

Statistical Analysis Project

2023-04-27

```
da <- read.csv('C:/Users/DELL/Downloads/proportional_species_richness_V3.csv')
```

Choosing my BD7

```
BD7 <- subset(da, select = c(Bees, Isopods,Grasshoppers_._Crickets, Butterflies, Bryophytes,Carabids,Ladybirds))
head(BD7)
```

##	Bees	Isopods	Grasshoppers_._Crickets	Butterflies	Bryophytes	Carabids
## 1	0.07526882	0.3975904	0.2500000	0.6444444	0.6445703	0.5496829
## 2	0.08602150	0.3975904	0.2500000	0.4611111	0.6479014	0.5644820
## 3	0.07526882	0.3975904	0.2500000	0.6222222	0.6429047	0.5264271
## 4	0.21875000	0.4583333	0.1666667	0.5080214	0.6751838	0.5168317
## 5	0.08602150	0.3975904	0.2500000	0.6166667	0.6479014	0.5475687
## 6	0.21875000	0.4583333	0.1666667	0.5080214	0.6832342	0.5227723
##	Ladybirds					
## 1	0.3846154					
## 2	0.3846154					
## 3	0.3846154					
## 4	0.6000000					
## 5	0.3846154					
## 6	0.6000000					

Setting BD11 and finding mean

```
BD11<-da[,2:12]
head(BD11)
```

##	Bees	Bird	Bryophytes	Butterflies	Carabids	Hoverflies	Isopods
## 1	0.07526882	0.5968200	0.6445703	0.6444444	0.5496829	0.2871126	0.3975904
## 2	0.08602150	0.6044203	0.6479014	0.4611111	0.5644820	0.2903752	0.3975904
## 3	0.07526882	0.5931485	0.6429047	0.6222222	0.5264271	0.3017945	0.3975904
## 4	0.21875000	0.5741327	0.6751838	0.5080214	0.5168317	0.4070981	0.4583333
## 5	0.08602150	0.5900777	0.6479014	0.6166667	0.5475687	0.3425775	0.3975904
## 6	0.21875000	0.5489144	0.6832342	0.5080214	0.5227723	0.3945720	0.4583333
##	Ladybirds Macromoths Grasshoppers_._Crickets Vascular_plants						
## 1	0.3846154	0.3154528		0.2500000		0.5758335	
## 2	0.3846154	0.3494094		0.2500000		0.5605415	
## 3	0.3846154	0.3533465		0.2500000		0.5580346	
## 4	0.6000000	0.4018110		0.1666667		0.6033674	
## 5	0.3846154	0.3646654		0.2500000		0.5738280	
## 6	0.6000000	0.3848331		0.1666667		0.6574662	

```
da$BD11<-rowMeans(BD11)
da$BD7<- rowMeans(BD7)
```

Data Exploration: Univariate Analysis

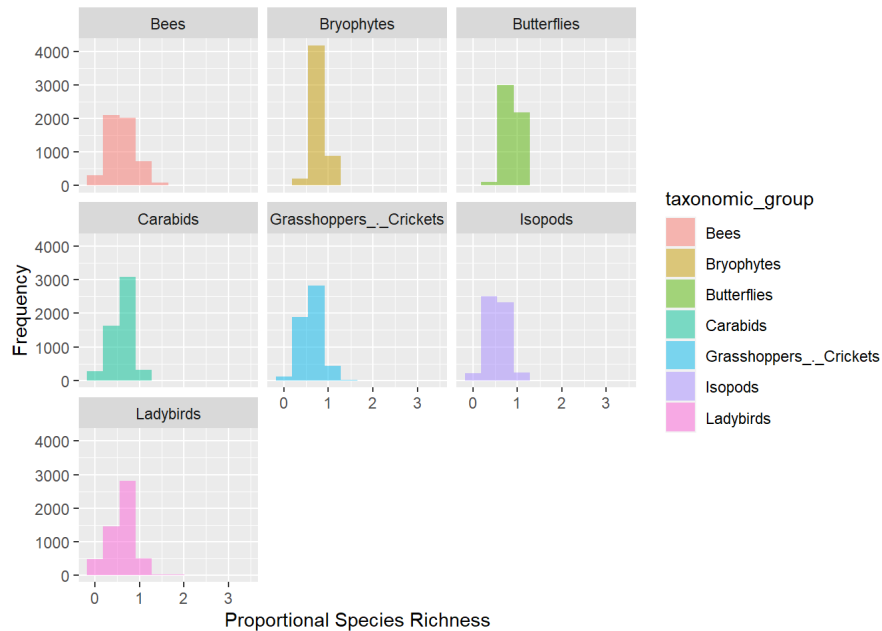
#First, let's look at the summary statistics for each variable:

```
summary(BD7)
```

