



Executive Summary Report 3

Course Name: ALY6000 Introduction to Analytics

CRN 22279

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A] Following an introduction, provide an analysis of descriptive characteristics of the data set provided by your instructor. This includes pertinent statistics including counts, cumulative counts, frequency, percentages, etc. Include R console screen snippets to support your observations and conclusions. Below is a sample excerpt.

INTRODUCTION:

The following project aims at providing you a comprehensive understanding of how to use statistics in R as well as a thorough understanding of various key RStudio visualizations.

The objective of the project is to learn:

- ➔ How to load and install several libraries.
- ➔ How to apply different functions on the table i.e. create, convert, extract values'.
- ➔ With a given set of instructions, how to create different visualizations like Bar plots and a Pareto chart.
- ➔ How to display unique elements of a column.

Implementation:

```
> str(bio)
'data.frame': 676 obs. of 7 variables:
 $ netID : int 12 12 12 12 12 12 12 13 13 13 ...
 $ fishID : int 16 23 30 44 50 65 66 68 69 70 ...
 $ species: chr "Bluegill" "Bluegill" "Bluegill" "Bluegill" ...
 $ tl : int 61 66 70 38 42 54 27 36 59 39 ...
 $ w : num 2.9 4.5 5.2 0.5 1 2.1 NA 0.5 2 0.5 ...
 $ tag : chr "" "" "" "" ...
 $ scale : logi FALSE FALSE FALSE FALSE FALSE FALSE ...
>
```

Explanation:

- ➔ The inchBio.csv file contains the unique net identification number, unique fish identification number, species, total length (mm), weight (g), tag "number," and whether scales were sampled or not for individual fish captured.
- ➔ The structure of the data.frame is seen with str(), as shown above. The structure shows the number of observations or rows (676) and variables or columns (7) in the

data.frame and the names (following \$), data type (or class), and the first few values of each variable. For example, netID is the first variable, it is an integer type, and the first three values are 12, 12, and 12 .

→ The Data frame has 676 observations and 7 variables.

Implementation :

```
> summary(bio)
 netID      fishID      species      tl      w
Min.   : 1.00   Min.   : 7.0   Length:676   Min.   : 27.0   Min.   : 0.2
1st Qu.: 13.00  1st Qu.:175.8   Class :character 1st Qu.: 66.0   1st Qu.: 2.0
Median : 37.00  Median :345.5   Mode  :character Median :189.5   Median : 54.5
Mean   : 67.65  Mean   :434.2           Mean :186.5   Mean   :126.8
3rd Qu.:109.00 3rd Qu.:695.5           3rd Qu.:295.0 3rd Qu.: 190.5
Max.   :206.00 Max.   :915.0           Max.   :429.0 Max.   :1070.0
                                     NA's   :165

 tag      scale
Length:676   Mode :logical
Class :character FALSE:213
Mode  :character  TRUE:463
```

Explanation :

- The summary() function produces a numerical summary of each variable in the dataset.
- From the above summary, we can see that the lightest fish weighs around 0.2 g whereas the heaviest weight 1070.0 g. Similarly the largest fish lengths 429.0 mm while, the shortest is just 27.0 mm.
- The mean weight of fishes is seen as 126.8 g and the median weight as 54.5 g.
- The mean total length of the fishes is 186.5 mm and the median length is 189.5 mm.
- 1st Quartile and 3rd Quartile weights are found to be 2.0 g and 190.5 g respectively for the fishes.
- 1st Quartile and 3rd Quartile for total length are 66.0 mm and 295.0 mm.
- Since species, tag and scale are not numerical, hence it does not give the mean, median, or quartile of data. It gives the count of a number of values in it. So we get to know that there are 676 entries in the species table, the tag also has 676 entries and the scale has 213 entries of false and 463 entries of true.

Implementation :

```
> table(counts)
counts
Black Crappie      Bluegill Bluntnose Minnow      Iowa Darter  Largemouth Bass
          36          220          103          32          228
Pumpkinseed  Tadpole Madtom      Yellow Perch
          13           6          38
> |
```

Explanation:

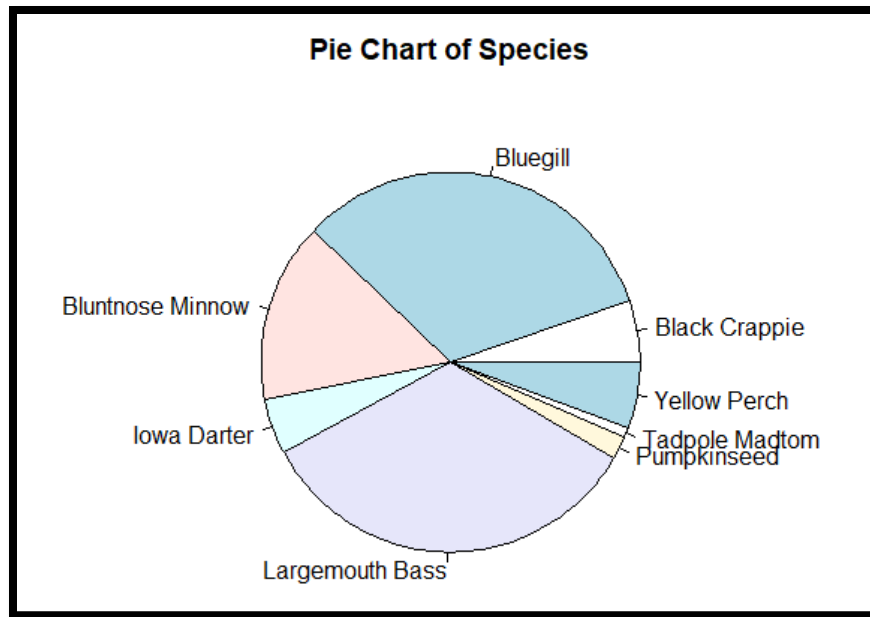
- ➔ In order to know a unique number of species in the dataset and their count, we use the count () function on the species entries.
- ➔ We get to know that the dataset contains 8 different species those are, Black Crappie, Bluegill, Bluntnose Minnow, Iowa Darter, Largemouth Bass, Pumpkinseed, Tadpole Madtom, and Yellow Perch.
- ➔ We also got to know that Largemouth Bass is the most with a count of 228 and Tadpole Madtom least with a count of 6.
- ➔ The second most are the Bluegill with a count of 220 and the second least with a count of 12 is the Pumpkinseed.

Implementation:

```
> cSpecPct
Black Crappie      Bluegill Bluntnose Minnow      Iowa Darter  Largemouth Bass
    5.325444      32.544379      15.236686      4.733728      33.727811
Pumpkinseed  Tadpole Madtom      Yellow Perch
    1.923077      0.887574      5.621302
> |
```

Explanation:

- ➔ To get the percentage of records for each species we divide the total length of species and multiply by 100.
- ➔ We store it in a new table cSpecPct.
- ➔ From the results, we can see that Largemouth Bass comprises 33.72% of the total species which is mostly occupied by a species.
- ➔ Bluegill is the second-largest occupying species with 32.54% of the total.
- ➔ The least occupancy is of Tadpole Madtom with just 0.88% of the total.
- ➔ Below is the pie chart representation of the above analysis.



Implementation:

```

> t$Freq
[1] 36 220 103 32 228 13 6 38
> desc <- t[order(-t$Freq),]
> desc$Freq
[1] 228 220 103 38 36 32 13 6
> d <- d %>% mutate(cumFreq=cumsum(d$RelFreq), counts=desc$Freq, cumcounts = cumsum(desc$Freq))
> d
  Species RelFreq cumFreq counts cumcounts
5 Largemouth Bass 33.727811 33.72781 228 228
2 Bluegill 32.544379 66.27219 220 448
3 Bluntnose Minnow 15.236686 81.50888 103 551
8 Yellow Perch 5.621302 87.13018 38 589
1 Black Crappie 5.325444 92.45562 36 625
4 Iowa Darter 4.733728 97.18935 32 657
6 Pumpkinseed 1.923077 99.11243 13 670
7 Tadpole Madtom 0.887574 100.00000 6 676

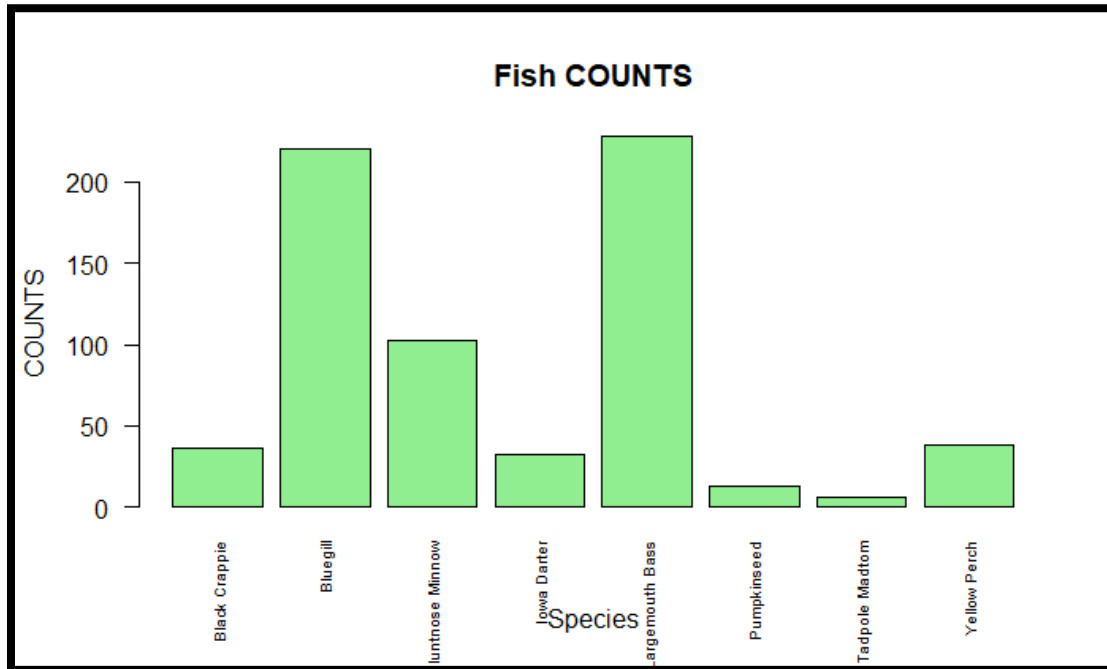
```

Explanation:

- ➔ To get a Pareto chart, we first have to find the count and then the cumulative count for all the species.
- ➔ We identified the counts of each species in the above output and then calculated cumulative counts (cumcounts) by adding the next count to the preceding.
- ➔ This is further used for a Pareto Chart.

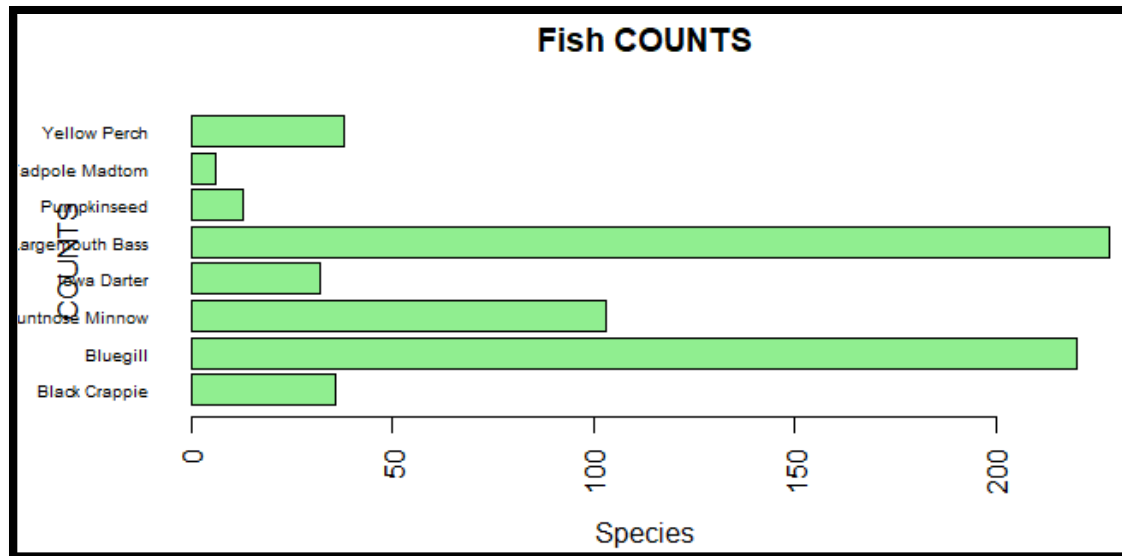
B] Provide the executive with visualizations (at least 3) that help them see the key characteristics you want to highlight. They can be boxplots, histograms, frequency and probability distributions, or bar plots (bar charts). A Pareto plot as illustrated below must be included in this part of your report. Include screen snippets of your plots to support your findings and conclusions. The goal is not only to present your visual results but also to explain their significance of them.

