USGS NAS

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The data source is the United States Geological Survey (USGS) Nonindigenous Aquatic Species (NAS) data set. It is a collection of zebra mussel sightings across the United States. Zebra mussels are an incredibly damaging invasive species in the continental US.

Zebra mussels belong to the phylum Mollusca, the group bivalves (unique among them—the only species that attaches to a hard substrate), which are native to the Black, Caspian, and Azov seas (USGS NAS). They were first observed in areas nearby Indiana, Michigan, and Ohio in 1986 (USGS NAS). Zebra mussels likely originated in Europe, establishing a population in Lakes St. Clair and Erie when ships dispersed ballast water (Hebert et al. 1989). The influence of bivalves on ecosystems includes the following: removal of particles from the water column, reducing populations of consumers depending on these particles as food source, increasing populations that use bivalves or their waste products, and making available particles previously used by phytoplankton (Strayer et al. 1999). As a result, their introduction into the United States as a result of commercial trade has had cascading effects on ecosystems, affecting multiple trophic levels and altering the abiotic conditions of the water bodies in which they reside.

The outcome of invasions depends in part on the tolerance of environmental conditions in the area of introduction. Introduced environments with conditions similar to those found in the native range may be more conducive to invasive establishment (Baker and Stebbins, 1965). Additionally, reproductive and physiological characteristics of the invading organism play a role in whether populations become established in the new range. In the case of zebra mussels, free swimming larvae and high fecundity (eggs per female) have been implicated in its proliferation success (Hebert et al. 1989). That zebra mussels are amenable to a wide range of habitats with a flexible reproductive system has aided their spread and establishment in North America where they were introduced via the release of ballast water from ships (Nichols 1996).

Physiological constraints of organisms determine the optimal range in which reproduction can occur, an important factor in the spread and establishment of an invasive species (or any species, for that matter). One such constraint is thermal tolerance, and temperature was shown to affect the timing of gametogenesis (Wacker & Elert 2003). The two experimental groups were raised at different depths to generate the temperature difference, and variation in environmental quality in the surrounding region may explain the differences in egg mass released at the two different depths (Wacker & Elert 2003). In addition, metabolic rates increase with temperature, so that differences in food availability at the two depths could further influence reproductive investment. Without respective to temperature differences, availability of polyunsaturated fatty acids and food quality in general resulted in changes to reproductive investment (Wacker & Elert 2003). An additional study confirms a threshold of 12 C for the onset of spawning and further identifies two spawning cohorts as the season proceeds from May to August (Borcherding 1991). Gametes are released into the water column over a period of 6 to 8 weeks, where they are fertilized and develop—30,000 to 40,000 eggs may be released by one female, though the actual number may be closer to 1.5 million. Egg release corresponds with temperature, beginning at 12C and peaking at 22C. As a result, juvenile proliferation should track these temperature changes (Hebert et al. 1988). High temperatures alongside low food availability results in reduction in the size of the gonads (Borcherding 1991).

Dreissena tends to colonize structures below 1.2 m. Zebra mussels may spread to larger structures through water intake pipes. Zebra mussels tend to increase water clarity through digestion of suspended sediments—for this reason zebra mussels have been intentionally stocked in lakes outside North America. Food deprived mussels fed indiscriminately on particles of all sizes, but satiated individuals felt only on those in a much smaller size range (MacIsaac 1996).

Ambient temperature, seston concentration, and mussel size frequency are three factors that influence Dreissena filtering impact. Maximal filtering rate has been hypothesized to be 5 and 20 C, declining outside

of this range. Ingestion rate may also be temperature dependent; however, the results of experiments investigating the effect of temperature are variable. It may still be an important performance factor, and impose a range limitation in southern states. Depending on size, zooplankton may succumb to or evade ingestion by zebra mussels—smaller individuals cannot escape the inflow current and are not rejected, whereas larger individuals may dodge the current or be expelled as a result of irritating the feeding apparatus (MacIsaac 1996).