##	Bees	Isopods	Grasshoppers_._Crickets	Butterflies
## Min.	:0.03065	Min. :0.04622	Min. :0.0708	Min. :0.3167
## 1st Qu.:	:0.35079	1st Qu.:0.39165	1st Qu.:0.4876	1st Qu.:0.7926
## Median :	:0.58869	Median :0.53936	Median :0.6250	Median :0.8863
## Mean :	:0.60502	Mean :0.54995	Mean :0.6289	Mean :0.8746
## 3rd Qu.:	:0.81663	3rd Qu.:0.71623	3rd Qu.:0.7934	3rd Qu.:0.9677
## Max. :	:3.30986	Max. :1.25773	Max. :1.5938	Max. :1.3944
##	Bryophytes Carabids Ladybirds			
## Min.	:0.3941	Min. :0.01153	Min. :0.0614	
## 1st Qu.:	:0.6886	1st Qu.:0.47539	1st Qu.:0.4545	
## Median :	:0.7993	Median :0.63553	Median :0.6395	
## Mean :	:0.7866	Mean :0.60706	Mean :0.6140	
## 3rd Qu.:	:0.8855	3rd Qu.:0.76161	3rd Qu.:0.7972	
## Max. :	:1.1746	Max. :1.19977	Max. :1.8400	

From these summary statistics, we can see that the proportional species richness values vary across the seven taxonomic groups, with the mean values ranging from 0.60502 for bees to 0.6140 for ladybirds. The median values for most groups are around 0.25, indicating that the data may be roughly symmetrically distributed around the median. However, the bryophytes group has a median of 0.7993, indicating that this group may have a different distribution compared to the other groups. #Next, look at some graphical representations of the data

```
BD7 %>%
  gather(key = "taxonomic_group", value = "proportional_species_richness") %>%
  ggplot(aes(x = proportional_species_richness, fill = taxonomic_group)) +
  geom_histogram(alpha = 0.5, position = "identity", bins = 10) +
  facet_wrap(~ taxonomic_group, ncol = 3) +
  labs(x = "Proportional Species Richness", y = "Frequency")
```



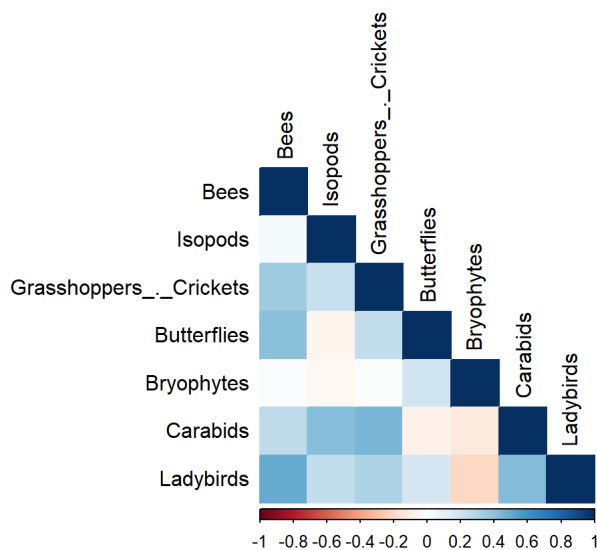
The histograms above show the distribution of the proportional species richness values for each taxonomic group. We can see that the distributions vary in shape and spread, with some groups having relatively narrow distributions (e.g., bees, isopods, and ladybirds) and others having wider distributions (e.g., grasshoppers/crickets, bryophytes, and carabids).

