Module 3 Project

Name: Parth Sawant

Course name: ALY6000

Due Date: 05/02/2022

Title: Executive summary report 3

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Introduction

This project is based on the different species of fishes, their comparison among each other by finding the cumulative counts and cumulative frequency and presenting them through barplots and pareto chart. Different functions are used like the count and unique. We get to learn how to read a pareto chart and how to use its parameters. Getting to learn how one can add columns to the existing dataset using the mutate function. Using pareto chart we will get to know which specie has the highest frequency and many other factors

A. Provide an analysis of descriptive characteristics of the data set provided by your instructor. This includes pertinent statistics including counts, cumulative counts, and frequency, percentages, etc. Include R console screen shots to support your observations and conclusions.

1] 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" "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 FALSE ...
```

Explanation: -

- >> The structure of the dataset bio contains in total 676 rows and 7 columns with unique names like unique netID, fishID, species, total length, weight, tag nature and the scale nature for each fish whether they are sampled or not.
- >> There are different data types of each variable as seen like fishID has integer datatype, scale has logical datatype containing only true or false values.
- >> There are some missing values as well in the dataset like here we can see in weight of the fish.

2] Implementation: -

```
> headtail(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
674
      110
             863 Black Crappie 307 415.0 1783
             870 Black Crappie 279 344.0 1789
675
      129
676
             879 Black Crappie 302 397.0 1792
      129
```

Explanation: -

>> A total of 6 rows are displayed with headtail function, to show the first 3 and last 3 values of each variable.

- >> Sometimes the data frame is too huge to analyze them, so just the first and last 3 rows are enough to get a rough idea about the dataset.
- >> By default, the n value for headtail function is 3 which otherwise can be changed.

3] Output: -

Explanation: -

- >> using the count() function we are able to know the different number of species in the dataset.
- >> There are 8 different types of species Black Crappie, Bluegill, Bluntnose Minnow, Iowa Darter, Largemouth Bass, pumpkinseed, Tadpole Madtom, Yellow Perch.

>> Largemouth Bass has largest number of species with a count of 228 whereas Tadpole Madtom has a count of 6 which is the least among all species.

4] Implementation: -

```
> cSpecPct <- (table(bio\species)*100)/length(bio\species)</pre>
> 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: -

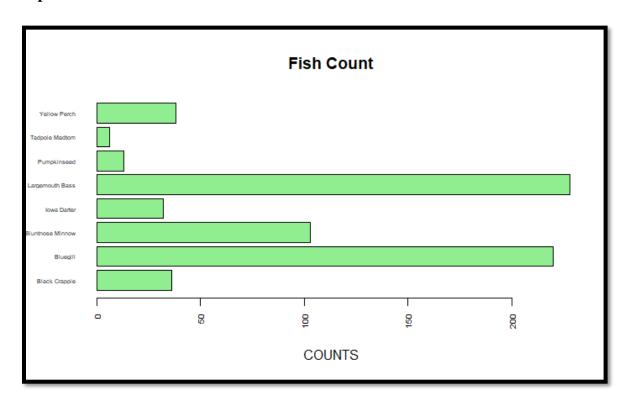
- >> We have calculated the percentage of each species using the total length of species and multiplying it by 100.
- >> From the output we can observe that Largemouth Bass has the highest percentage (33.72) of all species followed by Bluegill with 32.54% while Tadpole Madtom occupies 0.8% which is the lowest.

5] Implementation: -

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

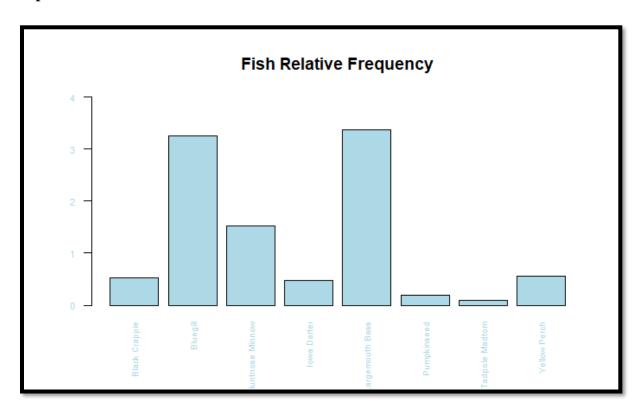
- >> To find cumulative frequency, we have first calculated counts and then cumulative counts for all individual species.
- >>> Cumcounts is calculated by adding the counts to the preceding one. This is how a pareto chart is calculated.
- >>> We have used the mutate function where we were able to add 3 columns for cumulative counts, cumulative frequency and counts.
- B. Provide the executive with visualizations (at least 3) in that help them see the key characteristics you want to highlight. They can be boxplots, histograms, frequency and probability distributions, or barplots (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 the significance of them.

Implementation: -



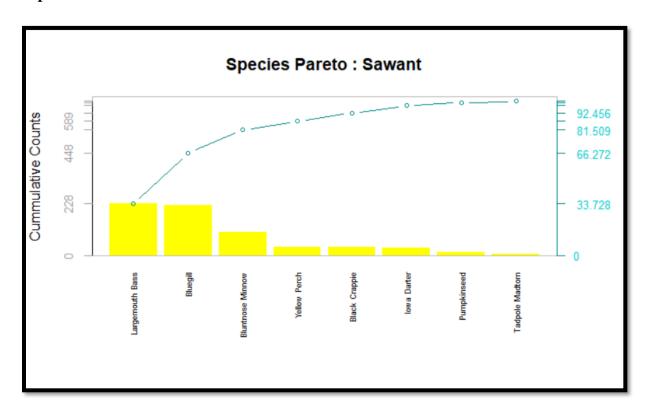
- >> The above graph is for the count of fishes.
- >> It can be noticed that Largemouth Bass and bluegill have most count of species as compared to others.
- >> Tadpole madtom and pumpkinseed have the least no of species.
- >> Though I kept a limit on the axis up to 200, we can see that Both Largemouth Bass and Bluegill have larger count than 200.

Implementation: -



- >> The above graph is for the relative frequency of the fish.
- >> From the above barplot it is noticed that Largemouth Bass and Bluegill have high relative frequency which means these species of fishes are to be seen more than other ones.
- >> On the other hand, the Tadpole Madtom are the least seen species as they have the lowest relative frequency, followed by Pumpkinseed.
- >> We can say that approximately Black Clappie, lowa Darter and Yellow Perch have approximately same frequency that is below 1%.

