

6. Worksheet: Among Site (Beta) Diversity – Part 1

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OVERVIEW

In this worksheet, we move beyond the investigation of within-site α -diversity. We will explore β -diversity, which is defined as the diversity that occurs among sites. This requires that we examine the compositional similarity of assemblages that vary in space or time.

After completing this exercise you will know how to:

1. formally quantify β -diversity
2. visualize β -diversity with heatmaps, cluster analysis, and ordination
3. test hypotheses about β -diversity using multivariate statistics

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, **push** this file to your GitHub repo.
6. For the assignment portion of the worksheet, follow the directions at the bottom of this file.
7. When you are done, **Knit** the text and code into a PDF file.
8. After Knitting, submit the completed exercise by creating a **pull request** via GitHub. Your pull request should include this file (**6.BetaDiversity_1_Worksheet.Rmd**) with all code blocks filled out and questions answered) and the PDF output of Knitr (**6.BetaDiversity_1_Worksheet.pdf**).

The completed exercise is due on **Wednesday, February 5th, 2025 before 12:00 PM (noon)**.

1) R SETUP

Typically, the first thing you will do in either an R script or an RMarkdown file is setup your environment. This includes things such as setting the working directory and loading any packages that you will need.

In the R code chunk below, please provide the code to:

- 1) Clear your R environment,
- 2) Print your current working directory,
- 3) Set your working directory to your **Week3-Beta/** folder folder, and
- 4) Load the **vegan** R package (be sure to install first if you have not already).

```
rm(list = ls())

getwd()

## [1] "/cloud/project/QB2025_Brown/Week3-Beta"
setwd("/cloud/project/QB2025_Brown/Week3-Beta")

package.list <- c('vegan', 'ade4', 'viridis', 'gplots', 'BiodiversityR', 'indicspecies')
library(vegan)

## Loading required package: permute
## Loading required package: lattice
## This is vegan 2.6-8
library(ade4)
library(viridis)

## Loading required package: viridisLite
library(gplots)

##
## Attaching package: 'gplots'
## The following object is masked from 'package:stats':
##
##      lowess
library(indicspecies)
```

2) LOADING DATA

Load dataset

In the R code chunk below, do the following:

1. load the doubs dataset from the `ade4` package, and
2. explore the structure of the dataset.

```
# note, please do not print the dataset when submitting
data(doubs)

length(doubs)

## [1] 4

str(doubs, max.level = 1)

## List of 4
## $ env      : 'data.frame': 30 obs. of  11 variables:
## $ fish      : 'data.frame': 30 obs. of  27 variables:
## $ xy         : 'data.frame': 30 obs. of  2 variables:
## $ species    : 'data.frame': 27 obs. of  4 variables:
head(doubs$env)

##   dfs alt   slo flo pH har pho nit amm oxy bdo
## 1    3 934 6.176 84 79 45   1 20   0 122 27
```

```
## 2  22 932 3.434 100 80  40  2  20  10 103  19
## 3 102 914 3.638 180 83  52  5  22   5 105  35
## 4 185 854 3.497 253 80  72 10  21   0 110  13
## 5 215 849 3.178 264 81  84 38  52  20  80  62
## 6 324 846 3.497 286 79  60 20  15   0 102  53
```

Question 1: Describe some of the attributes of the `doubs` dataset.

- How many objects are in `doubs`?
- How many fish species are there in the `doubs` dataset?
- How many sites are in the `doubs` dataset?

Answer 1a: There are 4 objects. **Answer 1b:** There are 27 fish species. **Answer 1c:** There are 30 sites.

Visualizing the Doubs River Dataset

Question 2: Answer the following questions based on the spatial patterns of richness (i.e., α -diversity) and Brown Trout (*Salmo trutta*) abundance in the Doubs River.

- How does fish richness vary along the sampled reach of the Doubs River?
- How does Brown Trout (*Salmo trutta*) abundance vary along the sampled reach of the Doubs River?
- What do these patterns say about the limitations of using richness when examining patterns of biodiversity?

Answer 2a: Fish richness seems to be the highest downstream and then slightly decreases as you approach the middle of the sampled reach. However, once you reach the middle of the stream, fish richness increases significantly. Then, as you approach upstream, fish richness decreases. **Answer 2b:** Brown trout abundance is almost non-existent downstream; however, it does increase slightly as you approach the middle of the sampled reach. Abundance then continues to increase, except for at a few sites, as you go upstream. From the graph you can clearly see that brown trout are much more abundant upstream than downstream. **Answer 2c:** When only using richness, you cannot tell exactly what species are present and in what quantities. You could have two different sites, both with a richness of 10; yet, the 10 species at each site could be completely different from one another. It also does not give you much insight into the structure of the community you are examining. It is best to use richness along with a combination of other methods of examining biodiversity.

3) QUANTIFYING BETA-DIVERSITY

In the R code chunk below, do the following:

- write a function (`beta.w()`) to calculate Whittaker's β -diversity (i.e., β_w) that accepts a site-by-species matrix with optional arguments to specify pairwise turnover between two sites, and
- use this function to analyze various aspects of β -diversity in the Doubs River.

```
beta.w <- function(site.by.species = "", sitenum1 = "", sitenum2 = "",
  pairwise = FALSE){
  if (pairwise == TRUE){
    if (sitenum1 == "" | sitenum2 == ""){
      print("Error: please specify sites to compare")
      return(NA)}
    site1 = site.by.species[sitenum1, ]
    site2 = site.by.species[sitenum2, ]
    site1 = subset(site1, select = site1 > 0)
    site2 = subset(site2, select = site2 > 0)
    gamma = union(colnames(site1), colnames(site2))
    s      = length(gamma)
```

```

a.bar = mean(c(specnumber(site1), specnumber(site2)))
b.w   = round(s/a.bar - 1, 3)
return(b.w)
}
else{
  SbyS.pa <- decostand(site.by.species, method = "pa")
  S <- ncol(SbyS.pa[,which(colSums(SbyS.pa) > 0)])
  a.bar <- mean(specnumber(SbyS.pa))
  b.w <- round(S/a.bar, 3)
  return(b.w)
}
}

```

```
beta.w(doubs$fish, 1, 5, pairwise = TRUE)
```

```
## [1] 0.833
```

```
beta.w(doubs$fish, 1, 2, pairwise = TRUE)
```

```
## [1] 0.5
```

```
beta.w(doubs$fish, 1, 10, pairwise = TRUE)
```

```
## [1] 0.714
```

Question 3: Using your `beta.w()` function above, answer the following questions:

- Describe how local richness (α) and turnover (β) contribute to regional (γ) fish diversity in the Doubs.
- Is the fish assemblage at site 1 more similar to the one at site 2 or site 10?
- Using your understanding of the equation $\beta_w = \gamma/\alpha$, how would your interpretation of β change if we instead defined beta additively (i.e., $\beta = \gamma - \alpha$)?

Answer 3a: Gamma diversity is defined as the total species diversity across an entire landscape and alpha and beta diversity contribute to this. Alpha diversity is a measure of the amount of species present in a particular site and beta diversity measures how similar and/or different sites are from one another. In the Doubs data, the alpha diversity of each site and the beta diversity between those sites will impact the gamma diversity. For example, if alpha diversity is extremely high but beta diversity is relatively low, gamma diversity may be lower. On the other hand, a very high beta diversity could result in a high gamma diversity, even if species richness is low. There can be a lot of variation in gamma diversity based on what the alpha and beta diversity for your site(s) are.

Answer 3b: The fish assemblage at site 1 is more similar to site 2 because a lower Whittaker's species turnover value corresponds to species that are more similar to one another. Site 1 and site 2 had a value of 0.5 and site 1 and site 10 had a value of 0.714. **Answer 3c:** If you subtracted gamma and alpha as opposed to dividing them, it would not provide as accurate information. Subtracting alpha diversity from gamma diversity would just subtract the species richness from the regional diversity. The ultimate goal of beta diversity is to determine differences between sites and subtracting alpha diversity would not allow you to infer anything about how sites differ from one another.

The Resemblance Matrix

In order to quantify β -diversity for more than two samples, we need to introduce a new primary ecological data structure: the **Resemblance Matrix**.

Question 4: How do incidence- and abundance-based metrics differ in their treatment of rare species?

Answer 4: Incidence-based metrics treat rare species the same as common species. The goal of incidence-based metrics is to determine if a species is present or absent in the community. It does not take into account how many individuals of that species are present or what species it is. On the other hand, abundance-based metrics examine how many individuals of a species are present. Therefore, a species with more individuals present will be seen differently, and in a greater context, than a species with only a few individuals present. If there is a rare species present in large numbers, it will be treated differently than a common species with only a few individuals.

In the R code chunk below, do the following:

1. make a new object, `fish`, containing the fish abundance data for the Doubs River,
2. remove any sites where no fish were observed (i.e., rows with sum of zero),
3. construct a resemblance matrix based on Sørensen's Similarity ("fish.ds"), and
4. construct a resemblance matrix based on Bray-Curtis Distance ("fish.db").

```
fish <- doubs$fish
fish <- fish[-8,]

fish.dj <- vegdist(fish, method = "jaccard", binary = TRUE)

fish.db <- vegdist(fish, method = "bray")

fish.ds <- vegdist(fish, method = "bray", binary = TRUE)

fish.ds
```

##	1	2	3	4	5	6	7
## 2	0.50000000						
## 3	0.60000000	0.14285714					
## 4	0.77777778	0.45454545	0.33333333				
## 5	0.83333333	0.57142857	0.46666667	0.15789474			
## 6	0.81818182	0.53846154	0.42857143	0.11111111	0.04761905		
## 7	0.66666667	0.25000000	0.33333333	0.38461538	0.37500000	0.33333333	
## 9	1.00000000	0.50000000	0.55555556	0.38461538	0.37500000	0.33333333	0.40000000
## 10	0.71428571	0.33333333	0.40000000	0.28571429	0.29411765	0.25000000	0.09090909
## 11	0.71428571	0.33333333	0.40000000	0.42857143	0.52941176	0.50000000	0.27272727
## 12	0.71428571	0.33333333	0.40000000	0.42857143	0.52941176	0.50000000	0.27272727
## 13	0.71428571	0.33333333	0.40000000	0.57142857	0.64705882	0.62500000	0.45454545
## 14	0.81818182	0.53846154	0.42857143	0.33333333	0.42857143	0.40000000	0.46666667
## 15	0.83333333	0.57142857	0.60000000	0.36842105	0.36363636	0.33333333	0.37500000
## 16	0.88888889	0.70000000	0.61904762	0.36000000	0.28571429	0.25925926	0.54545455
## 17	0.91304348	0.76000000	0.69230769	0.46666667	0.39393939	0.37500000	0.62962963
## 18	0.91666667	0.76923077	0.70370370	0.48387097	0.41176471	0.39393939	0.64285714
## 19	1.00000000	0.84615385	0.77777778	0.54838710	0.41176471	0.45454545	0.71428571
## 20	1.00000000	0.84000000	0.76923077	0.53333333	0.39393939	0.43750000	0.70370370
## 21	1.00000000	0.84615385	0.77777778	0.54838710	0.41176471	0.45454545	0.71428571
## 22	1.00000000	0.92000000	0.84615385	0.60000000	0.45454545	0.50000000	0.77777778
## 23	1.00000000	1.00000000	1.00000000	0.81818182	0.71428571	0.69230769	0.75000000
## 24	1.00000000	1.00000000	1.00000000	0.75000000	0.68421053	0.66666667	0.84615385
## 25	1.00000000	1.00000000	0.83333333	0.62500000	0.36842105	0.44444444	0.69230769
## 26	1.00000000	0.91666667	0.84000000	0.58620690	0.43750000	0.48387097	0.76923077
## 27	1.00000000	0.92000000	0.84615385	0.60000000	0.45454545	0.50000000	0.77777778
## 28	1.00000000	0.92000000	0.84615385	0.60000000	0.45454545	0.50000000	0.77777778
## 29	0.92592593	0.79310345	0.73333333	0.52941176	0.40540541	0.44444444	0.67741935
## 30	1.00000000	1.00000000	0.92000000	0.65517241	0.50000000	0.54838710	0.84615385