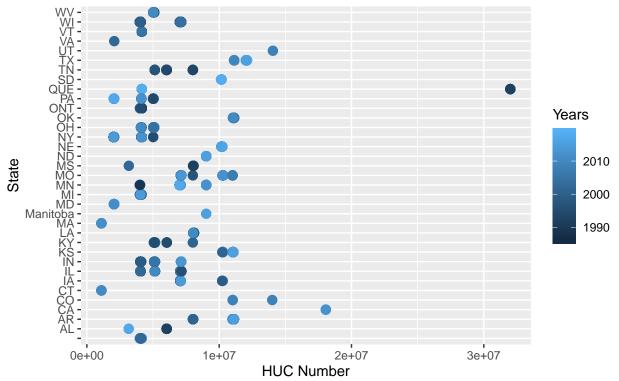
```
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
## filter, lag
## The following objects are masked from 'package:base':
##
## intersect, setdiff, setequal, union
```

Initial Visualization

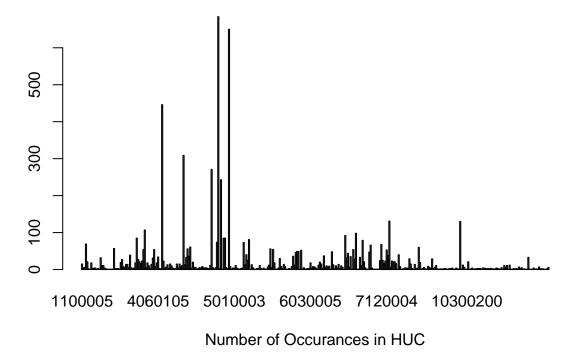
The graph below is one way in which the data set from the USGS zebra mussel data can be visualized. The horizontal axis represents the HUC Number and it is plotted again the state in which that HUC resides. The color of the point exhibits the time when the data was collected. There appears to be an even spread of years throughout the plot given by the various color gradients.

Warning: Removed 576 rows containing missing values (geom_point).

HUC Number vs. State From USGS Zebra Mussel Dataset



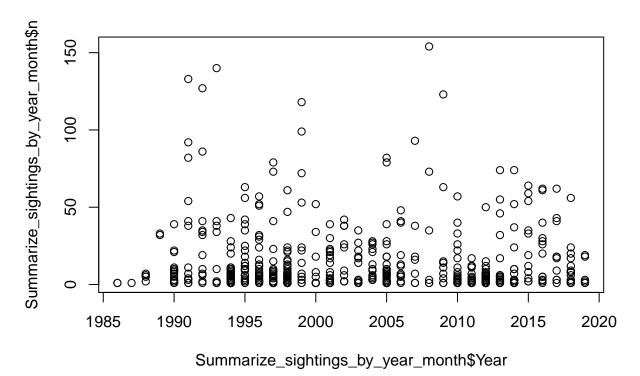
HUC Distribution



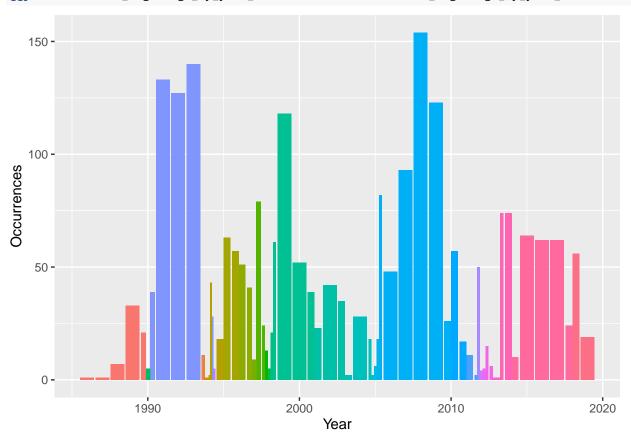
Maps and graph by state and year

These maps demonstrate the spread of zebra mussels in increments of one year.

```
#Plot number of zebra mussels sightings per year
Summarize_sightings_by_year_month <- dat %>% count(Year, Month,State)
plot(Summarize_sightings_by_year_month$Year,Summarize_sightings_by_year_month$n)
```



ggplot(Summarize_sightings_by_year_month, aes(fill=Summarize_sightings_by_year_month\$State, y=Summarize



This graph demonstrates the number of occurrences by state, month, and year. The spread from state to state is observable.

