3. Worksheet: Basic R

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OVERVIEW

This worksheet introduces some of the basic features of the R computing environment (http://www.r-project.org). It is designed to be used along side the **3. RStudio** handout in your binder. You will not be able to complete the exercises without the corresponding handout.

Directions:

- 1. In the Markdown version of this document in your cloned repo, change "Student Name" on line 3 (above) with your name.
- 2. Complete as much of the worksheet as possible during class.
- 3. Use the handout as a guide; it contains a more complete description of data sets along with examples of proper scripting needed to carry out the exercises.
- 4. Answer questions in the worksheet. Space for your answers is provided in this document and is indicated by the ">" character. If you need a second paragraph be sure to start the first line with ">". You should notice that the answer is highlighted in green by RStudio (color may vary if you changed the editor theme).
- 5. Before you leave the classroom today, it is *imperative* that you **push** this file to your GitHub repo, at whatever stage you are. The will enable you to pull your work onto your own computer.
- 6. When you have completed the worksheet, **Knit** the text and code into a single PDF file by pressing the Knit button in the RStudio scripting panel. This will save the PDF output in your '3.RStudio' folder.
- 7. After Knitting, please submit the worksheet by making a **push** to your GitHub repo and then create a **pull request** via GitHub. Your pull request should include this file (**3.RStudio_Worksheet.Rmd**) with all code blocks filled out and questions answered) and the PDF output of Knitr (**3.RStudio_Worksheet.pdf**).

The completed exercise is due on Wednesday, March 24th, 2021 before 12:00 PM (noon).

1) HOW WE WILL BE USING R AND OTHER TOOLS

You are working in an RMarkdown (.Rmd) file. This allows you to integrate text and R code into a single document. There are two major features to this document: 1) Markdown formatted text and 2) "chunks" of R code. Anything in an R code chunk will be interpreted by R when you *Knit* the document.

When you are done, you will *knit* your document together. However, if there are errors in the R code contained in your Markdown document, you will not be able to knit a PDF file. If this happens, you will need to review your code, locate the source of the error(s), and make the appropriate changes. Even if you are able to knit without issue, you should review the knitted document for correctness and completeness before you submit the Worksheet. Next to the Knit button in the RStudio scripting panel there is a spell checker button (ABC) button.

2) SETTING YOUR WORKING DIRECTORY

In the R code chunk below, please provide the code to: 1) clear your R environment, 2) print your current working directory, and 3) set your working directory to your '3.RStudio' folder.

```
rm(list=ls())
getwd()

## [1] "/Users/tbiewerh/GitHub/QB2021_Biewer-Heisler/2.Worksheets/3.RStudio"
setwd("~/GitHub/QB2021_Biewer-Heisler/2.Worksheets/3.RStudio")
```

3) USING R AS A CALCULATOR

To follow up on the pre-class exercises, please calculate the following in the R code chunk below. Feel free to reference the 1. Introduction to version control and computing tools handout.

- 1) the volume of a cube with length, $l_1 = 5$ (volume = l^3)
- 2) the area of a circle with radius, $r_1 = 2$ (area = $pi * r^2$).
- 3) the length of the opposite side of a right-triangle given that the angle, theta, = pi/4. (radians, a.k.a. 45°) and with hypotenuse length sqrt(2) (remember: $\sin(\text{theta}) = \text{opposite/hypotenuse}$).
- 4) the log (base e) of your favorite number.

```
## [1] 125
volume = 125
pi * 2^2
## [1] 12.56637
area = 12.56637
opposite = sin(pi/4)*sqrt(2)
opposite = 1
log(7)
## [1] 1.94591
answer = 1.94591
```

4) WORKING WITH VECTORS

To follow up on the pre-class exercises, please perform the requested operations in the R-code chunks below.