Implementation of Plot 1



Explanation:

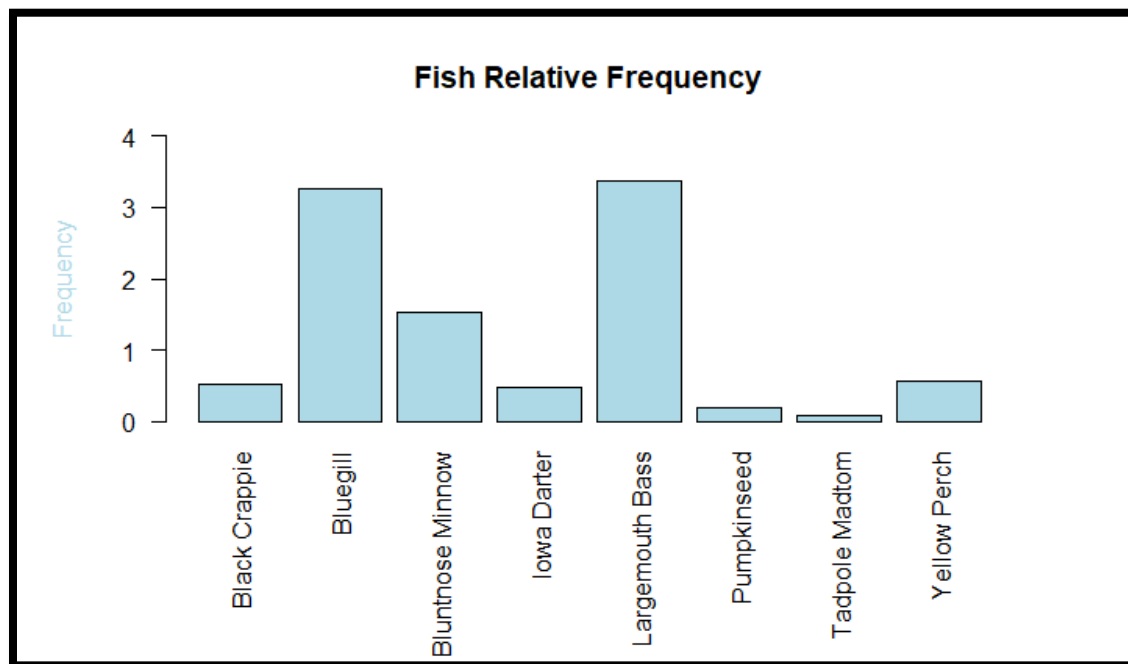
- The above graph is of the counts of species.
- From the bar, we can see that Largemouth Bass and Bluegill has more count as compared to others.
- Tadpole Madtom and Pumpkinseed have the least count than others.
- We can say that Tadpole Madtom and Pumpkinseed are extinct species whereas Largemouth Bass and Bluegill are dormant species.



→ This is the graph that was required as per the question.

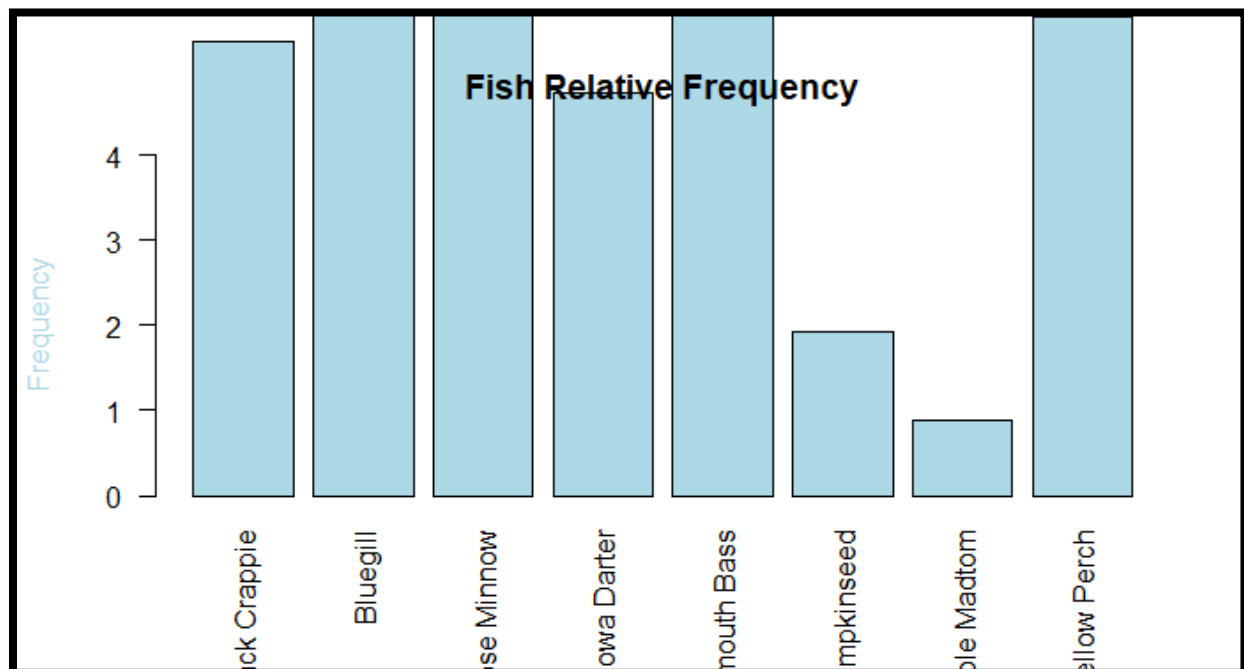
→ But this graph might be wrong due to various reasons. Hence I am submitting both the graphs.

Plot 2-



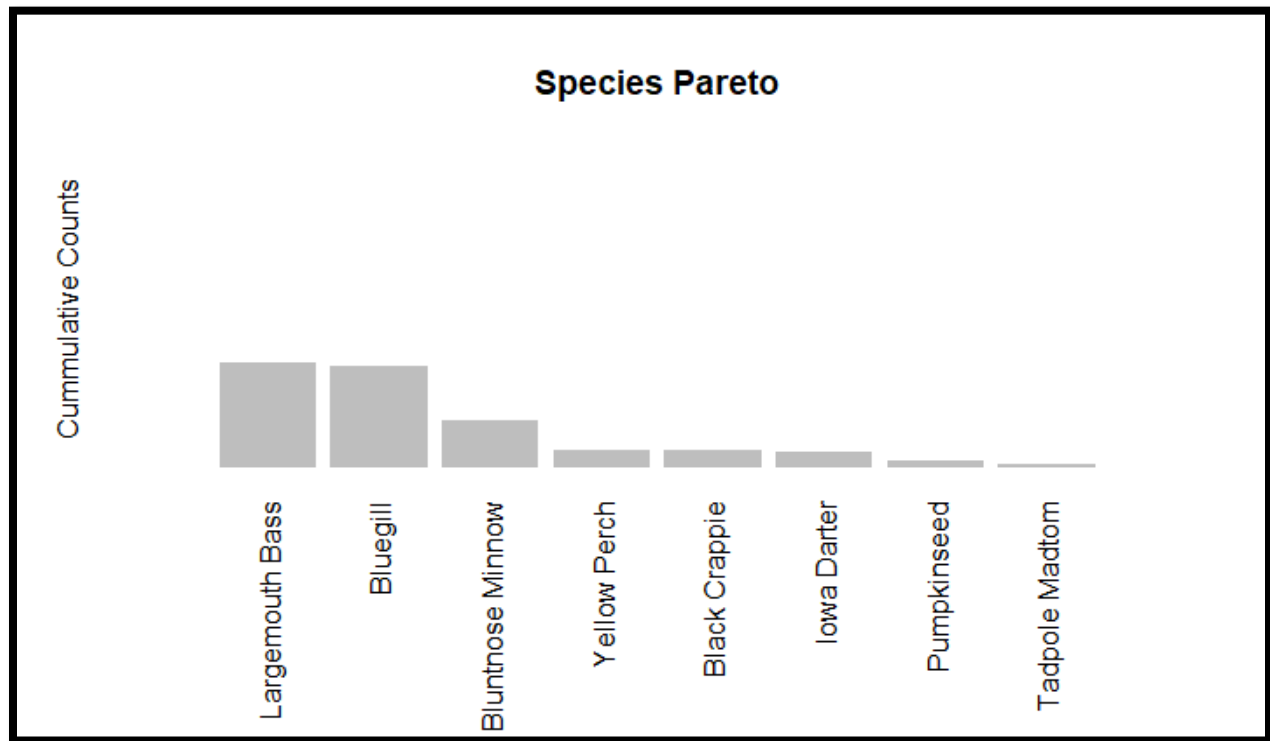
Explanation:

- ➔ The above graph is of the relative frequency of the fish.
- ➔ From the bar, we can see that Largemouth Bass and Bluegill have a high relative frequency.
- ➔ The chances we will see Largemouth Bass or Bluegill species is much higher than the others.
- ➔ Tadpole Madtom and Pumpkinseed have a less relative frequency.
- ➔ We have a significantly lower likelihood of seeing Tadpole Madtom or Pumpkinseed than the others.
- ➔ Apart from the above Bluntnose Minnow too have a good relative frequency.
- ➔ Since the question demanded the Y-axis limit to be from 0 to 4, we had to divide the dataset by 10 in order to meet the requirements.
- ➔ On not dividing the dataset by 10, we get the output shown below.



Implementation:

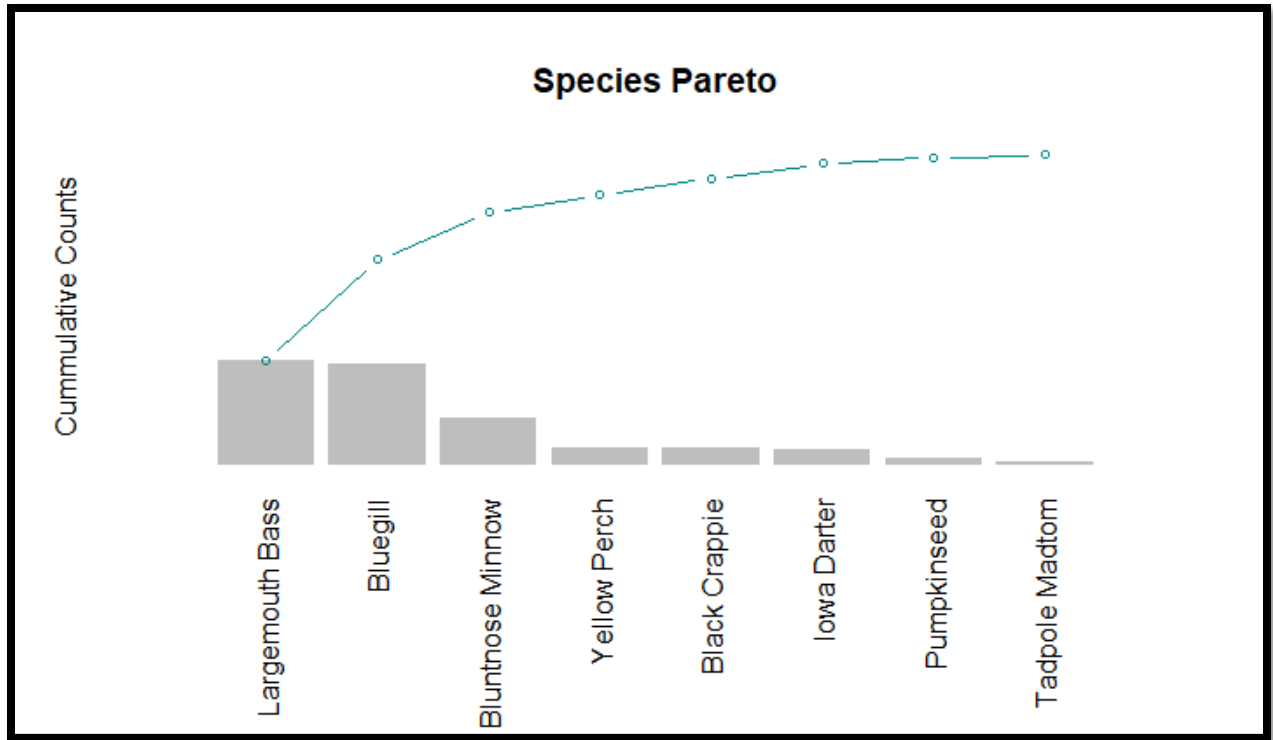
A stepwise procedure for representing a Pareto Chart is represented below.



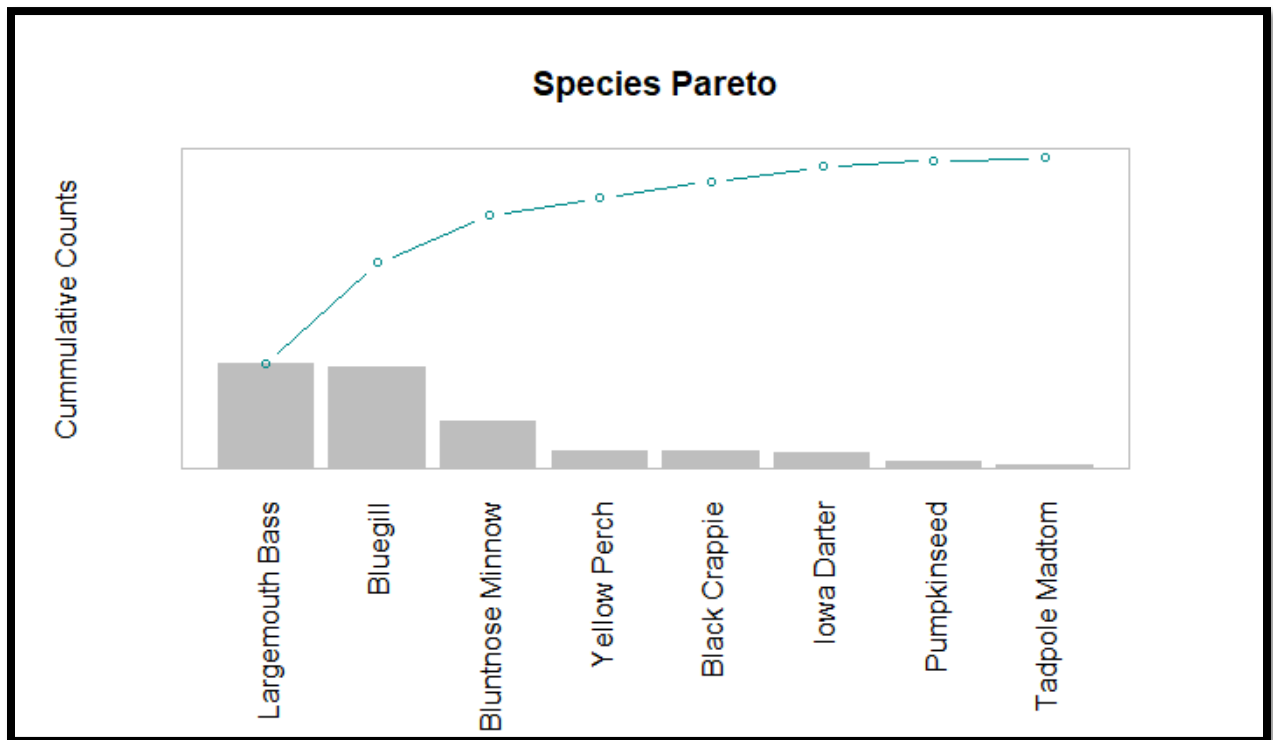
➔ This is a normal bar graph that has only 2 axes as of now i.e. X-axis and the Y-axis.