[&]quot;'r #filter data for only Great Lakes lat and long BorderingGreatLakes <- c("New York", "Pennsylvania", "Ohio", "Indiana", "Mich

PlottingBorder <- map_data("state", region=BorderingGreatLakes)

 $Borders <- ggplot(data = PlottingBorder) + geom_polygon(aes(x = long, y = lat, group = group), color = \\ "white") + coord_fixed(1.3) + guides(fill=FALSE) + scale_y_continuous(breaks=seq(from=30,to=50,by=1)) + scale_x_continuous(breaks=seq(from=30,to=50,by=1)) + scale_x_continuous(breaks=seq(fro$

Superior <- dat %>% filter(Latitude>=46 & Latitude<=50) %>% filter(Longitude<=-84.5 & Longitude>=-92) map4 <- Borders + geom point(aes(x=Longitude,y=Latitude,color=Status),data=Superior,alpha=0.2)

Michigan <- dat %>% filter(Latitude>=42.5 & Latitude<=46) %>% filter(Longitude<=-84.5 & Longitude>=-88) map5 <- Borders + geom_point(aes(x=Longitude,y=Latitude,color=Status),data=Michigan,alpha=0.2)

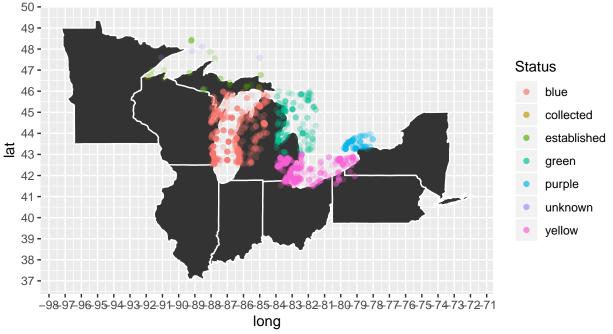
Huron <- dat %>% filter(Latitude>=43 & Latitude<=46) %>% filter(Longitude<=-81.5 & Longitude>=-84) map6 <- Borders + geom point(aes(x=Longitude,y=Latitude,color=Status),data=Huron,alpha=0.2)

Erie <- dat %>% filter(Latitude>=41.5 & Latitude<=43) %>% filter(Longitude<=-79 & Longitude>=-84) map7 <- Borders + geom point(aes(x=Longitude,y=Latitude,color=Status),data=Erie,alpha=0.2)

Ontario <- dat %>% filter(Latitude>=43.25 & Latitude<=44) %>% filter(Longitude<=-78 & Longitude>=-80) map8 <- Borders + geom point(aes(x=Longitude,y=Latitude,color=Status),data=Ontario,alpha=0.2)

 $\# all\ great\ Lakes\ plotted\ Great\ Lakes\ <-\ Borders\ +\ geom_point(aes(x=Longitude,y=Latitude,color=Status),data=Superior,alphageom_point(aes(x=Longitude,y=Latitude,color="blue"),data=Michigan,alpha=0.2) + geom_point(aes(x=Longitude,y=Latitude,color="yellow"),data=Erie,alpha=0.2) + geom_point(aes(x=Longitude,y=Latitude,color="yellow"),data=Erie$

GreatLakes "'



This map shows the occurrences for only the Great Lakes region. Notably, Lake Michigan has the most, and Lake Superior has the least.

```
#number of specimens found in each lake`
sup_occ <- unique(Superior$Specimen.Number)
mich_occ <- unique(Michigan$Specimen.Number)
huron_occ <- unique(Huron$Specimen.Number)
erie_occ <- unique(Erie$Specimen.Number)
ont_occ <- unique(Ontario$Specimen.Number)
zm_occ <- c(sup_occ, mich_occ, huron_occ, erie_occ, ont_occ)</pre>
```

```
SupOcc <- nrow(table(sup_occ))
MichOcc <- nrow(table(mich_occ))
HurOcc <- nrow(table(huron_occ))
EriOcc <- nrow(table(erie_occ))
OntOcc <- nrow(table(ont_occ))
zm_occ <- c(SupOcc, MichOcc, HurOcc, EriOcc, OntOcc)</pre>
```