Basic Features Of Vectors

In the R-code chunk below, do the following: 1) Create a vector \mathbf{x} consisting of any five numbers. 2) Create a new vector \mathbf{w} by multiplying \mathbf{x} by 14 (i.e., "scalar"). 3) Add \mathbf{x} and \mathbf{w} and divide by 15.

```
x \leftarrow c(2,4,6,8,10)

w \leftarrow x * 14

(x+w)/15
```

```
## [1] 2 4 6 8 10
```

Now, do the following: 1) Create another vector (k) that is the same length as w. 2) Multiply k by x. 3) Use the combine function to create one more vector, d that consists of any three elements from w and any four elements of k.

```
k <- c(1,2,3,4,5)
k*x
## [1] 2 8 18 32 50
d <- c(w[1:3], k[1:4])
```

Summary Statistics of Vectors

In the R-code chunk below, calculate the **summary statistics** (i.e., maximum, minimum, sum, mean, median, variance, standard deviation, and standard error of the mean) for the vector (v) provided.

```
sem <- function(x){</pre>
  sd(x)/sqrt(length(x))
}
v \leftarrow c(16.4, 16.0, 10.1, 16.8, 20.5, 20.2, 13.1, 24.8, 20.2, 25.0, 20.5, 30.5, 31.4, 27.1)
max(v)
## [1] 31.4
min(v)
## [1] 10.1
sum(v)
## [1] 292.6
mean(v)
## [1] 20.9
median(v)
## [1] 20.35
var(v)
## [1] 39.44
sd(v)
## [1] 6.280127
sem(v)
## [1] 1.678435
```

5) WORKING WITH MATRICES

In the R-code chunk below, do the following: Using a mixture of Approach 1 and 2 from the **3. RStudio** handout, create a matrix with two columns and five rows. Both columns should consist of random numbers. Make the mean of the first column equal to 8 with a standard deviation of 2 and the mean of the second column equal to 25 with a standard deviation of 10.

```
j <- c(rnorm(5, mean = 8, sd = 2))
z <- c(rnorm(5, mean = 25, sd = 10))
jz <- cbind(j,z)</pre>
```

Question 1: What does the rnorm function do? What do the arguments in this function specify? Remember to use help() or type?rnorm.

Answer 1: The rnorm function generates random numbers that are normally distributed. The first number gives the amount of numbers to be generated. 'mean' sets what the average of these numbers will be, and 'sd' determines what the standard deviation of the set of numbers will be.

In the R code chunk below, do the following: 1) Load matrix.txt from the 3.RStudio data folder as matrix m. 2) Transpose this matrix. 3) Determine the dimensions of the transposed matrix.

```
m <- as.matrix(read.table("data/matrix.txt"), sep = "\t", header = FALSE)
n <- t(m)</pre>
```

 $Question\ 2$: What are the dimensions of the matrix you just transposed?

Answer 2: There are 5 rows and 10 columns.

###Indexing a Matrix

In the R code chunk below, do the following: 1) Index matrix m by selecting all but the third column. 2) Remove the last row of matrix m.

```
n <- m[, c(1:2, 4:5)]
o <- m[1:9,]
```

6) BASIC DATA VISUALIZATION AND STATISTICAL ANALYSIS

Load Zooplankton Data Set

In the R code chunk below, do the following: 1) Load the zooplankton data set from the **3.RStudio** data folder. 2) Display the structure of this data set.

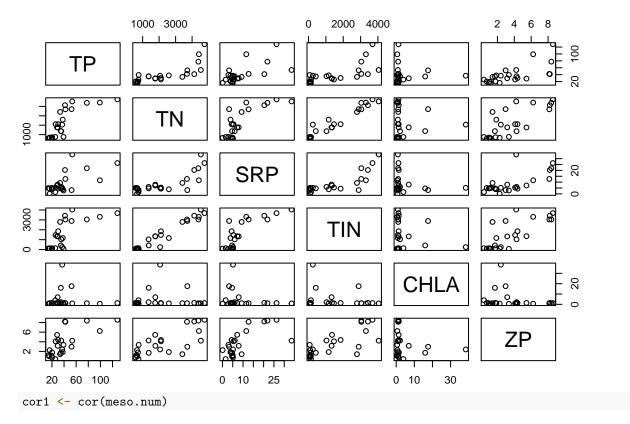
```
meso <- read.table("data/zoop_nuts.txt", sep = "\t", header = TRUE)
str(meso)</pre>
```

```
## 'data.frame':
                   24 obs. of 8 variables:
##
   $ TANK: int
                34 14 23 16 21 5 25 27 30 28 ...
                "L" "L" "L" "L" ...
##
   $ NUTS: chr
   $ TP : num
                20.3 25.6 14.2 39.1 20.1 ...
                720 750 610 761 570 ...
##
   $ TN : num
   $ SRP : num
                4.02 1.56 4.97 2.89 5.11 4.68 5 0.1 7.9 3.92 ...
   $ TIN : num 131.6 141.1 107.7 71.3 80.4 ...
  $ CHLA: num 1.52 4 0.61 0.53 1.44 1.19 0.37 0.72 6.93 0.94 ...
   $ ZP : num 1.781 0.409 1.201 3.36 0.733 ...
```