Implementation: -



- >> Pareto Chart is a type of chart that contains both bars and a line graph where values are represented in descending order by bars, and cumulative values by blue line as seen in the chart.
- >> It is better than pie charts and many other, as it reveals more information.
- >> To make a Pareto chart we must first sort data in descending order and then calculate cumulative sum and frequency of the counts.
- >> As noticed from the Pareto chart, the cumulative counts are on the left side whereas the frequency is on the other side. Also, the different species are placed on the X axis.
- >> It is observed that Largemouth Bass, Bluegill are the ruling species among all.
- >> Largemouth Bass constitute 33.72% where as Tadpole Madtom constitutes only nearly 1% of all species.
- >> It can be observed that first three species have a majority among all. It can also be noted that since Largemouth Bass, Bluegill have high frequency, they may be feeding on other fishes since their population is high.

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 provides a wide Explanation of different fishes with each fish having unique fishID and specie type. We arranged the data in descending order according to relative frequency of different species. Using the mutate function we calculated the cumulative frequency and cumulative counts needed for pareto chart. With the Pareto we concluded that Largemouth Bass and Bluegill have the highest frequency and are the most dangerous species among all.

Bibliography

[1] Rotating axis labels in R plots (Tender is the byte, May 2021)

Source: https://www.tenderisthebyte.com/blog/2019/04/25/rotating-axis-labels-in-r/

Last Accessed: 04th February,2022

[2] How do I find percentage of R in something. (Stack overflow)

Source: https://stackoverflow.com/questions/42379751/how-do-i-find-the-percentage-of-

something-in-r

Last Accessed: 04th February,2022

[3] Pareto Chart with Base R plotting System. (Amazon news)

Source: https://rstudio-pubs-

static.s3.amazonaws.com/72023 670962b57f444c04999fd1a0a393e113.html

Last Accessed: 04th February,2022

[4] How to use mutate in R (sharpsightlabs, August 2018)

Source: https://www.sharpsightlabs.com/blog/mutate-in-r/

Last Accessed: 05th February, 2022

My Github repository Link

https://github.com/parthh03/ALY6000_module3.git

module3.R

Parth

2022-02-05

```
#1. print your name at the top of the script and load these libraries:
FSA, FSAdata, magrittr, dplyr, tidyr plyr and tidyverse
print("Parth Sawant")
## [1] "Parth Sawant"
r=getOption("repos")
r["CRAN"]="http://cran.us.r-project.org"
options(repos=r)
install.packages('FSA')
## Installing package into 'C:/Users/Parth/OneDrive/Documents/R/win-li
brary/4.1'
## (as 'lib' is unspecified)
## package 'FSA' successfully unpacked and MD5 sums checked
##
## The downloaded binary packages are in
## C:\Users\Parth\AppData\Local\Temp\RtmpgPNLyt\downloaded_packages
install.packages('FSAdata')
## Installing package into 'C:/Users/Parth/OneDrive/Documents/R/win-li
brary/4.1'
## (as 'lib' is unspecified)
## package 'FSAdata' successfully unpacked and MD5 sums checked
##
## The downloaded binary packages are in
   C:\Users\Parth\AppData\Local\Temp\RtmpgPNLyt\downloaded packages
install.packages('magrittr')
## Installing package into 'C:/Users/Parth/OneDrive/Documents/R/win-li
brary/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\Parth\OneDrive\Documents\R\win-
## library\4.1\00LOCK\magrittr\libs\x64\magrittr.dll
## to C:\Users\Parth\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\Parth\AppData\Local\Temp\RtmpgPNLyt\downloaded packages
install.packages('dplyr')
## Installing package into 'C:/Users/Parth/OneDrive/Documents/R/win-li
brary/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\Parth\OneDrive\Documents\R\win-
## library\4.1\00LOCK\dplyr\libs\x64\dplyr.dll to C:
## \Users\Parth\OneDrive\Documents\R\win-library\4.1\dplyr\libs\x64\dp
lyr.dll:
## Permission denied
## Warning: restored 'dplyr'
##
## The downloaded binary packages are in
## C:\Users\Parth\AppData\Local\Temp\RtmpgPNLyt\downloaded packages
install.packages('tidyr')
## Installing package into 'C:/Users/Parth/OneDrive/Documents/R/win-li
brary/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\Parth\OneDrive\Documents\R\win-
## library\4.1\00LOCK\tidyr\libs\x64\tidyr.dll to C:
## \Users\Parth\OneDrive\Documents\R\win-library\4.1\tidyr\libs\x64\ti
```

```
dvr.dll:
## Permission denied
## Warning: restored 'tidyr'
##
## The downloaded binary packages are in
## C:\Users\Parth\AppData\Local\Temp\RtmpgPNLyt\downloaded packages
install.packages('plyr')
## Installing package into 'C:/Users/Parth/OneDrive/Documents/R/win-li
brary/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 cop
ying C:
## \Users\Parth\OneDrive\Documents\R\win-library\4.1\00LOCK\plyr\libs\
x64\plyr.dll
## to C:\Users\Parth\OneDrive\Documents\R\win-library\4.1\plyr\libs\x6
4\plyr.dll:
## Permission denied
## Warning: restored 'plyr'
##
## The downloaded binary packages are in
## C:\Users\Parth\AppData\Local\Temp\RtmpgPNLyt\downloaded packages
install.packages('tidyverse')
## Installing package into 'C:/Users/Parth/OneDrive/Documents/R/win-li
brarv/4.1'
## (as 'lib' is unspecified)
## package 'tidyverse' successfully unpacked and MD5 sums checked
##
## The downloaded binary packages are in
## C:\Users\Parth\AppData\Local\Temp\RtmpgPNLyt\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 bo
ok.
```

```
library(FSAdata)
## ## FSAdata v0.3.8. See ?FSAdata to find data for specific fisheries
analyses.
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 fi
rst, then 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 ----- tidyv
erse 1.3.1 --
## v ggplot2 3.3.5
                                 0.3.4
                       v purrr
## v tibble 3.1.6
                       v stringr 1.4.0
## v readr
             2.1.1
                       v forcats 0.5.1
## -- Conflicts -----
                                   ----- tidyverse c
onflicts() --
## 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()
#2. Import the inchBio.csv and name the table <bio>
bio <- read.csv("C:/Users/Parth/OneDrive/Desktop/Introduction to Analy</pre>
tics/ALY6000 module3/inchBio.csv")
bio
       netID fishID
##
                             species
                                      tl
                                                     tag scale
                                              W
## 1
          12
                 16
                            Bluegill
                                      61
                                            2.9
                                                          FALSE
## 2
          12
                 23
                            Bluegill
                                            4.5
                                                          FALSE
                                      66
          12
                            Bluegill
                                            5.2
## 3
                 30
                                      70
                                                          FALSE
                                            0.5
## 4
          12
                 44
                            Bluegill
                                      38
                                                          FALSE
## 5
          12
                 50
                                            1.0
                            Bluegill
                                      42
                                                          FALSE
## 6
          12
                 65
                            Bluegill
                                      54
                                            2.1
                                                          FALSE
          12
## 7
                 66
                            Bluegill
                                      27
                                             NA
                                                          FALSE
          13
## 8
                 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
                                            0.5
                                                          FALSE
                            Bluegill
                                      34
```