##	9	10	11	12	13	14	15
## 2							
## 3							
## 4							
## 5							
## 6							
## 7							
## 9							
## 10	0.45454545						
## 11	0.45454545	0.33333333					
## 12	0.45454545	0.33333333	0.00000000				
## 13	0.63636364	0.50000000	0.16666667	0.16666667			
## 14	0.60000000	0.37500000	0.25000000	0.25000000	0.25000000		
## 15	0.50000000	0.29411765	0.29411765	0.29411765	0.29411765	0.14285714	
## 16	0.54545455	0.47826087	0.56521739	0.56521739	0.56521739	0.33333333	0.28571429
## 17	0.62962963	0.57142857	0.57142857	0.57142857	0.57142857	0.37500000	0.33333333
## 18	0.64285714	0.58620690	0.58620690	0.58620690	0.58620690	0.39393939	0.35294118
## 19	0.64285714	0.65517241	0.79310345	0.79310345	0.79310345	0.57575758	0.52941176
## 20	0.62962963	0.64285714	0.78571429	0.78571429	0.85714286	0.62500000	0.57575758
## 21	0.64285714	0.65517241	0.79310345	0.79310345	0.86206897	0.63636364	0.58823529
## 22	0.70370370	0.71428571	0.85714286	0.85714286	0.92857143	0.68750000	0.63636364
## 23	0.50000000	0.77777778	0.77777778	0.77777778	1.00000000	0.84615385	0.85714286
## 24	0.69230769	0.71428571	0.85714286	0.85714286	1.00000000	0.77777778	0.78947368
## 25	0.69230769	0.57142857	0.85714286	0.85714286	1.00000000	0.66666667	0.68421053
## 26	0.69230769	0.70370370	0.85185185	0.85185185	0.92592593	0.67741935	0.62500000
## 27	0.70370370	0.71428571	0.85714286	0.85714286	0.92857143	0.68750000	0.63636364
## 28	0.70370370	0.71428571	0.85714286	0.85714286	0.92857143	0.68750000	0.63636364
## 29	0.67741935	0.62500000	0.68750000	0.68750000	0.68750000	0.50000000	0.45945946
## 30	0.76923077	0.77777778	0.92592593	0.92592593	1.00000000	0.74193548	0.68750000
##	16	17	18	19	20	21	22
## 2							
## 3							
## 4							
## 5							
## 6							
## 7							
## 9							
## 10							
## 11							
## 12							
## 13							
## 14							
## 15							
## 16							
## 17	0.12820513						
## 18	0.15000000	0.02222222					
## 19	0.25000000	0.15555556	0.13043478				
## 20	0.28205128	0.18181818	0.15555556	0.02222222			
## 21	0.30000000	0.20000000	0.17391304	0.04347826	0.02222222		
## 22	0.33333333	0.22727273	0.20000000	0.06666667	0.04545455	0.02222222	
## 23	0.80000000	0.76000000	0.76923077	0.76923077	0.76000000	0.76923077	0.76000000
## 24	0.68000000	0.60000000	0.54838710	0.48387097	0.46666667	0.48387097	0.46666667
## 25	0.60000000	0.60000000	0.54838710	0.48387097	0.46666667	0.48387097	0.46666667
## 26	0.36842105	0.25581395	0.22727273	0.09090909	0.06976744	0.04545455	0.02325581

```

## 27 0.33333333 0.22727273 0.20000000 0.06666667 0.04545455 0.02222222 0.00000000
## 28 0.33333333 0.22727273 0.20000000 0.06666667 0.04545455 0.02222222 0.00000000
## 29 0.25581395 0.12500000 0.10204082 0.06122449 0.08333333 0.06122449 0.08333333
## 30 0.36842105 0.25581395 0.22727273 0.09090909 0.06976744 0.04545455 0.02325581
##      23      24      25      26      27      28      29
## 2
## 3
## 4
## 5
## 6
## 7
## 9
## 10
## 11
## 12
## 13
## 14
## 15
## 16
## 17
## 18
## 19
## 20
## 21
## 22
## 23
## 24 0.45454545
## 25 0.45454545 0.37500000
## 26 0.75000000 0.44827586 0.44827586
## 27 0.76000000 0.46666667 0.46666667 0.02325581
## 28 0.76000000 0.46666667 0.46666667 0.02325581 0.00000000
## 29 0.79310345 0.52941176 0.52941176 0.10638298 0.08333333 0.08333333
## 30 0.75000000 0.44827586 0.44827586 0.04761905 0.02325581 0.02325581 0.10638298

```

fish.db

```

##      1      2      3      4      5      6      7
## 2 0.60000000
## 3 0.68421053 0.14285714
## 4 0.75000000 0.33333333 0.18918919
## 5 0.89189189 0.69565217 0.68000000 0.49090909
## 6 0.75000000 0.39393939 0.29729730 0.19047619 0.41818182
## 7 0.68421053 0.14285714 0.12500000 0.24324324 0.64000000 0.24324324
## 9 1.00000000 0.69230769 0.73333333 0.65714286 0.58333333 0.54285714 0.66666667
## 10 0.88235294 0.38461538 0.40000000 0.37142857 0.54166667 0.25714286 0.26666667
## 11 0.57142857 0.30434783 0.40740741 0.43750000 0.68888889 0.43750000 0.33333333
## 12 0.71428571 0.20000000 0.23529412 0.33333333 0.69230769 0.38461538 0.17647059
## 13 0.72727273 0.29032258 0.31428571 0.45000000 0.73584906 0.55000000 0.37142857
## 14 0.80645161 0.40000000 0.31818182 0.34693878 0.67741935 0.42857143 0.36363636
## 15 0.83333333 0.51111111 0.46938776 0.40740741 0.55223881 0.37037037 0.38775510
## 16 0.86046512 0.65384615 0.57142857 0.47540984 0.45945946 0.37704918 0.53571429
## 17 0.91489362 0.67857143 0.63333333 0.50769231 0.51282051 0.44615385 0.60000000
## 18 0.95555556 0.74074074 0.72413793 0.58730159 0.50000000 0.52380952 0.68965517
## 19 1.00000000 0.79310345 0.70967742 0.61194030 0.50000000 0.52238806 0.67741935
## 20 1.00000000 0.91176471 0.88888889 0.74025974 0.48888889 0.68831169 0.86111111

```

```

## 21 1.00000000 0.94594595 0.92307692 0.78313253 0.50000000 0.73493976 0.89743590
## 22 1.00000000 0.97619048 0.95454545 0.82795699 0.52830189 0.78494624 0.93181818
## 23 1.00000000 1.00000000 1.00000000 0.92000000 0.89473684 0.84000000 0.90000000
## 24 1.00000000 1.00000000 1.00000000 0.88888889 0.79591837 0.77777778 0.93548387
## 25 1.00000000 1.00000000 0.92592593 0.81250000 0.68888889 0.68750000 0.85185185
## 26 1.00000000 0.96363636 0.93220339 0.78125000 0.55844156 0.68750000 0.89830508
## 27 1.00000000 0.97333333 0.94936709 0.83333333 0.56701031 0.76190476 0.92405063
## 28 1.00000000 0.97560976 0.95348837 0.82417582 0.57692308 0.78021978 0.93023256
## 29 0.97777778 0.93939394 0.92233010 0.81481481 0.53719008 0.77777778 0.90291262
## 30 1.00000000 1.00000000 0.98095238 0.87272727 0.59349593 0.83636364 0.96190476
##          9          10          11          12          13          14          15
## 2
## 3
## 4
## 5
## 6
## 7
## 9
## 10 0.57142857
## 11 0.76000000 0.44000000
## 12 0.68750000 0.37500000 0.24137931
## 13 0.81818182 0.57575758 0.33333333 0.18918919
## 14 0.76190476 0.47619048 0.43589744 0.21739130 0.19148936
## 15 0.65957447 0.40425532 0.50000000 0.33333333 0.38461538 0.24590164
## 16 0.70370370 0.51851852 0.64705882 0.55172414 0.59322034 0.44117647 0.26027397
## 17 0.68965517 0.51724138 0.63636364 0.58064516 0.61904762 0.50000000 0.40259740
## 18 0.64285714 0.57142857 0.69811321 0.66666667 0.70491803 0.60000000 0.46666667
## 19 0.66666667 0.63333333 0.82456140 0.75000000 0.81538462 0.67567568 0.56962025
## 20 0.68571429 0.77142857 0.91044776 0.89189189 0.92000000 0.83333333 0.70786517
## 21 0.76315789 0.81578947 0.91780822 0.92500000 0.95061728 0.86666667 0.76842105
## 22 0.76744186 0.86046512 0.95180723 0.95555556 0.97802198 0.90000000 0.77142857
## 23 0.77777778 0.88888889 0.86666667 0.90909091 1.00000000 0.93750000 0.94594595
## 24 0.72413793 0.79310345 0.92307692 0.93939394 1.00000000 0.90697674 0.87500000
## 25 0.84000000 0.76000000 0.90909091 0.93103448 1.00000000 0.84615385 0.81818182
## 26 0.71929825 0.82456140 0.92592593 0.93442623 0.96774194 0.85915493 0.76315789
## 27 0.76623377 0.84415584 0.94594595 0.95061728 0.97560976 0.89010989 0.77083333
## 28 0.76190476 0.85714286 0.95061728 0.95454545 0.97752809 0.89795918 0.78640777
## 29 0.78217822 0.84158416 0.89795918 0.90476190 0.90566038 0.84347826 0.73333333
## 30 0.84466019 0.90291262 0.98000000 0.98130841 1.00000000 0.93162393 0.81967213
##          16          17          18          19          20          21          22
## 2
## 3
## 4
## 5
## 6
## 7
## 9
## 10
## 11
## 12
## 13
## 14
## 15
## 16

```



```

## 17 0.26190476
## 18 0.34146341 0.13953488
## 19 0.39534884 0.31111111 0.25000000
## 20 0.58333333 0.42000000 0.32653061 0.23529412
## 21 0.62745098 0.49056604 0.40384615 0.29629630 0.10169492
## 22 0.66071429 0.55172414 0.47368421 0.38983051 0.18750000 0.10447761
## 23 0.90909091 0.83333333 0.82608696 0.84000000 0.86666667 0.87878788 0.89473684
## 24 0.81818182 0.69491525 0.64912281 0.63934426 0.57746479 0.61038961 0.65517241
## 25 0.76470588 0.74545455 0.66037736 0.61403509 0.67164179 0.69863014 0.73493976
## 26 0.63855422 0.54022989 0.45882353 0.32584270 0.21212121 0.20000000 0.25217391
## 27 0.66990291 0.57009346 0.48571429 0.37614679 0.19327731 0.13600000 0.12592593
## 28 0.69090909 0.57894737 0.50000000 0.41379310 0.22222222 0.16666667 0.12676056
## 29 0.65354331 0.51145038 0.44186047 0.41353383 0.24475524 0.18120805 0.11949686
## 30 0.72093023 0.57894737 0.52671756 0.48148148 0.29655172 0.23178808 0.18012422
##      23      24      25      26      27      28      29
## 2
## 3
## 4
## 5
## 6
## 7
## 9
## 10
## 11
## 12
## 13
## 14
## 15
## 16
## 17
## 18
## 19
## 20
## 21
## 22
## 23
## 24 0.57894737
## 25 0.46666667 0.46153846
## 26 0.82978723 0.48275862 0.59259259
## 27 0.88059701 0.61538462 0.70270270 0.18867925
## 28 0.89189189 0.64705882 0.72839506 0.23893805 0.09774436
## 29 0.91208791 0.70588235 0.77551020 0.33846154 0.18666667 0.14649682
## 30 0.91397849 0.71153846 0.78000000 0.36363636 0.19736842 0.15723270 0.14772727
fish.db <- vegdist(fish, method = "bray", upper = TRUE, diag = TRUE)

```

Question 5: Using the distance matrices from above, answer the following questions:

- Does the resemblance matrix (`fish.db`) represent similarity or dissimilarity? What information in the resemblance matrix led you to arrive at your answer?
- Compare the resemblance matrices (`fish.db` or `fish.ds`) you just created. How does the choice of the Sørensen or Bray-Curtis distance influence your interpretation of site (dis)similarity?

