Assignment 7: GLMs (Linear Regressios, ANOVA, & t-tests)

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## OVERVIEW

This exercise accompanies the lessons in Environmental Data Analytics on generalized linear models.

## Directions

1. Change “Student Name” on line 3 (above) with your name.
2. Work through the steps, **creating code and output** that fulfill each instruction.
3. Be sure to **answer the questions** in this assignment document.
4. When you have completed the assignment, **Knit** the text and code into a single PDF file.
5. After Knitting, submit the completed exercise (PDF file) to the dropbox in Sakai. Add your last name into the file name (e.g., “Fay\_A06\_GLMs.Rmd”) prior to submission.

The completed exercise is due on Tuesday, March 2 at 1:00 pm.

## Set up your session

1. Set up your session. Check your working directory. Load the tidyverse, agricolae and other needed packages. Import the *raw* NTL-LTER raw data file for chemistry/physics (NTL-LTER\_Lake\_ChemistryPhysics\_Raw.csv). Set date columns to date objects.
2. Build a ggplot theme and set it as your default theme.

#1  
  
library(knitr)  
opts\_chunk$set(tidy.opts=list(width.cutoff=80),tidy=TRUE)  
  
# setup  
getwd()

## [1] "/Users/Aislinn/Documents/GitHub/Environmental\_Data\_Analytics\_2021/Assignments"

setwd("~/Documents/GitHub/Environmental\_Data\_Analytics\_2021")  
  
library(tidyverse)

## ── Attaching packages ─────────────────────────────────────── tidyverse 1.3.0 ──

## ✓ ggplot2 3.3.3 ✓ purrr 0.3.4  
## ✓ tibble 3.0.4 ✓ dplyr 1.0.3  
## ✓ tidyr 1.1.2 ✓ stringr 1.4.0  
## ✓ readr 1.4.0 ✓ forcats 0.5.0

## ── Conflicts ────────────────────────────────────────── tidyverse\_conflicts() ──  
## x dplyr::filter() masks stats::filter()  
## x dplyr::lag() masks stats::lag()

library(agricolae)  
library(lubridate)

##   
## Attaching package: 'lubridate'

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

library(viridis)

## Loading required package: viridisLite

# import  
lake.chem.phys <-  
 read.csv("./Data/Raw/NTL-LTER\_Lake\_ChemistryPhysics\_Raw.csv", stringsAsFactors = TRUE)  
  
# format dates  
lake.chem.phys$sampledate <-  
 as.Date(lake.chem.phys$sampledate, "%m/%d/%y")  
  
#2  
  
mytheme <-  
 theme\_gray(base\_size = 12) +  
 theme(legend.background = element\_rect(fill = "gray"), legend.position = "bottom", plot.title = element\_text(face = "bold", size = 14, color = "black", hjust = 0.5), plot.subtitle = element\_text(size = 10, color = "gray", hjust = 0.5))  
  
theme\_set(mytheme)

## Simple regression

Our first research question is: Does mean lake temperature recorded during July change with depth across all lakes?

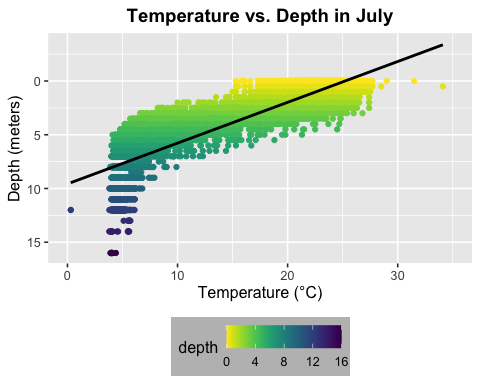
1. State the null and alternative hypotheses for this question: > Answer: H0: Mean lake temperature recorded during July does not change with depth across all lakes (mean lake temperature is the same at all depths during July). Ha: Mean lake temperature recorded during July does change with depth across all lakes.
2. Wrangle your NTL-LTER dataset with a pipe function so that the records meet the following criteria:

* Only dates in July.
* Only the columns: lakename, year4, daynum, depth, temperature\_C
* Only complete cases (i.e., remove NAs)

1. Visualize the relationship among the two continuous variables with a scatter plot of temperature by depth. Add a smoothed line showing the linear model, and limit temperature values from 0 to 35 °C. Make this plot look pretty and easy to read.