Correlation

In the R-code chunk below, do the following: 1) Create a matrix with the numerical data in the meso dataframe. 2) Visualize the pairwise **bi-plots** of the six numerical variables. 3) Conduct a simple **Pearson's correlation** analysis.

```
meso.num <- meso[,3:8]
pairs(meso.num)</pre>
```



Question 3: Describe some of the general features based on the visualization and correlation analysis above?

Answer 3:It seems that CHLA is not correlated with anything. The other values all seem to be

correlated with each other to some extent, but with some having logarithmic distributions.

In the R code chunk below, do the following: 1) Redo the correlation analysis using the corr.test() function in the psych package with the following options: method = "pearson", adjust = "BH". 2) Now, redo this correlation analysis using a non-parametric method. 3) Use the print command from the handout to see the results of each correlation analysis.

```
install.packages("psych", repos = "http://cran.rstudio.com/")
## Installing package into '/Users/tbiewerh/Library/R/4.0/library'
## (as 'lib' is unspecified)
##
## The downloaded binary packages are in
   /var/folders/rl/5fysqz596 v krmqljc490d52cnvv6/T//RtmphkKNCR/downloaded packages
require("psych")
## Loading required package: psych
cor2 <- corr.test(meso.num, method = "pearson", adjust = "BH")</pre>
print(cor2, digits = 3)
## Call:corr.test(x = meso.num, method = "pearson", adjust = "BH")
## Correlation matrix
##
            TP
                   TN
                         SRP
                                 TIN
                                       CHLA
                                                ΖP
                               0.717 -0.017
## TP
         1.000 0.787
                       0.654
                                             0.697
                               0.969 -0.004
  TN
         0.787
                1.000
                       0.784
                                             0.756
         0.654 0.784
                       1.000
                              0.801 - 0.189
                                             0.676
## SRP
```

```
0.717 0.969 0.801 1.000 -0.157 0.761
  CHLA -0.017 -0.004 -0.189 -0.157
                                     1.000 -0.183
               0.756  0.676  0.761 -0.183  1.000
         0.697
## Sample Size
## [1] 24
## Probability values (Entries above the diagonal are adjusted for multiple tests.)
##
           TP
                 TN
                      SRP
                            TIN
                                 CHLA
## TP
        0.000 0.000 0.001 0.000 0.983 0.000
##
  TN
        0.000 0.000 0.000 0.000 0.983 0.000
  SRP
       0.001 0.000 0.000 0.000 0.491 0.000
       0.000 0.000 0.000 0.000 0.536 0.000
  CHLA 0.938 0.983 0.376 0.464 0.000 0.491
##
        0.000 0.000 0.000 0.000 0.393 0.000
##
##
   To see confidence intervals of the correlations, print with the short=FALSE option
cor3 <- corr.test(meso.num, method = "kendall", adjust = "BH")</pre>
print(cor3, digits = 3)
## Call:corr.test(x = meso.num, method = "kendall", adjust = "BH")
## Correlation matrix
##
           TP
                 TN
                       SRP
                             TIN
                                    CHLA
                                             ZP
## TP
        1.000 0.739
                     0.391 0.577
                                  0.044
                                         0.536
        0.739 1.000
                                          0.551
## TN
                     0.478 0.809
                                  0.015
## SRP
       0.391 0.478
                     1.000 0.563 -0.066
## TIN
       0.577 0.809
                     0.563 1.000
                                  0.044
## CHLA 0.044 0.015 -0.066 0.044
                                  1.000 -0.051
        0.536 0.551 0.449 0.548 -0.051 1.000
## Sample Size
## [1] 24
## Probability values (Entries above the diagonal are adjusted for multiple tests.)
##
           TP
                      SRP
                            TIN CHLA
                 TN
                                          ZP
## TP
        0.000 0.000 0.088 0.014 0.899 0.015
  TN
        0.000 0.000 0.034 0.000 0.946 0.014
##
  SRP
       0.059 0.018 0.000 0.014 0.899 0.046
       0.003 0.000 0.004 0.000 0.899 0.014
## CHLA 0.839 0.946 0.760 0.839 0.000 0.899
        0.007 0.005 0.028 0.006 0.813 0.000
## ZP
##
   To see confidence intervals of the correlations, print with the short=FALSE option
```