Answer 5a: The resemblance matrix `fish.db` represents dissimilarity and this is known since the matrix was created using the Bray-Curtis index. This index is used specifically to measure dissimilarity. **Answer 5b:** When looking at the matrices, it is important to know what exactly

both are measuring to properly interpret it. The Sorensen matrix is determining if the sites you are comparing have the same species present. Therefore when looking at the matrix, if two sites have a score close to 1, then they share a lot of species. On the other hand, the Bray-Curtis matrix is also taking abundance into account along with richness. Meaning that two sites with similar species richness and even abundance will have a score closer to 1. Based on the questions you are asking, it is acceptable to use one or the other, but it is important to know what exactly you are inferring. If you are looking at two sites that have a high Sorensen index but a low Bray-Curtis index, then that means that species richness is high but the abundance is not evenly distributed.

4) VISUALIZING BETA-DIVERSITY

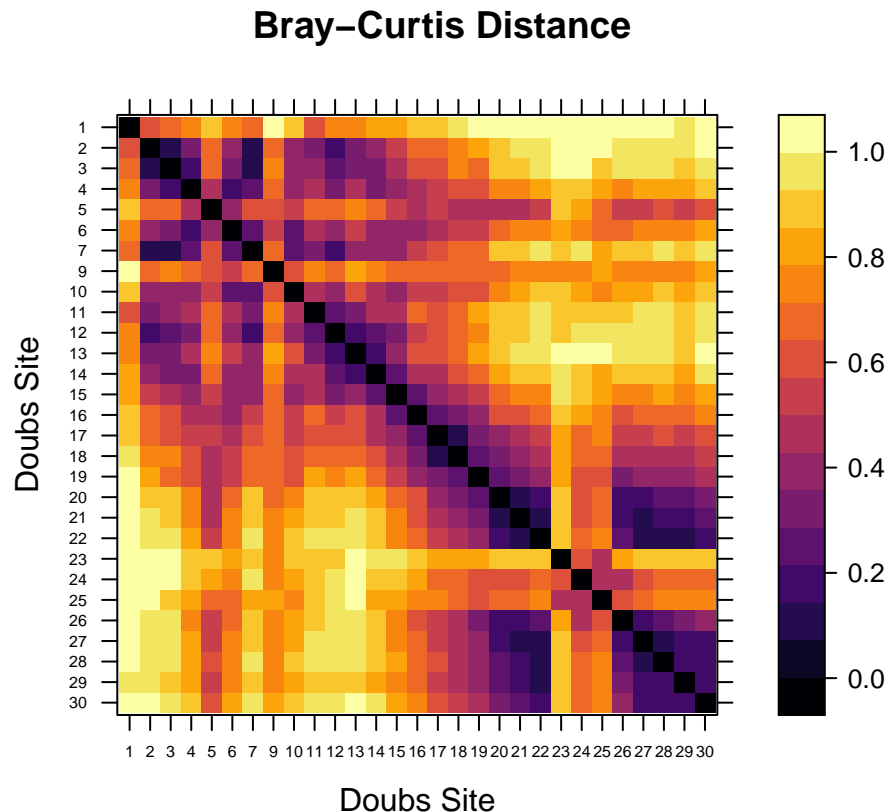
A. Heatmaps

In the R code chunk below, do the following:

1. define a color palette,
2. define the order of sites in the Doubs River, and
3. use the `levelplot()` function to create a heatmap of fish abundances in the Doubs River.

```
order <- rev(attr(fish.db, "Labels"))

levelplot(as.matrix(fish.db)[, order], aspect = "iso", col.regions = inferno,
          xlab = "Doubs Site", ylab = "Doubs Site", scales = list(cex = 0.5),
          main = "Bray-Curtis Distance")
```



B. Cluster Analysis

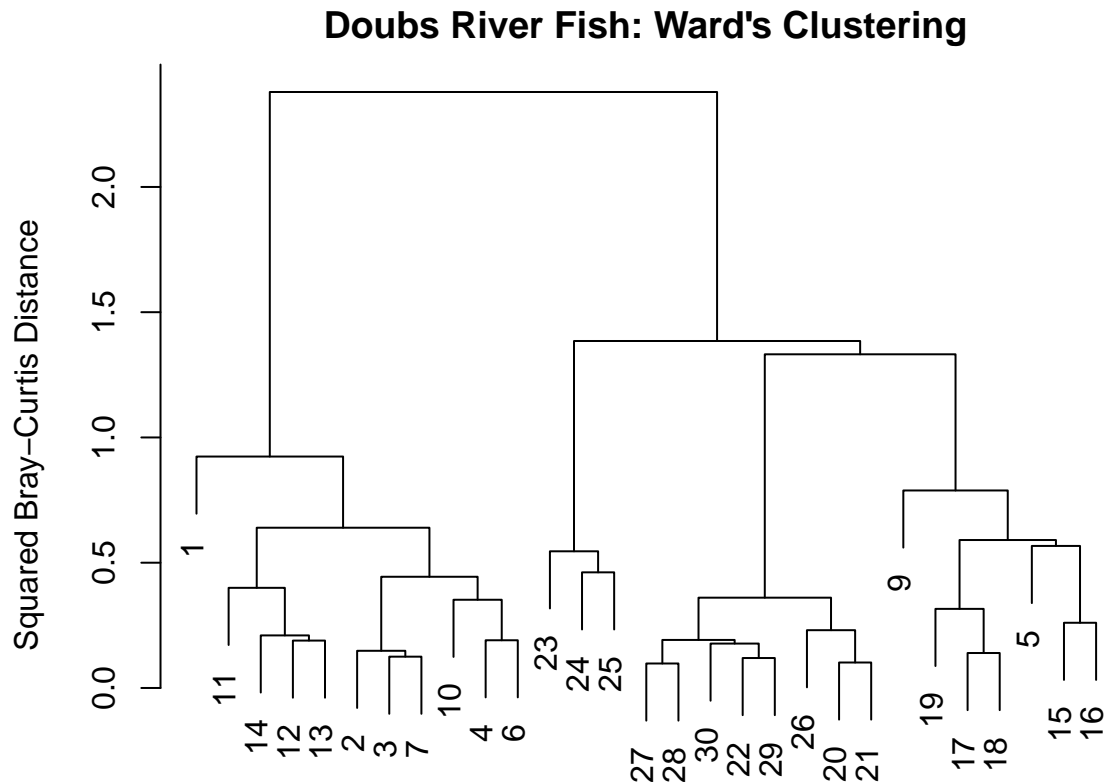
In the R code chunk below, do the following:

1. perform a cluster analysis using Ward's Clustering, and

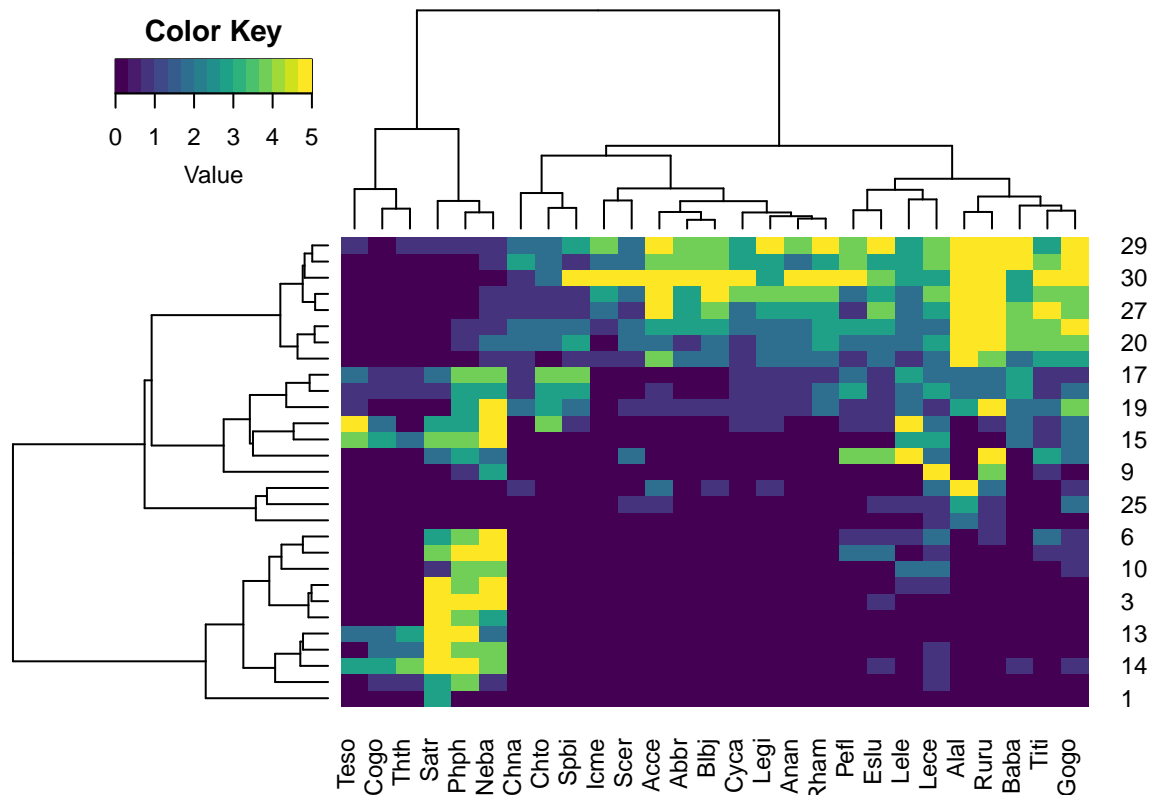
2. plot your cluster analysis (use either `hclust` or `heatmap.2`).

```
fish.ward <- hclust(fish.db, method = "ward.D2")

par(mar = c(1, 5, 2, 2) + 0.1)
plot(fish.ward, main = "Doubs River Fish: Ward's Clustering",
     ylab = "Squared Bray-Curtis Distance")
```



```
gplots::heatmap.2(as.matrix(fish),
  distfun = function(x) vegdist(x, method = "bray"),
  hclustfun = function(x) hclust(x, method = "ward.D2"),
  col = viridis, trace = "none", density.info = "none")
```



Question 6: Based on cluster analyses and the introductory plots that we generated after loading the data, develop an ecological hypothesis for fish diversity the doubts data set?

Answer 6: Fish diversity is relatively high within the doubts data set and species are more similar to one another in neighboring sites and become more dissimilar as site distance increases.