#4  
  
lake.chem.phys.July <-  
 lake.chem.phys %>%  
 mutate(month = month(sampledate)) %>%  
 filter(month == 7) %>%  
 select(lakename, year4, daynum, depth, temperature\_C) %>%  
 na.omit()  
  
#5  
  
lake.July.plot <-  
 ggplot(lake.chem.phys.July, aes(x = temperature\_C, y = depth, color = depth)) +  
 geom\_point() +   
 geom\_smooth(method = lm, color = "black") +  
 scale\_y\_reverse() +  
 scale\_color\_viridis(direction = -1) +  
 xlim(0, 35) +  
 labs(x = "Temperature (°C)", y = "Depth (meters)", title = "Temperature vs. Depth in July")  
  
print(lake.July.plot)

## `geom\_smooth()` using formula 'y ~ x'



1. Interpret the figure. What does it suggest with regards to the response of temperature to depth? Do the distribution of points suggest about anything about the linearity of this trend?

Answer: As depth decreases, temperature increases. The increase in temperature as it relates to depth accelerates around a depth of 8 m from the surface. The distribution of points suggests this is not a linear relationship. For the most part, there seem to be more data points on top of the linear regression line and the points are not evenly distributed with regards to distance from the line.

1. Perform a linear regression to test the relationship and display the results

#7  
  
temperature.regression <-  
 lm(data = lake.chem.phys.July, temperature\_C ~ depth)  
  
summary(temperature.regression)

##   
## Call:  
## lm(formula = temperature\_C ~ depth, data = lake.chem.phys.July)  
##   
## Residuals:  
## Min 1Q Median 3Q Max   
## -9.5173 -3.0192 0.0633 2.9365 13.5834   
##   
## Coefficients:  
## Estimate Std. Error t value Pr(>|t|)   
## (Intercept) 21.95597 0.06792 323.3 <2e-16 \*\*\*  
## depth -1.94621 0.01174 -165.8 <2e-16 \*\*\*  
## ---  
## Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1  
##   
## Residual standard error: 3.835 on 9726 degrees of freedom  
## Multiple R-squared: 0.7387, Adjusted R-squared: 0.7387   
## F-statistic: 2.75e+04 on 1 and 9726 DF, p-value: < 2.2e-16

1. Interpret your model results in words. Include how much of the variability in temperature is explained by changes in depth, the degrees of freedom on which this finding is based, and the statistical significance of the result. Also mention how much temperature is predicted to change for every 1m change in depth.

Answer: Changes in depth explain 73.87% of the variability in temperature (simple linear regression, df = 9726, p-value = 2.2e-16). Because our p-value is below 0.05, this regression is meaningful. For every 1m increase in depth, temperature will change by ~ 1.95°C.

## Multiple regression

Let’s tackle a similar question from a different approach. Here, we want to explore what might the best set of predictors for lake temperature in July across the monitoring period at the North Temperate Lakes LTER.

1. Run an AIC to determine what set of explanatory variables (year4, daynum, depth) is best suited to predict temperature.
2. Run a multiple regression on the recommended set of variables.

#9  
  
temp.AIC <-  
 lm(data = lake.chem.phys.July, temperature\_C ~ year4 + daynum + depth)  
  
step(temp.AIC)

## Start: AIC=26065.53  
## temperature\_C ~ year4 + daynum + depth  
##   
## Df Sum of Sq RSS AIC  
## <none> 141687 26066  
## - year4 1 101 141788 26070  
## - daynum 1 1237 142924 26148  
## - depth 1 404475 546161 39189

##   
## Call:  
## lm(formula = temperature\_C ~ year4 + daynum + depth, data = lake.chem.phys.July)  
##   
## Coefficients:  
## (Intercept) year4 daynum depth   
## -8.57556 0.01134 0.03978 -1.94644

#10  
  
temp.model <- lm(data = lake.chem.phys.July, temperature\_C ~ year4 + daynum + depth)  
summary(temp.model)