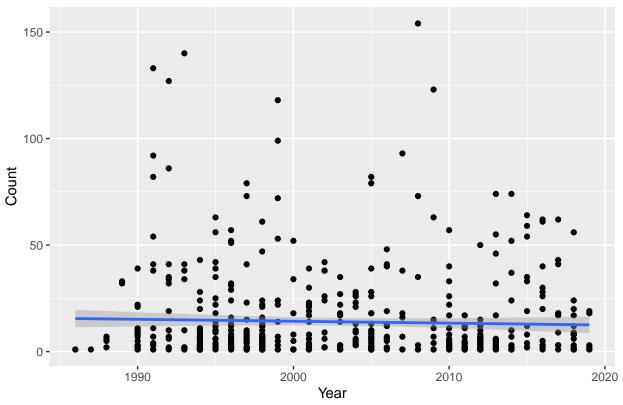
One occurrence indicates one collection effort—anywhere from one to thousands of zebra mussels could be included. Lake Michigan has the highest number of occurrences at 610, while Lake Superior has the least, at 36.

```
CountYear <- lm(n~Year, Summarize_sightings_by_year_month)
summary(CountYear)</pre>
```

Call:

```
## lm(formula = n ~ Year, data = Summarize_sightings_by_year_month)
##
## Residuals:
##
       Min
                1Q Median
                                3Q
                                       Max
                             3.464 140.459
## -14.519 -12.182 -8.721
##
## Coefficients:
##
              Estimate Std. Error t value Pr(>|t|)
                          217.9431
                                              0.374
## (Intercept) 194.0582
                                     0.890
                -0.0899
                            0.1088 -0.826
                                              0.409
##
## Residual standard error: 21.59 on 553 degrees of freedom
## Multiple R-squared: 0.001233,
                                    Adjusted R-squared:
                                                         -0.000573
## F-statistic: 0.6827 on 1 and 553 DF, p-value: 0.409
ggplot(Summarize_sightings_by_year_month, aes(x=Summarize_sightings_by_year_month$Year, y=Summarize_sig
```

Occurrences By Year

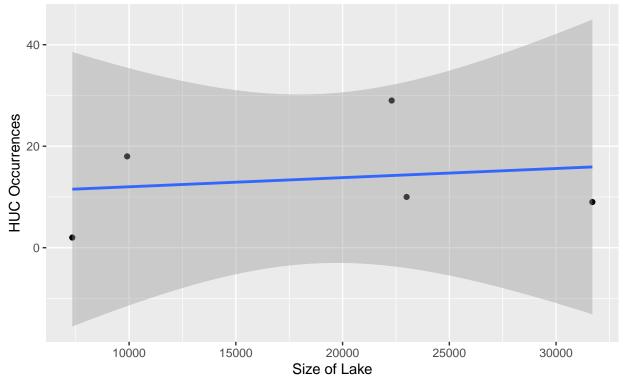


There is no relationship between number of occurrences and year. Number of sightings does not increase or decrease as time goes on.

Linear Regression of Size of Lake and HUC Counts

```
##
## lm(formula = zm_HUC_occurances ~ size_of_lake, data = sizeDat)
##
## Residuals:
## -6.915 14.778 -4.348 6.011 -9.526
##
## Coefficients:
##
                 Estimate Std. Error t value Pr(>|t|)
                                                0.462
## (Intercept) 1.020e+01 1.214e+01
                                        0.84
## size_of_lake 1.802e-04 5.811e-04
                                       0.31
                                                0.777
## Residual standard error: 11.72 on 3 degrees of freedom
## Multiple R-squared: 0.03104,
                                   Adjusted R-squared: -0.2919
## F-statistic: 0.0961 on 1 and 3 DF, p-value: 0.7768
```

Size of Lake vs. HUC Occurrences From USGS Zebra Mussel Dataset



A linear regression analysis was conducted after collecting data on the size of the lake and filtering out the data to only look at the HUC zebra mussel occurrences in the areas of the Great Lakes. The reason for this analysis is to understand if there is a relationship between the size of the lake in water volume, and the number of occurrences on certain HUCs.

