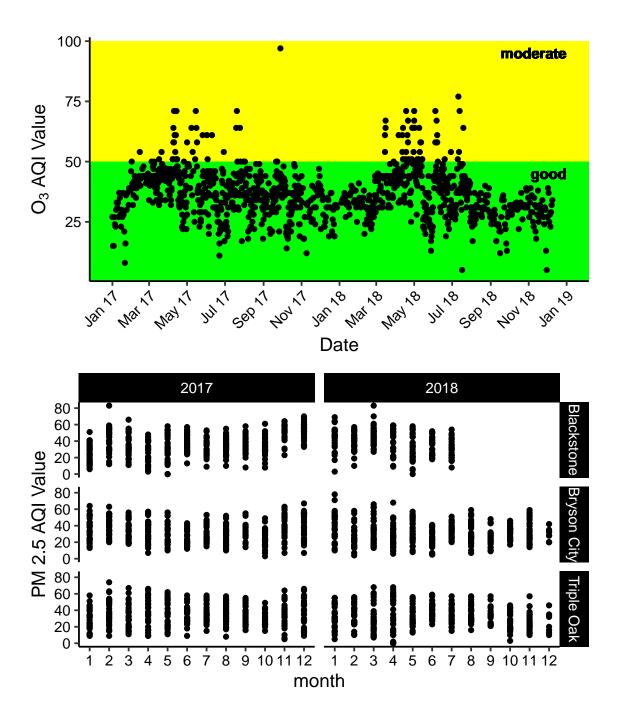
Warning: Removed 21648 rows containing missing values (geom_point).



grid.arrange(O3plot7, PMplot.faceted)

Warning: Removed 868 rows containing missing values (geom_point).

Warning: Removed 52 rows containing missing values (geom_point).



Saving plots

The ggsave function allows you to save plots in jpg, png, eps, pdf, tiff, and other formats. The following information can be supplied:

- filename, with file extension and in quotes (required)
- plot object (required)
- path, with file name
- width, height, units
- resolution (dpi)

For example: ggsave("PMplot.jpg", PMplot.faceted, path = "./Output/PMplot.jpg", height =

4, width = 6, units = "in", dpi = 300)