

Assignment 6: 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 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/ataliefischer/Desktop/EDA/Environmental_Data_Analytics_2022/Assignments"
setwd("/Users/ataliefischer/Desktop/EDA/Environmental_Data_Analytics_2022")

library(tidyverse)

## -- Attaching packages ----- tidyverse 1.3.1 --
## v ggplot2 3.3.5      v purrr  0.3.4
## v tibble  3.1.4      v dplyr  1.0.7
## v tidyr   1.1.3      v stringr 1.4.0
## v readr   2.0.1      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

LTERraw.data <- read.csv("./Data/Raw/NTL-LTER_Lake_ChemistryPhysics_Raw.csv", stringsAsFactors = TRUE)
LTERraw.data$sampleddate <- as.Date(LTERraw.data$sampleddate, format = "%m/%d/%y")

#2
mytheme <- theme_classic(base_size = 12) +
  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?

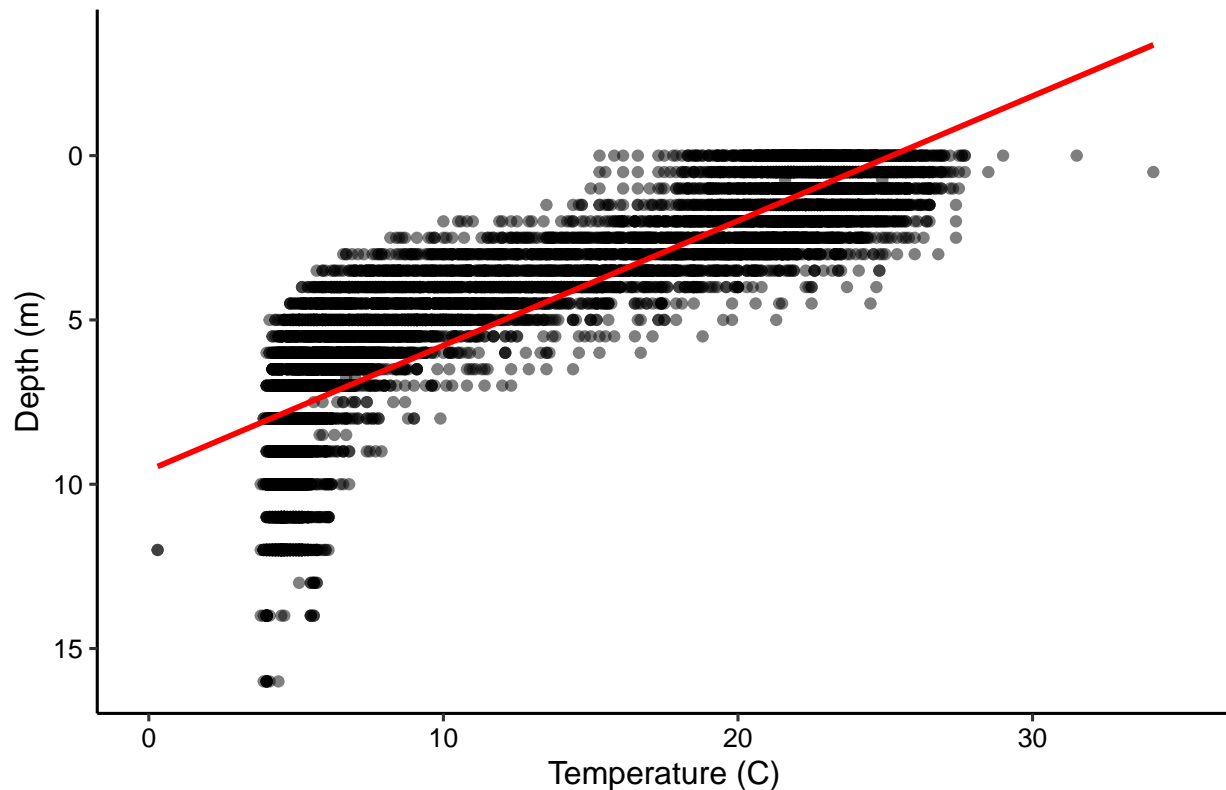
3. State the null and alternative hypotheses for this question: > Answer: H0: Mean lake temperature recorded during July has the equal depths across all lakes. Ha: Mean lake temperature recorded during July has different depths across all lakes.
4. Wrangle your NTL-LTER dataset with a pipe function so 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
LTERdata.filtered <- LTERraw.data %>%
  mutate(month = month(sampledate)) %>%
  filter(month == 7) %>%
  select(lakename, year4, daynum, depth, temperature_C) %>%
  na.omit()