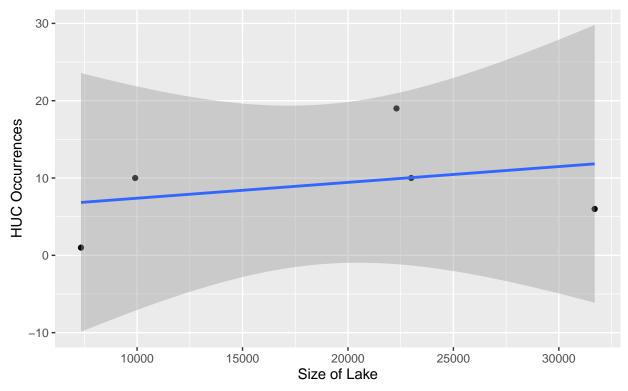
Once the analysis was conducted, the p-value associated with the coefficient is not a significant value, and therefore not statistically significant. From the model, its states that a unit increase in the size of the lake leads to a 1.8015439×10^{-4} increase in the HUC Occurrences. Therefore, there is not a significant relationship between the HUC Occurrences, and the size of each Great Lake. The plot also depicts the trend line with the slope of 1.8015439×10^{-4} and there are confidences bands around the trend line also. This is a small data set with only five data points though, given there are only five Great Lakes.

To conduct further analysis, that data used used the time span of the years 1991-2015 and the data was then split into three different time period. The time periods are [1991-1999], [2000-2008], and [2009-2015]. Each time period has a linear regression analysis associated with it and the details for each will be described along with a graph of each trend line.

```
##
## Call:
## lm(formula = zm_HUC_occurances_F8 ~ size_of_lake, data = sizeDat)
##
## Residuals:
                   2
##
                             3
##
   -5.83035 9.09380 -0.04949
                               2.62999 -5.84395
##
##
  Coefficients:
##
                 Estimate Std. Error t value Pr(>|t|)
## (Intercept) 5.3414746
                                        0.711
                                                 0.528
                          7.5144357
## size_of_lake 0.0002047 0.0003596
                                        0.569
                                                 0.609
```

```
##
## Residual standard error: 7.252 on 3 degrees of freedom
## Multiple R-squared: 0.09748, Adjusted R-squared: -0.2034
## F-statistic: 0.324 on 1 and 3 DF, p-value: 0.609
```

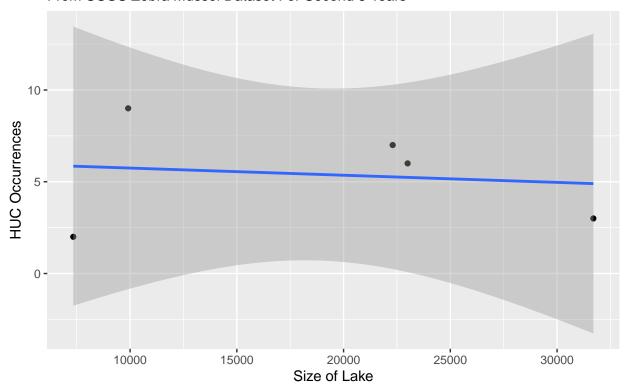
Size of Lake vs. HUC Occurrences From USGS Zebra Mussel Dataset For First 8 Years



