

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

Addie Navarro

## 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
library(tidyverse)

## -- Attaching packages ----- tidyverse 1.3.1 --
## v ggplot2 3.3.5      v purrr 0.3.4
## v tibble 3.1.6       v dplyr 1.0.7
## v tidyr 1.1.4        v stringr 1.4.0
## v readr 2.1.1       v forcats 0.5.1

## -- 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
```

```
getwd()

## [1] "Z:/EDA/Environmental_Data_Analytics_2022"

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

#set date columns as date objects
NTL_LTER_RAW$sampleddate <- as.Date(NTL_LTER_RAW$sampleddate, format = "%m/%d/%y")

#2 Build ggplot theme and set as default theme

my_theme <- theme_classic(base_size = 14) + theme(axis.text = element_text(color = "black"),
legend.position = "top")
theme_set(my_theme)
```

## Simple regression

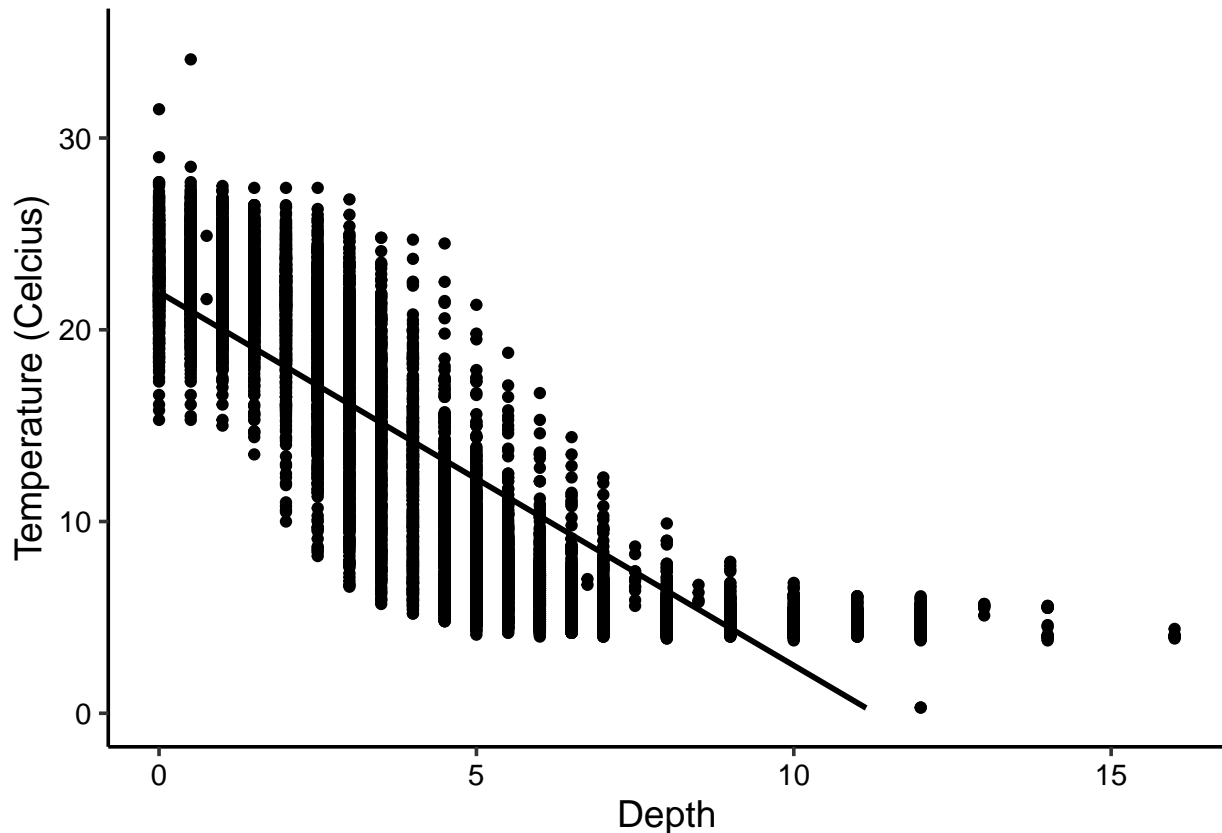
Our first research question is: Does mean lake temperature recorded during July change with depth across all lakes?

3. State the null and alternative hypotheses for this question: > Answer: H0: The mean lake temperature recorded during July does not change with depth across all lakes.  
Ha: The mean temperature recorded during July changes with depth across 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
NTL_temp_depth_July <- NTL_LTER_RAW %>%
  mutate(Month = month(sampleddate))%>%
  filter(Month %in% c("7"))%>%
  select('lakename', 'year4', 'daynum', 'depth', 'temperature_C')%>%
  na.omit()

#5
Temp_depth_Scatter <-
  ggplot(NTL_temp_depth_July, aes(x = depth, y = temperature_C))+
  geom_point()+
  geom_smooth(method = "lm", color = "black")+
  xlab("Depth")+
  ylab("Temperature (Celcius)")+
  ylim(0,35)
print(Temp_depth_Scatter)

## `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 anything about the linearity of this trend?

Answer: The scatterplot shows a negative correlation between temperature and depth. As depth increases, temperature decreases. The distribution of the points around the line, particularly at the shallower depths, demonstrate a linear relationship between temperature and depth.

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