#5
ggplot(LTERdata.filtered, aes(x = temperature_C, y = depth)) +
  geom_point(alpha = 0.50) +
  geom_smooth(method = lm, color = "red") +
  scale_y_reverse() +
  xlim(0, 35) +
  labs(title = "Temperature vs. Depth in North Temperate Lakes",
       x = "Temperature (C)",
       y = "Depth (m)") +
  theme(plot.title = element_text(hjust = 0.5))

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

Temperature vs. Depth in North Temperate Lakes



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: Temperature generally decreases as depth increases. The distribution of points do not appear evenly scattered about the linear model, suggesting that the relationship between temperature and depth is non-linear.

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

#7

```
LTER.lm <- lm(data = LTERdata.filtered, temperature_C ~ depth)
summary(LTER.lm)
```

```
##
## Call:
## lm(formula = temperature_C ~ depth, data = LTERdata.filtered)
##
## 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 resulting from the linear regression is $\text{temperature} = -1.946(\text{depth}) + 21.956$. In other words, with every meter increase in depth, temperature drops by 1.946 degrees celcius. With a p-value $< 2e-16$, this relationship is significant. The residual standard error is 3.84 on 9726 degrees of freedom. The multiple R-squared value is 0.739, suggesting that approximately 73.9% of the variation in temperature can be explained by 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

```
LTER.AIC <- lm(data = LTERdata.filtered, temperature_C ~ depth + year4 + daynum)
step(LTER.AIC)
```

```
## 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 = LTERdata.filtered)
##
## Coefficients:
## (Intercept)      depth      year4      daynum
##   -8.57556    -1.94644     0.01134     0.03978
```

#10

```
summary(LTER.AIC)

##
## Call:
## lm(formula = temperature_C ~ depth + year4 + daynum, data = LTERdata.filtered)
##
## 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 AIC method suggests we use all the variables (depth, year, and day) to predict temperature because the regression that included all these variables had the lowest AIC value. With an R-squared value of 0.7412, approximately 74.12% of the variation in temperature can be explained by the interaction between depth, year, and day. This value is slightly better than our original R-squared value.