➔ The X-axis is represented by the name of the species and on the Y-axis is the cumulative count.

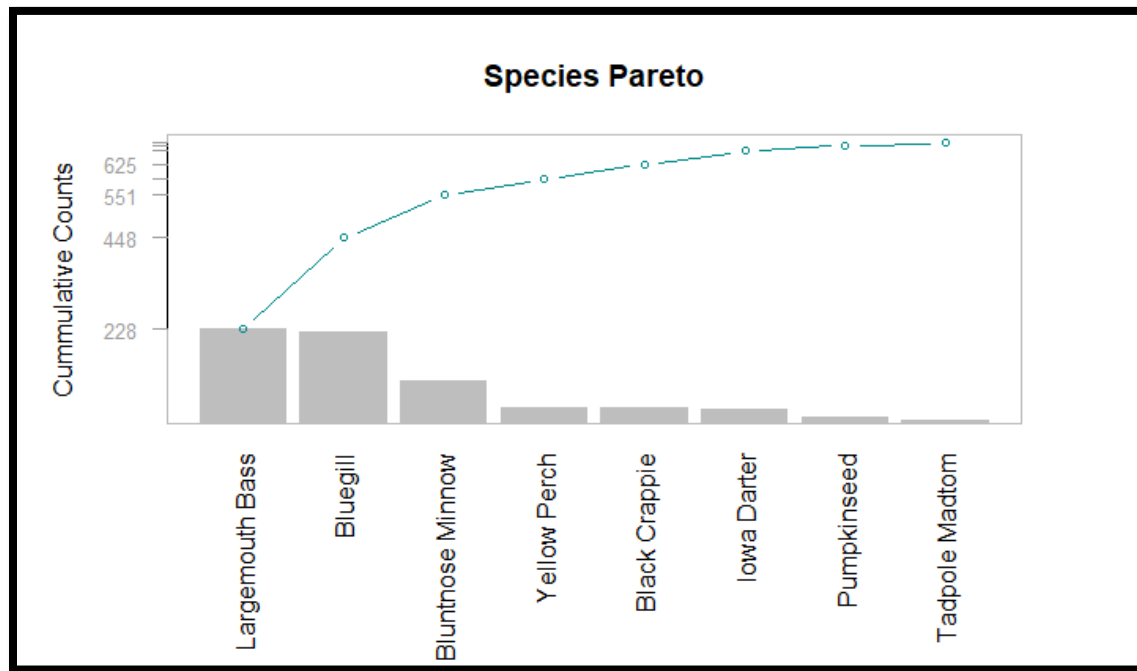
➔ To make it a Pareto chart, the bars are arranged in descending order.



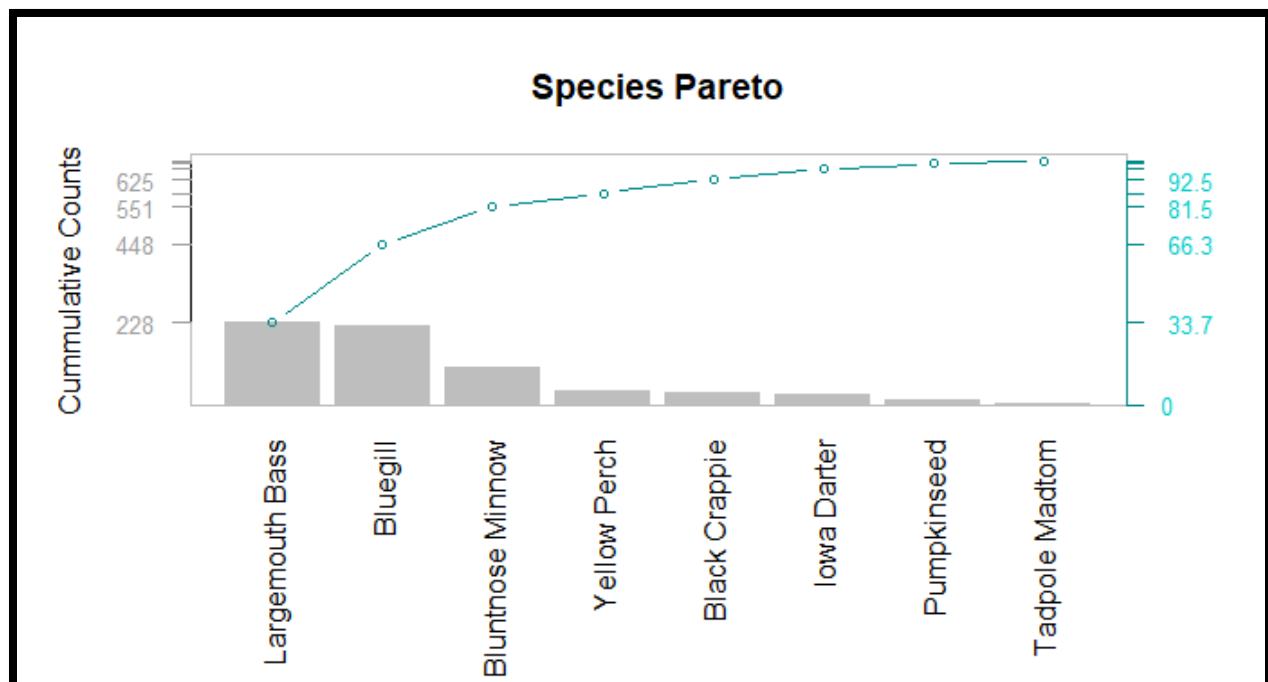
➔ A line graph is plotted here along with the bar graph.



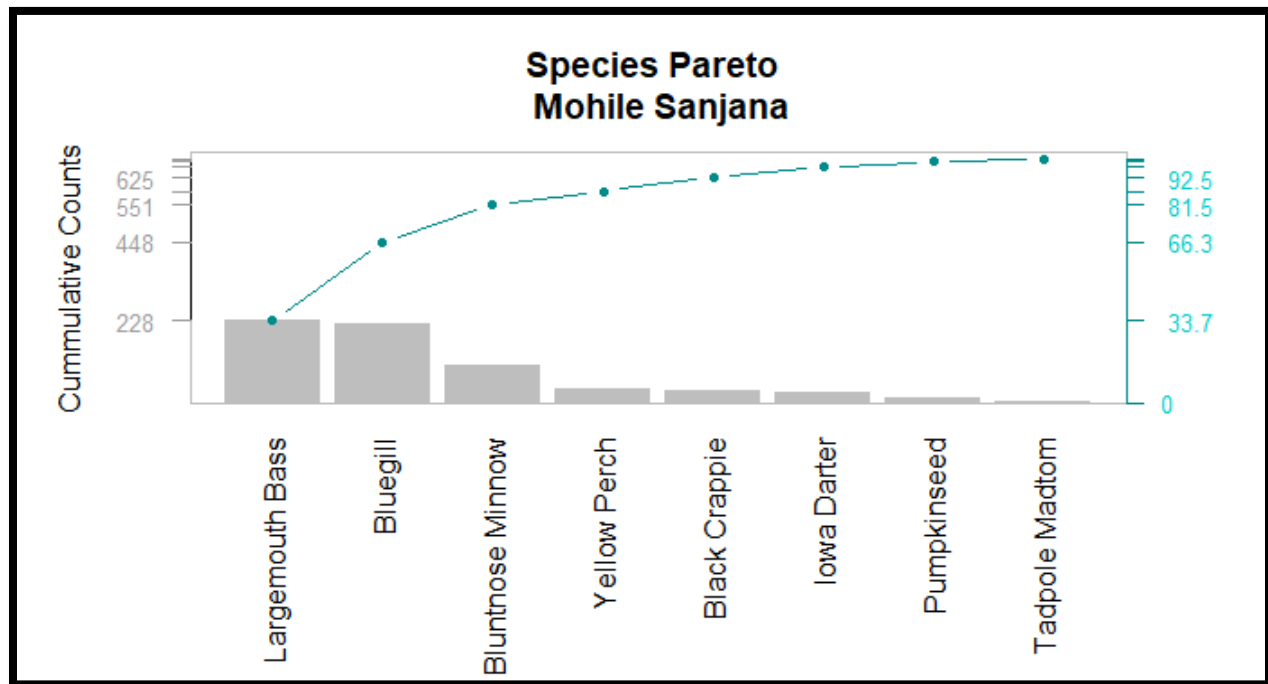
→ Here, only a box is plotted around the graph to give it a border.



→ The cumulative counts is shown on the Y-axis on the graph above.



➔ The right-side Y-axis is shown in this graph and it indicates the cumulative frequency.



➔ Pareto Charts are basically bar graphs whose lengths represent frequency and are arranged with longest bars on the left and shortest on the right. These graphs give statistical information on a bulk of information for each category.

➔ Fundamental properties of a Pareto Chart –

1. The Pareto distribution chart is a very long right-hand tail.
2. The Pareto chart is plotted on three axes, two y-axes, and one x-axis.
3. Pareto consists of a Bar and line chart.
4. The Bar Chart is plotted in the order of most frequent to least frequent occurrences.

➔ The X-axis stands for the 'Species' in the graph which is arranged from dominant to extinct species in numbers.

➔ From the Pareto chart, we see that the Largemouth Bass, Bluegill, and Bluntnose Minnow are the most dominant of all the species.

➔ As seen from the chart, Largemouth Bass comprises 33.72% of the total species, followed by Bluegill (32.54%) and Bluntnose Minnow (15.23%).

➔ The remaining species constitute 18.51% of the total species which is less than 20%.

- ➔ This means that three species constitute over 80% of the total species. Hence these three are the dominant species.
- ➔ From the research, we get to know that Largemouth Bass, Bluegill, and Bluntnose Minnow are carnivorous in nature and prey on fishes and their eggs.
- ➔ And on the other hand, Pumpkinseed and Tadpole Madtom survive on algae and insects.
- ➔ By this, we can conclude the dominance of these fishes(that Largemouth Bass, Bluegill, and Bluntnose Minnow) and the extinction of the others.

C. Finally, provide a clear two to three-sentence paragraph summary of the key points that you want the audience to walk away with regarding your analysis. This summary should present accurate analysis and be supported by the data presented in the rest of the report.

The inchBio dataset gave quite a few insights. The dataset showed that Largemouth Bass, Bluegill, and Bluntnose Minnow are widely found and are dominant species. Largemouth Bass, Bluegill, and Bluntnose Minnow round out the top three and account for over 80% of the total species. The remaining species account for less than 20% of the total species. Largemouth Bass and Bluegill alone constituted 66.26% of the total species. Apart from this, the maximum length of the fish is 429.0 mm and most weighted fish is 1070.0 g.

Bibliography

1. Finding unique values in a column in R (Nizamuddin Siddiqui, August 2020)
<https://www.tutorialspoint.com/how-to-find-the-unique-values-in-a-column-of-an-r-data-frame>
Last accessed: 4th February,2022
2. Extracting first n-rows in R (Statistics globe)
<https://statisticsglobe.com/extract-first-n-rows-of-data-frame-in-r>
Last accessed: 4th February,2022
3. Percentage of the column in R (DataScience made simple)
<https://www.datasciencemadesimple.com/percentage-of-the-column-in-r-2>
Last accessed: 4th February,2022
4. How to convert table to data frame (Statology, December 2021)
<https://www.statology.org/convert-table-to-data-frame-in-r>
Last accessed: 4th February,2022
5. Barplot in R (Tutorial Gateway)
<https://www.tutorialgateway.org/barplot-in-r-programming>
Last accessed: 4th February,2022
6. Sort Data Frame in R (Statistics Globe)
<https://statisticsglobe.com/sort-data-frame-in-r-example>
Last accessed: 4th February,2022

Mohile_RScript_Module3.R

mohil

2022-02-05

```
r = getOption("repos")
r["CRAN"]="http://cran.us.r-project.org"
options(repos=r)

#Q1.Print your name at the top of the script and load these Libraries: FSA,FS
Adata,magrittr,
#dplyr,tidyr plyr and tidyverse
print("Sanjana Mohile")

## [1] "Sanjana Mohile"

install.packages("FSA")

## Installing package into 'C:/Users/mohil/OneDrive/Documents/R/win-library/4
## .1'
## (as 'lib' is unspecified)

## package 'FSA' successfully unpacked and MD5 sums checked
##
## The downloaded binary packages are in
## C:\Users\mohil\AppData\Local\Temp\RtmpGW0030\downloaded_packages

install.packages("FSAdata")

## Installing package into 'C:/Users/mohil/OneDrive/Documents/R/win-library/4
## .1'
## (as 'lib' is unspecified)

## package 'FSAdata' successfully unpacked and MD5 sums checked
##
## The downloaded binary packages are in
## C:\Users\mohil\AppData\Local\Temp\RtmpGW0030\downloaded_packages

install.packages("magrittr")

## Installing package into 'C:/Users/mohil/OneDrive/Documents/R/win-library/4
## .1'
## (as 'lib' is unspecified)

## package 'magrittr' successfully unpacked and MD5 sums checked

## Warning: cannot remove prior installation of package 'magrittr'

## Warning in file.copy(savedcopy, lib, recursive = TRUE):
## problem copying C:\Users\mohil\OneDrive\Documents\R\win-
## library\4.1\00LOCK\magrittr\libs\x64\magrittr.dll
```

```
## to C:\Users\mohil\OneDrive\Documents\R\win-
## library\4.1\magrittr\libs\x64\magrittr.dll: Permission denied

## Warning: restored 'magrittr'

##
## The downloaded binary packages are in
## C:\Users\mohil\AppData\Local\Temp\RtmpGW0030\downloaded_packages

install.packages("dplyr")

## Installing package into 'C:/Users/mohil/OneDrive/Documents/R/win-library/4
.1'
## (as 'lib' is unspecified)

## package 'dplyr' successfully unpacked and MD5 sums checked

## Warning: cannot remove prior installation of package 'dplyr'

## Warning in file.copy(savedcopy, lib, recursive = TRUE):
## problem copying C:\Users\mohil\OneDrive\Documents\R\win-
## library\4.1\00LOCK\dplyr\libs\x64\dplyr.dll to C:
## \Users\mohil\OneDrive\Documents\R\win-library\4.1\dplyr\libs\x64\dplyr.dll
:
## Permission denied

## Warning: restored 'dplyr'

##
## The downloaded binary packages are in
## C:\Users\mohil\AppData\Local\Temp\RtmpGW0030\downloaded_packages

install.packages("tidyr")

## Installing package into 'C:/Users/mohil/OneDrive/Documents/R/win-library/4
.1'
## (as 'lib' is unspecified)

## package 'tidyr' successfully unpacked and MD5 sums checked

## Warning: cannot remove prior installation of package 'tidyr'

## Warning in file.copy(savedcopy, lib, recursive = TRUE):
## problem copying C:\Users\mohil\OneDrive\Documents\R\win-
## library\4.1\00LOCK\tidyr\libs\x64\tidyr.dll to C:
## \Users\mohil\OneDrive\Documents\R\win-library\4.1\tidyr\libs\x64\tidyr.dll
:
## Permission denied

## Warning: restored 'tidyr'
```



```

##
## The downloaded binary packages are in
## C:\Users\mohil\AppData\Local\Temp\RtmpGW0030\downloaded_packages

install.packages("plyr")

## Installing package into 'C:/Users/mohil/OneDrive/Documents/R/win-library/4
.1'
## (as 'lib' is unspecified)

## package 'plyr' successfully unpacked and MD5 sums checked

## Warning: cannot remove prior installation of package 'plyr'

## Warning in file.copy(savedcopy, lib, recursive = TRUE): problem copying C:
## \Users\mohil\OneDrive\Documents\R\win-library\4.1\00LOCK\plyr\libs\x64\ply
r.dll
## to C:\Users\mohil\OneDrive\Documents\R\win-library\4.1\plyr\libs\x64\plyr.
dll:
## Permission denied

## Warning: restored 'plyr'

##
## The downloaded binary packages are in
## C:\Users\mohil\AppData\Local\Temp\RtmpGW0030\downloaded_packages

install.packages("tidyverse")

## Installing package into 'C:/Users/mohil/OneDrive/Documents/R/win-library/4
.1'
## (as 'lib' is unspecified)

## package 'tidyverse' successfully unpacked and MD5 sums checked
##
## The downloaded binary packages are in
## C:\Users\mohil\AppData\Local\Temp\RtmpGW0030\downloaded_packages

library("FSA")

## ## FSA v0.9.1. See citation('FSA') if used in publication.
## ## Run fishR() for related website and fishR('IFAR') for related book.

library("FSAdata")

## ## FSAdata v0.3.8. See ?FSAdata to find data for specific fisheries analys
es.

library("magrittr")
library("dplyr")

##
## Attaching package: 'dplyr'