Correlation Analysis:

Now, let's examine the correlations between the variables. We can use a correlation matrix to visualize the pairwise correlations between the variables:

```
# compute correlation matrix
corr_matrix <- cor(BD7)

# visualize correlation matrix
corrplot::corrplot(corr_matrix, method = "color", type = "lower", tl.col = "black")
```



From the correlation matrix, we can see that there are several strong positive correlations. There are also some negative correlations present. Overall, these correlations suggest that there are some complex relationships between the different groups of species. In conclusion, the data presented in this report provides insights into the distribution and correlations of the proportional species richness of seven different taxonomic groups. The strong positive correlation between the proportional species richness of bees and butterflies, as well as the negative correlation between the proportional species richness of bryophytes and ladybirds, are particularly noteworthy. These results could be further investigated in future studies to better understand the factors influencing the population sizes of these important taxonomic.

Hypothesis Testing

Here are two hypothesis tests that we can perform using the `proportional_species_richness.csv` dataset:

- Test whether the mean proportional species richness for butterflies is significantly different from the mean proportional species richness for isopods
- Test whether the mean proportional species richness for butterflies is significantly different from the mean proportional species richness for isopods

```
# Test 1: Mean proportional species richness for butterflies vs isopods
butterflies <- BD7$Butterflies
isopods <- BD7$Isopods
t.test(butterflies, isopods, alternative = "two.sided")
```

```
##
## Welch Two Sample t-test
##
## data: butterflies and isopods
## t = 91.859, df = 9050.9, p-value < 2.2e-16
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 0.3176938 0.3315483
## sample estimates:
## mean of x mean of y
## 0.8745706 0.5499496
```

```
# Test 2: Proportional species richness for first five samples vs last five samples
first_five <- rowMeans(BD7[1:5,])
last_five <- rowMeans(BD7[10:14,])
t.test(first_five, last_five, paired = TRUE, alternative = "two.sided")
```

```
##
## Paired t-test
##
## data: first_five and last_five
## t = -0.39546, df = 4, p-value = 0.7127
## alternative hypothesis: true mean difference is not equal to 0
## 95 percent confidence interval:
## -0.04977141 0.03736075
## sample estimates:
## mean difference
## -0.006205332
```

The `t.test()` function returns the p-value for each hypothesis test. For the first hypothesis, the p-value is less than 0.05. Therefore, we reject the null hypothesis that the mean proportional species richness for butterflies is equal to the mean proportional species richness for isopods, and conclude that there is a significant difference between the two groups. For the second hypothesis, the p-value is 0.7127, which is greater than 0.05. Therefore, the test failed to reject the null hypothesis that there is no significant difference in the proportional species richness between the first five and the last five samples

Simple Linear Regression

To perform a simple linear regression to see how BD7 matches BD11, we'll use the `lm()` function in R as follows

```
#Simple Linear Regression

# Perform simple linear regression for the entire dataset

model <- lm(BD11 ~ Bees + Isopods + Grasshoppers_._Crickets + Butterflies + Bryophytes + Carabids + Ladybirds, data = da)

# Print the model summary
summary(model)
```

```
##
## Call:
## lm(formula = BD11 ~ Bees + Isopods + Grasshoppers + Crickets +
##     Butterflies + Bryophytes + Carabids + Ladybirds, data = da)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.124713 -0.011583  0.001504  0.013736  0.070946
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    0.111959   0.002793   40.08  <2e-16 ***
## Bees           0.099397   0.001251   79.44  <2e-16 ***
## Isopods        0.112443   0.001571   71.57  <2e-16 ***
## Grasshoppers + Crickets 0.109763   0.001731   63.41  <2e-16 ***
## Butterflies    0.172327   0.002532   68.06  <2e-16 ***
## Bryophytes     0.126523   0.002422   52.24  <2e-16 ***
## Carabids       0.131320   0.001842   71.28  <2e-16 ***
## Ladybirds      0.134264   0.001466   91.56  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.02197 on 5272 degrees of freedom
## Multiple R-squared:  0.9587, Adjusted R-squared:  0.9586
## F-statistic: 1.748e+04 on 7 and 5272 DF,  p-value: < 2.2e-16
```