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

```
LTERR.anova <- aov(data = LTERRdata.filtered, temperature_C ~ lakename)
summary(LTERR.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
```

```
LTERR.anova.lm <- lm(data = LTERRdata.filtered, temperature_C ~ lakename)
summary(LTERR.anova.lm)
```

```
##
## Call:
## lm(formula = temperature_C ~ lakename, data = LTERRdata.filtered)
##
## 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 ***
```

```
## lakenamCrampton Lake      -2.3145      0.7699   -3.006 0.002653 **
## lakenamEast Long Lake    -7.3987      0.6918  -10.695 < 2e-16 ***
## lakenamHummingbird Lake  -6.8931      0.9429   -7.311 2.87e-13 ***
## lakenamPaul Lake         -3.8522      0.6656   -5.788 7.36e-09 ***
## lakenamPeter Lake        -4.3501      0.6645   -6.547 6.17e-11 ***
## lakenamTuesday Lake     -6.5972      0.6769   -9.746 < 2e-16 ***
## lakenamWard Lake         -3.2078      0.9429   -3.402 0.000672 ***
## lakenamWest 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.

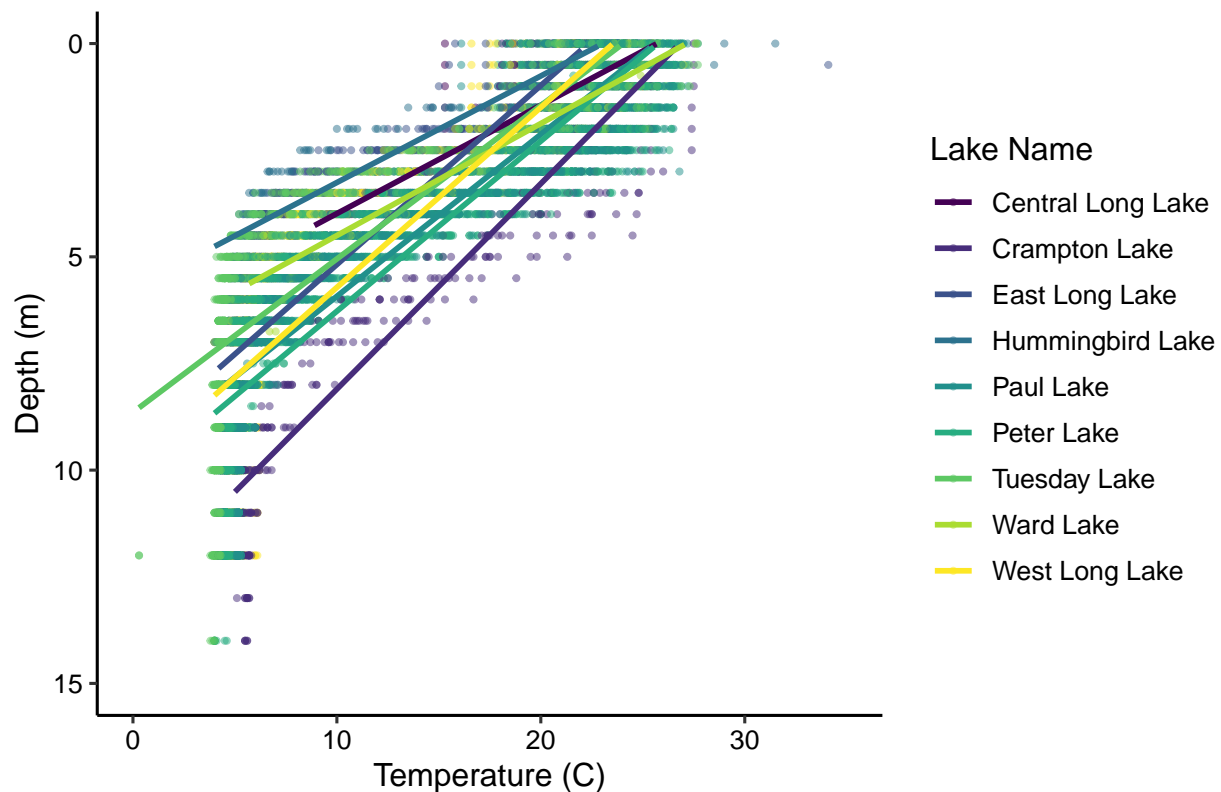
Answer: The ANOVA results show that there is a significant difference in mean temperature among the lakes with a p-value < 2e-16. The linear model results show that the mean temperature of each lake is significantly different the others', with p-value < 2e-16. The residual error is 7.355 on 9719 degrees of freedom. The multiple R-squared value is 0.0395, suggesting that 3.95% of the variance in temperature is explained by 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
ggplot(LTERdata.filtered, aes(x = temperature_C, y = depth, color = lakenam)) +
  geom_point(size = 0.75, alpha = 0.50) +
  geom_smooth(aes(color = lakenam), method = lm, se = FALSE) +
  xlim(0, 35) +
  scale_y_reverse(limits = c(15, 0)) +
  scale_color_viridis_d() +
  labs(title = "Temperature vs. Depth by Lake in North Temperate Lakes",
       x = "Temperature (C)",
       y = "Depth (m)",
       color = "Lake Name") +
  theme(plot.title = element_text(hjust = 0.5))
```

```
## `geom_smooth()` using formula 'y ~ x'
## Warning: Removed 9 rows containing non-finite values (stat_smooth).
## Warning: Removed 9 rows containing missing values (geom_point).
## Warning: Removed 99 rows containing missing values (geom_smooth).
```

Temperature vs. Depth by Lake in North Temperate Lakes



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

#15

```
LTER.tukey <- TukeyHSD(LTER.anova)
LTER.tukey
```

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

```
LTER.groups <- HSD.test(LTER.anova, "lakename", group = TRUE)
LTER.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: Paul Lake and Ward Lake statistically have the same mean temperature as Peter Lake. There are no lakes that have 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: t-test