```

```

## The following objects are masked from 'package:stats':
##
##   filter, lag

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

library("tidyr")

##
## Attaching package: 'tidyr'

## The following object is masked from 'package:magrittr':
##
##   extract

library("plyr")

## -----
## ----

## You have loaded plyr after dplyr - this is likely to cause problems.
## If you need functions from both plyr and dplyr, please load plyr first, th
en dplyr:
## library(plyr); library(dplyr)

## -----
## ----

##
## Attaching package: 'plyr'

## The following objects are masked from 'package:dplyr':
##
##   arrange, count, desc, failwith, id, mutate, rename, summarise,
##   summarize

## The following object is masked from 'package:FSA':
##
##   mapvalues

library("tidyverse")

## -- Attaching packages ----- tidyverse 1.
3.1 --

## v ggplot2 3.3.5      v purrr  0.3.4
## v tibble  3.1.6      v stringr 1.4.0
## v readr   2.1.1      v forcats 0.5.1

## -- Conflicts ----- tidyverse_conflict
s() --

```

```
## x plyr::arrange()      masks dplyr::arrange()
## x purrr::compact()    masks plyr::compact()
## x plyr::count()       masks dplyr::count()
## x tidyr::extract()    masks magrittr::extract()
## x plyr::failwith()    masks dplyr::failwith()
## x dplyr::filter()     masks stats::filter()
## x plyr::id()          masks dplyr::id()
## x dplyr::lag()        masks stats::lag()
## x plyr::mutate()      masks dplyr::mutate()
## x plyr::rename()      masks dplyr::rename()
## x purrr::set_names()  masks magrittr::set_names()
## x plyr::summarise()   masks dplyr::summarise()
## x plyr::summarize()   masks dplyr::summarize()
```

#Q2. Import the inchBio.csv and name the table <bio>

```
bio <- read.csv("C:/Users/mohil/OneDrive/Desktop/Studies/ALY-6000 Introduction to Analytics/Module 3/inchBio.csv")
```

```
bio
```

	netID	fishID	species	tl	w	tag	scale
## 1	12	16	Bluegill	61	2.9		FALSE
## 2	12	23	Bluegill	66	4.5		FALSE
## 3	12	30	Bluegill	70	5.2		FALSE
## 4	12	44	Bluegill	38	0.5		FALSE
## 5	12	50	Bluegill	42	1.0		FALSE
## 6	12	65	Bluegill	54	2.1		FALSE
## 7	12	66	Bluegill	27	NA		FALSE
## 8	13	68	Bluegill	36	0.5		FALSE
## 9	13	69	Bluegill	59	2.0		FALSE
## 10	13	70	Bluegill	39	0.5		FALSE
## 11	13	71	Bluegill	34	0.5		FALSE
## 12	13	73	Bluegill	40	1.0		FALSE
## 13	13	74	Bluegill	35	0.5		FALSE
## 14	13	75	Bluegill	32	1.0		FALSE
## 15	13	76	Bluegill	37	0.5		FALSE
## 16	13	77	Bluegill	38	1.0		FALSE
## 17	13	78	Bluegill	69	7.0		FALSE
## 18	13	80	Bluegill	39	1.0		FALSE
## 19	13	81	Bluegill	37	0.5		FALSE
## 20	13	82	Bluegill	38	1.0		FALSE
## 21	13	83	Bluegill	47	NA		FALSE
## 22	14	86	Bluegill	33	0.5		FALSE
## 23	14	87	Bluegill	31	1.5		FALSE
## 24	14	88	Bluegill	36	1.5		FALSE
## 25	4	118	Bluegill	150	60.0		TRUE
## 26	4	119	Bluegill	193	145.0		TRUE
## 27	4	120	Bluegill	185	123.0		TRUE
## 28	4	121	Bluegill	152	67.0		TRUE
## 29	4	122	Bluegill	160	75.0		TRUE
## 30	4	123	Bluegill	185	118.0		TRUE

## 31	4	124	Bluegill	170	100.0	TRUE
## 32	4	125	Bluegill	135	35.0	TRUE
## 33	4	126	Bluegill	183	120.0	TRUE
## 34	4	127	Bluegill	168	90.0	TRUE
## 35	4	128	Bluegill	165	85.0	TRUE
## 36	4	129	Bluegill	178	100.0	TRUE
## 37	4	130	Bluegill	193	155.0	TRUE
## 38	4	131	Bluegill	193	140.0	TRUE
## 39	4	132	Bluegill	201	180.0	TRUE
## 40	4	133	Bluegill	203	185.0	TRUE
## 41	4	134	Bluegill	99	15.0	TRUE
## 42	5	138	Bluegill	135	42.0	TRUE
## 43	5	139	Bluegill	38	2.0	FALSE
## 44	5	140	Bluegill	41	NA	FALSE
## 45	5	141	Bluegill	41	NA	FALSE
## 46	5	142	Bluegill	46	NA	FALSE
## 47	5	143	Bluegill	165	68.0	TRUE
## 48	5	144	Bluegill	43	NA	FALSE
## 49	5	145	Bluegill	51	4.0	FALSE
## 50	5	146	Bluegill	203	184.0	TRUE
## 51	5	147	Bluegill	168	98.0	TRUE
## 52	5	148	Bluegill	152	62.0	TRUE
## 53	5	149	Bluegill	64	NA	FALSE
## 54	5	150	Bluegill	157	76.0	TRUE
## 55	10	151	Bluegill	173	100.0	TRUE
## 56	10	152	Bluegill	173	95.0	TRUE
## 57	10	153	Bluegill	185	130.0	TRUE
## 58	10	154	Bluegill	218	250.0	TRUE
## 59	10	155	Bluegill	206	197.0	TRUE
## 60	10	156	Bluegill	165	78.0	TRUE
## 61	10	157	Bluegill	152	72.0	TRUE
## 62	10	158	Bluegill	170	98.0	TRUE
## 63	9	161	Bluegill	206	175.0	TRUE
## 64	9	162	Bluegill	191	144.0	TRUE
## 65	9	163	Bluegill	193	148.0	TRUE
## 66	9	164	Bluegill	183	130.0	TRUE
## 67	9	165	Bluegill	201	185.0	TRUE
## 68	9	166	Bluegill	221	225.0	TRUE
## 69	9	167	Bluegill	165	80.0	FALSE
## 70	9	168	Bluegill	206	180.0	TRUE
## 71	9	169	Bluegill	203	175.0	TRUE
## 72	9	170	Bluegill	165	NA	TRUE
## 73	9	171	Bluegill	193	160.0	FALSE
## 74	9	172	Bluegill	173	NA	FALSE
## 75	6	176	Bluegill	213	200.0	TRUE
## 76	6	177	Bluegill	155	74.0	TRUE
## 77	6	178	Bluegill	157	62.0	TRUE
## 78	6	179	Bluegill	211	220.0	TRUE
## 79	6	180	Bluegill	188	149.0	TRUE
## 80	6	181	Bluegill	188	139.0	TRUE

## 81	6	182	Bluegill	196	132.0	TRUE
## 82	6	183	Bluegill	188	139.0	TRUE
## 83	6	184	Bluegill	160	73.0	TRUE
## 84	6	185	Bluegill	196	120.0	TRUE
## 85	6	186	Bluegill	221	242.0	TRUE
## 86	6	187	Bluegill	180	130.0	TRUE
## 87	6	188	Bluegill	152	70.0	TRUE
## 88	6	189	Bluegill	140	40.0	TRUE
## 89	6	190	Bluegill	203	170.0	TRUE
## 90	6	191	Bluegill	145	52.0	TRUE
## 91	6	192	Bluegill	147	32.0	TRUE
## 92	11	193	Bluegill	211	218.0	TRUE
## 93	11	194	Bluegill	147	60.0	TRUE
## 94	11	195	Bluegill	152	70.0	TRUE
## 95	17	196	Bluegill	203	192.0	TRUE
## 96	17	197	Bluegill	132	31.0	TRUE
## 97	17	199	Bluegill	142	59.0	TRUE
## 98	20	201	Bluegill	140	54.0	TRUE
## 99	15	203	Bluegill	142	40.0	TRUE
## 100	15	206	Bluegill	147	30.0	TRUE
## 101	15	207	Bluegill	119	20.0	TRUE
## 102	16	210	Bluegill	229	280.0	TRUE
## 103	16	211	Bluegill	224	260.0	TRUE
## 104	16	212	Bluegill	224	260.0	TRUE
## 105	16	213	Bluegill	224	240.0	TRUE
## 106	16	214	Bluegill	150	60.0	TRUE
## 107	16	215	Bluegill	137	60.0	TRUE
## 108	21	217	Bluegill	94	14.0	TRUE
## 109	21	219	Bluegill	130	38.0	TRUE
## 110	26	220	Bluegill	132	49.0	TRUE
## 111	26	221	Bluegill	137	41.0	TRUE
## 112	23	224	Bluegill	114	20.0	TRUE
## 113	27	226	Bluegill	127	20.0	TRUE
## 114	27	228	Bluegill	122	20.0	TRUE
## 115	28	230	Bluegill	137	50.0	TRUE
## 116	28	231	Bluegill	234	280.0	TRUE
## 117	37	322	Bluegill	152	NA	TRUE
## 118	37	356	Bluegill	201	NA	TRUE
## 119	206	501	Bluegill	38	0.7	FALSE
## 120	205	502	Bluegill	43	1.4	FALSE
## 121	205	503	Bluegill	56	1.5	FALSE
## 122	205	504	Bluegill	53	1.4	FALSE
## 123	205	505	Bluegill	38	1.0	FALSE
## 124	205	506	Bluegill	48	1.8	FALSE
## 125	205	507	Bluegill	48	1.4	FALSE
## 126	205	508	Bluegill	36	0.6	FALSE
## 127	205	509	Bluegill	30	0.3	FALSE
## 128	205	510	Bluegill	36	0.8	FALSE
## 129	205	511	Bluegill	51	1.3	FALSE
## 130	205	512	Bluegill	58	2.4	FALSE

## 131	205	513	Bluegill	33	0.7		FALSE
## 132	205	514	Bluegill	38	1.0		FALSE
## 133	205	515	Bluegill	33	0.6		FALSE
## 134	205	516	Bluegill	56	2.8		FALSE
## 135	205	517	Bluegill	33	1.1		FALSE
## 136	205	518	Bluegill	53	2.0		FALSE
## 137	205	519	Bluegill	66	4.5		FALSE
## 138	205	520	Bluegill	71	4.9		FALSE
## 139	101	533	Bluegill	213	190.0		TRUE
## 140	101	538	Bluegill	216	198.0	1021	TRUE
## 141	101	539	Bluegill	216	210.0	1022	TRUE
## 142	101	540	Bluegill	231	258.0	1023	TRUE
## 143	101	541	Bluegill	193	138.0		TRUE
## 144	101	542	Bluegill	226	236.0	1024	TRUE
## 145	101	543	Bluegill	163	75.0		TRUE
## 146	101	544	Bluegill	224	229.0	1025	TRUE
## 147	101	545	Bluegill	178	101.0		TRUE
## 148	101	546	Bluegill	180	110.0		TRUE
## 149	101	547	Bluegill	239	295.0	1015	TRUE
## 150	101	548	Bluegill	183	113.0		TRUE
## 151	101	549	Bluegill	211	191.0	1018	TRUE
## 152	101	550	Bluegill	191	139.0		TRUE
## 153	101	552	Bluegill	191	137.0		TRUE
## 154	101	553	Bluegill	201	165.0		TRUE
## 155	101	554	Bluegill	178	103.0		TRUE
## 156	101	555	Bluegill	201	166.0		TRUE
## 157	101	556	Bluegill	180	115.0		TRUE
## 158	101	557	Bluegill	165	89.0		TRUE
## 159	101	558	Bluegill	191	121.0		TRUE
## 160	101	559	Bluegill	157	78.0		TRUE
## 161	101	560	Bluegill	152	63.0		TRUE
## 162	101	561	Bluegill	206	181.0	1075	TRUE
## 163	101	562	Bluegill	147	60.0		TRUE
## 164	101	563	Bluegill	163	74.0		TRUE
## 165	101	564	Bluegill	180	101.0		TRUE
## 166	101	565	Bluegill	193	141.0		TRUE
## 167	101	566	Bluegill	216	228.0	1074	TRUE
## 168	101	567	Bluegill	229	266.0	1073	TRUE
## 169	101	568	Bluegill	168	93.0		TRUE
## 170	101	569	Bluegill	163	74.0		TRUE
## 171	101	570	Bluegill	165	82.0		TRUE
## 172	101	571	Bluegill	231	274.0	1072	TRUE
## 173	101	572	Bluegill	201	151.0		TRUE
## 174	101	573	Bluegill	188	129.0		TRUE
## 175	101	574	Bluegill	175	98.0		TRUE
## 176	101	575	Bluegill	145	57.0		TRUE
## 177	101	576	Bluegill	152	62.0		TRUE
## 178	101	577	Bluegill	175	107.0		TRUE
## 179	101	578	Bluegill	221	240.0	1071	TRUE
## 180	101	580	Bluegill	178	110.0		TRUE