The p-value associated with the coefficient from the data in the range of 1991-1999 is not a significant value, and therefore not statistically significant. From the model, its states that a unit increase in the size of the lake leads to a 2.0469631×10^{-4} increase in the HUC Occurrences between the years of 1991-1999. Therefore, there is not a significant relationship between the HUC Occurrences, and the size of each Great Lake during that time frame. The plot also depicts the trend line with the slope of 2.0469631×10^{-4} and there are confidences bands around the trend line also.

```
##
  lm(formula = zm_HUC_occurances_S8 ~ size_of_lake, data = sizeDat)
##
##
  Residuals:
##
         1
                         3
  -1.8982 1.7347 0.7621 3.2509 -3.8495
##
##
##
  Coefficients:
##
                  Estimate Std. Error t value Pr(>|t|)
   (Intercept)
                 6.136e+00
                            3.415e+00
                                         1.797
                                                  0.170
  size_of_lake -3.905e-05
                            1.634e-04
                                       -0.239
##
##
## Residual standard error: 3.295 on 3 degrees of freedom
## Multiple R-squared: 0.01868,
                                    Adjusted R-squared:
## F-statistic: 0.0571 on 1 and 3 DF, p-value: 0.8265
```

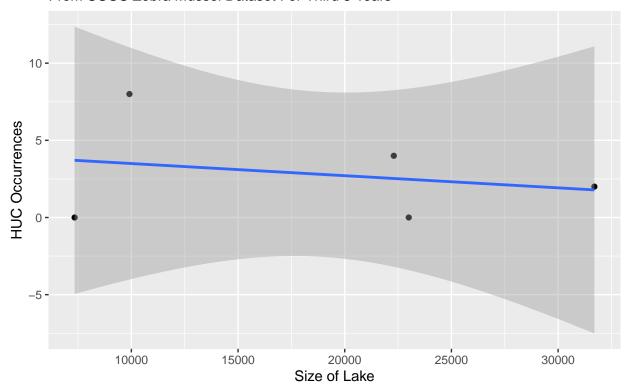
Size of Lake vs. HUC Occurrences From USGS Zebra Mussel Dataset For Second 8 Years



The p-value associated with the coefficient from the data in the range of 2000-2008 is not a significant value, and therefore not statistically significant. From the model, its states that a unit increase in the size of the lake leads to a $-3.9050666 \times 10^{-5}$ increase in the HUC Occurrences between the years of 2000-2008. Therefore, there is not a significant relationship between the HUC Occurrences, and the size of each Great Lake during that time frame. The plot also depicts the trend line with the slope of $-3.9050666 \times 10^{-5}$ and there are confidences bands around the trend line also. This relationship is also negative compared the the first 8 years, which had a positive relationship. Even though neither of the slopes were significant, there could be a change occurring in those time periods that need to be explored.

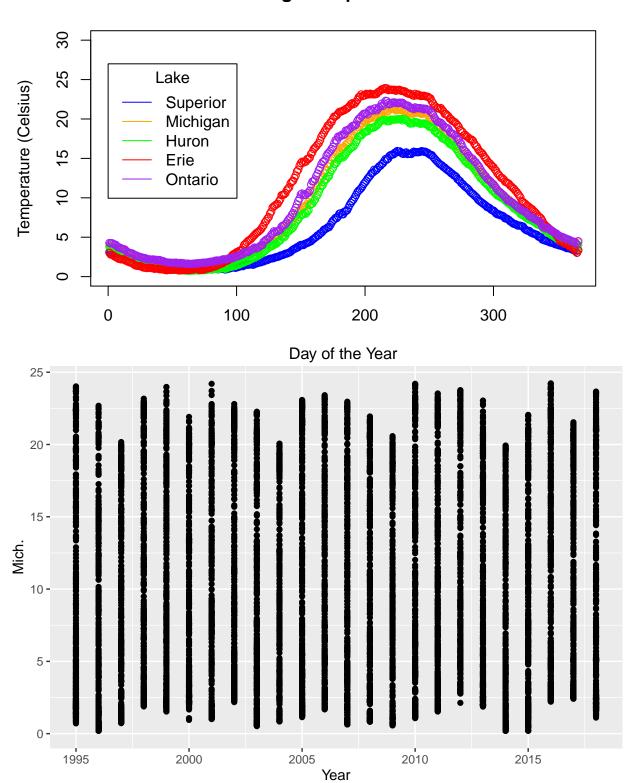
```
##
## Call:
  lm(formula = zm_HUC_occurances_T8 ~ size_of_lake, data = sizeDat)
##
##
## Residuals:
##
                 2
                         3
   0.2118 1.4717 -2.4732 4.4961 -3.7063
##
##
##
  Coefficients:
##
                  Estimate Std. Error t value Pr(>|t|)
                 4.284e+00 3.890e+00
                                         1.101
                                                  0.351
##
  (Intercept)
## size of lake -7.874e-05 1.862e-04
                                                  0.701
\#\# Residual standard error: 3.754 on 3 degrees of freedom
## Multiple R-squared: 0.05628,
                                    Adjusted R-squared:
## F-statistic: 0.1789 on 1 and 3 DF, p-value: 0.7008
```