##   
## Call:  
## lm(formula = temperature\_C ~ year4 + daynum + depth, data = lake.chem.phys.July)  
##   
## Residuals:  
## Min 1Q Median 3Q Max   
## -9.6536 -3.0000 0.0902 2.9658 13.6123   
##   
## Coefficients:  
## Estimate Std. Error t value Pr(>|t|)   
## (Intercept) -8.575564 8.630715 -0.994 0.32044   
## year4 0.011345 0.004299 2.639 0.00833 \*\*   
## daynum 0.039780 0.004317 9.215 < 2e-16 \*\*\*  
## depth -1.946437 0.011683 -166.611 < 2e-16 \*\*\*  
## ---  
## Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1  
##   
## Residual standard error: 3.817 on 9724 degrees of freedom  
## Multiple R-squared: 0.7412, Adjusted R-squared: 0.7411   
## F-statistic: 9283 on 3 and 9724 DF, p-value: < 2.2e-16

1. What is the final set of explanatory variables that the AIC method suggests we use to predict temperature in our multiple regression? How much of the observed variance does this model explain? Is this an improvement over the model using only depth as the explanatory variable?

Answer: The AIC method recommends we use year, day number and depth to predict temperature in our multiple regression. This new model explains 74.11% of variance in temperature. This is an improvement over our model using depth as the sole explanatory variable because in that model depth only explained 73.87% of the variance in temperature.

## Analysis of Variance

1. Now we want to see whether the different lakes have, on average, different temperatures in the month of July. Run an ANOVA test to complete this analysis. (No need to test assumptions of normality or similar variances.) Create two sets of models: one expressed as an ANOVA models and another expressed as a linear model (as done in our lessons).

#12  
  
temp.lakes.July.ANOVA <-  
 aov(data = lake.chem.phys.July, temperature\_C ~ lakename)  
summary(temp.lakes.July.ANOVA)

## Df Sum Sq Mean Sq F value Pr(>F)   
## lakename 8 21642 2705.2 50 <2e-16 \*\*\*  
## Residuals 9719 525813 54.1   
## ---  
## Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

temp.lakes.July.ANOVA2 <-  
 lm(data = lake.chem.phys.July, temperature\_C ~lakename)  
summary(temp.lakes.July.ANOVA2)

##   
## Call:  
## lm(formula = temperature\_C ~ lakename, data = lake.chem.phys.July)  
##   
## Residuals:  
## Min 1Q Median 3Q Max   
## -10.769 -6.614 -2.679 7.684 23.832   
##   
## Coefficients:  
## Estimate Std. Error t value Pr(>|t|)   
## (Intercept) 17.6664 0.6501 27.174 < 2e-16 \*\*\*  
## lakenameCrampton Lake -2.3145 0.7699 -3.006 0.002653 \*\*   
## lakenameEast Long Lake -7.3987 0.6918 -10.695 < 2e-16 \*\*\*  
## lakenameHummingbird Lake -6.8931 0.9429 -7.311 2.87e-13 \*\*\*  
## lakenamePaul Lake -3.8522 0.6656 -5.788 7.36e-09 \*\*\*  
## lakenamePeter Lake -4.3501 0.6645 -6.547 6.17e-11 \*\*\*  
## lakenameTuesday Lake -6.5972 0.6769 -9.746 < 2e-16 \*\*\*  
## lakenameWard Lake -3.2078 0.9429 -3.402 0.000672 \*\*\*  
## lakenameWest Long Lake -6.0878 0.6895 -8.829 < 2e-16 \*\*\*  
## ---  
## Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1  
##   
## Residual standard error: 7.355 on 9719 degrees of freedom  
## Multiple R-squared: 0.03953, Adjusted R-squared: 0.03874   
## F-statistic: 50 on 8 and 9719 DF, p-value: < 2.2e-16

1. Is there a significant difference in mean temperature among the lakes? Report your findings.

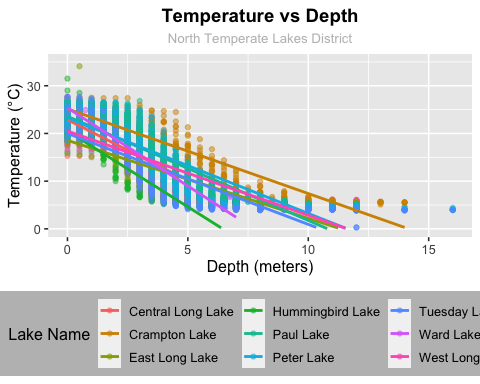
Answer: Our null hypothesis is that the mean temperature is the same across all lakes. Both the ANOVA and linear models give us p-values < 0.05 so we can reject the null hypothesis.