## 181	101	581	Bluegill	211	208.0	1069	TRUE
## 182	101	583	Bluegill	140	48.0		TRUE
## 183	101	585	Bluegill	173	96.0		TRUE
## 184	101	586	Bluegill	152	73.0		TRUE
## 185	101	587	Bluegill	213	198.0	1066	TRUE
## 186	107	621	Bluegill	203	156.0	1065	TRUE
## 187	107	622	Bluegill	221	242.0	1064	TRUE
## 188	107	623	Bluegill	203	187.0	1063	TRUE
## 189	107	624	Bluegill	229	242.0	1062	TRUE
## 190	107	625	Bluegill	170	83.0		TRUE
## 191	104	658	Bluegill	213	197.0	1099	TRUE
## 192	104	659	Bluegill	229	255.0	1098	TRUE
## 193	104	660	Bluegill	185	111.0		TRUE
## 194	103	689	Bluegill	206	NA	1107	TRUE
## 195	103	695	Bluegill	137	45.3		TRUE
## 196	103	697	Bluegill	147	50.9		TRUE
## 197	103	699	Bluegill	66	3.4		TRUE
## 198	103	700	Bluegill	150	54.5		TRUE
## 199	103	701	Bluegill	142	48.8		TRUE
## 200	103	703	Bluegill	150	50.9		TRUE
## 201	111	717	Bluegill	130	35.0		TRUE
## 202	111	718	Bluegill	137	45.0		TRUE
## 203	111	719	Bluegill	132	39.0		TRUE
## 204	111	722	Bluegill	137	43.0		TRUE
## 205	111	726	Bluegill	137	50.0		TRUE
## 206	113	742	Bluegill	135	39.0		TRUE
## 207	114	755	Bluegill	66	3.1		FALSE
## 208	116	761	Bluegill	97	10.9		FALSE
## 209	116	762	Bluegill	91	10.6		FALSE
## 210	112	774	Bluegill	203	171.0	1045	TRUE
## 211	112	780	Bluegill	76	6.1		FALSE
## 212	112	781	Bluegill	109	19.0		FALSE
## 213	120	802	Bluegill	94	10.7		TRUE
## 214	120	803	Bluegill	84	8.9		TRUE
## 215	120	804	Bluegill	91	10.6		TRUE
## 216	120	805	Bluegill	84	8.7		TRUE
## 217	120	806	Bluegill	86	9.3		TRUE
## 218	120	807	Bluegill	69	4.0		TRUE
## 219	119	824	Bluegill	119	25.0		TRUE
## 220	122	826	Bluegill	132	33.0		TRUE
## 221	12	7	Bluntnose Minnow	59	2.0		FALSE
## 222	12	8	Bluntnose Minnow	58	1.7		FALSE
## 223	12	10	Bluntnose Minnow	54	1.5		FALSE
## 224	12	11	Bluntnose Minnow	60	1.7		FALSE
## 225	12	12	Bluntnose Minnow	78	2.3		FALSE
## 226	12	14	Bluntnose Minnow	62	2.2		FALSE
## 227	12	17	Bluntnose Minnow	59	2.0		FALSE
## 228	12	18	Bluntnose Minnow	63	2.5		FALSE
## 229	12	19	Bluntnose Minnow	58	2.0		FALSE
## 230	12	21	Bluntnose Minnow	71	3.8		FALSE

## 231	12	22 Bluntnose Minnow	55	2.5	FALSE
## 232	12	24 Bluntnose Minnow	60	2.2	FALSE
## 233	12	25 Bluntnose Minnow	51	2.0	FALSE
## 234	12	26 Bluntnose Minnow	52	1.8	FALSE
## 235	12	27 Bluntnose Minnow	64	2.5	FALSE
## 236	12	28 Bluntnose Minnow	67	4.0	FALSE
## 237	12	29 Bluntnose Minnow	63	2.0	FALSE
## 238	12	31 Bluntnose Minnow	65	2.0	FALSE
## 239	12	32 Bluntnose Minnow	67	3.0	FALSE
## 240	12	33 Bluntnose Minnow	71	4.8	FALSE
## 241	12	34 Bluntnose Minnow	66	3.0	FALSE
## 242	12	35 Bluntnose Minnow	73	4.0	FALSE
## 243	12	36 Bluntnose Minnow	67	4.0	FALSE
## 244	12	37 Bluntnose Minnow	51	2.5	FALSE
## 245	12	38 Bluntnose Minnow	58	1.5	FALSE
## 246	12	39 Bluntnose Minnow	71	3.8	FALSE
## 247	12	40 Bluntnose Minnow	52	1.5	FALSE
## 248	12	41 Bluntnose Minnow	61	1.5	FALSE
## 249	12	42 Bluntnose Minnow	58	1.5	FALSE
## 250	12	43 Bluntnose Minnow	72	3.0	FALSE
## 251	12	45 Bluntnose Minnow	68	NA	FALSE
## 252	12	46 Bluntnose Minnow	57	1.5	FALSE
## 253	12	47 Bluntnose Minnow	58	1.5	FALSE
## 254	12	48 Bluntnose Minnow	58	2.0	FALSE
## 255	12	49 Bluntnose Minnow	58	2.5	FALSE
## 256	12	51 Bluntnose Minnow	61	3.0	FALSE
## 257	12	52 Bluntnose Minnow	57	1.0	FALSE
## 258	12	53 Bluntnose Minnow	70	2.0	FALSE
## 259	12	54 Bluntnose Minnow	64	2.5	FALSE
## 260	12	55 Bluntnose Minnow	61	1.0	FALSE
## 261	12	56 Bluntnose Minnow	56	1.0	FALSE
## 262	12	57 Bluntnose Minnow	58	NA	FALSE
## 263	12	58 Bluntnose Minnow	66	4.0	FALSE
## 264	12	59 Bluntnose Minnow	59	2.0	FALSE
## 265	12	60 Bluntnose Minnow	62	1.5	FALSE
## 266	12	61 Bluntnose Minnow	64	2.5	FALSE
## 267	12	62 Bluntnose Minnow	63	1.5	FALSE
## 268	12	63 Bluntnose Minnow	61	1.8	FALSE
## 269	12	64 Bluntnose Minnow	64	3.0	FALSE
## 270	13	79 Bluntnose Minnow	72	5.0	FALSE
## 271	13	84 Bluntnose Minnow	71	4.0	FALSE
## 272	205	522 Bluntnose Minnow	61	2.1	FALSE
## 273	201	523 Bluntnose Minnow	56	NA	FALSE
## 274	101	589 Bluntnose Minnow	64	1.4	FALSE
## 275	101	590 Bluntnose Minnow	46	0.5	FALSE
## 276	101	591 Bluntnose Minnow	53	1.0	FALSE
## 277	101	592 Bluntnose Minnow	58	1.7	FALSE
## 278	101	593 Bluntnose Minnow	56	1.0	FALSE
## 279	101	594 Bluntnose Minnow	58	1.1	FALSE
## 280	101	595 Bluntnose Minnow	58	1.0	FALSE

## 281	101	596 Bluntnose Minnow	56	1.0	FALSE
## 282	101	597 Bluntnose Minnow	53	1.0	FALSE
## 283	101	598 Bluntnose Minnow	51	0.8	FALSE
## 284	101	599 Bluntnose Minnow	56	0.8	FALSE
## 285	101	600 Bluntnose Minnow	56	1.0	FALSE
## 286	101	601 Bluntnose Minnow	58	1.0	FALSE
## 287	101	603 Bluntnose Minnow	64	1.4	FALSE
## 288	101	605 Bluntnose Minnow	58	1.2	FALSE
## 289	101	606 Bluntnose Minnow	61	1.3	FALSE
## 290	101	607 Bluntnose Minnow	69	1.8	FALSE
## 291	101	608 Bluntnose Minnow	64	1.4	FALSE
## 292	101	609 Bluntnose Minnow	53	0.8	FALSE
## 293	101	610 Bluntnose Minnow	64	1.4	FALSE
## 294	101	611 Bluntnose Minnow	61	1.2	FALSE
## 295	101	612 Bluntnose Minnow	51	0.7	FALSE
## 296	101	613 Bluntnose Minnow	56	1.1	FALSE
## 297	101	614 Bluntnose Minnow	56	1.0	FALSE
## 298	101	615 Bluntnose Minnow	53	0.9	FALSE
## 299	101	616 Bluntnose Minnow	58	1.2	FALSE
## 300	101	617 Bluntnose Minnow	53	0.8	FALSE
## 301	101	618 Bluntnose Minnow	53	0.9	FALSE
## 302	101	619 Bluntnose Minnow	66	1.5	FALSE
## 303	101	620 Bluntnose Minnow	61	1.3	FALSE
## 304	102	638 Bluntnose Minnow	74	3.9	FALSE
## 305	102	639 Bluntnose Minnow	61	2.2	FALSE
## 306	102	640 Bluntnose Minnow	84	1.3	FALSE
## 307	102	642 Bluntnose Minnow	53	1.3	FALSE
## 308	102	643 Bluntnose Minnow	56	2.1	FALSE
## 309	102	644 Bluntnose Minnow	56	1.4	FALSE
## 310	102	645 Bluntnose Minnow	61	2.0	FALSE
## 311	111	727 Bluntnose Minnow	79	3.8	FALSE
## 312	116	763 Bluntnose Minnow	74	3.4	FALSE
## 313	121	816 Bluntnose Minnow	51	1.2	FALSE
## 314	121	817 Bluntnose Minnow	79	3.9	FALSE
## 315	121	818 Bluntnose Minnow	74	3.8	FALSE
## 316	121	819 Bluntnose Minnow	64	1.9	FALSE
## 317	121	820 Bluntnose Minnow	51	1.0	FALSE
## 318	121	821 Bluntnose Minnow	51	1.1	FALSE
## 319	121	822 Bluntnose Minnow	64	2.1	FALSE
## 320	121	823 Bluntnose Minnow	51	1.2	FALSE
## 321	125	832 Bluntnose Minnow	81	2.8	FALSE
## 322	125	833 Bluntnose Minnow	81	3.2	FALSE
## 323	125	834 Bluntnose Minnow	79	4.1	FALSE
## 324	13	72 Iowa Darter	61	NA	FALSE
## 325	205	521 Iowa Darter	53	1.1	FALSE
## 326	203	524 Iowa Darter	46	0.6	FALSE
## 327	203	525 Iowa Darter	44	0.6	FALSE
## 328	203	526 Iowa Darter	46	0.6	FALSE
## 329	203	527 Iowa Darter	43	0.4	FALSE
## 330	101	602 Iowa Darter	56	0.9	FALSE

## 331	101	604	Iowa Darter	51	0.8		FALSE
## 332	102	641	Iowa Darter	48	1.0		FALSE
## 333	106	646	Iowa Darter	38	0.6		FALSE
## 334	106	647	Iowa Darter	46	0.9		FALSE
## 335	106	648	Iowa Darter	43	0.8		FALSE
## 336	106	649	Iowa Darter	48	1.0		FALSE
## 337	106	650	Iowa Darter	41	0.5		FALSE
## 338	106	651	Iowa Darter	41	0.7		FALSE
## 339	106	652	Iowa Darter	43	0.6		FALSE
## 340	106	653	Iowa Darter	43	0.6		FALSE
## 341	106	654	Iowa Darter	46	0.9		FALSE
## 342	105	711	Iowa Darter	48	0.7		FALSE
## 343	105	712	Iowa Darter	46	0.7		FALSE
## 344	115	730	Iowa Darter	48	0.8		FALSE
## 345	115	731	Iowa Darter	48	0.8		FALSE
## 346	115	732	Iowa Darter	51	1.0		FALSE
## 347	115	733	Iowa Darter	51	1.1		FALSE
## 348	116	764	Iowa Darter	51	1.2		FALSE
## 349	116	765	Iowa Darter	48	0.9		FALSE
## 350	116	766	Iowa Darter	51	0.9		FALSE
## 351	116	767	Iowa Darter	43	0.5		FALSE
## 352	116	768	Iowa Darter	51	1.2		FALSE
## 353	116	769	Iowa Darter	56	1.2		FALSE
## 354	116	770	Iowa Darter	53	1.1		FALSE
## 355	116	771	Iowa Darter	43	0.9		FALSE
## 356	14	85	Largemouth Bass	68	3.5		TRUE
## 357	3	89	Largemouth Bass	338	NA	g0996	TRUE
## 358	1	90	Largemouth Bass	356	NA	y06073	TRUE
## 359	1	91	Largemouth Bass	328	NA	o0449	TRUE
## 360	1	92	Largemouth Bass	305	NA	o0448	TRUE
## 361	1	93	Largemouth Bass	386	NA	o0447	TRUE
## 362	1	94	Largemouth Bass	310	NA	g0985	TRUE
## 363	1	95	Largemouth Bass	262	NA		TRUE
## 364	1	96	Largemouth Bass	305	NA	g0986	TRUE
## 365	1	97	Largemouth Bass	315	NA	g0987	TRUE
## 366	1	98	Largemouth Bass	371	NA	g1000	TRUE
## 367	1	99	Largemouth Bass	338	NA	g0999	TRUE
## 368	1	100	Largemouth Bass	239	NA		TRUE
## 369	1	101	Largemouth Bass	236	NA		TRUE
## 370	1	102	Largemouth Bass	348	NA	g0997	TRUE
## 371	2	104	Largemouth Bass	307	NA	o0501	TRUE
## 372	2	105	Largemouth Bass	323	NA	o0506	TRUE
## 373	2	106	Largemouth Bass	231	NA		TRUE
## 374	2	107	Largemouth Bass	216	NA		TRUE
## 375	2	108	Largemouth Bass	325	NA	o0507	TRUE
## 376	2	109	Largemouth Bass	399	NA	o0510	TRUE
## 377	2	110	Largemouth Bass	333	NA	o0511	TRUE
## 378	2	111	Largemouth Bass	356	NA	o0512	TRUE
## 379	2	112	Largemouth Bass	274	NA		TRUE
## 380	17	198	Largemouth Bass	259	190.0		TRUE