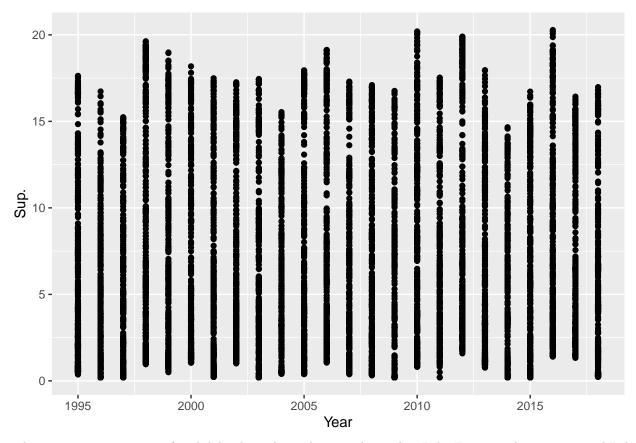
Size of Lake vs. HUC Occurrences From USGS Zebra Mussel Dataset For Third 8 Years



The p-value associated with the coefficient from the data in the range of 2009-2015 is not a significant value, and therefore not statistically significant. From the model, its states that a unit increase in the size of the lake leads to a -7.87407×10^{-5} increase in the HUC Occurrences between the years of 2009-2015. Therefore, there is not a significant relationship between the HUC Occurrences, and the size of each Great Lake during that time frame. The plot also depicts the trend line with the slope of -7.87407×10^{-5} and there are confidences bands around the trend line also. This coefficient is negative and larger than the previous 8 year period. However, the overall regression analysis shows a positive non significant trend, which insinuates that the first 8 year period has a large affect on the regression analysis since it is the only positive trend out of the three time periods.

Average Temp 1992–2018

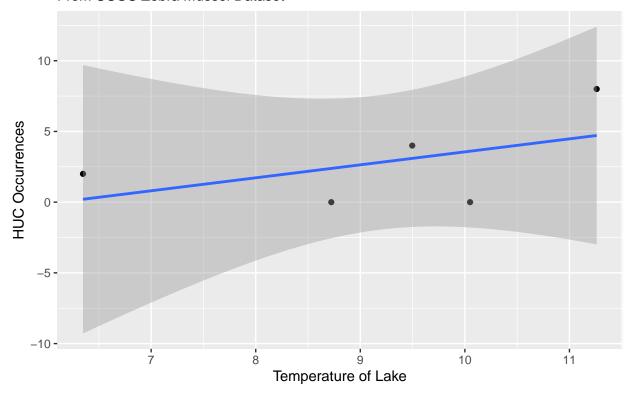




The average temperature of each lake throughout the year shows that Lake Erie get sthe warmest and Lake Superior is the coldest. When specifically looked at, Lake Superior has a high temperature that is about 5 degrees lower than the high temperature of Lake Michigan. There is a lower density of zebra mussels in Lake Superior due to its lower temperatures and harsher conditions.

```
##
##
  Call:
   lm(formula = zm_HUC_occurances ~ avgTemp, data = TempDat)
##
##
##
  Residuals:
##
                             3
                                               5
          1
                   2
                                      4
##
    -0.6765 14.9544
                      -2.9716
                                 1.5067 -12.8131
##
##
   Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
                  0.867
                             29.363
                                       0.03
                                               0.978
   (Intercept)
## avgTemp
                  1.388
                              3.150
                                       0.44
                                               0.689
##
## Residual standard error: 11.54 on 3 degrees of freedom
## Multiple R-squared: 0.06075,
                                     Adjusted R-squared:
## F-statistic: 0.194 on 1 and 3 DF, p-value: 0.6894
```