1. Create a graph that depicts temperature by depth, with a separate color for each lake. Add a geom\_smooth (method = “lm”, se = FALSE) for each lake. Make your points 50 % transparent. Adjust your y axis limits to go from 0 to 35 degrees. Clean up your graph to make it pretty.

#14.  
  
temp.depth.lakes.plot <-  
 ggplot(data = lake.chem.phys.July, aes(x = depth, y = temperature\_C, color = lakename)) +  
 geom\_point(alpha = 0.5) +  
 geom\_smooth(method = "lm", se = FALSE) +  
 ylim(0, 35) +  
 labs(x = "Depth (meters)", y = "Temperature (°C)", title = "Temperature vs Depth", subtitle = "North Temperate Lakes District", color = "Lake Name") +  
 guides(color = guide\_legend(nrow = 3))  
  
temp.depth.lakes.plot

## `geom\_smooth()` using formula 'y ~ x'

## Warning: Removed 73 rows containing missing values (geom\_smooth).



1. Use the Tukey’s HSD test to determine which lakes have different means.

#15  
  
TukeyHSD(temp.lakes.July.ANOVA)

## Tukey multiple comparisons of means  
## 95% family-wise confidence level  
##   
## Fit: aov(formula = temperature\_C ~ lakename, data = lake.chem.phys.July)  
##   
## $lakename  
## diff lwr upr p adj  
## Crampton Lake-Central Long Lake -2.3145195 -4.7031913 0.0741524 0.0661566  
## East Long Lake-Central Long Lake -7.3987410 -9.5449411 -5.2525408 0.0000000  
## Hummingbird Lake-Central Long Lake -6.8931304 -9.8184178 -3.9678430 0.0000000  
## Paul Lake-Central Long Lake -3.8521506 -5.9170942 -1.7872070 0.0000003  
## Peter Lake-Central Long Lake -4.3501458 -6.4115874 -2.2887042 0.0000000  
## Tuesday Lake-Central Long Lake -6.5971805 -8.6971605 -4.4972005 0.0000000  
## Ward Lake-Central Long Lake -3.2077856 -6.1330730 -0.2824982 0.0193405  
## West Long Lake-Central Long Lake -6.0877513 -8.2268550 -3.9486475 0.0000000  
## East Long Lake-Crampton Lake -5.0842215 -6.5591700 -3.6092730 0.0000000  
## Hummingbird Lake-Crampton Lake -4.5786109 -7.0538088 -2.1034131 0.0000004  
## Paul Lake-Crampton Lake -1.5376312 -2.8916215 -0.1836408 0.0127491  
## Peter Lake-Crampton Lake -2.0356263 -3.3842699 -0.6869828 0.0000999  
## Tuesday Lake-Crampton Lake -4.2826611 -5.6895065 -2.8758157 0.0000000  
## Ward Lake-Crampton Lake -0.8932661 -3.3684639 1.5819317 0.9714459  
## West Long Lake-Crampton Lake -3.7732318 -5.2378351 -2.3086285 0.0000000  
## Hummingbird Lake-East Long Lake 0.5056106 -1.7364925 2.7477137 0.9988050  
## Paul Lake-East Long Lake 3.5465903 2.6900206 4.4031601 0.0000000  
## Peter Lake-East Long Lake 3.0485952 2.2005025 3.8966879 0.0000000  
## Tuesday Lake-East Long Lake 0.8015604 -0.1363286 1.7394495 0.1657485  
## Ward Lake-East Long Lake 4.1909554 1.9488523 6.4330585 0.0000002  
## West Long Lake-East Long Lake 1.3109897 0.2885003 2.3334791 0.0022805  
## Paul Lake-Hummingbird Lake 3.0409798 0.8765299 5.2054296 0.0004495  
## Peter Lake-Hummingbird Lake 2.5429846 0.3818755 4.7040937 0.0080666  
## Tuesday Lake-Hummingbird Lake 0.2959499 -1.9019508 2.4938505 0.9999752  
## Ward Lake-Hummingbird Lake 3.6853448 0.6889874 6.6817022 0.0043297  
## West Long Lake-Hummingbird Lake 0.8053791 -1.4299320 3.0406903 0.9717297  
## Peter Lake-Paul Lake -0.4979952 -1.1120620 0.1160717 0.2241586  
## Tuesday Lake-Paul Lake -2.7450299 -3.4781416 -2.0119182 0.0000000  
## Ward Lake-Paul Lake 0.6443651 -1.5200848 2.8088149 0.9916978  
## West Long Lake-Paul Lake -2.2356007 -3.0742314 -1.3969699 0.0000000  
## Tuesday Lake-Peter Lake -2.2470347 -2.9702236 -1.5238458 0.0000000  
## Ward Lake-Peter Lake 1.1423602 -1.0187489 3.3034693 0.7827037  
## West Long Lake-Peter Lake -1.7376055 -2.5675759 -0.9076350 0.0000000  
## Ward Lake-Tuesday Lake 3.3893950 1.1914943 5.5872956 0.0000609  
## West Long Lake-Tuesday Lake 0.5094292 -0.4121051 1.4309636 0.7374387  
## West Long Lake-Ward Lake -2.8799657 -5.1152769 -0.6446546 0.0021080