	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		120.0	TRUE
##	34	4	127	Bluegill	168	90.0	TRUE
##	35	4	128	Bluegill	165	85.0	TRUE
##	36	4	129	Bluegill		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		42.0	TRUE
##	43	5	139	Bluegill	38	2.0	FALSE
##	44	5	140	Bluegill		NA	FALSE
##	45	5	141	Bluegill		NA	FALSE
##	46	5	142	Bluegill	46	NA	FALSE
##	47	5	143	Bluegill	165	68.0	TRUE
##	48	5	144	Bluegill		NA	FALSE
##	49	5	145	Bluegill		4.0	FALSE
##	50	5	146	Bluegill		184.0	TRUE
##	51	5	147	Bluegill		98.0	TRUE
##	52	5	148	Bluegill		62.0	TRUE
	53	5	149	Bluegill		NA	FALSE
##	54	5	150	Bluegill		76.0	TRUE
##	55	10	151	Bluegill		100.0	TRUE
##	56	10	152	Bluegill		95.0	TRUE
##	57	10	153	Bluegill		130.0	TRUE

##	58	10	154	Bluegill	218	250.0	TRUE
	59	10	155	Bluegill		197.0	TRUE
	60	10	156	Bluegill		78.0	TRUE
##		10	157	Bluegill		72.0	TRUE
##		10	158	Bluegill		98.0	TRUE
##		9	161	Bluegill		175.0	TRUE
	64	9	162	Bluegill		144.0	TRUE
	65	9	163	Bluegill		148.0	TRUE
##		9	164	Bluegill		130.0	TRUE
##		9	165	Bluegill		185.0	TRUE
	68	9	166	Bluegill		225.0	TRUE
	69	9	167	Bluegill		80.0	FALSE
	70	9	168	Bluegill		180.0	TRUE
	71	9	169	Bluegill		175.0	TRUE
	72	9	170	Bluegill		NA	TRUE
##		9	171	Bluegill		160.0	FALSE
	74	9	172	Bluegill		NA	FALSE
	75	6	176	Bluegill		200.0	TRUE
##		6	177	Bluegill		74.0	TRUE
	77	6	178	Bluegill		62.0	TRUE
	78	6	179	Bluegill		220.0	TRUE
	79	6	180	Bluegill		149.0	TRUE
##	80	6	181	Bluegill		139.0	TRUE
	81	6	182	Bluegill		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		32.0	TRUE
	92	11	193	Bluegill		218.0	TRUE
##		11	194	Bluegill		60.0	TRUE
	94	11	195	Bluegill		70.0	TRUE
	95	17	196	Bluegill		192.0	TRUE
	96	17	197	Bluegill		31.0	TRUE
	97	17	199	Bluegill		59.0	TRUE
	98	20	201	Bluegill		54.0	TRUE
	99	15	203	Bluegill		40.0	TRUE
	100	15	206	Bluegill		30.0	TRUE
	101	15	207	Bluegill		20.0	TRUE
	102	16	210	Bluegill		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		60.0	TRUE
##	107	16	215	Bluegill		60.0	TRUE
	108	21	217	Bluegill		14.0	TRUE
	109	21	219	Bluegill		38.0	TRUE
	110	26	220	Bluegill		49.0	TRUE
	111	26	221	Bluegill		41.0	TRUE
	112	23	224	Bluegill		20.0	TRUE
	113	27	226	Bluegill		20.0	TRUE
	114	27	228	Bluegill		20.0	TRUE
	115	28	230	Bluegill		50.0	TRUE
	116	28	231	Bluegill		280.0	TRUE
	117	37	322	Bluegill		NA	TRUE
	118	37 37	356	Bluegill		NA NA	TRUE
	119	206	501	Bluegill	38	0.7	FALSE
	120	205	502	Bluegill		1.4	FALSE
				_	43		
	121	205	503	Bluegill		1.5	FALSE
	122	205	504	Bluegill	53	1.4	FALSE
	123	205	505	Bluegill	38	1.0	FALSE
	124	205	506	Bluegill		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		229.0	1025 TRUE
##	147	101	545	Bluegill		101.0	TRUE
##	148	101	546	Bluegill		110.0	TRUE
##	149	101	547	Bluegill		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		137.0		TRUE
	154	101	553	Bluegill		165.0		TRUE
	155	101	554	Bluegill		103.0		TRUE
	156	101	555	Bluegill		166.0		TRUE
	157	101	556	Bluegill		115.0		TRUE
	158	101	557	Bluegill		89.0		TRUE
	159	101	558	Bluegill		121.0		TRUE
	160	101	559	Bluegill		78.0		TRUE
	161	101	560	Bluegill		63.0		TRUE
	162	101	561	Bluegill		181.0	1075	TRUE
	163	101	562	Bluegill		60.0	10/5	TRUE
	164	101	563	Bluegill		74.0		TRUE
	165	101	564	Bluegill		101.0		TRUE
	166	101	565	Bluegill		141.0		TRUE
	167	101	566	Bluegill		228.0	1074	TRUE
	168			Bluegill				TRUE
		101	567	_		266.0	1073	
	169	101	568	Bluegill		93.0		TRUE
	170	101	569	Bluegill		74.0		TRUE
	171	101	570	Bluegill		82.0	4070	TRUE
	172	101	571	Bluegill		274.0	1072	TRUE
	173	101	572	Bluegill		151.0		TRUE
	174	101	573	Bluegill		129.0		TRUE
	175	101	574	Bluegill		98.0		TRUE
	176	101	575	Bluegill		57.0		TRUE
	177	101	576	Bluegill		62.0		TRUE
	178	101	577	Bluegill		107.0		TRUE
	179	101	578	Bluegill		240.0	1071	TRUE
	180	101	580	Bluegill		110.0		TRUE
	181	101	581	Bluegill		208.0	1069	TRUE
	182	101	583	Bluegill		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		255.0	1098	TRUE
##	193	104	660	Bluegill		111.0		TRUE
	194	103	689	Bluegill		NA	1107	TRUE
##	195	103	695	Bluegill		45.3		TRUE
				0				