## 381	27	227	Largemouth Bass	137	20.0		TRUE
## 382	29	232	Largemouth Bass	279	NA	o0516	TRUE
## 383	29	233	Largemouth Bass	351	NA	o0517	TRUE
## 384	29	234	Largemouth Bass	307	NA	o0518	TRUE
## 385	29	235	Largemouth Bass	325	NA	o0520	TRUE
## 386	29	237	Largemouth Bass	330	NA	o0521	TRUE
## 387	29	239	Largemouth Bass	282	NA	o0523	TRUE
## 388	29	240	Largemouth Bass	262	NA		TRUE
## 389	29	241	Largemouth Bass	391	NA	o0524	TRUE
## 390	29	242	Largemouth Bass	262	NA		TRUE
## 391	29	243	Largemouth Bass	295	NA	o0525	TRUE
## 392	30	244	Largemouth Bass	305	NA	o0550	TRUE
## 393	30	245	Largemouth Bass	193	NA		TRUE
## 394	30	246	Largemouth Bass	254	NA		TRUE
## 395	30	247	Largemouth Bass	246	NA		TRUE
## 396	30	248	Largemouth Bass	330	NA	o0549	TRUE
## 397	30	249	Largemouth Bass	300	NA		TRUE
## 398	30	250	Largemouth Bass	284	NA		TRUE
## 399	30	251	Largemouth Bass	371	NA	o0548	TRUE
## 400	30	252	Largemouth Bass	335	NA	o0547	TRUE
## 401	30	253	Largemouth Bass	305	NA	o0546	TRUE
## 402	30	254	Largemouth Bass	254	NA		TRUE
## 403	30	255	Largemouth Bass	318	NA	o0545	TRUE
## 404	30	256	Largemouth Bass	348	NA	o0544	TRUE
## 405	30	257	Largemouth Bass	343	NA	o0543	TRUE
## 406	30	258	Largemouth Bass	325	NA	o0542	TRUE
## 407	30	259	Largemouth Bass	333	NA	o0541	TRUE
## 408	30	260	Largemouth Bass	330	NA	o0540	TRUE
## 409	30	261	Largemouth Bass	381	NA	o0539	TRUE
## 410	30	262	Largemouth Bass	320	NA	o0538	TRUE
## 411	30	263	Largemouth Bass	272	NA		TRUE
## 412	30	264	Largemouth Bass	312	NA	o0536	TRUE
## 413	30	265	Largemouth Bass	310	NA	o0535	TRUE
## 414	30	266	Largemouth Bass	297	NA		TRUE
## 415	31	270	Largemouth Bass	307	NA	o0533	TRUE
## 416	31	271	Largemouth Bass	310	NA	o0532	TRUE
## 417	31	272	Largemouth Bass	277	NA		TRUE
## 418	33	273	Largemouth Bass	130	NA		TRUE
## 419	32	274	Largemouth Bass	140	NA		TRUE
## 420	35	275	Largemouth Bass	239	NA		TRUE
## 421	35	276	Largemouth Bass	236	NA		TRUE
## 422	35	277	Largemouth Bass	320	NA	o0519	TRUE
## 423	35	278	Largemouth Bass	241	NA		TRUE
## 424	35	279	Largemouth Bass	343	NA	g0988	TRUE
## 425	35	280	Largemouth Bass	226	NA		TRUE
## 426	35	281	Largemouth Bass	211	NA		TRUE
## 427	35	282	Largemouth Bass	409	NA	g0990	TRUE
## 428	35	284	Largemouth Bass	371	NA	o0991	TRUE
## 429	35	285	Largemouth Bass	312	NA	o0989	TRUE
## 430	35	286	Largemouth Bass	333	NA	y00003	TRUE

## 431	35	287	Largemouth Bass	338	NA	y00228	TRUE
## 432	35	288	Largemouth Bass	282	NA		TRUE
## 433	35	289	Largemouth Bass	302	NA		TRUE
## 434	35	290	Largemouth Bass	305	NA	y160016	TRUE
## 435	35	291	Largemouth Bass	376	NA	o0534	TRUE
## 436	35	292	Largemouth Bass	251	NA		TRUE
## 437	35	293	Largemouth Bass	338	NA	y2519	TRUE
## 438	35	294	Largemouth Bass	318	NA	y00034	TRUE
## 439	35	295	Largemouth Bass	312	NA	y00123	TRUE
## 440	35	296	Largemouth Bass	272	NA		TRUE
## 441	35	297	Largemouth Bass	394	NA	o0531	TRUE
## 442	35	298	Largemouth Bass	363	NA	o0530	TRUE
## 443	35	299	Largemouth Bass	320	NA	y01525	FALSE
## 444	36	300	Largemouth Bass	307	NA	o0529	TRUE
## 445	36	301	Largemouth Bass	330	NA	o0528	TRUE
## 446	36	302	Largemouth Bass	318	NA	o0527	TRUE
## 447	36	303	Largemouth Bass	302	NA		TRUE
## 448	36	304	Largemouth Bass	277	NA		TRUE
## 449	36	305	Largemouth Bass	241	NA		TRUE
## 450	36	306	Largemouth Bass	318	NA	o0526	TRUE
## 451	36	307	Largemouth Bass	297	NA		TRUE
## 452	36	308	Largemouth Bass	318	NA	y01598	TRUE
## 453	36	309	Largemouth Bass	373	NA	y01600	TRUE
## 454	36	310	Largemouth Bass	290	NA		TRUE
## 455	36	311	Largemouth Bass	300	NA		TRUE
## 456	36	312	Largemouth Bass	318	NA	y01603	TRUE
## 457	36	313	Largemouth Bass	333	NA	y01604	TRUE
## 458	36	314	Largemouth Bass	356	NA	y01605	TRUE
## 459	36	315	Largemouth Bass	180	NA		TRUE
## 460	36	316	Largemouth Bass	305	NA	y01606	TRUE
## 461	37	317	Largemouth Bass	211	NA		TRUE
## 462	37	318	Largemouth Bass	320	NA	y00385	TRUE
## 463	37	320	Largemouth Bass	236	NA		TRUE
## 464	37	321	Largemouth Bass	193	NA		TRUE
## 465	37	323	Largemouth Bass	249	NA		TRUE
## 466	37	324	Largemouth Bass	318	NA	y01601	TRUE
## 467	37	325	Largemouth Bass	211	NA		TRUE
## 468	37	326	Largemouth Bass	254	NA		TRUE
## 469	37	327	Largemouth Bass	318	NA	y00023	TRUE
## 470	37	328	Largemouth Bass	427	NA	y00316	TRUE
## 471	37	329	Largemouth Bass	345	NA	y00220	TRUE
## 472	37	330	Largemouth Bass	330	NA	y00011	TRUE
## 473	37	331	Largemouth Bass	320	NA	y01534	TRUE
## 474	37	332	Largemouth Bass	356	NA	y00028	TRUE
## 475	37	333	Largemouth Bass	325	NA	y00009	TRUE
## 476	37	336	Largemouth Bass	297	NA		TRUE
## 477	37	337	Largemouth Bass	201	NA		TRUE
## 478	37	338	Largemouth Bass	257	NA		TRUE
## 479	37	339	Largemouth Bass	206	NA		TRUE
## 480	37	340	Largemouth Bass	201	NA		TRUE

## 481	37	341	Largemouth Bass	251	NA		TRUE
## 482	37	342	Largemouth Bass	239	NA		TRUE
## 483	37	343	Largemouth Bass	203	NA		TRUE
## 484	37	344	Largemouth Bass	262	NA		TRUE
## 485	37	345	Largemouth Bass	272	NA		TRUE
## 486	37	346	Largemouth Bass	353	NA	y00025	TRUE
## 487	37	347	Largemouth Bass	318	NA	y00229	TRUE
## 488	37	348	Largemouth Bass	330	NA	y00409	TRUE
## 489	37	349	Largemouth Bass	292	NA		TRUE
## 490	37	352	Largemouth Bass	318	NA	y00006	TRUE
## 491	37	353	Largemouth Bass	335	NA	y00015	TRUE
## 492	37	354	Largemouth Bass	295	NA		TRUE
## 493	37	355	Largemouth Bass	323	NA	y00027	TRUE
## 494	101	536	Largemouth Bass	318	407.0	1019	TRUE
## 495	101	551	Largemouth Bass	178	60.0		TRUE
## 496	101	588	Largemouth Bass	221	117.0		TRUE
## 497	102	630	Largemouth Bass	429	1070.0	1058	TRUE
## 498	102	631	Largemouth Bass	394	737.0	1057	TRUE
## 499	102	632	Largemouth Bass	361	548.0	1056	TRUE
## 500	104	656	Largemouth Bass	251	172.0		TRUE
## 501	104	657	Largemouth Bass	386	688.0	1100	TRUE
## 502	103	702	Largemouth Bass	361	579.0	1115	TRUE
## 503	111	713	Largemouth Bass	348	556.0	1117	TRUE
## 504	114	759	Largemouth Bass	168	48.0		TRUE
## 505	116	760	Largemouth Bass	71	3.1		FALSE
## 506	112	801	Largemouth Bass	168	52.0		FALSE
## 507	109	836	Largemouth Bass	424	864.0	1518	TRUE
## 508	109	837	Largemouth Bass	328	410.0	1519	TRUE
## 509	109	838	Largemouth Bass	315	385.0	1520	TRUE
## 510	109	839	Largemouth Bass	325	338.0	1521	TRUE
## 511	109	840	Largemouth Bass	274	235.0		TRUE
## 512	109	841	Largemouth Bass	348	NA	1522	TRUE
## 513	109	842	Largemouth Bass	318	386.0	o6860	TRUE
## 514	109	843	Largemouth Bass	333	460.0	1523	TRUE
## 515	109	844	Largemouth Bass	386	547.0	1524	TRUE
## 516	109	845	Largemouth Bass	384	540.0	1525	TRUE
## 517	109	846	Largemouth Bass	249	NA		TRUE
## 518	109	847	Largemouth Bass	307	320.0	1532	TRUE
## 519	109	848	Largemouth Bass	323	420.0	1776	TRUE
## 520	109	849	Largemouth Bass	315	406.0	1777	TRUE
## 521	109	850	Largemouth Bass	351	486.0	1778	TRUE
## 522	109	851	Largemouth Bass	249	274.0		TRUE
## 523	109	852	Largemouth Bass	287	265.0		TRUE
## 524	109	853	Largemouth Bass	307	338.0	1779	TRUE
## 525	109	854	Largemouth Bass	302	358.0		TRUE
## 526	109	855	Largemouth Bass	297	333.0		TRUE
## 527	109	856	Largemouth Bass	320	386.0	o0526	TRUE
## 528	109	857	Largemouth Bass	333	483.0	1780	TRUE
## 529	109	858	Largemouth Bass	333	426.0	1781	TRUE
## 530	109	859	Largemouth Bass	307	295.0	1782	TRUE

## 531	109	860	Largemouth Bass	229	132.0		TRUE
## 532	109	861	Largemouth Bass	201	92.0		TRUE
## 533	109	862	Largemouth Bass	201	88.0		TRUE
## 534	110	864	Largemouth Bass	325	436.0	1784	TRUE
## 535	110	865	Largemouth Bass	325	412.0	1785	TRUE
## 536	110	866	Largemouth Bass	323	444.0	1786	TRUE
## 537	110	867	Largemouth Bass	272	233.0		TRUE
## 538	129	868	Largemouth Bass	363	624.0	1787	TRUE
## 539	129	869	Largemouth Bass	353	561.0	1788	TRUE
## 540	129	871	Largemouth Bass	264	211.0		TRUE
## 541	129	872	Largemouth Bass	315	367.0	1790	TRUE
## 542	129	873	Largemouth Bass	239	157.0		TRUE
## 543	129	874	Largemouth Bass	330	433.0	1791	TRUE
## 544	129	875	Largemouth Bass	239	154.0		TRUE
## 545	129	876	Largemouth Bass	300	292.0		TRUE
## 546	129	877	Largemouth Bass	310	351.0		TRUE
## 547	129	878	Largemouth Bass	330	436.0	o0507	TRUE
## 548	129	880	Largemouth Bass	267	244.0		TRUE
## 549	129	881	Largemouth Bass	330	437.0	1793	TRUE
## 550	129	882	Largemouth Bass	234	133.0		TRUE
## 551	129	883	Largemouth Bass	315	409.0	1794	TRUE
## 552	129	884	Largemouth Bass	310	330.0	1795	TRUE
## 553	129	885	Largemouth Bass	251	187.0		TRUE
## 554	129	886	Largemouth Bass	226	132.0		TRUE
## 555	129	887	Largemouth Bass	254	168.0		TRUE
## 556	129	888	Largemouth Bass	320	385.0	1796	TRUE
## 557	129	889	Largemouth Bass	305	324.0	1797	TRUE
## 558	129	890	Largemouth Bass	333	461.0	1798	TRUE
## 559	129	891	Largemouth Bass	224	110.0		TRUE
## 560	130	892	Largemouth Bass	310	346.0	1799	TRUE
## 561	130	893	Largemouth Bass	345	450.0	1800	TRUE
## 562	130	894	Largemouth Bass	333	443.0	1701	TRUE
## 563	130	895	Largemouth Bass	348	537.0	1702	TRUE
## 564	130	896	Largemouth Bass	305	306.0	1703	TRUE
## 565	130	897	Largemouth Bass	343	485.0	o0997	TRUE
## 566	130	898	Largemouth Bass	318	383.0	1019	FALSE
## 567	130	899	Largemouth Bass	310	348.0	1704	TRUE
## 568	130	900	Largemouth Bass	292	NA		TRUE
## 569	130	901	Largemouth Bass	272	242.0		TRUE
## 570	130	902	Largemouth Bass	257	173.0		TRUE
## 571	130	903	Largemouth Bass	312	371.0	y00229	TRUE
## 572	130	904	Largemouth Bass	257	207.0		TRUE
## 573	130	905	Largemouth Bass	305	320.0	1705	TRUE
## 574	130	906	Largemouth Bass	312	349.0	o0529	TRUE
## 575	130	907	Largemouth Bass	378	723.0	1706	TRUE
## 576	130	908	Largemouth Bass	348	571.0	1707	TRUE
## 577	130	909	Largemouth Bass	295	315.0		TRUE
## 578	130	910	Largemouth Bass	338	462.0	1708	TRUE
## 579	131	911	Largemouth Bass	300	342.0		TRUE
## 580	131	912	Largemouth Bass	264	215.0		TRUE

