

# Assignment 6: GLMs (Linear Regressios, ANOVA, & t-tests)

Sam Saltman

## 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 Monday, February 28 at 7:00 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
getwd()

## [1] "/Users/samsaltman/Documents/R/Environmental_Data_Analytics_2022"

library(tidyverse)

## — Attaching packages ————— tidyverse
1.3.1 —

## ✓ ggplot2 3.3.5      ✓ purrr   0.3.4
## ✓ tibble  3.1.4      ✓ dplyr   1.0.7
## ✓ tidyr   1.1.3      ✓ stringr 1.4.0
## ✓ readr   2.0.1      ✓ forcats 0.5.1

## — Conflicts —————
tidyverse_conflicts() —
```

```
## x dplyr::filter() masks stats::filter()
## x dplyr::lag() masks stats::lag()

library(lubridate)

##
## Attaching package: 'lubridate'

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

library(agricolae)

raw_chemistry_physics <- read.csv("./Data/Raw/NTL-
LTER_Lake_ChemistryPhysics_Raw.csv", stringsAsFactors = TRUE)

raw_chemistry_physics$sampdate <- as.Date(raw_chemistry_physics$sampdate
, format = "%m/%d/%y")

#2
mytheme <- theme_classic(base_size = 14) +
  theme(axis.text = element_text(color = "black"),
        legend.position = "right")
theme_set(mytheme)
```

## Simple regression

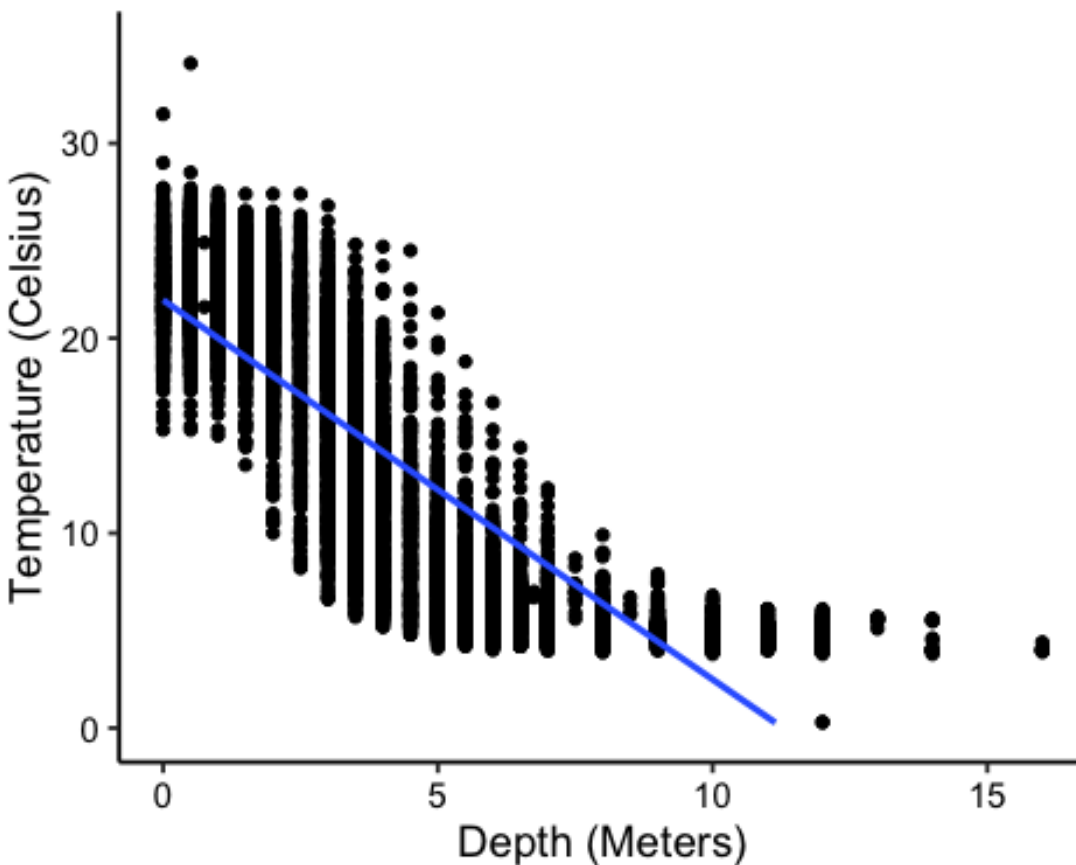
Our first research question is: Does mean lake temperature recorded during July change with depth across all lakes?

3. State the null and alternative hypotheses for this question: > Answer: H0: Lake mean temperature during July remains the same across depths in all lakes Ha: Lake mean temperature during July changes across depths in all lakes
4. 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)
5. 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
wrangled_chem_physics <- raw_chemistry_physics %>%
  mutate(month = month(sampdate)) %>%
  filter(month == '7') %>%
  select(`lakename`, `year4`, `daynum`, `depth`, `temperature_C`) %>%
  na.omit()
```

```
#5
wrangled_visualized <- ggplot(wrangled_chem_physics) +
  aes(x = depth, y = temperature_C) +
  geom_point() +
  geom_smooth(method = 'lm') +
  ylim(0, 35) +
  ylab("Temperature (Celsius)") +
  xlab("Depth (Meters)")
print(wrangled_visualized)

## `geom_smooth()` using formula 'y ~ x'
## Warning: Removed 24 rows containing missing values (geom_smooth).
```