##	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		50.9	TRUE
	201	111	717	Bluegill		35.0	TRUE
	202	111	718	Bluegill		45.0	TRUE
	203	111	719	Bluegill		39.0	TRUE
	204	111	722	Bluegill			TRUE
	205	111	726	Bluegill		50.0	TRUE
	206	113	742	Bluegill		39.0	TRUE
	207	114	755	Bluegill		3.1	FALSE
	208	116	761	Bluegill		10.9	FALSE
	209	116	762	Bluegill		10.6	FALSE
	210	112	774	Bluegill		171.0	1045 TRUE
	211	112	780	Bluegill		6.1	FALSE
				•			
	212	112	781	Bluegill		19.0	FALSE
	213	120	802	Bluegill		10.7	TRUE
	214	120	803	Bluegill		8.9	TRUE
	215	120	804	Bluegill		10.6	TRUE
	216	120	805	Bluegill		8.7	TRUE
	217	120	806	Bluegill		9.3	TRUE
	218	120	807	Bluegill		4.0	TRUE
	219	119	824	Bluegill		25.0	TRUE
	220	122	826	Bluegill		33.0	TRUE
	221	12	7	Bluntnose Minnow		2.0	FALSE
	222	12	8	Bluntnose Minnow		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		2.0	FALSE
	234	12		Bluntnose Minnow		1.8	FALSE
	235	12		Bluntnose Minnow		2.5	FALSE
	236	12		Bluntnose Minnow		4.0	FALSE
	237	12		Bluntnose Minnow		2.0	FALSE
	238	12		Bluntnose Minnow		2.0	FALSE
	239	12		Bluntnose Minnow		3.0	FALSE
	240	12		Bluntnose Minnow		4.8	FALSE
	241	12		Bluntnose Minnow			FALSE
##	Z41	12	54	PIULIUOSE MILIUOM	00	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		Bluntnose		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		Bluntnose		72	5.0	FALSE
##	271	13		Bluntnose		71	4.0	FALSE
	272	205		Bluntnose		61	2.1	FALSE
	273	201		Bluntnose			NA	FALSE
	274	101		Bluntnose		64	1.4	FALSE
	275	101		Bluntnose		46	0.5	FALSE
	276	101		Bluntnose		53	1.0	FALSE
	277	101		Bluntnose		58	1.7	FALSE
	278	101		Bluntnose		56	1.0	FALSE
	279	101		Bluntnose		58	1.1	FALSE
	280	101		Bluntnose		58	1.0	FALSE
	281	101		Bluntnose		56	1.0	FALSE
	282	101		Bluntnose		53	1.0	FALSE
	283	101		Bluntnose		51	0.8	FALSE
	284	101		Bluntnose		56	0.8	FALSE
	285	101		Bluntnose		56 50	1.0	FALSE
	286	101		Bluntnose		58	1.0	FALSE
##	287	101	603	Bluntnose	MITHUM	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		Bluntnose		56	1.0	FALSE
	298	101	615	Bluntnose	Minnow	53	0.9	FALSE
##	299	101		Bluntnose		58	1.2	FALSE
	300	101		Bluntnose		53	0.8	FALSE
##	301	101		Bluntnose		53	0.9	FALSE
##	302	101		Bluntnose		66	1.5	FALSE
	303	101		Bluntnose		61	1.3	FALSE
	304	102		Bluntnose		74	3.9	FALSE
##	305							
	306	102		Bluntnose		61	2.2	FALSE FALSE
		102		Bluntnose		84	1.3	
##	307	102		Bluntnose		53	1.3	FALSE
##	308	102		Bluntnose		56	2.1	FALSE
	309	102		Bluntnose		56	1.4	FALSE
	310	102		Bluntnose		61	2.0	FALSE
##	311	111		Bluntnose		79	3.8	FALSE
	312	116		Bluntnose		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		Bluntnose		79	4.1	FALSE
	324	13	72		Darter	61	NA	FALSE
	325	205	521		Darter	53	1.1	FALSE
	326	203	524		Darter		0.6	FALSE
	327	203	525		Darter	44	0.6	FALSE
	328	203	526		Darter	46	0.6	FALSE
	329	203	527		Darter	43	0.4	FALSE
	330	101	602		Darter	43 56	0.9	FALSE
	331	101	604		Darter		0.8	FALSE
	332	102	641		Darter		1.0	FALSE
##	333	106	646	Iowa	Darter	38	0.6	FALSE

	224	100	- 4-	T D .		0 0			
	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		0.8		FALSE	
	345	115	731			0.8		FALSE	
				Iowa Darter					
	346	115	732	Iowa Darter		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		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		NA	g0996	TRUE	
	358	1	90	Largemouth Bass		NA	y06073	TRUE	
	359	1	91	Largemouth Bass		NA	00449	TRUE	
	360	1	92	Largemouth Bass		NA	00448	TRUE	
	361	1	93	Largemouth Bass		NA	00447	TRUE	
	362	1	94	_			g0985	TRUE	
				Largemouth Bass		NA NA	gosos		
	363	1	95	Largemouth Bass		NA	~0000	TRUE	
	364	1	96	Largemouth Bass		NA	g0986	TRUE	
	365	1	97	Largemouth Bass		NA	g0987	TRUE	
	366	1	98	Largemouth Bass		NA	g1000	TRUE	
	367	1	99	Largemouth Bass		NA	g0999	TRUE	
	368	1	100	Largemouth Bass		NA		TRUE	
	369	1	101	Largemouth Bass		NA		TRUE	
##	370	1	102	Largemouth Bass	348	NA	g0997	TRUE	
##	371	2	104	Largemouth Bass	307	NA	00501	TRUE	
##	372	2	105	Largemouth Bass	323	NA	00506	TRUE	
##	373	2	106	Largemouth Bass	231	NA		TRUE	
##	374	2	107	Largemouth Bass		NA		TRUE	
##	375	2	108	Largemouth Bass		NA	o0507	TRUE	
	376	2	109	Largemouth Bass		NA	00510	TRUE	
	377	2	110	Largemouth Bass		NA	00511	TRUE	
	378	2	111	Largemouth Bass		NA	00512	TRUE	
	379	2	112	Largemouth Bass		NA		TRUE	
шт		_		Lai Bemodeli Dass	4 / T	11/7		INOL	