## 581	131	913	Largemouth Bass	251	180.0		TRUE
## 582	131	914	Largemouth Bass	378	646.0	1709	TRUE
## 583	131	915	Largemouth Bass	325	377.0	1785	FALSE
## 584	12	9	Pumpkinseed	75	6.5		FALSE
## 585	12	13	Pumpkinseed	53	2.2		FALSE
## 586	12	15	Pumpkinseed	28	1.0		FALSE
## 587	12	20	Pumpkinseed	51	1.5		FALSE
## 588	12	67	Pumpkinseed	61	4.0		FALSE
## 589	9	159	Pumpkinseed	188	120.0		TRUE
## 590	9	160	Pumpkinseed	191	185.0		TRUE
## 591	20	200	Pumpkinseed	191	200.0		TRUE
## 592	21	216	Pumpkinseed	152	80.0		TRUE
## 593	23	223	Pumpkinseed	201	210.0		TRUE
## 594	27	225	Pumpkinseed	130	20.0		TRUE
## 595	24	229	Pumpkinseed	229	320.0		TRUE
## 596	124	835	Pumpkinseed	175	130.0		TRUE
## 597	203	528	Tadpole Madtom	41	0.5		FALSE
## 598	203	529	Tadpole Madtom	33	0.2		FALSE
## 599	203	530	Tadpole Madtom	30	0.2		FALSE
## 600	203	531	Tadpole Madtom	30	0.3		FALSE
## 601	106	655	Tadpole Madtom	46	1.3		FALSE
## 602	115	734	Tadpole Madtom	38	1.0		FALSE
## 603	4	113	Yellow Perch	239	150.0		TRUE
## 604	4	114	Yellow Perch	267	170.0		TRUE
## 605	4	115	Yellow Perch	262	175.0		TRUE
## 606	4	116	Yellow Perch	257	157.0		TRUE
## 607	4	117	Yellow Perch	257	160.0		TRUE
## 608	9	173	Yellow Perch	221	105.0		TRUE
## 609	9	174	Yellow Perch	226	150.0		TRUE
## 610	9	175	Yellow Perch	282	230.0		TRUE
## 611	20	202	Yellow Perch	226	122.0		TRUE
## 612	15	204	Yellow Perch	244	130.0		TRUE
## 613	15	205	Yellow Perch	272	200.0		TRUE
## 614	23	222	Yellow Perch	229	110.0		TRUE
## 615	102	633	Yellow Perch	249	164.0	1055	TRUE
## 616	102	634	Yellow Perch	264	208.0	1054	TRUE
## 617	102	635	Yellow Perch	254	173.0	1053	TRUE
## 618	102	636	Yellow Perch	249	158.0	1052	TRUE
## 619	105	704	Yellow Perch	114	14.1		TRUE
## 620	105	705	Yellow Perch	102	9.7		TRUE
## 621	105	706	Yellow Perch	119	13.6		TRUE
## 622	105	707	Yellow Perch	79	4.0		TRUE
## 623	105	708	Yellow Perch	81	4.3		TRUE
## 624	105	709	Yellow Perch	79	4.1		TRUE
## 625	105	710	Yellow Perch	74	3.4		TRUE
## 626	111	714	Yellow Perch	307	340.0	1118	TRUE
## 627	115	728	Yellow Perch	94	6.3		FALSE
## 628	115	729	Yellow Perch	79	4.2		FALSE
## 629	113	735	Yellow Perch	267	201.0	1026	TRUE
## 630	114	756	Yellow Perch	297	327.0	1043	TRUE

## 631	116	772	Yellow Perch	91	6.8		TRUE
## 632	116	773	Yellow Perch	84	4.5		TRUE
## 633	112	796	Yellow Perch	229	123.0	1046	TRUE
## 634	112	797	Yellow Perch	239	135.0	1047	TRUE
## 635	122	825	Yellow Perch	249	152.0	1156	TRUE
## 636	125	827	Yellow Perch	221	113.0	1696	TRUE
## 637	125	828	Yellow Perch	84	4.6		TRUE
## 638	125	829	Yellow Perch	91	7.1		TRUE
## 639	125	830	Yellow Perch	81	4.1		TRUE
## 640	125	831	Yellow Perch	71	2.1		TRUE
## 641	1	103	Black Crappie	284	NA	g0998	TRUE
## 642	5	135	Black Crappie	305	NA	o0513	TRUE
## 643	5	136	Black Crappie	305	NA	o0514	TRUE
## 644	5	137	Black Crappie	267	275.0	o0515	TRUE
## 645	16	208	Black Crappie	295	380.0	g0995	TRUE
## 646	16	209	Black Crappie	274	260.0	g0994	TRUE
## 647	21	218	Black Crappie	152	46.0		TRUE
## 648	29	238	Black Crappie	279	NA	o0522	TRUE
## 649	30	267	Black Crappie	290	NA	o0537	TRUE
## 650	24	268	Black Crappie	330	580.0		TRUE
## 651	24	269	Black Crappie	307	440.0		TRUE
## 652	35	283	Black Crappie	290	NA	o0408	TRUE
## 653	37	319	Black Crappie	300	NA	y00384	TRUE
## 654	37	334	Black Crappie	297	NA	y00020	TRUE
## 655	37	335	Black Crappie	297	NA	y00318	TRUE
## 656	37	350	Black Crappie	297	NA	yunknown	TRUE
## 657	37	351	Black Crappie	290	NA	y09000	TRUE
## 658	101	532	Black Crappie	320	508.0	1014	TRUE
## 659	101	534	Black Crappie	305	443.0	1016	TRUE
## 660	101	535	Black Crappie	307	440.0	1017	TRUE
## 661	101	537	Black Crappie	287	379.0	1020	TRUE
## 662	102	626	Black Crappie	307	461.0	y00318	TRUE
## 663	102	627	Black Crappie	310	466.0	1061	TRUE
## 664	102	628	Black Crappie	315	536.0	1060	TRUE
## 665	102	629	Black Crappie	323	565.0	1059	TRUE
## 666	113	753	Black Crappie	155	47.0		TRUE
## 667	113	754	Black Crappie	147	37.0		TRUE
## 668	112	798	Black Crappie	188	102.0		FALSE
## 669	112	799	Black Crappie	330	606.0	1048	TRUE
## 670	112	800	Black Crappie	292	362.0	1049	TRUE
## 671	121	808	Black Crappie	323	509.0	1050	TRUE
## 672	121	809	Black Crappie	282	352.0	1700	TRUE
## 673	121	812	Black Crappie	142	37.0		TRUE
## 674	110	863	Black Crappie	307	415.0	1783	TRUE
## 675	129	870	Black Crappie	279	344.0	1789	TRUE
## 676	129	879	Black Crappie	302	397.0	1792	TRUE

#Q3. Display the head, tail and structure of <bio>
head(bio)


```
## netID fishID species tl w tag scale
## 1 12 16 Bluegill 61 2.9 FALSE
## 2 12 23 Bluegill 66 4.5 FALSE
## 3 12 30 Bluegill 70 5.2 FALSE
## 4 12 44 Bluegill 38 0.5 FALSE
## 5 12 50 Bluegill 42 1.0 FALSE
## 6 12 65 Bluegill 54 2.1 FALSE
```

```
tail(bio)
```

```
## netID fishID species tl w tag scale
## 671 121 808 Black Crappie 323 509 1050 TRUE
## 672 121 809 Black Crappie 282 352 1700 TRUE
## 673 121 812 Black Crappie 142 37 TRUE
## 674 110 863 Black Crappie 307 415 1783 TRUE
## 675 129 870 Black Crappie 279 344 1789 TRUE
## 676 129 879 Black Crappie 302 397 1792 TRUE
```

```
str(bio)
```

```
## 'data.frame': 676 obs. of 7 variables:
## $ netID : int 12 12 12 12 12 12 12 13 13 13 ...
## $ fishID : int 16 23 30 44 50 65 66 68 69 70 ...
## $ species: chr "Bluegill" "Bluegill" "Bluegill" "Bluegill" ...
## $ tl : int 61 66 70 38 42 54 27 36 59 39 ...
## $ w : num 2.9 4.5 5.2 0.5 1 2.1 NA 0.5 2 0.5 ...
## $ tag : chr "" "" "" "" ...
## $ scale : logi FALSE FALSE FALSE FALSE FALSE FALSE ...
```

#Q4. Create an object, <counts>, that counts and lists all the species records

```
counts <- bio$species
table(counts)
```

```
## counts
## Black Crappie Bluegill Bluntnose Minnow Iowa Darter
## 36 220 103 32
## Largemouth Bass Pumpkinseed Tadpole Madtom Yellow Perch
## 228 13 6 38
```

#Q5. Display just the 8 levels (names) of the species

```
unique(bio$species)
```

```
## [1] "Bluegill" "Bluntnose Minnow" "Iowa Darter" "Largemouth Bass"
## [5] "Pumpkinseed" "Tadpole Madtom" "Yellow Perch" "Black Crappie"
```

#Q6. Create a <tmp> object that displays the different species and the number of record of each species in the dataset. Include this information in your report.

```

tmp <- table(counts)
tmp

## counts
##      Black Crappie      Bluegill Bluntnose Minnow      Iowa Darter
##           36           220           103           32
##  Largemouth Bass      Pumpkinseed      Tadpole Madtom      Yellow Perch
##           228           13           6           38

#Q7. Create a subset, <tmp2>, of just the species variable and display the first five records
tmp2 <- bio$species
head(tmp2,5)

## [1] "Bluegill" "Bluegill" "Bluegill" "Bluegill" "Bluegill"

#Q8. Create a table, <w>, of the species variable. Display the class of w
w <- table(bio$species)
class(w)

## [1] "table"

#Q9. Convert <w> to a data frame named <t> and display the results
t <- data.frame(w)
t

##           Var1 Freq
## 1  Black Crappie   36
## 2    Bluegill  220
## 3 Bluntnose Minnow 103
## 4    Iowa Darter   32
## 5 Largemouth Bass 228
## 6    Pumpkinseed   13
## 7  Tadpole Madtom    6
## 8    Yellow Perch  38

class(t)

## [1] "data.frame"

#Q10. Extract and display the frequency values from the <t> data frame
## lpyr package used for this
t$Freq

## [1] 36 220 103 32 228 13 6 38

#Q11. Create a table named <cSpec> from the bio species attribute (variable) and confirm that
#you created a table which displays the number of species in the dataset <bio>
cSpec <- table(bio$species)
cSpec

```

```
##
##      Black Crappie      Bluegill Bluntnose Minnow      Iowa Darter
##           36           220           103           32
##  Largemouth Bass      Pumpkinseed  Tadpole Madtom      Yellow Perch
##           228           13           6           38
```

#Q12. Create a table named <cSpecPct> that displays the species and percentage of records for

#each species. Confirm you created a table class.

```
cSpecPct <- (table(bio$species)*100)/length(bio$species)
cSpecPct
```

```
##
##      Black Crappie      Bluegill Bluntnose Minnow      Iowa Darter
##           5.325444      32.544379      15.236686      4.733728
##  Largemouth Bass      Pumpkinseed  Tadpole Madtom      Yellow Perch
##           33.727811      1.923077      0.887574      5.621302
```

```
class(cSpecPct)
```

```
## [1] "table"
```

#Q13. Convert the table, <cSpecPct>, to a data frame named <u> and confirm that <u> is a data

#frame

```
u <- as.data.frame(cSpecPct)
class(u)
```

```
## [1] "data.frame"
```

```
u
```

```
##           Var1      Freq
## 1  Black Crappie  5.325444
## 2    Bluegill  32.544379
## 3 Bluntnose Minnow 15.236686
## 4    Iowa Darter  4.733728
## 5 Largemouth Bass 33.727811
## 6    Pumpkinseed  1.923077
## 7  Tadpole Madtom  0.887574
## 8    Yellow Perch  5.621302
```

#Q14. Create a barplot of <cSpec> with the following: titled Fish Count with the following

#specifications:

??? Title: Fish Count

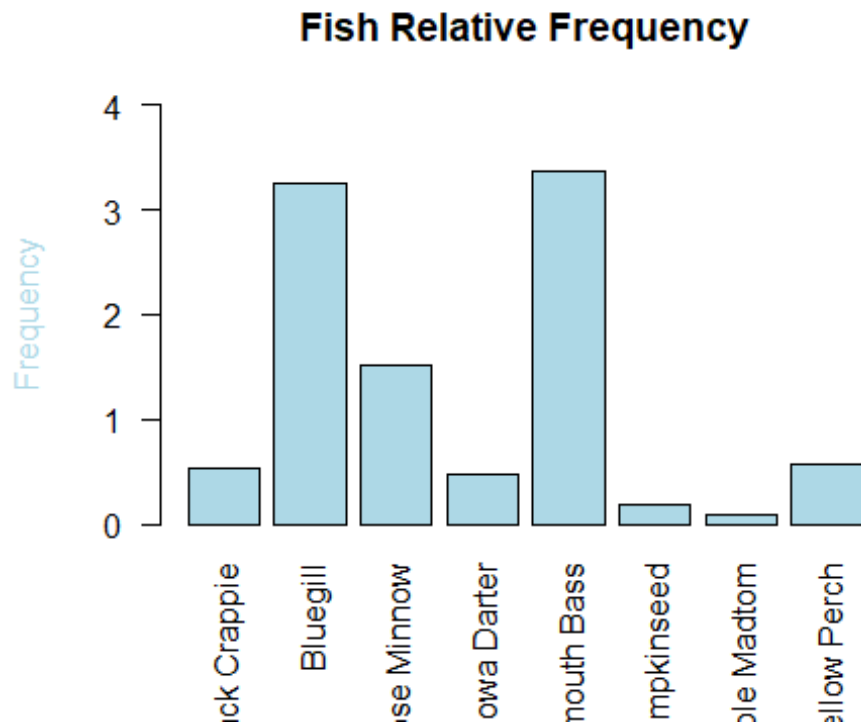
#??? Y axis is Labeled "COUNTS"

