# Assignment 5: Data Visualization

Student Name

#### **OVERVIEW**

This exercise accompanies the lessons in Environmental Data Analytics (ENV872L) on data wrangling.

#### **Directions**

- 1. Change "Student Name" on line 3 (above) with your name.
- 2. Use the lesson as a guide. It contains code that can be modified to complete the assignment.
- 3. Work through the steps, creating code and output that fulfill each instruction.
- 4. Be sure to **answer the questions** in this assignment document. Space for your answers is provided in this document and is indicated by the ">" character. If you need a second paragraph be sure to start the first line with ">". You should notice that the answer is highlighted in green by RStudio.
- 5. When you have completed the assignment, **Knit** the text and code into a single PDF file. You will need to have the correct software installed to do this (see Software Installation Guide) Press the **Knit** button in the RStudio scripting panel. This will save the PDF output in your Assignments folder.
- 6. After Knitting, please submit the completed exercise (PDF file) to the dropbox in Sakai. Please add your last name into the file name (e.g., "Salk\_A04\_DataWrangling.pdf") prior to submission.

The completed exercise is due on Tuesday, 19 February, 2019 before class begins.

### Set up your session

- 1. Set up your session. Upload the NTL-LTER processed data files for chemistry/physics for Peter and Paul Lakes (tidy and gathered), the USGS stream gauge dataset, and the EPA Ecotox dataset for Neonicotinoids.
- 2. Make sure R is reading dates as date format, not something else (hint: remember that dates were an issue for the USGS gauge data).

```
#1
getwd()
```

## [1] "/Users/xiaqianyi/Documents/current semester/ENV 872/ENV 872/Assignments"

```
library(tidyverse)
PeterPaul_Processed <-read.csv("../Data/Processed/NTL-LTER_Lake_Nutrients_PeterPaul_Processed.csv")
PeterPaul_Gathered <- read.csv("../Data/Processed/NTL-LTER_Lake_Nutrients_PeterPaulGathered_Processed.c
EPAtox <- read.csv("../Data/Raw/ECOTOX_Neonicotinoids_Mortality_raw.csv")
USGS.flow.data <- read.csv("../Data/Raw/USGS_Site02085000_Flow_Raw.csv")
#2 Format Date
USGS.flow.data$datetime <- as.Date(USGS.flow.data$datetime, format = "%m/%d/%y")
USGS.flow.data$datetime <- format(USGS.flow.data$datetime, "%y%m%d")

create.early.dates <- (function(d) {
    paste0(ifelse(d > 181231, "19", "20"), d)
    })
USGS.flow.data$datetime <- create.early.dates(USGS.flow.data$datetime)
USGS.flow.data$datetime <- as.Date(USGS.flow.data$datetime, format = "%Y%m%d")</pre>
```

```
class(PeterPaul_Gathered$sampledate)

## [1] "factor"

PeterPaul_Processed$sampledate<-as.Date(PeterPaul_Processed$sampledate, format = "%Y%m%d")

PeterPaul_Gathered$sampledate<-as.Date(PeterPaul_Gathered$sampledate, format = "%Y-%m-%d")

# EPAtox$Pub..Year <- factor(EPAtox$Pub..Year, format = '%Y')</pre>
```

#### Define your theme

3. Build a theme and set it as your default theme.

```
#3
QYtheme <- theme_classic(base_size= 14) +
theme(axis.text = element_text(color = "black"))</pre>
```

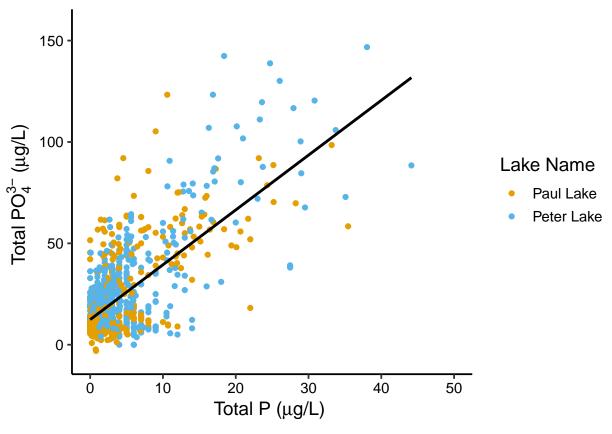
## Create graphs

For numbers 4-7, create graphs that follow best practices for data visualization. To make your graphs "pretty," ensure your theme, color palettes, axes, and legends are edited to your liking.