##	200	17	100	Langomouth	Dacc	250	100 0		TDUE
	380	17	198	Largemouth			190.0		TRUE
	381	27	227	Largemouth			20.0	0546	TRUE
	382	29	232	Largemouth			NA	00516	TRUE
	383	29	233	Largemouth			NA	o0517	TRUE
	384	29	234	Largemouth			NA	o0518	TRUE
##	385	29	235	Largemouth	Bass	325	NA	o0520	TRUE
##	386	29	237	Largemouth	Bass	330	NA	00521	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			NA	o0525	TRUE
##	392	30	244	Largemouth			NA	o0550	TRUE
	393	30	245	Largemouth			NA		TRUE
	394	30	246	Largemouth			NA		TRUE
	395	30	247	Largemouth			NA		TRUE
	396	30	248	Largemouth			NA	o0549	TRUE
	397	30	249	Largemouth			NA	00545	TRUE
	398	30	250	Largemouth			NA		TRUE
	399	30	251	Largemouth			NA NA	o0548	TRUE
				•					
	400	30	252	Largemouth			NA	00547	TRUE
	401	30	253	Largemouth			NA	00546	TRUE
	402	30	254	Largemouth			NA	05.45	TRUE
	403	30	255	Largemouth			NA	00545	TRUE
	404	30	256	Largemouth			NA	00544	TRUE
	405	30	257	Largemouth			NA	o0543	TRUE
	406	30	258	Largemouth			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	00536	TRUE
##	413	30	265	Largemouth	Bass	310	NA	o0535	TRUE
##	414	30	266	Largemouth			NA		TRUE
##	415	31	270	Largemouth			NA	o0533	TRUE
##	416	31	271	Largemouth			NA	o0532	TRUE
	417	31	272	Largemouth			NA		TRUE
	418	33	273	Largemouth			NA		TRUE
	419	32	274	Largemouth			NA		TRUE
	420	35	275	Largemouth			NA		TRUE
	421	35	276	Largemouth			NA		TRUE
	421	35	277	Largemouth			NA	o0519	TRUE
				_				00313	
	423	35	278	Largemouth			NA NA	~ 0000	TRUE
	424	35	279	Largemouth			NA	g0988	TRUE
##	425	35	280	Largemouth	Rass	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	00991	TRUE	
##	429	35	285	Largemouth	Bass	312	NA	00989	TRUE	
##	430	35	286	Largemouth	Bass	333	NA	y00003	TRUE	
##	431	35	287	Largemouth	Bass	338	NA	y00228	TRUE	
##	432	35	288	Largemouth			NA	•	TRUE	
##	433	35	289	Largemouth			NA		TRUE	
##	434	35	290	Largemouth			NA	y160016	TRUE	
##	435	35	291	Largemouth	Bass	376	NA	00534	TRUE	
##	436	35	292	Largemouth			NA		TRUE	
##	437	35	293	Largemouth			NA	y2519	TRUE	
##	438	35	294	Largemouth			NA	y00034	TRUE	
##	439	35	295	Largemouth			NA	y00123	TRUE	
##	440	35	296	Largemouth			NA	,	TRUE	
##	441	35	297	Largemouth			NA	o0531	TRUE	
##	442	35	298	Largemouth			NA	00530	TRUE	
##	443	35	299	Largemouth			NA	y01525		
	444	36	300	Largemouth			NA	00529	TRUE	
	445	36	301	Largemouth			NA	o0528	TRUE	
	446	36	302	Largemouth			NA	o0527	TRUE	
	447	36	303	Largemouth			NA		TRUE	
	448	36	304	Largemouth			NA		TRUE	
	449	36	305	Largemouth			NA		TRUE	
##	450	36	306	Largemouth			NA	o0526	TRUE	
##	451	36	307	Largemouth			NA		TRUE	
##	452	36	308	Largemouth			NA	y01598	TRUE	
##	453	36	309	Largemouth	Bass	373	NA	y01600	TRUE	
##	454	36	310	Largemouth	Bass	290	NA	-	TRUE	
##	455	36	311	Largemouth			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			NA	y01601	TRUE	
##	467	37	325	Largemouth			NA	-	TRUE	
##	468	37	326	Largemouth			NA		TRUE	
##	469	37	327	Largemouth			NA	y00023	TRUE	
##	470	37	328	Largemouth			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			NA		TRUE	
##	479	37	339	Largemouth			NA		TRUE	
##	480	37	340	Largemouth			NA		TRUE	
	481	37	341	Largemouth			NA		TRUE	
	482	37	342	Largemouth			NA		TRUE	
	483	37	343	Largemouth			NA		TRUE	
	484	37	344	Largemouth			NA		TRUE	
	485	37	345	Largemouth			NA		TRUE	
	486	37	346	Largemouth			NA	y00025	TRUE	
	487	37	347	Largemouth			NA	y00229	TRUE	
	488	37	348	Largemouth			NA	y00409	TRUE	
	489	37	349	Largemouth			NA	y 00 103	TRUE	
	490	37	352	Largemouth			NA	y00006	TRUE	
	491	37	353	Largemouth			NA	y00015	TRUE	
	492	37	354	Largemouth			NA NA	y00013	TRUE	
	493	37	355	Largemouth			NA	y00027	TRUE	
	494	101	536	Largemouth			407.0	1019	TRUE	
	495	101	551	Largemouth			60.0	1019	TRUE	
	496	101	588	Largemouth			117.0		TRUE	
	497	101		_				1050	TRUE	
			630	Largemouth				1058		
	498	102	631	Largemouth			737.0	1057	TRUE	
	499	102	632	Largemouth			548.0	1056	TRUE	
	500	104	656	Largemouth			172.0	1100	TRUE	
	501	104	657	Largemouth			688.0	1100	TRUE	
	502	103	702	Largemouth			579.0	1115	TRUE	
	503	111	713	Largemouth			556.0	1117	TRUE	
	504	114	759	Largemouth			48.0		TRUE	
	505	116	760	Largemouth			3.1		FALSE	
	506	112	801	Largemouth			52.0	4=40	FALSE	
	507	109	836	Largemouth			864.0	1518	TRUE	
	508	109	837	Largemouth			410.0	1519	TRUE	
	509	109	838	Largemouth				1520	TRUE	
	510	109	839	Largemouth				1521	TRUE	
	511	109	840	Largemouth			235.0		TRUE	
	512	109	841	Largemouth			NA	1522	TRUE	
	513	109	842	Largemouth			386.0	06860	TRUE	
	514	109	843	Largemouth			460.0	1523	TRUE	
	515	109	844	Largemouth			547.0	1524	TRUE	
	516	109	845	Largemouth			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			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					TRUE
##	523	109	852	Largemouth					TRUE
	524	109	853	Largemouth				1779	TRUE
	525	109	854	Largemouth			358.0		TRUE
	526	109	855	Largemouth					TRUE
	527	109	856	Largemouth				00526	TRUE
	528	109	857	Largemouth				1780	TRUE
	529	109	858	Largemouth				1781	TRUE
	530	109	859	Largemouth				1782	TRUE
	531	109	860	Largemouth				_,	TRUE
	532	109	861	Largemouth			92.0		TRUE
	533	109	862	Largemouth					TRUE
	534	110	864	Largemouth				1784	TRUE
	535	110	865	Largemouth				1785	TRUE
	536	110	866	Largemouth				1786	TRUE
	537	110	867	Largemouth			233.0	1700	TRUE
	538	129	868	Largemouth				1787	TRUE
	539	129	869	Largemouth				1788	TRUE
	540	129	871	Largemouth				1700	TRUE
	541	129	872	Largemouth				1790	TRUE
	542	129	873	Largemouth				1/50	TRUE
	543	129	873 874	Largemouth			433.0	1791	TRUE
	544	129	875	Largemouth				1//1	TRUE
	545	129	875 876	Largemouth					TRUE
	546	129	877	Largemouth					TRUE
	547	129	877 878	•				o0507	TRUE
	548	129	880	Largemouth Largemouth				00307	TRUE
	549	129	881	_				1793	TRUE
	_			Largemouth			437.0	1/93	
	550 EE1	129	882	Largemouth				1704	TRUE
	551	129	883	Largemouth				1794	TRUE
	552	129	884	Largemouth				1795	TRUE
	553	129	885	Largemouth					TRUE
	554	129	886	Largemouth			132.0		TRUE
	555	129	887	Largemouth			168.0	1706	TRUE
	556	129	888	Largemouth			385.0	1796	TRUE
	557	129	889	Largemouth			324.0	1797	TRUE
	558	129	890	Largemouth			461.0	1798	TRUE
	559	129	891	Largemouth			110.0	4700	TRUE
	560	130	892	Largemouth			346.0	1799	TRUE
	561	130	893	Largemouth				1800	TRUE
	562	130	894	Largemouth				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		349.0	o0529	TRUE	
##	575	130	907	Largemouth Bass		723.0	1706	TRUE	
##	576	130	908	Largemouth Bass		571.0	1707	TRUE	
##	577	130	909	Largemouth Bass		315.0		TRUE	
	578	130	910	Largemouth Bass		462.0	1708	TRUE	
##	579	131	911	Largemouth Bass		342.0		TRUE	
	580	131	912	Largemouth Bass		215.0		TRUE	
	581	131	913	Largemouth Bass		180.0		TRUE	
	582	131	914	Largemouth Bass		646.0	1709	TRUE	
	583	131	915	Largemouth Bass		377.0		FALSE	
	584	12	9	Pumpkinseed		6.5		FALSE	
	585	12	13	Pumpkinseed	53	2.2		FALSE	
	586	12	15	Pumpkinseed		1.0		FALSE	
	587	12	20	Pumpkinseed		1.5		FALSE	
	588	12	67	Pumpkinseed	61	4.0		FALSE	
	589	9	159	Pumpkinseed		120.0		TRUE	
	590	9	160	Pumpkinseed		185.0		TRUE	
	591	20	200	Pumpkinseed		200.0		TRUE	
	592	21	216	Pumpkinseed		80.0		TRUE	
	593	23	223	Pumpkinseed		210.0		TRUE	
	594	27	225	Pumpkinseed		20.0		TRUE	
	595	24	229	Pumpkinseed				TRUE	
	596	124	835	Pumpkinseed		130.0		TRUE	
	597	203	528	Tadpole Madtom	41	0.5		FALSE	
	598	203	529	Tadpole Madtom		0.2		FALSE	
	599	203	530	Tadpole Madtom		0.2		FALSE	
	600	203	531	Tadpole Madtom		0.3		FALSE	
	601	106	655	Tadpole Madtom		1.3		FALSE	
	602	115	734	Tadpole Madtom		1.0		FALSE	
	603	4	113	Yellow Perch		150.0		TRUE	
	604	4	114	Yellow Perch		170.0		TRUE	
	605	4	115	Yellow Perch		175.0		TRUE	
	606	4	116	Yellow Perch		157.0		TRUE	
	607	4	117	Yellow Perch		160.0		TRUE	
	608	9	173	Yellow Perch		105.0		TRUE	
	609	9	174	Yellow Perch		150.0		TRUE	
πĦ	003	9	1/4	IETTOM LELCII	220	T)0.0		TNUL	