C. Ordination

Principal Coordinates Analysis (PCoA)

In the R code chunk below, do the following:

1. perform a Principal Coordinates Analysis to visualize beta-diversity
2. calculate the variation explained by the first three axes in your ordination
3. plot the PCoA ordination,
4. label the sites as points using the Doubs River site number, and
5. identify influential species and add species coordinates to PCoA plot.

```
fish.pcoa <- cmdscale(fish.db, eig = TRUE, k = 3)

explainvar1 <- round(fish.pcoa$eig[1] / sum(fish.pcoa$eig), 3) * 100
explainvar2 <- round(fish.pcoa$eig[2] / sum(fish.pcoa$eig), 3) * 100
explainvar3 <- round(fish.pcoa$eig[3] / sum(fish.pcoa$eig), 3) * 100
sum.eig <- sum(explainvar1, explainvar2, explainvar3)

par(mar = c(5,5,1,2) + 0.1)

plot(fish.pcoa$points[,1], fish.pcoa$points[,2], ylim = c(-0.2, 0.7),
     xlab = paste("PCoA 1 (", explainvar1, "%)", sep = ""),
     ylab = paste("PCoA 2 (", explainvar2, "%)", sep = ""),
     pch = 16, cex = 2.0, type = "n", cex.lab = 1.5,
```

```

    cex.axis = 1.2, axes = FALSE)

axis(side = 1, labels = T, lwd.ticks = 2, cex.axis = 1.2, las = 1)
axis(side = 2, labels = T, lwd.ticks = 2, cex.axis = 1.2, las = 1)
abline(h = 0, v = 0, lty = 3)
box(lwd = 2)

points(fish.pcoa$points[,1], fish.pcoa$points[,2],
       pch = 19, cex = 3, bg = "gray", col = "gray")
text(fish.pcoa$points[,1], fish.pcoa$points[,2],
     labels = row.names(fish.pcoa$points))

fishREL <- fish
for(i in 1:nrow(fish)){
  fishREL[i, ] = fish[i, ] / sum(fish[i, ])
}

library(vegan)

`add.spec.scores.class` <-
function(ordi,comm,method="cor.scores",multi=1,Rscale=F,scaling="1") {
  ordiscores <- scores(ordi,display="sites")
  n <- ncol(comm)
  p <- ncol(ordiscores)
  specscores <- array(NA,dim=c(n,p))
  rownames(specscores) <- colnames(comm)
  colnames(specscores) <- colnames(ordiscores)
  if (method == "cor.scores") {
    for (i in 1:n) {
      for (j in 1:p) {specscores[i,j] <- cor(comm[,i],ordiscores[,j],method="pearson")}
    }
  }
  if (method == "wa.scores") {specscores <- wascores(ordiscores,comm)}
  if (method == "pcoa.scores") {
    rownames(ordiscores) <- rownames(comm)
    eigenv <- ordi$eig
    accounted <- sum(eigenv)
    tot <- 2*(accounted/ordi$GOF[2])-(accounted/ordi$GOF[1])
    eigen.var <- eigenv/(nrow(comm)-1)
    neg <- length(eigenv[eigenv<0])
    pos <- length(eigenv[eigenv>0])
    tot <- tot/(nrow(comm)-1)
    eigen.percen <- 100*eigen.var/tot
    eigen.cumpercen <- cumsum(eigen.percen)
    constant <- ((nrow(comm)-1)*tot)^0.25
    ordiscores <- ordiscores * (nrow(comm)-1)^-0.5 * tot^-0.5 * constant
    p1 <- min(p, pos)
    for (i in 1:n) {
      for (j in 1:p1) {
        specscores[i,j] <- cor(comm[,i],ordiscores[,j])*sd(comm[,i])/sd(ordiscores[,j])
        if(is.na(specscores[i,j])) {specscores[i,j]<-0}
      }
    }
  }
}

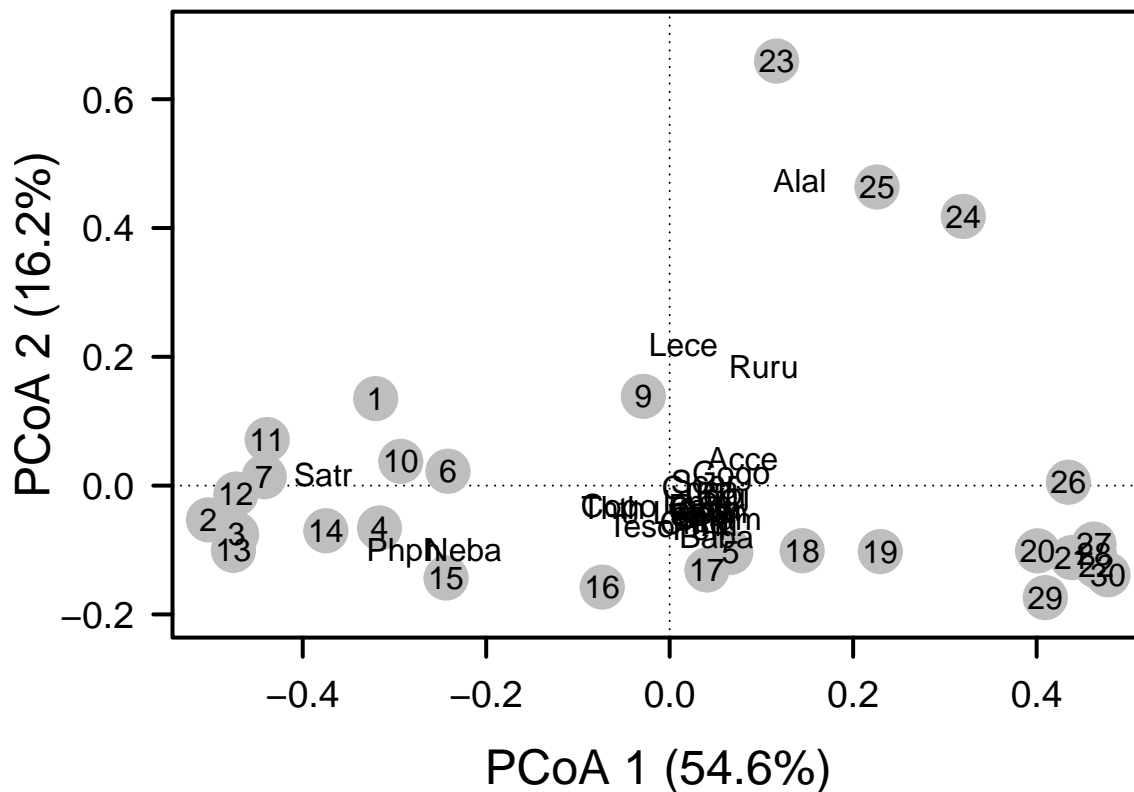
```

```

if (Rscale==T && scaling=="2") {
  percen <- eigen.var/tot
  percen <- percen^0.5
  ordiscores <- sweep(ordiscores,2,percen,"/")
  specscores <- sweep(specscores,2,percen,"*")
}
if (Rscale==F) {
  specscores <- specscores / constant
  ordiscores <- ordi$points
}
ordi$points <- ordiscores
ordi$eig <- eigen.var
ordi$eig.percen <- eigen.percen
ordi$eig.cumpercen <- eigen.cumpercen
ordi$eigen.total <- tot
ordi$R.constant <- constant
ordi$Rscale <- Rscale
ordi$scaling <- scaling
}
specscores <- specscores * multi
ordi$cproj <- specscores
return(ordi)
}

fish.pcoa <- add.spec.scores.class(fish.pcoa,fishREL,method = "pcoa.scores")
text(fish.pcoa$cproj[,1], fish.pcoa$cproj[,2],
     labels = row.names(fish.pcoa$cproj), col = "black")

```



In the R code chunk below, do the following:

1. identify influential species based on correlations along each PCoA axis (use a cutoff of 0.70), and
2. use a permutation test (999 permutations) to test the correlations of each species along each axis.

```
spe.corr <- add.spec.scores.class(fish.pcoa, fishREL, method = "cor.scores")$cproj
corrcut <- 0.7
imp.spp <- spe.corr[abs(spe.corr[, 1]) >= corrcut | abs(spe.corr[, 2]) >= corrcut, ]

fit <- envfit(fish.pcoa, fishREL, perm = 999)
```

Question 7: Address the following questions about the ordination results of the doubs data set:

- a. Describe the grouping of sites in the Doubs River based on fish community composition.
- b. Generate a hypothesis about which fish species are potential indicators of river quality.

Answer 7a: Based on the ordination plot, sites 2, 3, 7, 11, 12, and 13 share very similar abundances and fish communities. Sites 1, 4, 6, 10, 14, and 15 are all also graphed close to those sites as well. These sites likely have similar fish compositions that vary slightly from the first grouping listed. Following these two groupings, there is a gap in the plot, which signifies an increase in variance between the sites. The next grouping is sites 5, 16, 17, 18, and 19; with 9 being a slight outlier of that group. I find this interesting that site 5 has a more similar fish community than sites 1-4, 6, and 7. Sites 23, 24, and 25 appear to be outliers of the entire graph, and although being somewhat close in proximity to one another, they are far away from other sites. This likely means that there is high variance between their fish communities and the fish communities of the other sites. Lastly, the most tight grouping consists of sites 8, 20, 21, 22, 26, 27, 29, and 30. These groups overlap significantly on the graph, meaning they likely have an extremely small amount of variance between each other and have very similar fish communities. Additionally, with them being located on the far bottom right corner of the plot, their fish communities appear to vary significantly compared to the first two groupings listed earlier that appear on the far bottom left of the plot. **Answer 7b:** Given their strong negative values in Dim 1, Phph and Neba appear to be strong indicator species of low quality river conditions. In contrast, Rham, Legi, Cyca, Abbr, Acce, Blbj, and Anan have strong positive values in Dim 1, which suggests they are good indicator species that river quality conditions are high.

SYNTHESIS

Load the dataset from that you and your partner are using for the team project. Use one of the tools introduced in the beta diversity module to visualize your data. Describe any interesting patterns and identify a hypothesis is relevant to the principles of biodiversity.

```
fish_data <- read.csv("/cloud/project/QB2025_Brown/Fish_Dataset.csv")

fish_data$Latitude <- cut(fish_data$Latitude, breaks=c(20, 25, 30, 35, 40, 45, 50),
                          labels=c("20-25", "25-30", "30-35", "35-40", "40-45", "45-50"),
                          include.lowest=TRUE)

Fish_lat <- cbind(fish_data[,2], fish_data[,23:658])
Fish_lat <- data.frame(Latitude = fish_data$Latitude, Fish_lat)

result <- aggregate(. ~ Latitude, data=Fish_lat, sum, na.rm = TRUE)

print(result)
```

##	Latitude	fish_data...2.	Acantharchus.pomotis	Acanthogobius.flavimanus
## 1	25-30	96	0	0
## 2	30-35	1413	20	0

## 3	35-40	3612	6	9
## 4	40-45	4975	0	0
## 5	45-50	2016	0	0
##	Acipenser.brevirostrum	Acipenser.fulvescens	Acrocheilus.alutaceus	
## 1	0	0	0	
## 2	0	0	0	
## 3	0	0	0	
## 4	1	0	45	
## 5	0	2	51	
##	Agosia.chrysogaster	Alosa.aestivalis	Alosa.alabamae	Alosa.chrysochloris
## 1	0	0	2	5
## 2	616	136	0	47
## 3	0	622	0	198
## 4	0	323	0	3
## 5	0	0	0	6
##	Alosa.mediocris	Alosa.pseudoharengus	Alosa.sapidissima	Ambloplites.ariommus
## 1	0	0	0	0
## 2	6	140	138	77
## 3	5	116	51	61
## 4	0	1661	1640	0
## 5	0	2	0	0
##	Ambloplites.cavifrons	Ambloplites.constellatus	Ambloplites.rupestris	
## 1	0	0	0	
## 2	0	0	147	
## 3	23	207	5623	
## 4	0	0	6579	
## 5	0	0	723	
##	Ameiurus.brunneus	Ameiurus.catus	Ameiurus.melas	Ameiurus.natalis
## 1	0	4	1	37
## 2	570	16	362	1053
## 3	262	220	215	1175
## 4	0	52	1577	1048
## 5	0	0	749	60
##	Ameiurus.nebulosus	Ameiurus.platycephalus	Ameiurus.serracanthus	Amia.calva
## 1	5	0	0	132
## 2	8	50	4	314
## 3	85	62	0	96
## 4	977	0	0	126
## 5	30	0	0	17
##	Ammocrypta.beanii	Ammocrypta.clara	Ammocrypta.pellucida	Ammocrypta.vivax
## 1	0	0	0	0
## 2	45	0	0	13
## 3	0	0	4	1
## 4	0	6	1	0
## 5	0	18	0	0
##	Anchoa.mitchilli	Anguilla.rostrata	Aphredoderus.sayanus	Aplodinotus.grunniens
## 1	575	10	13	60
## 2	314	460	977	942
## 3	53	1601	570	2831
## 4	73	4569	25	1972
## 5	0	35	0	150
##	Archoplites.interruptus	Archosargus.probatoccephalus	Ascaphus.truei	
## 1	0	2	0	
## 2	0	0	0	