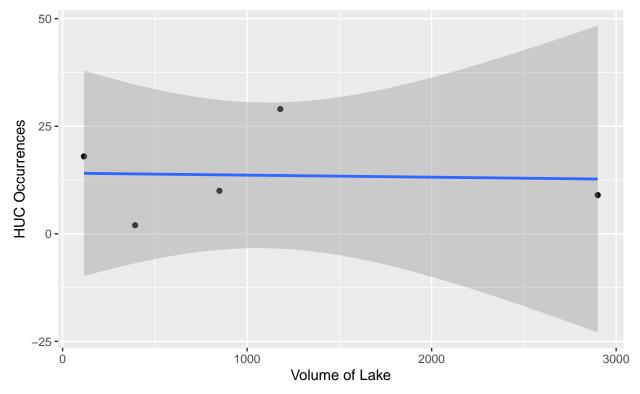
Temperature of Lake vs. HUC Occurrences From USGS Zebra Mussel Dataset



Analysis to be done on this, not significant values here.

```
##
## Call:
## lm(formula = zm_HUC_occurances ~ volume_of_lake, data = volumeDat)
##
## Residuals:
##
                 2
                         3
    -3.736 15.444 -3.713
                             3.937 -11.931
##
##
## Coefficients:
                    Estimate Std. Error t value Pr(>|t|)
##
                  14.1184532 7.9560595
                                          1.775
                                                   0.174
## (Intercept)
## volume_of_lake -0.0004766 0.0054405 -0.088
                                                   0.936
## Residual standard error: 11.89 on 3 degrees of freedom
## Multiple R-squared: 0.002552,
                                    Adjusted R-squared: -0.3299
## F-statistic: 0.007674 on 1 and 3 DF, p-value: 0.9357
```

Volume of Lake vs. HUC Occurrences From USGS Zebra Mussel Dataset



This analysis was mostly a descriptive visualization—where, and when, are zebra mussels found? Additional linear models demonstrate no significant relationship of zebra mussel occurrences to the size (area or volume) of the lake, or the time period.

Easton comments

In general, you did a nice job of finding a dataset, learning about the system, and running some analyses. Most of my comments below have to deal with the presentation of the results and organization of the text.

Things to do for final draft:

- In general the figures look nice, but they can be cleaned up with little things like the axes titles
- Reorder the text at the beginning so it reads with a traditional introduction, methods, and results
- add appropriate figure captions
- include formatting to make R markdown output look nicer (e.g. section headers, remove chunk messages)
- Have a short paragraph that summarizes and puts your results in context

References Hebert, P. D. N., Mackie, G. L., & Muncaster, B. W. (1989). Ecological and Genetic Studies on Dreissena polymorpha (Pallas): A New Mollusc in the Great Lakes. Candian Journal of Fisheries and Aquatic Sciences. https://doi.org/10.1139/f89-202

Macisaac, H. J. (2014). Potential Abiotic and Biotic Impacts of Zebra Mussels on the Inland Waters of North America Potential Abiotic and Biotic Impacts of Zebra Mussels on the, (March). https://doi.org/10.1093/icb/36.3.287

Mackie, G. L. (2016). Biology of the exotic zebra mussel, Dreissena polymporph, in relation to native bivalves and its potential impact in Lake St. Clair, (July 1991). https://doi.org/10.1007/BF00024759

Nichols, S. J. (1996). Variations in the Reproductive Cycle of Dreissena Polymorpha in Europe, Russia, and North America. American Zoologist, 36, 311–325.

Strayer, D. L., Caraco, N. F., Cole, J. J., Findlay, S., & Pace, M. L. (1999). Transformation of Freshwater A case study of zebra mussels in the Hudson River. BioScience, 49(1).

Wacker, E. A. (2003). Food quality controls reproduction of the zebra mussel (Dreissena polymorpha). Oecologia, 135, 332–338. https://doi.org/10.1007/s00442-003-1208-5