		•	4	V 11 B I	202	222.0		TRUE	
	610	9	175	Yellow Perch				TRUE	
	611	20	202	Yellow Perch	226			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		158.0			
	619	105	704	Yellow Perch		14.1		TRUE	
	620	105	705	Yellow Perch				TRUE	
	621	105	706	Yellow Perch				TRUE	
	622	105	700 707	Yellow Perch		4.0		TRUE	
	623	105	708	Yellow Perch		4.3		TRUE	
	624	105	709	Yellow Perch		4.1		TRUE	
	625	105	710	Yellow Perch		3.4		TRUE	
	626	111	714	Yellow Perch			1118		
##	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				TRUE	
	635	122	825	Yellow Perch		152.0		TRUE	
	636	125	827	Yellow Perch		113.0		TRUE	
	637	125	828	Yellow Perch		4.6	2020	TRUE	
	638	125	829	Yellow Perch		7.1		TRUE	
	639	125	830	Yellow Perch		4.1		TRUE	
	640	125	831	Yellow Perch		2.1		TRUE	
							~A009		
	641	1		Black Crappie			•		
	642	5	135	Black Crappie		NA	00513	TRUE	
	643	5	136	Black Crappie		NA 275 0	00514	TRUE	
	644	5	137	Black Crappie		275.0	00515	TRUE	
	645	16	208	Black Crappie		380.0	g0995	TRUE	
	646	16	209	Black Crappie		260.0	g0994	TRUE	
	647	21	218	Black Crappie		46.0		TRUE	
##	648	29	238	Black Crappie		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		NA	00408	TRUE	
	653	37	319	Black Crappie		NA	y00384	TRUE	
	654	37	334	Black Crappie		NA	y00020	TRUE	
	655	37	335	Black Crappie		NA	y00318	TRUE	
		٠,		Diack Cruppic	,	14/-1	,00510		

```
## 656
          37
                 350
                        Black Crappie 297
                                                NA yunknown
                                                              TRUE
## 657
          37
                 351
                        Black Crappie 290
                                                NA
                                                     y09000
                                                              TRUE
## 658
         101
                 532
                        Black Crappie 320
                                                             TRUE
                                            508.0
                                                       1014
## 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
                        Black Crappie 307
## 662
         102
                 626
                                            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
                 754
## 667
         113
                        Black Crappie 147
                                             37.0
                                                              TRUE
## 668
                 798
                        Black Crappie 188
                                            102.0
         112
                                                             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
                                            352.0
                        Black Crappie 282
                                                       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
#3. display the head, tail and structure of <bio>
headtail(bio)
##
       netID fishID
                           species
                                     tl
                                               tag scale
                                            W
## 1
          12
                  16
                          Bluegill
                                     61
                                          2.9
                                                    FALSE
## 2
          12
                  23
                          Bluegill
                                     66
                                          4.5
                                                    FALSE
## 3
          12
                  30
                          Bluegill
                                     70
                                          5.2
                                                    FALSE
## 674
         110
                 863 Black Crappie 307 415.0 1783
                                                     TRUE
         129
## 675
                 870 Black Crappie 279 344.0 1789
                                                     TRUE
## 676
         129
                 879 Black Crappie 302 397.0 1792
                                                     TRUE
str(bio)
## 'data.frame':
                     676 obs. of 7 variables:
##
    $ netID : int
                     12 12 12 12 12 12 13 13 13 ...
##
    $ fishID : int
                     16 23 30 44 50 65 66 68 69 70 ...
                     "Bluegill" "Bluegill" "Bluegill" ...
##
    $ species: chr
                     61 66 70 38 42 54 27 36 59 39 ...
##
    $ tl
              : int
    $ w
                     2.9 4.5 5.2 0.5 1 2.1 NA 0.5 2 0.5 ...
##
              : num
                     ... ... ... ...
##
    $ tag
              : chr
##
                     FALSE FALSE FALSE FALSE FALSE ...
    $ scale
              : logi
#4. Create an object, <counts>, that counts and lists all the species
```