#??? Color the bars Light Green

#??? Rotate Y axis to be horizontal

#??? Set the X axis font magnification to 60% of nominal

```
?barplot
```

#Q16. Rearrange the <u> cSpec Pct data frame in descending order of relative frequency. Save

#the rearranged data frame as the object <d>

```
d <- u[order(u$Freq, decreasing = TRUE),]
d
```

```
##           Var1      Freq
## 5  Largemouth Bass 33.727811
## 2      Bluegill 32.544379
## 3 Bluntnose Minnow 15.236686
## 8   Yellow Perch  5.621302
## 1   Black Crappie  5.325444
## 4     Iowa Darter  4.733728
## 6   Pumpkinseed  1.923077
## 7   Tadpole Madtom  0.887574
```

#Q17. Rename the <d> columns Var 1 to Species, and Freq to RelFreq

```
colnames(d)[1] <- "Species"
```

```
colnames(d)[2] <- "RelFreq"
```

```
d
```

```
##           Species  RelFreq
## 5  Largemouth Bass 33.727811
## 2      Bluegill 32.544379
## 3 Bluntnose Minnow 15.236686
## 8   Yellow Perch  5.621302
## 1   Black Crappie  5.325444
```

```
## 4      Iowa Darter  4.733728
## 6      Pumpkinseed 1.923077
## 7      Tadpole Madtom 0.887574
```

#Q18. Add new variables to <d> and call them cumfreq, counts, and cumcounts
t\$Freq

```
## [1] 36 220 103 32 228 13 6 38
```

```
desc <- t[order(-t$Freq),]
desc$Freq
```

```
## [1] 228 220 103 38 36 32 13 6
```

```
d <- d %>% mutate(cumfreq=cumsum(d$RelFreq), counts=desc$Freq, cumcounts = cum
sum(desc$Freq))
d
```

```
##           Species  RelFreq  cumfreq counts cumcounts
## 5  Largemouth Bass 33.727811 33.72781    228      228
## 2      Bluegill 32.544379 66.27219    220      448
## 3 Bluntnose Minnow 15.236686 81.50888    103      551
## 8   Yellow Perch  5.621302 87.13018     38      589
## 1   Black Crappie  5.325444 92.45562     36      625
## 4      Iowa Darter  4.733728 97.18935     32      657
## 6      Pumpkinseed  1.923077 99.11243     13      670
## 7      Tadpole Madtom 0.887574 100.00000     6      676
```

#Q19. Create a parameter variable <def_par> to store parameter variables

```
def_par <- as.data.frame(names(d))
def_par
```

```
##      names(d)
## 1   Species
## 2   RelFreq
## 3   cumfreq
## 4    counts
## 5 cumcounts
```

```
colnames(def_par) <- c('Parameters')
class(def_par)
```

```
## [1] "data.frame"
```

#Q20. Create a barplot, <pc>, with the following specifications:

####d\$counts of width 1, spacing of .15

####no boarder

####Axes: F

*####Yaxis Limit 0,3.05*max*

####d\$counts na.rm is true

####y Label is Cumulative Counts

####scale x axis to 70%

```

#???names.arg: d$Species
#???Title of the barplot is "Species Pareto"
#???las: 2)

pc <- barplot(d$counts, width = 1, space = 0.15, border = NA, axes = F, ylim =
c(0, 3.05*228),
            ylab = "Cumulative Counts", cex.axis = 0.7, names.arg = d$Species,
            main = "Species Pareto", las = 2)

#Q21. Add a cumulative counts line to the <pc> plot with the following:
#??? Spec line type is b
#??? Scale plotting text at 70%
#??? Data values are solid circles with color cyan4

lines(pc, d$cumcounts, type = "b", cex = 0.7, col = "cyan4")

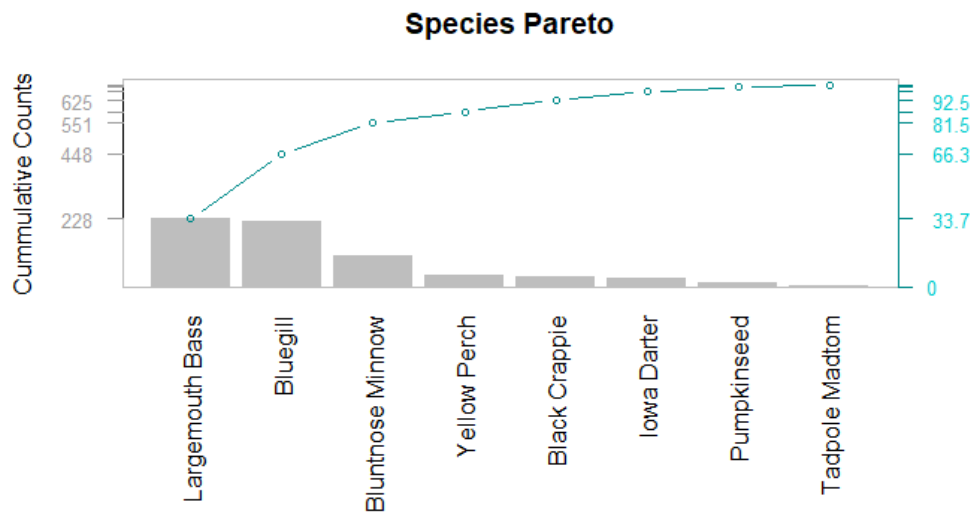
#Q22. Place a grey box around the pareto plot
box(which = "plot", lty = "solid", col = "grey", par(mar = c(8, 5, 4, 4)))

#Q23. Add a left side axis with the following specifications
#??? Horizontal values at tick marks at cumcounts on side 2
#??? Tickmark color of grey62
#??? Color of axis is grey62
#??? Axis scaled to 80% of normal

axis(side = 2, at = d$cumcounts, labels = d$cumcounts, las = 2, col.axis = "grey62",
      col.ticks = "grey62", tick = TRUE, cex.axis = 0.8, par(mar = c(8, 5, 4, 4)))

#Q24. Add axis details on right side of box with the specifications:
#??? Spec: Side 4
#??? Tickmarks at cumcounts with labels from 0 to cumfreq with %,
#??? Axis color of cyan5 and label color of cyan4
#??? Axis font scaled to 80% of nominal
axis(side = 4, at = c(0, d$cumcounts),
      labels = c(0, d$cumfreq),
      las = 1, col.axis = 'cyan3', col = 'cyan4',
      cex.axis = 0.8, par(mar = c(8, 5, 4, 4)))

```



#Q25. Display the finished Species Pareto Plot (without the star watermarks). Have your last #name on the plot

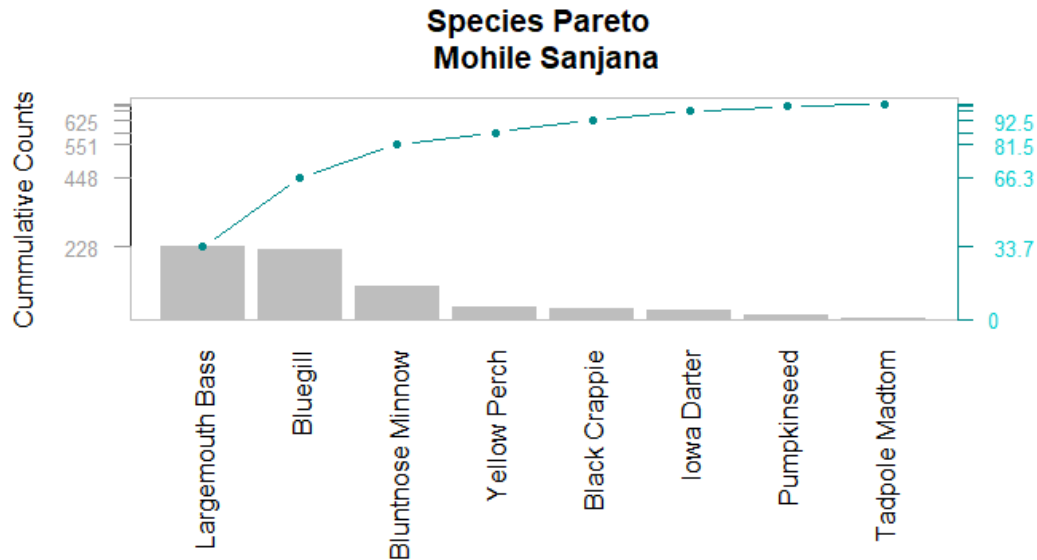
```
pc <- barplot(d$counts, width = 1, space = 0.15, border = NA, axes = F, ylim =
c(0, 3.05*228),
            ylab = "Cumulative Counts", cex.axis = 0.8, names.arg = d$Species,
            main = "Species Pareto \n Mohile Sanjana", las = 2)
lines(pc, d$cumcounts, type = "b", pch = 19, cex = 0.7, col = "cyan4")
box(which = "plot", lty = "solid", col = "grey")

axis(side = 2, at = d$cumcounts, labels = d$cumcounts, las = 2, col.axis = "grey62",
     col.ticks = "grey62", tick = TRUE, cex.axis = 0.8)

d$cumfreq <- format(round(d$cumfreq, 3), nsmall = 1)
d$cumfreq

## [1] " 33.728" " 66.272" " 81.509" " 87.130" " 92.456" " 97.189" " 99.112"
## [8] "100.000"

axis(side = 4, at = c(0, d$cumcounts),
     labels = c(0, d$cumfreq),
     las = 1, col.axis = 'cyan3', col = 'cyan4',
     cex.axis = 0.8, par(mar = c(8, 5, 4, 4)))
```

Explanatory supplements

```
str(bio)
```

```
## 'data.frame': 676 obs. of 7 variables:
## $ netID : int 12 12 12 12 12 12 12 13 13 13 ...
## $ fishID : int 16 23 30 44 50 65 66 68 69 70 ...
## $ species: chr "Bluegill" "Bluegill" "Bluegill" "Bluegill" ...
## $ tl : int 61 66 70 38 42 54 27 36 59 39 ...
## $ w : num 2.9 4.5 5.2 0.5 1 2.1 NA 0.5 2 0.5 ...
## $ tag : chr "" "" "" "" ...
## $ scale : logi FALSE FALSE FALSE FALSE FALSE FALSE ...
```

```
summary(bio)
```

```
##      netID      fishID      species      tl
## Min.   : 1.00   Min.   : 7.0   Length:676   Min.   : 27.0
## 1st Qu.: 13.00  1st Qu.:175.8   Class :character  1st Qu.: 66.0
## Median : 37.00  Median :345.5   Mode  :character  Median :189.5
## Mean   : 67.65  Mean   :434.2           Mean   :186.5
## 3rd Qu.:109.00  3rd Qu.:695.5           3rd Qu.:295.0
## Max.   :206.00  Max.   :915.0           Max.   :429.0
##
##      w      tag      scale
## Min.   : 0.2   Length:676   Mode :logical
## 1st Qu.: 2.0   Class :character  FALSE:213
## Median : 54.5  Mode  :character  TRUE :463
## Mean   :126.8
## 3rd Qu.:190.5
## Max.   :1070.0
## NA's   :165
```

#A pie chart for better visual representation

```
pie(cSpecPct, hole = 0, values = "%", main = "Pie Chart of Species")

## Warning in text.default(1.1 * P$x, 1.1 * P$y, labels[i], xpd = TRUE, adj =
## ifelse(P$x < : "hole" is not a graphical parameter

## Warning in text.default(1.1 * P$x, 1.1 * P$y, labels[i], xpd = TRUE, adj =
## ifelse(P$x < : "values" is not a graphical parameter

## Warning in text.default(1.1 * P$x, 1.1 * P$y, labels[i], xpd = TRUE, adj =
## ifelse(P$x < : "hole" is not a graphical parameter

## Warning in text.default(1.1 * P$x, 1.1 * P$y, labels[i], xpd = TRUE, adj =
## ifelse(P$x < : "values" is not a graphical parameter

## Warning in text.default(1.1 * P$x, 1.1 * P$y, labels[i], xpd = TRUE, adj =
## ifelse(P$x < : "hole" is not a graphical parameter

## Warning in text.default(1.1 * P$x, 1.1 * P$y, labels[i], xpd = TRUE, adj =
## ifelse(P$x < : "values" is not a graphical parameter

## Warning in text.default(1.1 * P$x, 1.1 * P$y, labels[i], xpd = TRUE, adj =
## ifelse(P$x < : "hole" is not a graphical parameter

## Warning in text.default(1.1 * P$x, 1.1 * P$y, labels[i], xpd = TRUE, adj =
## ifelse(P$x < : "values" is not a graphical parameter

## Warning in text.default(1.1 * P$x, 1.1 * P$y, labels[i], xpd = TRUE, adj =
## ifelse(P$x < : "hole" is not a graphical parameter

## Warning in text.default(1.1 * P$x, 1.1 * P$y, labels[i], xpd = TRUE, adj =
## ifelse(P$x < : "values" is not a graphical parameter

## Warning in text.default(1.1 * P$x, 1.1 * P$y, labels[i], xpd = TRUE, adj =
## ifelse(P$x < : "hole" is not a graphical parameter

## Warning in text.default(1.1 * P$x, 1.1 * P$y, labels[i], xpd = TRUE, adj =
## ifelse(P$x < : "values" is not a graphical parameter

## Warning in title(main = main, ...): "hole" is not a graphical parameter
```

```
## Warning in title(main = main, ...): "values" is not a graphical parameter
```

Pie Chart of Species