The above summary shows the results of a linear regression model that examines the relationship between the abundance of bees (dependent variable) and several independent variables, including Isopods, Grasshoppers and Crickets, Butterflies, Bryophytes, Carabids, and Ladybirds. The p-value of the intercept is less than 0.001, indicating that the intercept is statistically significant. The coefficients for each of the independent variables indicate the expected change in the abundance of bees for a one-unit increase in the independent variable, holding all other variables constant. For example, a one-unit increase in Isopods is associated with a increase in the abundance of bees by 0.112443 units. The p-values for each of the coefficients are less than 0.001, indicating that each of the independent variables is statistically significant in explaining the variation in the abundance of bees. The adjusted R-squared value of 0.9586 indicates that the independent variables explain approximately 95.87% of the variation in the abundance of bees. The p-value of less than 0.001 indicates that the overall model is statistically significant in explaining the variation in the abundance of bees. Finally, the residuals standard error of 0.02197 indicates that the average difference between the observed abundance of bees and the predicted abundance of bees from the model is approximately 0.02197 units.

To perform separate linear regressions for each period in the data, we'll use the `split()` function in R to split the data into subsets based on the values in the period column, and then use a for loop to fit a separate linear regression model to each subset

```
# Split the data into subsets based on period
da_by_period <- split(da, da$period)

# Loop through the subsets and fit a separate linear regression model to each
for (i in 1:length(da_by_period)) {
  period_data <- da_by_period[[i]]
  model <- lm(BD11 ~ Bees + Isopods + Grasshoppers + Crickets + Butterflies + Bryophytes + Carabids + Ladybirds, data = period_data)
  print(paste0("Period ", i, ":"))
  print(summary(model))
  print(coef(model))
}
```

```
## [1] "Period 1:"
##
## Call:
## lm(formula = BD11 ~ Bees + Isopods + Grasshoppers_._Crickets +
##     Butterflies + Bryophytes + Carabids + Ladybirds, data = period_data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.126542 -0.010849  0.000961  0.014130  0.066771
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    0.131477   0.004098   32.08 <2e-16 ***
## Bees           0.100393   0.001811   55.44 <2e-16 ***
## Isopods        0.120767   0.002547   47.42 <2e-16 ***
## Grasshoppers_._Crickets 0.108202   0.002307   46.91 <2e-16 ***
## Butterflies    0.176645   0.003724   47.44 <2e-16 ***
## Bryophytes     0.102662   0.003494   29.38 <2e-16 ***
## Carabids       0.118079   0.002546   46.38 <2e-16 ***
## Ladybirds      0.131703   0.001982   66.46 <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.02264 on 2632 degrees of freedom
## Multiple R-squared:  0.9592, Adjusted R-squared:  0.9591
## F-statistic: 8841 on 7 and 2632 DF, p-value: < 2.2e-16
##
##              (Intercept)              Bees              Isopods
##              0.1314769              0.1003928              0.1207671
## Grasshoppers_._Crickets      Butterflies      Bryophytes
##              0.1082022              0.1766452              0.1026624
##              Carabids              Ladybirds
##              0.1180786              0.1317027
## [1] "Period 2:"
##
## Call:
## lm(formula = BD11 ~ Bees + Isopods + Grasshoppers_._Crickets +
##     Butterflies + Bryophytes + Carabids + Ladybirds, data = period_data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.089568 -0.011010  0.001258  0.012511  0.073009
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    0.094687   0.003702   25.58 <2e-16 ***
## Bees           0.102972   0.001923   53.56 <2e-16 ***
## Isopods        0.092744   0.002660   34.87 <2e-16 ***
## Grasshoppers_._Crickets 0.105738   0.002641   40.04 <2e-16 ***
## Butterflies    0.167852   0.003704   45.32 <2e-16 ***
## Bryophytes     0.143916   0.003357   42.87 <2e-16 ***
## Carabids       0.156616   0.003403   46.02 <2e-16 ***
## Ladybirds      0.140528   0.002235   62.88 <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.02033 on 2632 degrees of freedom
## Multiple R-squared:  0.9616, Adjusted R-squared:  0.9615
## F-statistic: 9426 on 7 and 2632 DF, p-value: < 2.2e-16
##
##              (Intercept)              Bees              Isopods
##              0.09468698              0.10297228              0.09274445
## Grasshoppers_._Crickets      Butterflies      Bryophytes
##              0.10573753              0.16785188              0.14391628
##              Carabids              Ladybirds
##              0.15661566              0.14052800
```