Question 4: Describe what you learned from corr.test. Specifically, are the results sensitive to whether you use parametric (i.e., Pearson's) or non-parametric methods? When should one use non-parametric methods instead of parametric methods? With the Pearson's method, is there evidence for false discovery rate due to multiple comparisons? Why is false discovery rate important?

Answer 4: Some of the values do decrease in correlation when using non-parametric vs parametric. One should use non-parametric on non-normal data or with large sample sizes. It doesn't look like there's evidence for false discovery rate because the corrected values equal the uncorrected values. False discovery rate is important because you can determine if you may get false positives.

Linear Regression

In the R code chunk below, do the following: 1) Conduct a linear regression analysis to test the relationship between total nitrogen (TN) and zooplankton biomass (ZP). 2) Examine the output of the regression analysis. 3) Produce a plot of this regression analysis including the following: categorically labeled points, the predicted

regression line with 95% confidence intervals, and the appropriate axis labels.

```
fitreg <- lm(ZP ~ TN, data = meso)
summary(fitreg)
##
## Call:
## lm(formula = ZP ~ TN, data = meso)
##
## Residuals:
##
       Min
                 1Q Median
                                 3Q
                                         Max
##
   -3.7690 -0.8491 -0.0709
                             1.6238
                                      2.5888
##
## Coefficients:
##
                 Estimate Std. Error t value Pr(>|t|)
                                        1.074
## (Intercept) 0.6977712
                           0.6496312
               0.0013181
                           0.0002431
                                        5.421 1.91e-05 ***
##
                   0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Signif. codes:
##
## Residual standard error: 1.75 on 22 degrees of freedom
## Multiple R-squared: 0.5719, Adjusted R-squared: 0.5525
## F-statistic: 29.39 on 1 and 22 DF, p-value: 1.911e-05
plot(meso$TN, meso$ZP, ylim = c(0,10), xlim = c(500, 5000), xlab = expression(paste("Total Nitrogen (",
      10
Zooplankton Biomass (mg/L)
                                                                                 0
                                                                0
                                                                       0
       8
                                                                            0
       6
                                 0
                           0
                                       8
                                                                             0
       4
                  0
                                  0
                                                                0
                                      0
       2
                           0
                                                            0
                 0
       0
                    1000
                                   2000
                                                   3000
                                                                   4000
                                                                                   5000
                                       Total Nitrogen (μg/L)
```

Question 5: Interpret the results from the regression model

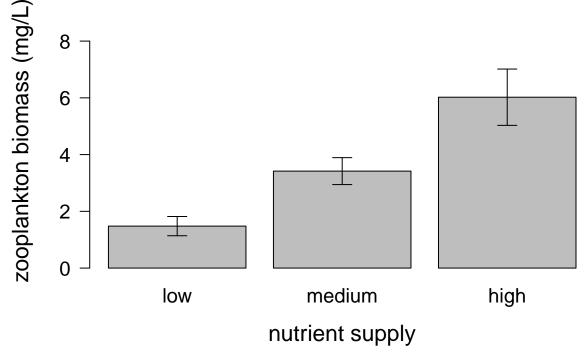
Answer 5: As total nitrogen increases zooplankton mass increases linearly.