temp.lakes.totals <-   
 HSD.test(temp.lakes.July.ANOVA, "lakename", group = TRUE)  
temp.lakes.totals

## $statistics  
## MSerror Df Mean CV  
## 54.1016 9719 12.72087 57.82135  
##   
## $parameters  
## test name.t ntr StudentizedRange alpha  
## Tukey lakename 9 4.387504 0.05  
##   
## $means  
## temperature\_C std r Min Max Q25 Q50 Q75  
## Central Long Lake 17.66641 4.196292 128 8.9 26.8 14.400 18.40 21.000  
## Crampton Lake 15.35189 7.244773 318 5.0 27.5 7.525 16.90 22.300  
## East Long Lake 10.26767 6.766804 968 4.2 34.1 4.975 6.50 15.925  
## Hummingbird Lake 10.77328 7.017845 116 4.0 31.5 5.200 7.00 15.625  
## Paul Lake 13.81426 7.296928 2660 4.7 27.7 6.500 12.40 21.400  
## Peter Lake 13.31626 7.669758 2872 4.0 27.0 5.600 11.40 21.500  
## Tuesday Lake 11.06923 7.698687 1524 0.3 27.7 4.400 6.80 19.400  
## Ward Lake 14.45862 7.409079 116 5.7 27.6 7.200 12.55 23.200  
## West Long Lake 11.57865 6.980789 1026 4.0 25.7 5.400 8.00 18.800  
##   
## $comparison  
## NULL  
##   
## $groups  
## temperature\_C groups  
## Central Long Lake 17.66641 a  
## Crampton Lake 15.35189 ab  
## Ward Lake 14.45862 bc  
## Paul Lake 13.81426 c  
## Peter Lake 13.31626 c  
## West Long Lake 11.57865 d  
## Tuesday Lake 11.06923 de  
## Hummingbird Lake 10.77328 de  
## East Long Lake 10.26767 e  
##   
## attr(,"class")  
## [1] "group"

16.From the findings above, which lakes have the same mean temperature, statistically speaking, as Peter Lake? Does any lake have a mean temperature that is statistically distinct from all the other lakes?

Answer: Paul Lake and Ward Lake do not have significantly different mean temperatures from Peter Lake. No lakes have a mean temperature that is statistically distinct from all the other lakes.

1. If we were just looking at Peter Lake and Paul Lake. What’s another test we might explore to see whether they have distinct mean temperatures?

Answer: We could use a two-way ANOVA so that we can include two categorical variables.