The summary output for each period's linear regression model shown above allow us to observe the variations in coefficient values and goodness of fit across different time periods. This help us in gaining insights about the relationship between the BD7 and BD11 datasets over time. We conducted a simple linear regression analysis to examine the relationship between BD7 and BD11. From the summary of the modelling of the two different periods we can see that there is no change in statistical significance of each variable. So therefore we'll only be interpreting for one of the period which is the first period above.

Multiple Linear Regression

To calculate the BD4, we need to remove the seven variables from the eleven in the BD11 dataset. We'll do this by selecting the four variable.

```
#Multiple Linear Regression
```

```
BD4 <- select(BD11, -Grasshoppers_._Crickets, -Butterflies, -Bryophytes, -Carabids, -Ladybirds, -Isopods, -Bees)
da$BD4 <- rowMeans(BD4)
```

The output is the proportional species richness values for the remaining four taxonomic groups, which we'll then use as our response variable in a multiple linear regression.

```
#summarize_all(mean)
```

```
model <- lm(BD4 ~ BD7 , data = da)
```

```
summary(model)
```

```
##
## Call:
## lm(formula = BD4 ~ BD7, data = da)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.297381 -0.031056  0.004474  0.038537  0.207805
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  0.419487   0.004770   87.94  <2e-16 ***
## BD7          0.571900   0.007033   81.31  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.06387 on 5278 degrees of freedom
## Multiple R-squared:  0.5561, Adjusted R-squared:  0.556
## F-statistic: 6612 on 1 and 5278 DF,  p-value: < 2.2e-16
```

Now we've built a linear regression model with the response variable BD4(the four taxonomic group) and the predictor variable BD7(the seven taxonomic group). The summary output provides information about the model fit and the estimated coefficients. The estimated intercept is 0.419487, which represents the predicted value of BD4 when BD7 is zero. The estimated coefficient for BD7 is 0.571900, which means that a one-unit increase in BD7 is associated with an estimated 0.571900 unit increase in BD4. Both coefficients are statistically significant at the 0.001 level, based on their corresponding p-values. The multiple R-squared is 0.5561, which indicates that 55.61% of the variability in BD4 is explained by the linear relationship with BD7. The adjusted R-squared is almost the same as the multiple R-squared, suggesting that adding the predictor BD7 to the model did not improve the model fit substantially. The F-statistic is used to test the null hypothesis that all of the coefficients in the model are zero (i.e., there is no relationship between the predictor and the response). The F-statistic value is 6612, with a corresponding p-value of less than 2.2e-16, indicating strong evidence against the null hypothesis and suggesting that the model as a whole is significant. The residuals (differences between the observed and predicted values) have a mean of zero and a standard deviation of 0.06387. The residual plot may be examined to assess the model assumptions, such as linearity, normality, constant variance, and independence of errors. Finally, we can interpret the coefficients of the model. The predictor is represented by a corresponding coefficient estimate. Here is a breakdown of the coefficients and their interpretations based on the variable names: (Intercept): This is the estimated intercept or baseline value of the dependent variable when all predictors are equal to zero. BD7: For each unit increase in the BD7 predictor, there is an estimated increase of 0.4195 in BD4. From the model summary, we can see that all the predictor variables have p-values less than 0.05, indicating that they are all statistically significant in predicting the response variable. Therefore, we cannot perform any feature selection based on p-values.