6. 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: There appears to be a negative linear correlation between temperature and depth. The temperature is warmer closer to the surface of the lakes.

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

```

#7
wrangled_linear_regression <- lm(data = wrangled_chem_physics, temperature_C
~ depth)
summary(wrangled_linear_regression)

##
## Call:
## lm(formula = temperature_C ~ depth, data = wrangled_chem_physics)
##
## 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

```

8. 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: The model indicates that there is a significant relationship between temperature and depth. The adjusted  $R^2$  value calculates that 73.87% of the variability in temperature is explained by changes in depth. This is based on 9726 degrees of freedom and a p-value less than 5%. Assuming that there is not any bias in the sampling, the relationship between depth and temperature is statistically significant. Temperature is predicted to change -1.94621 for every meter increase in depth.

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## 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.

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

#9

```
wrangled_LM_AIC <- lm(data = wrangled_chem_physics, temperature_C ~ year4 +  
daynum + depth)  
wrangled_AIC <- step(wrangled_LM_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
```

#10

```
Recommended_multiple_regression <- lm(data = wrangled_chem_physics,  
temperature_C ~ year4 + daynum + depth) #no changes are made through AIC step  
method.  
summary(Recommended_multiple_regression)
```

```
##  
## Call:  
## lm(formula = temperature_C ~ year4 + daynum + depth, data =  
wrangled_chem_physics)  
##  
## 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
```

11. 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 keeps year4, daynum and depth. This model explains 74.11% of the variance. This is a slight improvement over just accounting for

depth, but it is clear that depth is the most important variable in predicting temperature.

---

## Analysis of Variance

12. 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
anova_laketemps <- aov(data = wrangled_chem_physics, temperature_C ~
lakename)
summary(anova_laketemps)

##              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

lm_laketemps <- lm(data = wrangled_chem_physics, temperature_C ~ lakename)
summary(lm_laketemps)