```
#7
Temp_Depth_Regression <- lm(data = NTL_temp_depth_July, temperature_C ~ depth)
summary(Temp_Depth_Regression)
```

```
##
## Call:
## lm(formula = temperature_C ~ depth, data = NTL_temp_depth_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
```

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 p-value of the model summary is less than .05, so is statistically significant. Therefore, we can reject the null hypothesis that states that the mean recorded lake temperature does not change with depth across all lakes. The R-squared value shows that 73.87% of the variability in temperature data is explained by changes in depth. There are 9726 degrees of freedom on which this finding is based, with a low standard error of 3.835. Temperature is expected to decrease by 1.95 degrees celcius for every 1m change in depth

---

## 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
TempAIC <- lm(data = NTL_temp_depth_July, temperature_C ~ depth + year4 + daynum)

step(TempAIC)

## Start:  AIC=26065.53
## temperature_C ~ depth + year4 + daynum
##
##           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 ~ depth + year4 + daynum, data = NTL_temp_depth_July)
##
## Coefficients:
## (Intercept)      depth      year4      daynum
##   -8.57556    -1.94644     0.01134     0.03978

summary(TempAIC)

##
## Call:
## lm(formula = temperature_C ~ depth + year4 + daynum, data = NTL_temp_depth_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
## depth       -1.946437   0.011683 -166.611 < 2e-16 ***
## year4        0.011345   0.004299   2.639  0.00833 **
## daynum       0.039780   0.004317   9.215 < 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

#10
temp_final_variables <- lm(data = NTL_temp_depth_July, temperature_C ~ depth + year4 + daynum)
summary(temp_final_variables)

##
## Call:
## lm(formula = temperature_C ~ depth + year4 + daynum, data = NTL_temp_depth_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
## depth       -1.946437   0.011683 -166.611 < 2e-16 ***
## year4        0.011345   0.004299   2.639  0.00833 **
## daynum       0.039780   0.004317   9.215 < 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 final set of explanatory variables suggested by the AIC method is year4, daynum, and depth. This model explains 74% of the observed variance. Numerically, this is a slight improvement over the previous model using only depth as the explanatory variable, but perhaps adding in additional explanatory variables for such a small gain in explanation of observed variance adds more complication to the model than it's worth. It might be better to keep it simple, or use just depth and daynum as subtracting year4 from the regression showed only a slight increase in AIC (from 26,066 to 26,070).

---