Analysis of Variance (ANOVA)

Using the R code chunk below, do the following: 1) Order the nutrient treatments from low to high (see handout). 2) Produce a barplot to visualize zooplankton biomass in each nutrient treatment. 3) Include

error bars (+/-1 sem) on your plot and label the axes appropriately. 4) Use a one-way analysis of variance (ANOVA) to test the null hypothesis that zooplankton biomass is affected by the nutrient treatment.

```
NUTS <- factor(meso$NUTS, levels = c('L', 'M', 'H'))
zp.means <- tapply(meso$ZP, NUTS, mean)
zp.sem <- tapply(meso$ZP,NUTS,sem)
bp <- barplot(zp.means, ylim = c(0, round(max(meso$ZP), digits = 0)), pch = 15, cex = 1.25, las = 1, ce
arrows(x0 = bp, y0 = zp.means, y1 = zp.means - zp.sem, angle = 90, length = 0.1, lwd = 1)
arrows(x0 = bp, y0 = zp.means, y1 = zp.means + zp.sem, angle = 90, length = 0.1, lwd = 1)</pre>
```



SYNTHESIS: SITE-BY-SPECIES MATRIX

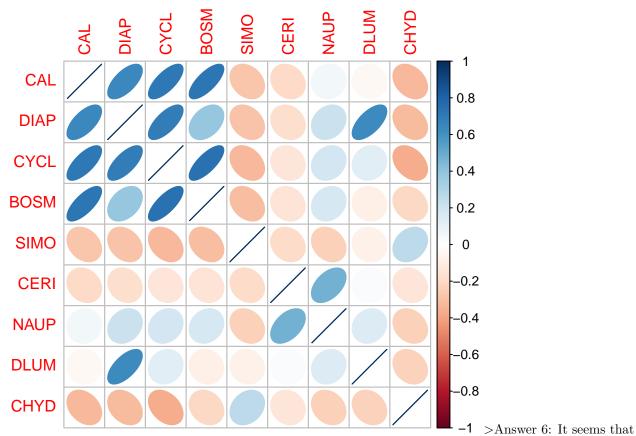
In the R code chunk below, load the zoops.txt data set in your **3.RStudio** data folder. Create a site-by-species matrix (or dataframe) that does *not* include TANK or NUTS. The remaining columns of data refer to the biomass (μ g/L) of different zooplankton taxa:

- \bullet CAL = calanoid copepods
- DIAP = Diaphanasoma sp.
- CYL = cyclopoid copepods
- BOSM = Bosmina sp.
- SIMO = Simocephallus sp.
- CERI = Ceriodaphnia sp.

- NAUP = naupuli (immature copepod)
- DLUM = Daphnia lumholtzi
- CHYD = Chydorus sp.

Question 6: With the visualization and statistical tools that we learned about in the **3. RStudio** handout, use the site-by-species matrix to assess whether and how different zooplankton taxa were responsible for the total biomass (ZP) response to nutrient enrichment. Describe what you learned below in the "Answer" section and include appropriate code in the R chunk.

```
mesonew <-read.table("data/zoops.txt", sep = "\t", header = TRUE)</pre>
mesonew.noNUTSTANK <- mesonew[,3:11]</pre>
pairs(mesonew.noNUTSTANK)
            0 150
                              0 4 8
                                               0 300
                                                                 0 3 6
     CAL
             DIAP
                      CYCL
                              BOSM
                                       SIMO
500
                                                CERI
                                  0 6
                                                         NAUP
                                                                  DLUM
                                  ٥d
   0 150
                     0 200
                                      0 1500
                                                       0.0
                                                          2.0
                                                                         0
                                                                             6000
cor4 <- cor(mesonew.noNUTSTANK)</pre>
install.packages("corrplot", repos="http://cran.rstudio.com/")
## Installing package into '/Users/tbiewerh/Library/R/4.0/library'
## (as 'lib' is unspecified)
##
## The downloaded binary packages are in
   /var/folders/rl/5fysqz596_v_krmqljc490d52cnvv6/T//RtmphkKNCR/downloaded_packages
require("corrplot")
## Loading required package: corrplot
## corrplot 0.84 loaded
corrplot(cor4, method = "ellipse")
```



the biomass of CAL, DIAP, CYCL, BOSM are all positively correlated. NAUP and CERI are positively correlated. DLUM and DIAP are also positively correlated for biomass. Some are negatively correlated, but none very strongly.

SUBMITTING YOUR WORKSHEET

Use Knitr to create a PDF of your completed **3.RStudio_Worksheet.Rmd** document, push the repo to GitHub, and create a pull request. Please make sure your updated repo include both the PDF and RMarkdown files.

This assignment is due on Wednesday, January 24th, 2021 at 12:00 PM (noon).