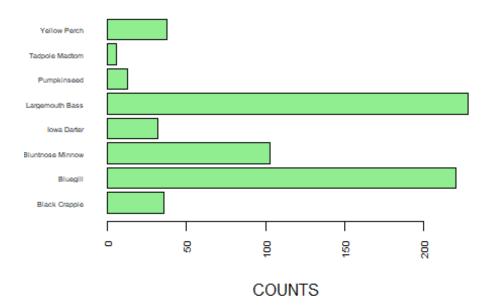
records

```
counts <- table(bio$species)</pre>
counts
##
##
      Black Crappie
                             Bluegill Bluntnose Minnow
                                                             Iowa Darter
                                                    103
##
                                  220
                                                                       32
##
    Largemouth Bass
                          Pumpkinseed
                                        Tadpole Madtom
                                                            Yellow Perch
##
                228
                                   13
                                                                       38
#5. Display just the 8 levels (names) of the species
unique(bio$species)
                           "Bluntnose Minnow" "Iowa Darter"
## [1] "Bluegill"
                                                                   "Large
mouth Bass"
## [5] "Pumpkinseed"
                           "Tadpole Madtom"
                                               "Yellow Perch"
                                                                   "Black
Crappie"
#6. Create a <tmp> object that displays the different species and the
number of record of each species in the dataset. Include this informat
ion in your report.-
tmp <- counts
tmp
##
##
      Black Crappie
                             Bluegill Bluntnose Minnow
                                                             Iowa Darter
##
                                  220
                                                    103
                                                            Yellow Perch
##
    Largemouth Bass
                          Pumpkinseed
                                        Tadpole Madtom
##
                228
                                                                       38
#7. Create a subset, <tmp2>, of just the species variable and display
the first five records
tmp2 <- bio$species</pre>
head(tmp2, n=5)
## [1] "Bluegill" "Bluegill" "Bluegill" "Bluegill" "Bluegill"
#8. Create a table, <w>, of the species variable. Display the class of
W
w <- table(bio$species)</pre>
##
##
      Black Crappie
                             Bluegill Bluntnose Minnow
                                                             Iowa Darter
##
                 36
                                  220
                                                    103
                                                                       32
```

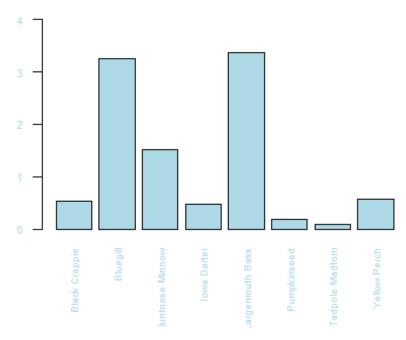
```
Yellow Perch
##
    Largemouth Bass
                          Pumpkinseed
                                        Tadpole Madtom
##
                228
                                   13
                                                      6
                                                                       38
#9. Convert <w> to a data frame named <t> and display the results
t <- as.data.frame(w)
##
                 Var1 Freq
## 1
        Black Crappie
                         36
             Bluegill
## 2
                        220
## 3 Bluntnose Minnow
                        103
## 4
          Iowa Darter
                         32
## 5 Largemouth Bass
                        228
## 6
          Pumpkinseed
                         13
## 7
       Tadpole Madtom
                          6
         Yellow Perch
                         38
## 8
      Extract and display the frequency values from the <t> data frame
t$Freq
## [1]
       36 220 103
                    32 228
                             13
#11. Create a table named <cSpec> from the bio species attribute (vari
able) and confirm that you created a table which displays the number o
f species in the dataset <bio>
cSpec <- table(bio$species)</pre>
cSpec
##
##
      Black Crappie
                             Bluegill Bluntnose Minnow
                                                             Iowa Darter
##
                  36
                                  220
                                                    103
                                                                       32
##
    Largemouth Bass
                          Pumpkinseed
                                        Tadpole Madtom
                                                            Yellow Perch
##
                228
                                                                       38
#12. Create a table named <cSpecPct> that displays the species and per
centage of records for each species. Confirm you created a table class
?prop.table
## starting httpd help server ...
##
    done
cSpecPct <- (table(bio$species)*100)/length(bio$species)</pre>
cSpecPct
```

```
##
##
      Black Crappie
                            Bluegill Bluntnose Minnow
                                                            Iowa Darter
           5.325444
                           32.544379
##
                                            15.236686
                                                               4.733728
                         Pumpkinseed
                                       Tadpole Madtom
##
    Largemouth Bass
                                                           Yellow Perch
##
          33.727811
                            1.923077
                                             0.887574
                                                               5.621302
class(cSpecPct)
## [1] "table"
#13. Convert the table, <cSpecPct>, to a data frame named <u> and conf
irm that <u> is a data frame
u <- as.data.frame(cSpecPct)</pre>
u
##
                 Var1
                           Frea
## 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
         Yellow Perch 5.621302
## 8
class(u)
## [1] "data.frame"
#14. Create a barplot of <cSpec> with the following: titled Fish Coun
t 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
barplot(cSpec, xlab="COUNTS", main = "Fish Count",horiz = TRUE, col="L
ight Green",
        las=2, cex.names = 0.45, cex.axis = 0.60)
```