## 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 model
lake_temp.anova <- aov(data=NTL_temp_depth_July, temperature_C ~ lakename)
summary(lake_temp.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

#linear model
lake_temp.anova2 <- lm(data = NTL_temp_depth_July, temperature_C ~ lakename)
summary(lake_temp.anova2)

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

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

Answer: According to the ANOVA test formatted as aov and formatted as a linear model, there is a significant difference in mean temperature among the lakes. The p-value is less than .05, so we can reject the null hypothesis that the mean temperature is the same across all lakes.

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.

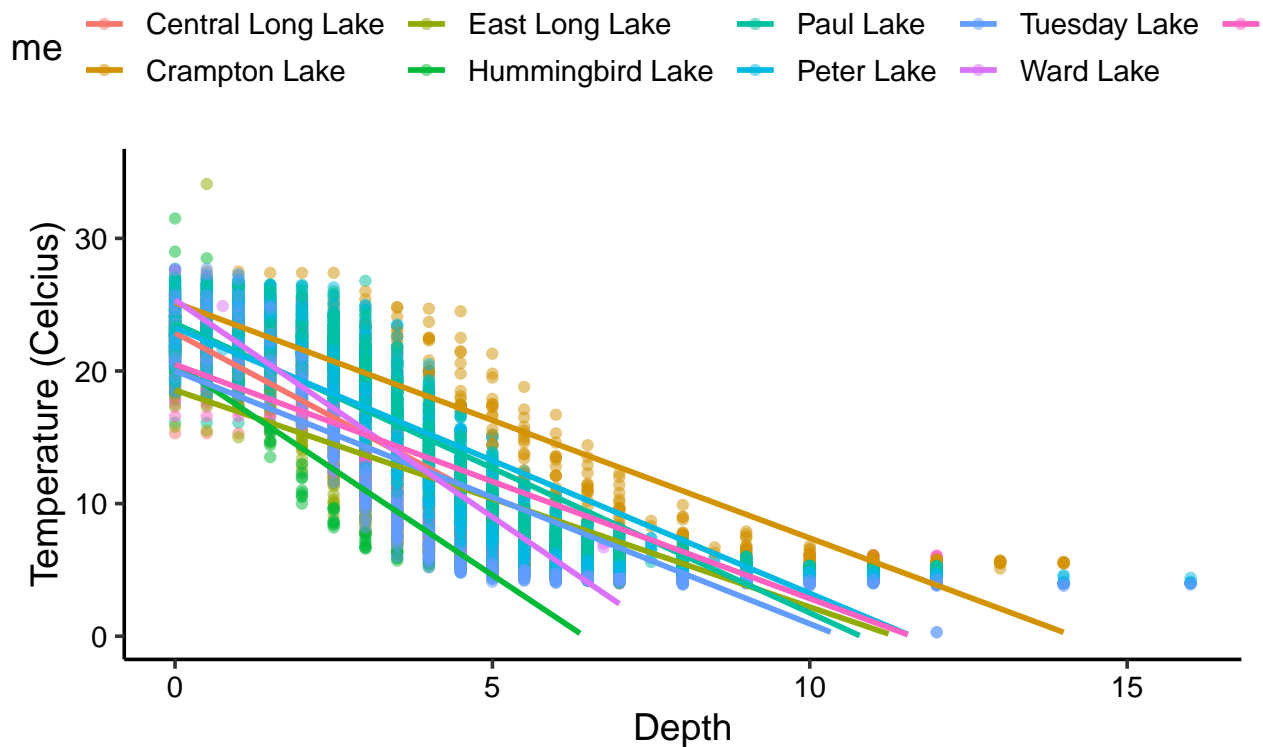
```
#14.
Temp_Lakename.plot <- ggplot(NTL_temp_depth_July, aes(x = depth, y = temperature_C, color = lakename))+
  geom_point(alpha = .5)+ #50% transparent
  geom_smooth(method = "lm", se = FALSE)+
```

```
ylim(0,35)+
ylab("Temperature (Celcius)")+
xlab("Depth")+
labs(color = "Lake Name", title = "Lake Temperature by Depth")
print(Temp_Lakename.plot)
```

```
## `geom_smooth()` using formula 'y ~ x'
```

```
## Warning: Removed 73 rows containing missing values (geom_smooth).
```

## Lake Temperature by Depth



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

```
#15
```

```
TukeyHSD(lake_temp.anova)
```

```
## Tukey multiple comparisons of means
```

```
## 95% family-wise confidence level
```

```
##
```

```
## Fit: aov(formula = temperature_C ~ lakename, data = NTL_temp_depth_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
```

```
Lake_Temp_Groups <- HSD.test(lake_temp.anova, "lakename", group = TRUE)
Lake_Temp_Groups
```

```
## $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: From the Tukey HSD test, we find that Ward Lake and Paul Lake have the same mean temperature, statistically speaking, as Peter Lake. None of the lakes have a mean temperature that is statistically distinct from all 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: We could possibly use a two-way ANOVA test to see whether Peter and Paul lake have distinct mean temperatures because there would be two categorical explanatory variables (lakename) and one continuous variable.