Hint: a good way to build graphs is to make them ugly first and then create more code to make them pretty.

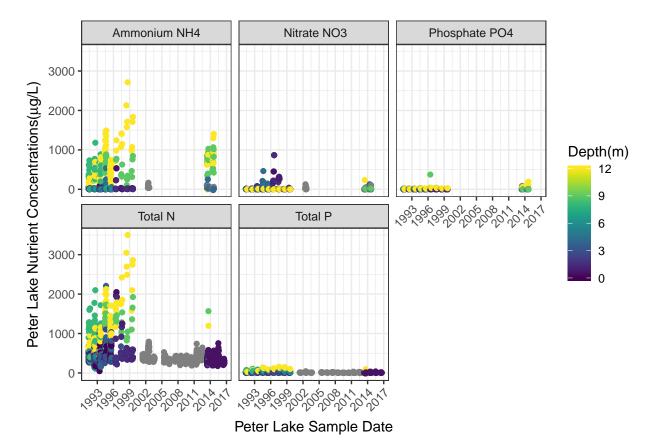
4. [NTL-LTER] Plot total phosphorus by phosphate, with separate aesthetics for Peter and Paul lakes. Add a line of best fit and color it black.

```
#4 P04 (x) is phosphate, and TP (y) is phosphorus
library(viridis)
PeterPaulP_P04 <-
ggplot(PeterPaul_Processed, aes(x= po4, y=tp_ug,color=lakename))+
xlim(0, 50)+
geom_point()+
scale_color_manual(values=c("#E69F00", "#56B4E9"))+
# scale_shape_manual(values = c(21, 23))+
# scale_fill_viridis(option = 'magma', direction = -1) +
geom_smooth(method=lm, color="black", se=F)+
labs(y=expression(paste('Total ',P0[4]^{'3-'}, ' (',mu, 'g/L) ')), x=expression(paste('Total P ', '(', #scale_color_discrete(name="Lake Name")+
QYtheme
#defaul color</pre>
print(PeterPaulP_P04)
```



5. [NTL-LTER] Plot nutrients by date for Peter Lake, with separate colors for each depth. Facet your graph by the nutrient type.

```
#5
nutrient_name <- c(</pre>
  'nh34' = "Ammonium NH4",
  'no23' =" Nitrate NO3",
  'po4' = "Phosphate PO4",
  'tn_ug' = "Total N",
   'tp_ug' = "Total P"
PeterNutrients <-
  ggplot(subset(PeterPaul_Gathered, lakename = 'Paul Lake'))+
  geom_point(aes(x=PeterPaul_Gathered$sampledate, y= PeterPaul_Gathered$concentration, color=depth))+
   facet_wrap(vars(nutrient), labeller = as_labeller(nutrient_name))+
  labs(x= expression("Peter Lake Sample Date"), y= expression(paste('Peter Lake Nutrient Concentrations
  scale_x_date(limits = as.Date(c("1991-05-01", "2016-08-31")),
date_breaks = "36 months", date_labels = " %Y")+
  scale_color_viridis()+
   theme_bw()+
  theme(axis.text.x=element_text(angle = 45, hjust = 1))
print(PeterNutrients)
```