##
## Call:
## lm(formula = temperature_C ~ lakename, data = wrangled_chem_physics)
##
## 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
```

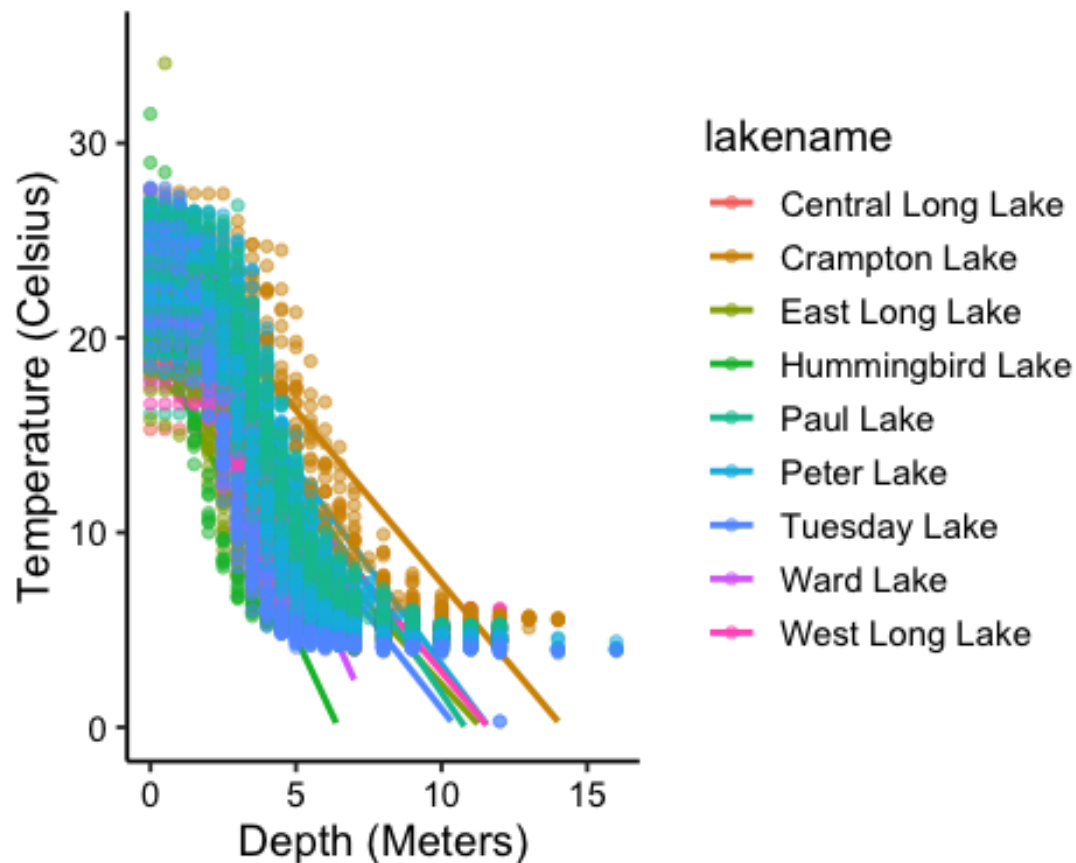
13. Is there a significant difference in mean temperature among the lakes? Report your findings.

Answer: Yes, the ANOVA shows that overall there is a statistically significant difference in temperatures between lakes. The LM shows the same overall statistical difference as well as each lake individually. The LM shows that each lake on its own has a statistically significant temperature difference.

14. 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.

```
Depth_Temp_Graph <- ggplot(wrangled_chem_physics) +  
  aes(x = depth, y = temperature_C, color = lakename) +  
  geom_smooth(se = FALSE, method = "lm") +  
  geom_point(alpha = 0.5) +  
  ylim(0, 35) +  
  ylab("Temperature (Celsius)") +  
  xlab("Depth (Meters)")  
print(Depth_Temp_Graph)  
  
## `geom_smooth()` using formula 'y ~ x'  
  
## Warning: Removed 73 rows containing missing values (geom_smooth).
```



15. Use the Tukey's HSD test to determine which lakes have different means.

```
#15
TukeyHSD(anova_laketemps)

##    Tukey multiple comparisons of means
##      95% family-wise confidence level
##
## Fit: aov(formula = temperature_C ~ lakename, data = wrangled_chem_physics)
##
## $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

HSDTest <- HSD.test(anova_laketemps, "lakename", group = TRUE)
HSDTest

## $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: Ward Lake and Paul Lake have the same mean temperature statistically speaking. No lake has a mean temperature that is statistically distinct from all the other lakes.

17. 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: A simple T test could work