Fish Count



Fish Relative Frequency



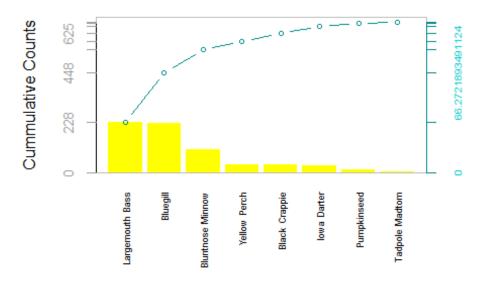
#16. Rearrange the <u> cSpec Pct data frame in descending order of relative frequency. Save the rearranged data frame as the object <d>

```
d <- u[order(-u$Freq),]</pre>
d
##
                  Var1
                            Frea
      Largemouth Bass 33.727811
## 2
             Bluegill 32.544379
## 3 Bluntnose Minnow 15.236686
         Yellow Perch 5.621302
## 8
## 1
        Black Crappie 5.325444
## 4
          Iowa Darter 4.733728
## 6
          Pumpkinseed 1.923077
## 7
       Tadpole Madtom 0.887574
#17. Rename the <d> columns Var 1 to Species, and Freq to RelFreq
colnames(d)[1] <- "Species"</pre>
colnames(d)[2] <- "RelFreq"</pre>
d
##
              Species
                         RelFreq
## 5 Largemouth Bass 33.727811
## 2
             Bluegill 32.544379
```

```
## 3 Bluntnose Minnow 15.236686
         Yellow Perch 5.621302
## 8
## 1
        Black Crappie 5.325444
          Iowa Darter 4.733728
## 4
          Pumpkinseed 1.923077
## 6
## 7
       Tadpole Madtom 0.887574
#18. Add new variables to <d> and call them cumfreq, counts, and cumco
unts
counts
##
##
      Black Crappie
                            Bluegill Bluntnose Minnow
                                                            Iowa Darter
##
                 36
                                 220
                                                                     32
##
    Largemouth Bass
                         Pumpkinseed
                                       Tadpole Madtom
                                                           Yellow Perch
##
                228
                                  13
                                                                     38
                                                     6
tdesc <- t[order(-t$Freq),]
tdesc$Freq
## [1] 228 220 103
                   38 36
                            32
d <- d %>% mutate(cumfreq=cumsum(d$RelFreq), counts=tdesc$Freq, cumcou
nts=cumsum(tdesc$Freq))
d
##
              Species
                        RelFreq
                                  cumfreq counts cumcounts
## 5 Largemouth Bass 33.727811 33.72781
                                              228
                                                        228
## 2
             Bluegill 32.544379 66.27219
                                             220
                                                        448
## 3 Bluntnose Minnow 15.236686 81.50888
                                             103
                                                        551
## 8
         Yellow Perch 5.621302 87.13018
                                               38
                                                        589
## 1
        Black Crappie 5.325444 92.45562
                                               36
                                                        625
## 4
          Iowa Darter 4.733728 97.18935
                                               32
                                                        657
## 6
          Pumpkinseed 1.923077 99.11243
                                               13
                                                        670
       Tadpole Madtom 0.887574 100.00000
                                               6
## 7
                                                        676
#19. Create a parameter variable <def_par> to store parameter variable
def_par <- as.data.frame(names(d))</pre>
#20. Create a barplot, <pc>, with the following specifications:
#d$counts of width 1, spacing of .15
#no boarder
#Axes: F
#Yaxis limit 0,3.05*max
#d$counts na.rm is true
```

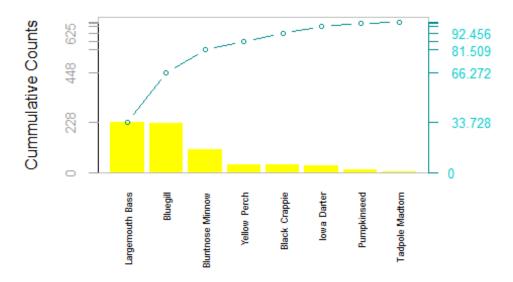
```
#v label is Cummulative Counts
#scale x axis to 70%
#names.arg: d$Species
#Title of the barplot is "Species Pareto"
par(mar=c(8,4,4,4))
pc <- barplot(d$counts, width = 1, space = 0.15, border = NA, axes = F
ylim = c(0,3.05*228), ylab = "Cummulative Counts", cex.axis = 0.7, cex.na
mes = 0.55,
names.arg = d$Species,las=2,col="Yellow", main = "Species Pareto")
#21. Add a cumulative counts line to the <pc> plot with the following:
#Spec line type is b
#Scale plotting text at 70%
#Data values are solid circles with color cyan4
lines(pc, d$cumcounts, type = 'b', cex=0.70, col="cyan4")
#22. Place a grey box around the pareto plot
box(which = "plot", lty = "solid", col="grey")
#23. 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,col.axis="grey62", col.ticks = "grey62", cex.axis=0.80,
at=c(0,d$cumcounts ))
#24. Add axis details on right side of box with the specifications:
#Spec: Side 4
#Tickmarks at cumcounts with labels from 0 to cumfreq with %,
#Axis color of cyan5 and label color of cyan4
#Axis font scaled to 80% of nominal
axis(side = 4, at=c(0,d$cumcounts), labels =c(0,d$cumfreq), col.axis="
cyan3", col="cyan4",
cex.axis=0.60)
```

Species Pareto



```
#25. Display the finished Species Pareto Plot (without the star water
marks). Have your last name on the plot
d$cumfreq <- format(round(d$cumfreq, 3), nsmall = 3)</pre>
d$cumfreq
## [1] " 33.728" " 66.272" " 81.509" " 87.130" " 92.456" " 97.189" " 9
9.112"
## [8] "100.000"
par(mar=c(8,4,4,4))
pc <- barplot(d$counts, width = 1, space = 0.15, border = NA, axes = F
              ylim = c(0,3.05*max(d$counts, na.rm = TRUE)), ylab = "Cumm"
ulative Counts",cex.axis = 0.7,
              names.arg = d$Species,cex.names = 0.55, las=2,col="Yello"
w",main = "Species Pareto : Sawant")
lines(pc, d$cumcounts, type = 'b', cex=0.70, col="cyan4" )
box(which = "plot", lty = "solid", col="grey")
axis(side = 2,col.axis="grey62", col.ticks = "grey62", cex.axis=0.80,
at=c(0,d$cumcounts ))
axis(side = 4, las=1, at=c(0,dscumcounts), labels = c(0,dscumfreq), col.
axis="cyan3", col="cyan4",
     cex.axis=0.80)
```

Species Pareto: Sawant



```
install.packages('tinytex')
## Installing package into 'C:/Users/Parth/OneDrive/Documents/R/win-li
brary/4.1'
## (as 'lib' is unspecified)
## package 'tinytex' successfully unpacked and MD5 sums checked
##
## The downloaded binary packages are in
## C:\Users\Parth\AppData\Local\Temp\RtmpgPNLyt\downloaded_packages
tinytex::install_tinytex()
```