## 3	0	0	0
## 4	0	0	0
## 5	0	0	0
##	Astyanax.mexicanus	Atractosteus.spatula	Awaous.banana Brevoortia.patronus
## 1	108	5	0 1127
## 2	2	138	0 420
## 3	0	13	0 0
## 4	0	0	0 0
## 5	0	0	0 0
##	Brevoortia.tyrannus	Campostoma.anomalum	Campostoma.oligolepis
## 1	0	214	0
## 2	0	4516	6921
## 3	3	27654	6962
## 4	2	11407	36
## 5	0	837	10
##	Campostoma.pauciradii	Carassius.auratus	Carpiodes.carpio Carpiodes.cyprinus
## 1	0	0	0 1
## 2	398	105	1113 153
## 3	0	114	1672 184
## 4	0	61	3204 512
## 5	0	1	224 10
##	Carpiodes.velifer	Catostomus.ardens	Catostomus.catostomus Catostomus.clarkii
## 1	0	0	0 0
## 2	164	0	0 29
## 3	25	39	186 51
## 4	140	251	403 0
## 5	3	0	556 0
##	Catostomus.columbianus	Catostomus.commersonii	Catostomus.discobolus
## 1	0	0	0
## 2	0	264	24
## 3	0	4889	425
## 4	36	25840	9
## 5	130	6652	0
##	Catostomus.insignis	Catostomus.latipinnis	Catostomus.macrocheilus
## 1	0	0	0
## 2	49	0	0
## 3	0	755	0
## 4	0	21	209
## 5	0	0	387
##	Catostomus occidentalis	Catostomus.platyrrhynchus	Catostomus.plebeius
## 1	0	0	0
## 2	0	0	0
## 3	419	656	25
## 4	185	1217	0
## 5	0	347	0
##	Catostomus.rimiculus	Catostomus.santaanae	Catostomus.snyderi
## 1	0	0	0
## 2	0	0	0
## 3	0	0	0
## 4	52	0	23
## 5	0	0	0
##	Catostomus.tahoensis	Centrarchus.macropterus	Centropomus.parallelus
## 1	0	2	0
## 2	0	273	0

## 3	16	73	0		
## 4	0	0	0		
## 5	0	0	0		
##	Centropomus.undecimalis	Chaenobryttus.gulosus	Channa.argus		
## 1	4	26	0		
## 2	0	243	0		
## 3	0	129	1		
## 4	0	2	0		
## 5	0	0	0		
##	Chologaster.cornuta	Chrosomus.cumberlandensis	Chrosomus.eos		
## 1	0	0	0		
## 2	15	0	0		
## 3	3	0	0		
## 4	0	0	246		
## 5	0	0	435		
##	Chrosomus.erythrogaster	Chrosomus.neogaeus	Chrosomus.oreas		
## 1	0	0	0		
## 2	0	0	0		
## 3	917	0	443		
## 4	192	1	0		
## 5	9	7	0		
##	Chrosomus.tennesseensis	Cichlasoma.cyanoguttatum	Cichlasoma.urophthalmus		
## 1	0	204	7		
## 2	0	35	0		
## 3	84	0	0		
## 4	0	0	0		
## 5	0	0	0		
##	Citharichthys.spilopterus	Clarias.batrachus	Clinostomus.elongatus		
## 1	2	28	0		
## 2	5	0	0		
## 3	0	0	106		
## 4	0	0	251		
## 5	0	0	0		
##	Clinostomus.funduloides	Conger.conger	Coregonus.clupeaformis	Cottus.aleuticus	
## 1	0	0	0	0	
## 2	11	0	0	0	
## 3	2695	0	0	0	
## 4	0	0	0	44	
## 5	0	0	2	3175	
##	Cottus.asper	Cottus.bairdii	Cottus.beldingii	Cottus.caeruleomentum	
## 1	0	0	0	0	
## 2	0	224	0	0	
## 3	539	5432	66	1337	
## 4	108	5608	1476	0	
## 5	645	1824	73	0	
##	Cottus.carolinae	Cottus.chattahoochee	Cottus.cognatus	Cottus.confusus	
## 1	0	0	0	0	
## 2	752	1	0	0	
## 3	4550	0	0	0	
## 4	1	0	1609	73	
## 5	0	0	143	1581	
##	Cottus.girardi	Cottus.gulosus	Cottus.hubbsi	Cottus.hypselurus	Cottus.kanawhae
## 1	0	0	0	0	0
## 2	0	0	0	0	0

## 3	191	198	0	396	68
## 4	0	395	0	0	0
## 5	0	174	0	0	0
##	Cottus.klamathensis	Cottus.leiopomus	Cottus.perplexus	Cottus.rhotheus	
## 1	0	0	0	0	
## 2	0	0	0	0	
## 3	0	0	0	0	
## 4	21	0	1790	572	
## 5	0	0	7796	2988	
##	Cottus.tenuis	Couesius.plumbeus	Crystallaria.asprella	Ctenogobius.shufeldti	
## 1	0	0	0	1	
## 2	0	0	41	8	
## 3	0	0	0	0	
## 4	1	2	0	0	
## 5	0	461	0	0	
##	Ctenopharyngodon.idella	Culaea.inconstans	Cycleptus.elongatus		
## 1	2	0	0		
## 2	62	0	35		
## 3	33	2	24		
## 4	47	3223	75		
## 5	0	1412	4		
##	Cycleptus.meridionalis	Cyprinella.analostana	Cyprinella.callisema		
## 1	0	0	0		
## 2	1	160	187		
## 3	0	1332	0		
## 4	0	700	0		
## 5	0	0	0		
##	Cyprinella.callistia	Cyprinella.camura	Cyprinella.chloristia		
## 1	0	0	0		
## 2	130	71	234		
## 3	0	174	179		
## 4	0	0	0		
## 5	0	0	0		
##	Cyprinella.formosa	Cyprinella.galactura	Cyprinella.gibbsi	Cyprinella.leedsii	
## 1	0	0	0	2	
## 2	0	68	0	521	
## 3	0	3416	0	0	
## 4	0	0	0	0	
## 5	0	0	0	0	
##	Cyprinella.lepida	Cyprinella.lutrensis	Cyprinella.nivea		
## 1	5	941	0		
## 2	1	16061	1767		
## 3	0	24936	40		
## 4	0	10189	0		
## 5	0	34	0		
##	Cyprinella.pyrrhomelas	Cyprinella.spiloptera	Cyprinella.trichroistia		
## 1	0	0	0		
## 2	69	313	407		
## 3	1	7581	0		
## 4	0	16334	0		
## 5	0	2364	0		
##	Cyprinella.venusta	Cyprinella.whipplei	Cyprinella.xaenura		
## 1	669	0	0		
## 2	9828	550	42		

## 3	813	455	0	
## 4	0	23	0	
## 5	0	0	0	
##	Cyprinodon.rubrofluviatilis	Cyprinodon.variegatus	Cyprinus.carpio	
## 1	0	71	65	
## 2	2671	178	798	
## 3	1672	0	2999	
## 4	0	0	5718	
## 5	0	0	1791	
##	Dicamptodon.tenebrosus	Dionda.nigrotaeniata	Dionda.serena	
## 1	0	0	19	
## 2	0	43	0	
## 3	0	0	0	
## 4	0	0	0	
## 5	0	0	0	
##	Dormitator.maculatus	Dorosoma.cephedianum	Dorosoma.petenense	
## 1	1	655	791	
## 2	0	8636	15304	
## 3	0	14290	2116	
## 4	0	6728	6	
## 5	0	771	0	
##	Elassoma.evergladei	Elassoma.zonatum	Elops.saurus	Elotris.amblyopsis
## 1	0	9	153	0
## 2	0	117	6	1
## 3	0	3	0	0
## 4	0	0	0	0
## 5	0	0	0	0
##	Enneacanthus.chaetodon	Enneacanthus.gloriosus	Enneacanthus.obesus	
## 1	0	0	0	
## 2	0	31	4	
## 3	0	230	3	
## 4	0	1	3	
## 5	0	0	0	
##	Entosphenus.lethophagus	Entosphenus.tridentatus	Erimonax.monachus	
## 1	0	0	0	
## 2	0	0	0	
## 3	0	0	16	
## 4	6	61	0	
## 5	0	1	0	
##	Erimystax.dissimilis	Erimystax.harryi	Erimystax.insignis	
## 1	0	0	0	
## 2	1	0	9	
## 3	212	21	146	
## 4	283	0	0	
## 5	0	0	0	
##	Erimystax.x.punctatus	Erimyzon.oblongus	Erimyzon.sucetta	Erimyzon.tenuis
## 1	0	0	15	0
## 2	0	336	39	4
## 3	230	616	9	0
## 4	44	129	17	0
## 5	0	0	0	0
##	Esox.americanus	Esox.americanus.americanus	Esox.americanus.vermiculatus	
## 1	0	0	0	
## 2	321	73	0	

## 3	59		114	24
## 4	571		62	24
## 5	0		0	0
##	<i>Esox.lucius</i>	<i>Esox.masquinongy</i>	<i>Esox.niger</i>	<i>Etheostoma.acuticeps</i>
## 1	0	0	9	0
## 2	0	0	148	0
## 3	14	9	101	34
## 4	396	37	781	0
## 5	359	30	18	0
##	<i>Etheostoma.aquali</i>	<i>Etheostoma.artesiaie</i>	<i>Etheostoma.asprigene</i>	
## 1	0	0	0	
## 2	0	59	4	
## 3	29	22	12	
## 4	0	0	29	
## 5	0	0	0	
##	<i>Etheostoma.autumnale</i>	<i>Etheostoma.baileyi</i>	<i>Etheostoma.barrenense</i>	
## 1	0	0	0	
## 2	0	0	0	
## 3	39	4	7	
## 4	0	0	0	
## 5	0	0	0	
##	<i>Etheostoma.bellum</i>	<i>Etheostoma.blennioides</i>	<i>Etheostoma.blennius</i>	
## 1	0	0	0	
## 2	0	99	0	
## 3	47	3805	457	
## 4	0	2804	0	
## 5	0	0	0	
##	<i>Etheostoma.boschungii</i>	<i>Etheostoma.caeruleum</i>	<i>Etheostoma.camurum</i>	
## 1	0	0	0	
## 2	1	390	0	
## 3	0	6085	321	
## 4	0	2672	16	
## 5	0	239	0	
##	<i>Etheostoma.chlorobranchium</i>	<i>Etheostoma.chlorosomum</i>	<i>Etheostoma.chuckwachatte</i>	
## 1	0	10	0	
## 2	0	29	0	
## 3	240	6	0	
## 4	0	0	0	
## 5	0	0	0	
##	<i>Etheostoma.cinereum</i>	<i>Etheostoma.collettei</i>	<i>Etheostoma.collis</i>	
## 1	0	0	0	
## 2	0	15	8	
## 3	4	0	10	
## 4	0	0	0	
## 5	0	0	0	
##	<i>Etheostoma.colorosum</i>	<i>Etheostoma.coosae</i>	<i>Etheostoma.cragini</i>	
## 1	0	0	0	
## 2	6	129	0	
## 3	0	0	181	
## 4	0	0	0	
## 5	0	0	0	
##	<i>Etheostoma.crossopterygum</i>	<i>Etheostoma.derivativum</i>	<i>Etheostoma.ditrema</i>	
## 1	0	0	0	
## 2	1	0	31	