6. [USGS gauge] Plot discharge by date. Create two plots, one with the points connected with geom\_line and one with the points connected with geom\_smooth (hint: do not use method = "lm"). Place these graphs on the same plot (hint: ggarrange or something similar)

```
#6
library(ggpubr)
library(gridExtra)
colnames(USGS.flow.data) <- c("agency_cd", "site_no", "datetime", "discharge.max", "discharge.max.appro</pre>
"discharge.mean", "discharge.mean.approval",
"gage.height.max", "gage.height.max.approval",
"gage.height.min", "gage.height.min.approval",
"gage.height.mean", "gage.height.mean.approval")
DischargeLine <-
  ggplot(USGS.flow.data, aes(x= datetime, y=discharge.mean, na.rm=TRUE))+
  ylim(0,500)+
  scale_x_date(limits = as.Date(c("2004-05-01", "2018-12-31")),
date_breaks = "10 months", date_labels = "%Y-%m")+
  geom_point(alpha = 0.4, size=0.6)+
  geom_line(color='black', alpha=0.6)+
  labs(x= expression("Date"), y= expression(paste('Mean Discharge (', ft^3,'/s)')))+
theme_bw(base_size= 12)+
    theme(axis.text.x=element_text(angle = 45, hjust = 1))
DischargeSmooth <-
    ggplot(USGS.flow.data, aes(x= datetime, y=discharge.mean, na.rm=TRUE))+
  ylim(0,500) +
  scale_x_date(limits = as.Date(c("2004-05-01", "2018-12-31")),
```

```
date_breaks = "10 months", date_labels = "%Y-%m")+
  geom_point(alpha = 0.4, size=0.6)+
  geom_smooth(alpha=0.6)+
  labs(x= expression("Date"), y= expression(paste('Mean Discharge (', ft^3,'/s)')))+
theme_bw(base_size= 12)+
    theme(axis.text.x=element_text(angle = 45, hjust = 1))
ggarrange(DischargeLine, DischargeSmooth, nrow=2, labels = c("
                                                                      Lines for Discharge Mean", "LOESS Smoo
           font.label = list(size = 10, color = "red3"), vjust=2)
Mean Discharge (ft³/s
    500
                Lines for Discharge Mean
    400
    300
    200
    100
                                          2017.02
       2001-06
                        201-10
                             208.08
                                                   2012-10
                                                                         2016-12
                    2005/12
                                 209.06
                                      2010-04
                                                            2014-06
                                                                2015.04
                                                                             2017,70
                                                                                      2019-06
                                                                     2016-02
                                                                                  2018-08
                                                 Date
Mean Discharge (ft³/s)
    500
                 LOESS Smooth For Discharge Mean
    400
    300
    200
    100
                             2008-08
                                          2017.02
```

Question: How do these two types of lines affect your interpretation of the data?

Answer: From the plot data, we know that most of the points are concentrate at a low value of y axis, therefore, when using geom line to connect them from point to point, we can hardly see the distinction at the bottom of the graph; When applying the geom smooth function, it generates the locally estimated scatterplot smoothing (LOESS), which allows to see a moving regression of the data point trend. The geom smooth may provide better interpretation of the data.

Date

7. [ECOTOX Neonicotinoids] Plot the concentration, divided by chemical name. Choose a geom that accurately portrays the distribution of data points.

```
#7
class(EPAtox$Conc..Mean..Std.)
## [1] "numeric"
ChemicalConc <- EPAtox %>%
    group_by(Chemical.Name, Conc..Mean..Std.) %>%
    summarise(sd=sd(Conc..Mean..Std.),
```

```
mean=mean(Conc..Mean..Std.))
ChemicalC <-
ggplot(EPAtox, aes(x=Chemical.Name, y=Conc..Mean..Std.))+
geom_boxplot(fill='#A4A4A4', color="black", outlier.size = 1)+
ylim(0,1500)+
labs(x= expression("Chemicals"), y= expression('Mean Concentration (mg/L)'), hjust=1)+
QYtheme+
theme(axis.text.x=element_text(angle = 45, hjust = 1))

# plot1 <-
# ggplot(ChemicalConc, aes(x=Chemical.Name, y=mean))+
# geom_bar(stat = "identity", position = "dodge") +
# geom_errorbar(aes(ymin = mean-sd, ymax = mean+sd), position = "dodge")
print(ChemicalC)</pre>
```