## 3	77	1	0	
## 4	0	0	0	
## 5	0	0	0	
##	Etheostoma.duryi	Etheostoma.edwini	Etheostoma.euzonum	Etheostoma.exile
## 1	0	0	0	0
## 2	399	8	0	0
## 3	385	0	24	2
## 4	0	0	0	121
## 5	0	0	0	194
##	Etheostoma.flabellare	Etheostoma.flavum	Etheostoma.fragi	Etheostoma.fricksium
## 1	0	0	0	0
## 2	23	0	0	0
## 3	5754	609	11	0
## 4	2140	0	0	0
## 5	60	0	0	0
##	Etheostoma.fusiforme	Etheostoma.gracile	Etheostoma.gutselli	
## 1	0	18	0	
## 2	4	133	0	
## 3	1	41	1	
## 4	1	0	0	
## 5	0	0	0	
##	Etheostoma.histrio	Etheostoma.hopkinsi	Etheostoma.inscriptum	
## 1	0	0	0	
## 2	31	186	273	
## 3	12	0	0	
## 4	0	0	0	
## 5	0	0	0	
##	Etheostoma.jessiae	Etheostoma.jordani	Etheostoma.juliae	
## 1	0	0	0	
## 2	16	76	0	
## 3	65	0	60	
## 4	0	0	0	
## 5	0	0	0	
##	Etheostoma.kantuckeense	Etheostoma.kennicotti	Etheostoma.lachneri	
## 1	0	0	0	
## 2	0	56	5	
## 3	5	0	0	
## 4	0	0	0	
## 5	0	0	0	
##	Etheostoma.lepidum	Etheostoma.longimanum	Etheostoma.luteovinctum	
## 1	27	0	0	
## 2	27	0	0	
## 3	0	35	63	
## 4	0	0	0	
## 5	0	0	0	
##	Etheostoma.lynceum	Etheostoma.maculatum	Etheostoma.mariae	Etheostoma.mihileze
## 1	0	0	0	0
## 2	21	0	0	0
## 3	1	27	0	27
## 4	0	10	0	0
## 5	0	0	0	0
##	Etheostoma.neopterum	Etheostoma.nigripinne	Etheostoma.nigrum	
## 1	0	0	0	
## 2	0	48	12	

## 3	1	77	2039
## 4	0	0	6833
## 5	0	0	1360
##	Ettheostoma.occidentale	Ettheostoma.olmstedii	Ettheostoma.oophylax
## 1	0	0	0
## 2	0	221	0
## 3	9	1983	24
## 4	0	4994	0
## 5	0	0	0
##	Ettheostoma.osburni	Ettheostoma.parvipinne	Ettheostoma.podostemone
## 1	0	0	0
## 2	0	27	0
## 3	1	2	21
## 4	0	0	0
## 5	0	0	0
##	Ettheostoma.proeliare	Ettheostoma.punctulatum	Ettheostoma.radiosum
## 1	0	0	0
## 2	16	0	270
## 3	0	66	0
## 4	0	0	0
## 5	0	0	0
##	Ettheostoma.rafinesquei	Ettheostoma.ramseyi	Ettheostoma.rufilineatum
## 1	0	0	0
## 2	0	68	148
## 3	7	0	7504
## 4	0	0	0
## 5	0	0	0
##	Ettheostoma.rupestre	Ettheostoma.sanguifluum	Ettheostoma.scotti
## 1	0	0	0
## 2	34	0	100
## 3	0	24	0
## 4	0	0	0
## 5	0	0	0
##	Ettheostoma.serrifer	Ettheostoma.simoterum	Ettheostoma.spectabile
## 1	0	0	1
## 2	4	183	358
## 3	3	745	4531
## 4	0	0	156
## 5	0	0	0
##	Ettheostoma.squamiceps	Ettheostoma.stigmaeum	Ettheostoma.swaini
## 1	0	0	0
## 2	0	103	154
## 3	140	63	6
## 4	0	0	0
## 5	0	0	0
##	Ettheostoma.swannanoa	Ettheostoma.tallapoosae	Ettheostoma.tetrazonum
## 1	0	0	0
## 2	0	0	0
## 3	158	0	67
## 4	0	0	0
## 5	0	0	0
##	Ettheostoma.thalassinum	Ettheostoma.tippecanoe	Ettheostoma.variatum
## 1	0	0	0
## 2	56	0	0

## 3	11	35	181	
## 4	0	2	135	
## 5	0	0	0	
##	<i>Etheostoma.virgatum</i>	<i>Etheostoma.vitreum</i>	<i>Etheostoma.vulneratum</i>	
## 1	0	0	0	
## 2	0	0	0	
## 3	19	80	25	
## 4	0	0	0	
## 5	0	0	0	
##	<i>Etheostoma.whipplei</i>	<i>Etheostoma.zonale</i>	<i>Etheostoma.zonistium</i>	
## 1	0	0	0	
## 2	68	0	0	
## 3	77	1585	2	
## 4	0	1986	0	
## 5	0	3	0	
##	<i>Eucinostomus.argenteus</i>	<i>Exoglossum.laurae</i>	<i>Exoglossum.maxilllingua</i>	
## 1	0	0	0	
## 2	14	0	0	
## 3	0	4	366	
## 4	0	15	2828	
## 5	0	0	0	
##	<i>Forbesichthys.agassizii</i>	<i>Fundulus.catenatus</i>	<i>Fundulus.chrysotus</i>	
## 1	0	0	18	
## 2	0	31	8	
## 3	1	824	0	
## 4	0	0	0	
## 5	0	0	0	
##	<i>Fundulus.diaphanus</i>	<i>Fundulus.dispar</i>	<i>Fundulus.euryzonus</i>	<i>Fundulus.grandis</i>
## 1	0	0	0	42
## 2	2	0	0	84
## 3	605	0	0	0
## 4	1196	33	0	0
## 5	25	0	0	0
##	<i>Fundulus.heteroclitus</i>	<i>Fundulus.kansae</i>	<i>Fundulus.lineolatus</i>	<i>Fundulus.notatus</i>
## 1	0	0	0	26
## 2	4	0	0	219
## 3	22	36	19	312
## 4	6	0	0	280
## 5	0	0	0	0
##	<i>Fundulus.olivaceus</i>	<i>Fundulus.rathbuni</i>	<i>Fundulus.sciadicus</i>	<i>Fundulus.seminolis</i>
## 1	0	0	0	128
## 2	623	0	0	1
## 3	552	65	7	0
## 4	0	0	139	0
## 5	0	0	0	0
##	<i>Fundulus.stellifer</i>	<i>Fundulus.stellifera</i>	<i>Fundulus.zebrinus</i>	<i>Gambusia.affinis</i>
## 1	0	0	0	2036
## 2	34	15	705	7148
## 3	0	0	1353	7250
## 4	0	0	497	310
## 5	0	0	0	13
##	<i>Gambusia.holbrooki</i>	<i>Gasterosteus.aculeatus</i>	<i>Gerres.cinereus</i>	<i>Gila.atraria</i>
## 1	1094	0	0	0
## 2	385	0	0	0

## 3	255		200	0	188	
## 4	47		20	0	7	
## 5	0		167	0	0	
##	Gila.bicolor	Gila.cypha	Gila.elegans	Gila.orcuttii	Gila.pandora	Gila.robusta
## 1	0	0	0	0	0	0
## 2	0	0	0	0	287	0
## 3	0	7	0	0	28	269
## 4	49	0	0	0	0	7
## 5	0	0	0	0	0	0
##	Gila.seminuda	Gobioides.broussonnetii	Gobiomorus.dormitor	Gobiosoma.bosc		
## 1	0		1	13		0
## 2	0		0	0		23
## 3	0		0	0		0
## 4	0		0	0		0
## 5	0		0	0		0
##	Hemitremia.flammea	Herichthys.cyanoguttatum	Hesperoleucus.symmetricus			
## 1	0		25			0
## 2	2		0			0
## 3	11		0			1108
## 4	0		0			133
## 5	0		0			0
##	Heterandria.formosa	Hiodon.alosoides	Hiodon.tergisus	Hoplosternum.littorale		
## 1	30	0	0			6
## 2	71	253	0			0
## 3	0	148	53			0
## 4	0	156	63			0
## 5	0	419	5			0
##	Hybognathus.amarus	Hybognathus.argyritis	Hybognathus.hankinsoni			
## 1	0	0	0			0
## 2	6	0	0			0
## 3	0	0	50			
## 4	0	182	1057			
## 5	0	99	442			
##	Hybognathus.hayi	Hybognathus.nuchalis	Hybognathus.placitus	Hybognathus.regius		
## 1	0	0	0	0		0
## 2	4	998	743			1001
## 3	130	1047	1582			714
## 4	0	483	324			39
## 5	0	0	3			0
##	Hybopsis.amblops	Hybopsis.amnis	Hybopsis.dorsalis	Hybopsis.hypsinotus		
## 1	0	0	0	0		0
## 2	87	3	0			46
## 3	2278	17	463			162
## 4	282	0	5754			0
## 5	0	0	841			0
##	Hybopsis.lineapunctata	Hybopsis.rubrifrons	Hybopsis.winchelli	Hybopsis.zanema		
## 1	0	0	0	0		0
## 2	0	325	197			45
## 3	0	4	0			0
## 4	0	0	0			0
## 5	0	0	0			0
##	Hypentelium.etowanum	Hypentelium.nigricans	Hypentelium.roanokense			
## 1	0	0	0			0
## 2	829	381	0			0

## 3	0	4948	8	
## 4	0	4960	0	
## 5	0	345	0	
##	Hypophthalmichthys.molitrix	Hypophthalmichthys.nobilis	Hypostomus.plecostomus	
## 1	3	0	2	
## 2	83	0	0	
## 3	454	5	0	
## 4	48	14	0	
## 5	0	0	0	
##	Hysterocarpus.traskii	Ichthyomyzon.bdelium	Ichthyomyzon.castaneus	
## 1	0	0	0	
## 2	0	0	6	
## 3	83	37	24	
## 4	0	26	62	
## 5	0	0	27	
##	Ichthyomyzon.fossor	Ichthyomyzon.gagei	Ichthyomyzon.greeleyi	
## 1	0	0	0	
## 2	0	180	0	
## 3	1	0	128	
## 4	5	0	0	
## 5	11	0	0	
##	Ichthyomyzon.unicuspis	Ictalurus.furcatus	Ictalurus.punctatus	
## 1	0	18	244	
## 2	0	455	1264	
## 3	4	176	3463	
## 4	12	1	3578	
## 5	12	0	976	
##	Ictiobus.bubalus	Ictiobus.cyprinellus	Ictiobus.niger	Iotichthys.phlegethontis
## 1	68	3	0	0
## 2	867	103	39	0
## 3	1379	207	88	0
## 4	352	157	28	4
## 5	29	24	0	0
##	Labidesthes.sicculus	Lampetra.aepyptera	Lethenteron.appendix	
## 1	96	0	0	
## 2	1331	12	5	
## 3	627	98	185	
## 4	263	2	135	
## 5	3	0	25	
##	Lampetra.richardsoni	Lampetra.tridentata	Lavinia.exilicauda	
## 1	0	0	0	
## 2	0	0	0	
## 3	0	0	2	
## 4	2	0	0	
## 5	9	23	0	
##	Leiostomus.xanthurus	Lepidomeda.copei	Lepidomeda.mollispinis	
## 1	0	0	0	
## 2	4	0	0	
## 3	1	370	0	
## 4	0	0	0	
## 5	0	0	0	
##	Lepidomeda.vittata	Lepisosteus.oculatus	Lepisosteus.osseus	
## 1	0	120	78	
## 2	1	758	778	

## 3	0	115	700	
## 4	0	10	148	
## 5	0	0	0	
##	Lepisosteus.platostomus	Lepisosteus.platyrhincus	Lepomis.auritus	
## 1	1	248	213	
## 2	147	26	6312	
## 3	588	0	9422	
## 4	258	0	5270	
## 5	0	0	10	
##	Lepomis.cyanellus	Lepomis.gibbosus	Lepomis.gulosus	Lepomis.humilis
## 1	400	0	279	4
## 2	5496	89	953	2106
## 3	9914	954	197	1777
## 4	6798	6044	16	1693
## 5	1034	151	0	165
##	Lepomis.macrochirus	Lepomis.marginatus	Lepomis.megalotis	Lepomis.microlophus
## 1	1821	0	1421	714
## 2	14025	534	13583	1288
## 3	11990	35	13841	539
## 4	10269	0	2114	17
## 5	726	0	0	12
##	Lepomis.miniatus	Lepomis.punctatus	Lepomis.symmetricus	Lota.lota
## 1	42	326	76	0
## 2	412	694	36	0
## 3	35	1	1	0
## 4	0	0	0	128
## 5	0	0	0	174
##	Lucania.goodei	Lucania.parva	Luxilus.albeolus	Luxilus.cardinalis
## 1	42	1	0	0
## 2	0	0	0	0
## 3	0	0	1214	1654
## 4	0	0	0	0
## 5	0	0	0	0
##	Luxilus.cerasinus	Luxilus.chrysocephalus	Luxilus.coccogenis	Luxilus.cornutus
## 1	0	0	0	0
## 2	0	1700	4	168
## 3	269	6000	2326	1182
## 4	0	1966	0	10826
## 5	0	9	0	8379
##	Luxilus.pilsbryi	Luxilus.zonatus	Luxilus.zonistius	Lythrurus.ardens
## 1	0	0	0	0
## 2	0	0	97	84
## 3	1637	3316	0	297
## 4	0	0	0	30
## 5	0	0	0	0
##	Lythrurus.atrapiculus	Lythrurus.bellus	Lythrurus.fasciolaris	Lythrurus.fumeus
## 1	0	0	0	2
## 2	1	40	949	93
## 3	0	0	1347	139
## 4	0	0	168	0
## 5	0	0	0	0
##	Lythrurus.lirus	Lythrurus.roseipinnis	Lythrurus.umbratilis	
## 1	0	0	0	
## 2	45	80	133	

## 3	731	0	972	
## 4	0	0	279	
## 5	0	0	0	
##	Macrhybopsis.aestivalis	Macrhybopsis.australis	Macrhybopsis.hyostoma	
## 1	0	0	0	
## 2	16	35	157	
## 3	1	0	16	
## 4	6	0	4	
## 5	0	0	0	
##	Macrhybopsis.marconis	Macrhybopsis.meeki	Macrhybopsis.storeriana	
## 1	1	0	0	
## 2	0	0	271	
## 3	0	0	69	
## 4	0	0	149	
## 5	0	0	2	
##	Margariscus.margarita	Margariscus.nachtriebi	Membras.martinica	Menidia.audens
## 1	0	0	55	3
## 2	0	0	0	79
## 3	0	0	0	119
## 4	2528	0	0	0
## 5	156	21	0	0
##	Menidia.beryllina	Menidia.menidia	Microgadus.tomcod	Microgobius.gulosus
## 1	155	0	0	0
## 2	934	0	0	2
## 3	464	3	0	0
## 4	2	0	2	0
## 5	0	0	0	0
##	Micropterus.cataractae	Micropterus.coosae	Micropterus.dolomieu	
## 1	0	0	7	
## 2	43	215	75	
## 3	0	5	4401	
## 4	0	0	12200	
## 5	0	0	2885	
##	Micropterus.notius	Micropterus.punctulatus	Micropterus.salmoides	
## 1	26	230	763	
## 2	2	2192	3407	
## 3	0	2427	2855	
## 4	0	129	4154	
## 5	0	0	279	
##	Micropterus.treculii	Minytrema.melanops	Misgurnus.anguillicaudatus	
## 1	108	103	1	
## 2	31	720	0	
## 3	0	385	0	
## 4	0	247	1	
## 5	0	0	1	
##	Morone.americana	Morone.chrysops	Morone.mississippiensis	Morone.saxatilis
## 1	0	7	11	4
## 2	52	218	37	69
## 3	1512	428	62	158
## 4	891	374	8	413
## 5	1	73	0	0
##	Moxostoma.anisurum	Moxostoma.breviceps	Moxostoma.carinatum	Moxostoma.cervinum
## 1	0	0	0	0
## 2	13	0	67	0

## 3	339	501	377	178
## 4	1378	222	141	0
## 5	435	0	19	0
##	Moxostoma.collapsum	Moxostoma.congestum	Moxostoma.duquesnii	
## 1	0	62	0	
## 2	221	32	152	
## 3	138	0	1745	
## 4	0	0	797	
## 5	0	0	0	
##	Moxostoma.erythrurum	Moxostoma.lachneri	Moxostoma.macrolepidotum	
## 1	0	0	0	
## 2	407	33	43	
## 3	4562	0	640	
## 4	3919	0	3145	
## 5	877	0	1490	
##	Moxostoma.pappillosum	Moxostoma.pisolabrum	Moxostoma.poecilurum	
## 1	0	0	0	
## 2	0	0	493	
## 3	44	20	37	
## 4	0	0	0	
## 5	0	0	0	
##	Moxostoma.robustum	Moxostoma.rupiscartes	Moxostoma.valenciennesi	
## 1	0	0	0	
## 2	1	122	0	
## 3	0	76	26	
## 4	0	0	99	
## 5	0	0	124	
##	Mugil.cephalus	Mugil.curema	Mylocheilus.caurinus	Mylopharodon.conocephalus
## 1	460	0	0	0
## 2	1475	0	0	0
## 3	10	0	0	152
## 4	1	0	2	65
## 5	0	0	7	0
##	Mylopharyngodon.piceus	Myrophis.punctatus	Neogobius.melanostomus	
## 1	0	0	0	
## 2	0	4	0	
## 3	0	0	0	
## 4	5	0	432	
## 5	0	0	0	
##	Nocomis.asper	Nocomis.biguttatus	Nocomis.effusus	Nocomis.leptocephalus
## 1	0	0	0	0
## 2	0	142	0	2066
## 3	207	587	49	2614
## 4	0	1300	0	0
## 5	0	1622	0	0
##	Nocomis.micropogon	Nocomis.platyrhynchus	Nocomis.raneyi	
## 1	0	0	0	
## 2	0	0	0	
## 3	2901	460	188	
## 4	1445	0	0	
## 5	41	0	0	
##	Notemigonus.crysoleucas	Notropis.alborus	Notropis.altipinnis	
## 1	55	0	0	
## 2	814	4	20	

## 3	516	61	105	
## 4	2095	0	0	
## 5	170	0	0	
##	Notropis.amabilis	Notropis.ammophilus	Notropis.amoenus	Notropis.amplamala
## 1	30	0	0	0
## 2	11	62	0	13
## 3	0	0	149	0
## 4	0	0	51	0
## 5	0	0	0	0
##	Notropis.ariommus	Notropis.atherinoides	Notropis.atrocaudalis	
## 1	0	3	0	
## 2	0	3730	49	
## 3	162	9300	0	
## 4	0	29384	0	
## 5	0	1352	0	
##	Notropis.baileyi	Notropis.bairdi	Notropis.bifrenatus	Notropis.blennius
## 1	0	0	0	0
## 2	88	480	0	3
## 3	0	634	1	937
## 4	0	0	30	2255
## 5	0	0	1	0
##	Notropis.boops	Notropis.braytoni	Notropis.buccatus	Notropis.buccula
## 1	0	3	0	0
## 2	536	0	169	69
## 3	1312	0	872	0
## 4	88	0	402	0
## 5	0	0	0	0
##	Notropis.buchanani	Notropis.cahabae	Notropis.candidus	Notropis.chalybaeus
## 1	0	0	0	0
## 2	108	39	147	56
## 3	57	0	0	3
## 4	2	0	0	8
## 5	0	0	0	0
##	Notropis.chiliticus	Notropis.chlorocephalus	Notropis.chrosomus	
## 1	0	0	0	
## 2	0	122	131	
## 3	72	2	2	
## 4	0	0	0	
## 5	0	0	0	
##	Notropis.cummingsae	Notropis.edwardraneyi	Notropis.girardi	Notropis.greeniei
## 1	0	0	0	0
## 2	257	31	0	0
## 3	26	0	18	111
## 4	0	0	0	0
## 5	0	0	0	0
##	Notropis.harperi	Notropis.heterodon	Notropis.heterolepis	Notropis.hudsonius
## 1	0	0	0	0
## 2	242	0	0	438
## 3	0	0	2	3678
## 4	0	0	98	12309
## 5	0	86	591	312
##	Notropis.hypsilepis	Notropis.leuciodus	Notropis.longirostris	
## 1	0	0	0	
## 2	31	5	662	

## 3	0	2663	6	
## 4	0	0	0	
## 5	0	0	0	
##	Notropis.lutipinnis	Notropis.maculatus	Notropis.micropteryx	Notropis.nubilus
## 1	0	0	0	0
## 2	1910	15	0	13
## 3	74	0	167	1322
## 4	0	0	0	5
## 5	0	0	0	0
##	Notropis.oxyrhynchus	Notropis.ozarcanus	Notropis.percobromus	
## 1	0	0	0	
## 2	143	0	0	
## 3	0	1	239	
## 4	0	0	173	
## 5	0	0	84	
##	Notropis.perpallidus	Notropis.petersoni	Notropis.photogenis	Notropis.potteri
## 1	0	204	0	0
## 2	0	1336	9	0
## 3	0	0	575	2
## 4	0	0	407	0
## 5	0	0	0	0
##	Notropis.procne	Notropis.rafalesquei	Notropis.rubellus	Notropis.rubricroceus
## 1	0	0	0	0
## 2	53	5	6	0
## 3	1230	0	3036	91
## 4	975	0	1437	0
## 5	0	0	42	0
##	Notropis.sabinae	Notropis.scabriceps	Notropis.scepticus	Notropis.semperasper
## 1	0	0	0	0
## 2	57	0	186	0
## 3	0	5	57	20
## 4	0	0	0	0
## 5	0	0	0	0
##	Notropis.shumardi	Notropis.simus	Notropis.spectrunculus	Notropis.stilbicus
## 1	0	0	0	0
## 2	215	0	0	105
## 3	4	8	109	0
## 4	0	0	0	0
## 5	0	0	0	0
##	Notropis.stramineus	Notropis.suttkusi	Notropis.telescopus	Notropis.texasus
## 1	3	0	0	0
## 2	63	12	10	2640
## 3	7835	0	4835	1
## 4	16648	0	0	12
## 5	2594	0	0	0
##	Notropis.topeka	Notropis.uranoscopus	Notropis.volucellus	Notropis.wickliffi
## 1	0	0	45	0
## 2	0	161	574	0
## 3	36	0	2835	619
## 4	4	0	10368	1253
## 5	0	0	961	0
##	Notropis.xaenoccephalus	Noturus.albater	Noturus.elegans	Noturus.eleutherus
## 1	0	0	0	0
## 2	150	0	0	5

## 3	0	158	52	139
## 4	0	0	0	10
## 5	0	0	0	0
##	Noturus.exilis	Noturus.flavater	Noturus.flavus	Noturus.funebris
## 1	0	0	0	0
## 2	110	0	0	27
## 3	1008	12	219	0
## 4	1	0	388	0
## 5	0	0	140	0
##	Noturus.gyrinus	Noturus.hildebrandi	Noturus.insignis	Noturus.lachneri
## 1	2	0	0	0
## 2	73	11	173	3
## 3	27	0	1224	0
## 4	100	0	1407	0
## 5	121	0	155	0
##	Noturus.leptacanthus	Noturus.miurus	Noturus.munitus	Noturus.nocturnus
## 1	1	0	0	0
## 2	181	46	3	124
## 3	0	62	0	42
## 4	0	16	0	13
## 5	0	0	0	0
##	Noturus.phaeus	Noturus.stigmosus	Oncorhynchus.clarkii	
## 1	0	0	0	
## 2	67	0	0	
## 3	12	0	18	
## 4	0	2	191	
## 5	0	0	2218	
##	Oncorhynchus.clarkii.bouvieri	Oncorhynchus.clarkii.clarkii		
## 1		0	0	
## 2		0	0	
## 3		0	0	
## 4		116	0	
## 5		2	1	
##	Oncorhynchus.clarkii.henshawi	Oncorhynchus.clarkii.lewisi		
## 1		0	0	
## 2		0	0	
## 3		20	0	
## 4		0	0	
## 5		0	2	
##	Oncorhynchus.clarkii.pleuriticus	Oncorhynchus.clarkii.utah		
## 1		0	0	
## 2		0	0	
## 3		8	0	
## 4		0	11	
## 5		0	0	
##	Oncorhynchus.clarkii.virginalis	Oncorhynchus.kisutch	Oncorhynchus.mykiss	
## 1		0	0	0
## 2		0	0	353
## 3		0	0	761
## 4		0	104	2308
## 5		0	1792	1432
##	Oncorhynchus.mykiss.gairdnerii	Oncorhynchus.nerka	Oncorhynchus.tshawytscha	
## 1		0	0	0
## 2		0	0	0

## 3		0	20	5
## 4		0	0	245
## 5		0	0	1093
##	Opsopoeodus.emiliae	Oreochromis.aureus	Oreochromis.mossambicus	
## 1	78	50	0	
## 2	297	0	0	
## 3	5	0	0	
## 4	0	0	0	
## 5	0	0	0	
##	Orthodon.microlepidotus	Paralichthys.lethostigma	Perca.flavescens	
## 1	0	0	0	
## 2	0	16	119	
## 3	12	1	560	
## 4	0	0	6705	
## 5	0	0	1661	
##	Percina.aurantiaca	Percina.aurolineata	Percina.burtoni	Percina.caprodes
## 1	0	0	0	0
## 2	0	6	1	218
## 3	147	0	15	1651
## 4	0	0	0	1541
## 5	0	0	0	879
##	Percina.carbonaria	Percina.copelandi	Percina.crassa	Percina.cymatotaenia
## 1	49	0	0	0
## 2	37	1	47	0
## 3	0	29	28	5
## 4	0	15	0	0
## 5	0	0	0	0
##	Percina.evides	Percina.gymnocephala	Percina.kathae	Percina.lenticula
## 1	0	0	0	0
## 2	1	0	96	3
## 3	340	46	0	0
## 4	24	0	0	0
## 5	13	0	0	0
##	Percina.macrocephala	Percina.macrolepida	Percina.maculata	Percina.nasuta
## 1	0	6	0	0
## 2	0	155	16	0
## 3	74	21	152	36
## 4	330	0	1424	0
## 5	0	0	835	0
##	Percina.nevisense	Percina.nigrofasciata	Percina.notogramma	
## 1	0	3	0	
## 2	0	3659	0	
## 3	11	0	26	
## 4	0	0	1	
## 5	0	0	0	
##	Percina.oxyrhynchus	Percina.palmaris	Percina.pantherina	Percina.peltata
## 1	0	0	0	0
## 2	0	19	2	0
## 3	46	0	0	60
## 4	0	0	0	692
## 5	0	0	0	0
##	Percina.phoxocephala	Percina.rex	Percina.roanoka	Percina.sciera
## 1	0	0	0	0
## 2	25	0	0	209

## 3	209	5	477	174
## 4	179	0	0	11
## 5	10	0	0	0
##	Percina.shumardi	Percina.squamata	Percina.stictogaster	Percina.suttkusi
## 1	0	0	0	0
## 2	16	0	0	14
## 3	15	6	2	0
## 4	3	0	0	0
## 5	5	0	0	0
##	Percina.tanasi	Percina.uranidea	Percina.vigil	Percopsis.omiscomaycus
## 1	0	0	0	0
## 2	0	1	1	0
## 3	7	0	2	166
## 4	0	0	0	287
## 5	0	0	0	494
##	Percopsis.transmontana	Petromyzon.marinus	Phenacobius.catostomus	
## 1	0	0	0	
## 2	0	0	12	
## 3	0	60	0	
## 4	9	135	0	
## 5	2	33	0	
##	Phenacobius.crassilabrum	Phenacobius.mirabilis	Phenacobius.teretulus	
## 1	0	0	0	
## 2	0	112	0	
## 3	23	1159	11	
## 4	0	367	0	
## 5	0	0	0	
##	Phenacobius.uranops	Phoxinus.eos	Phoxinus.erythrogaster	Phoxinus.oreas
## 1	0	0	0	0
## 2	0	0	0	0
## 3	69	0	543	58
## 4	0	0	484	0
## 5	0	13	0	0
##	Pimephales.notatus	Pimephales.promelas	Pimephales.tenellus	Pimephales.vigilax
## 1	0	0	0	215
## 2	980	840	1	6330
## 3	16819	5430	165	3763
## 4	18156	8186	0	509
## 5	755	4632	0	0
##	Plagopterus.argentissimus	Platichthys.stellatus	Platygobio.gracilis	
## 1	0	0	0	
## 2	0	0	10	
## 3	0	0	594	
## 4	0	0	302	
## 5	0	0	979	
##	Poecilia.latipinna	Poecilia.sphenops	Pogonichthys.macrolepidotus	
## 1	785	1	0	
## 2	294	0	0	
## 3	0	0	2	
## 4	0	0	0	
## 5	0	0	0	
##	Polyodon.spathula	Pomoxis.annularis	Pomoxis.nigromaculatus	
## 1	0	67	26	
## 2	2	460	310	

## 3	0	261	188
## 4	2	140	637
## 5	0	9	223
##	Prosopium.cylindraceum	Prosopium.williamsoni	Pteronotropis.euryzonus
## 1	0	0	0
## 2	0	0	1
## 3	0	100	0
## 4	0	515	0
## 5	0	672	0
##	Pteronotropis.grandipinnis	Pteronotropis.hypsellopterus	
## 1	0	0	
## 2	8	395	
## 3	0	0	
## 4	0	0	
## 5	0	0	
##	Pteronotropis.signipinnis	Pteronotropis.stonei	Pterygoplichthys.disjunctivus
## 1	0	0	46
## 2	0	121	0
## 3	0	0	0
## 4	0	0	0
## 5	0	0	0
##	Pterygoplichthys.multiradiatus	Ptychocheilus.grandis	Ptychocheilus.lucius
## 1	1	0	0
## 2	0	0	0
## 3	0	404	2
## 4	0	139	0
## 5	0	0	0
##	Ptychocheilus.oregonensis	Pungitius.pungitius	Pylodictis.olivaris
## 1	0	0	20
## 2	0	0	502
## 3	0	0	1009
## 4	66	0	248
## 5	474	20	0
##	Rhinichthys.atratulus	Rhinichthys.cataractae	Rhinichthys.cobitis
## 1	0	0	0
## 2	21	2	4
## 3	5987	3721	0
## 4	22513	12148	0
## 5	1597	3224	0
##	Rhinichthys.falcatus	Rhinichthys.obtusius	Rhinichthys.osculus
## 1	0	0	0
## 2	0	0	533
## 3	36	2051	2318
## 4	0	2663	2804
## 5	0	1145	2964
##	Rhinichthys.umatilla	Richardsonius.balteatus	Richardsonius.egregius
## 1	0	0	0
## 2	0	0	0
## 3	0	408	47
## 4	103	1740	27
## 5	2	1017	0
##	Salmo.salar	Salmo.trutta	Salvelinus.confluentus
## 1	0	0	0
## 2	0	342	0

```

## 3      0      1314      0      214
## 4     410     4360      0     2733
## 5      8     1434      8     893
##  Salvelinus.malma Sander.canadensis Sander.vitreus Scaphirhynchus.platorynchus
## 1      0      0      0      0
## 2      0      7     20      0
## 3      0     212     55     18
## 4      0     161    1255     59
## 5      0     121     402      8
##  Sciaenops.ocellatus Semotilus.atromaculatus Semotilus.corporalis
## 1      0      0      0
## 2      4     1857      0
## 3     44    14206     448
## 4      0    23365     9261
## 5      0     3858     1243
##  Semotilus.lumbee Semotilus.thoreauianus Strongylura.marina Taricha.granulosa
## 1      0      0      9      0
## 2      0     101     39      0
## 3      5      0     11      0
## 4      0      0      0      0
## 5      0      0      0      0
##  Thaleichthys.pacificus Thoburnia.atripinnis Thoburnia.rhothoeca
## 1      0      0      0
## 2      0      0      0
## 3      0      5     402
## 4      0      0      0
## 5      0      0      0
##  Thymallus.arcticus Tinca.tinca Trinectes.maculatus Umbra.limi Umbra.pygmaea
## 1      0      0     23      0      0
## 2      0      0    121      0      3
## 3      0      0     21      1    139
## 4      0      0      3    3000     67
## 5      0      4      0     594      0
##  Xyrauchen.texanus
## 1      0
## 2      0
## 3      5
## 4      0
## 5      0

```

```

result <- cbind(result[,1], result[,3:638])

new_result <- result[,2:637]

colnames(result)[1] <- "Latitude Group"

new_result <- result[,2:637]

fish.db1 <- vegdist(new_result, method = "bray", upper = TRUE, diag = TRUE)

Fishdbmatrix <- as.matrix(fish.db1)

order2 <- rev(rownames(Fishdbmatrix))

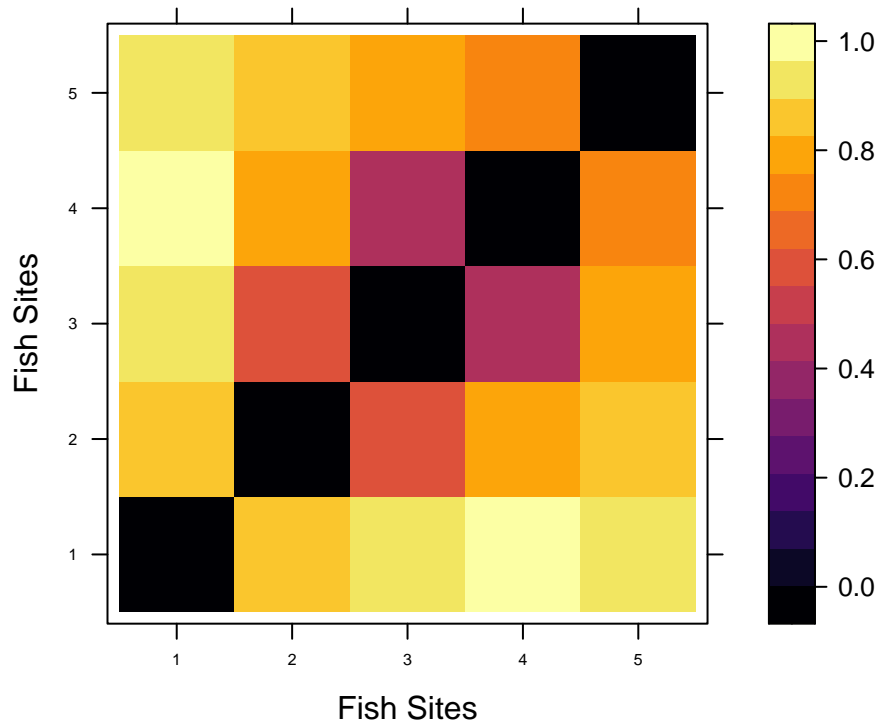
```

```

levelplot(Fishdbmatrix,
          aspect = "iso",
          col.regions = inferno,
          xlab = "Fish Sites",
          ylab = "Fish Sites",
          scales = list(cex = 0.5),
          main = "Bray-Curtis Distance")

```

Bray-Curtis Distance



Synthesis Question > Answer

answer : The graph above represents the five sites that we divided our data into. We did this because having over 2,700 sites made it very difficult to properly visualize the data. We grouped the data by the latitude at which the fish data was collected. Site 1 represents latitudes 25-30. Site 2 represents latitudes 30-35. Site 3 represents latitudes 35-40. Site 4 represents latitudes 40-45. Site 5 represents latitudes 45-50. The figure shows that the fish communities were most similar between sites 3 and 4. Sites 2 and 3 were somewhat similar. Comparisons between the rest of the combinations of other sites show much dissimilarity. This implies that fish communities are most similar at latitudes between 35 and 45, which corresponds to the central region of the United States. On the other hand, fish communities in the north are more dissimilar compared to the southern communities. Something I found interesting was that fish communities at sites 1 and 2 and 4 and 5 were more dissimilar. Meaning that even though they are in the same geographic region, they are not as similar. This could be since species living in very northern or southern regions must have the ability to endure very hot and cold temperatures, resulting in less species being able to live in those areas. Based on this information, I hypothesize that biodiversity is lower in the northern southern latitudes due to extreme weather conditions preventing many species from inhabiting that area.

```

#fish.cluster <- hclust(fish.db1, method = "ward.D2")

#par(mar = c(1, 5, 2, 2) + 0.1)
#plot(fish.cluster, main = "Fish Sites by Latitude",
      #ylab = "Squared Bray-Curtis Distance")

```