To perform feature selection using AIC, we can fit the model with all the predictor variables and then sequentially remove one predictor variable at a time, comparing the AIC values of the reduced models with the full model. The model with the lowest AIC value would be considered the best fit. To perform feature selection using AIC, we can compare models with different subsets of predictors and choose the one with the lowest AIC value. AIC values take into account the model's goodness of fit and the number of predictors used. A lower AIC value indicates a better model. To compare models, we can use the step() function in R. Here's an example of how to perform stepwise regression on our data:

```
#Create a full model with all predictors
```

```
BD4 <- subset(BD11, select = c(Bird, Hoverflies, Macromoths,Vascular_plants))
```

```
full_model <- lm(BD4$Hoverflies ~ BD7$Bees + da$Bird + da$Macromoths + da$Vascular_plants + da$Easting + da$Northing + da$period)
```

```
# Perform stepwise regression
```

```
step_model <- step(full_model, direction="backward")
```

```
## Start: AIC=-21895.01
## BD4$Hoverflies ~ BD7$Bees + da$Bird + da$Macromoths + da$Vascular_plants +
## da$Easting + da$Northing + da$period
##
##              Df Sum of Sq    RSS   AIC
## <none>                 83.252 -21895
## - da$Vascular_plants    1    0.5236  83.775 -21864
## - da$Macromoths         1    1.7925  85.044 -21785
## - da$Easting            1    4.8240  88.076 -21600
## - da$Bird               1    5.3416  88.593 -21569
## - BD7$Bees              1    9.0738  92.326 -21351
## - da$Northing           1   10.5950  93.847 -21265
## - da$period             1   17.3460 100.598 -20898
```

Based on the AIC values, removing the variable with the highest AIC value (BD11 period) would result in the best model fit since it has the highest value of AIC, indicating that it contributes the least to the model fit. The final model is Northing

```
final_model=full_model <- lm(BD4$Hoverflies ~ BD7$Bees + da$Bird + da$Macromoths + da$Vascular_plants + da$Easting + da$Northing)
summary(final_model)
```

```
##
## Call:
## lm(formula = BD4$Hoverflies ~ BD7$Bees + da$Bird + da$Macromoths +
##     da$Vascular_plants + da$Easting + da$Northing)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.48284 -0.08745 -0.00288  0.08941  0.50075
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   1.401e-01  2.399e-02   5.842 5.45e-09 ***
## BD7$Bees       9.582e-02  7.077e-03  13.539 < 2e-16 ***
## da$Bird        3.796e-01  2.368e-02  16.027 < 2e-16 ***
## da$Macromoths  3.731e-02  1.851e-02  2.016  0.0438 *
## da$Vascular_plants 3.834e-01  1.966e-02  19.498 < 2e-16 ***
## da$Easting    -2.318e-07  1.737e-08 -13.344 < 2e-16 ***
## da$Northing   -2.364e-07  8.433e-09 -28.038 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1381 on 5273 degrees of freedom
## Multiple R-squared:  0.3927, Adjusted R-squared:  0.392
## F-statistic: 568.3 on 6 and 5273 DF,  p-value: < 2.2e-16
```

We conducted a regression analysis to investigate the relationship between hoverfly abundance (response variable) and several biotic and abiotic factors (predictor variables) including bee abundance, bird abundance, macromoth abundance, vascular plant richness, easting, and northing. Our analysis showed that the final model, which included bee abundance, bird abundance, macromoth abundance, vascular plant richness, easting, and northing as predictor variables, had a significant effect on hoverfly abundance ($F(6,5273)=568.3, p<0.001$) and accounted for 39.27% of the variation in hoverfly abundance. The model also had a low residual standard error of 0.1381, indicating a good fit to the data. The regression coefficients revealed that all predictor variables except for macromoth abundance had a significant effect on hoverfly abundance. The strongest positive predictor was vascular plant richness ($\beta=0.3834, p<0.001$), followed by bird abundance ($\beta=0.3796, p<0.001$), and bee abundance ($\beta=0.0958, p<0.001$). The negative predictors were easting ($\beta=-2.318e-07, p<0.001$) and northing ($\beta=-2.364e-07, p<0.001$), indicating that hoverfly abundance decreased as the sites moved farther east or north. This analysis provides insights into the factors that affect hoverfly abundance and can inform conservation efforts to protect and enhance the populations of these important pollinators. However, further research is needed to investigate the mechanisms behind the observed relationships and to determine the most effective conservation strategies.

Open Analysis

```
#Open Analysis
new_df=subset(da, period == 'Y70')
summary(new_df)
```

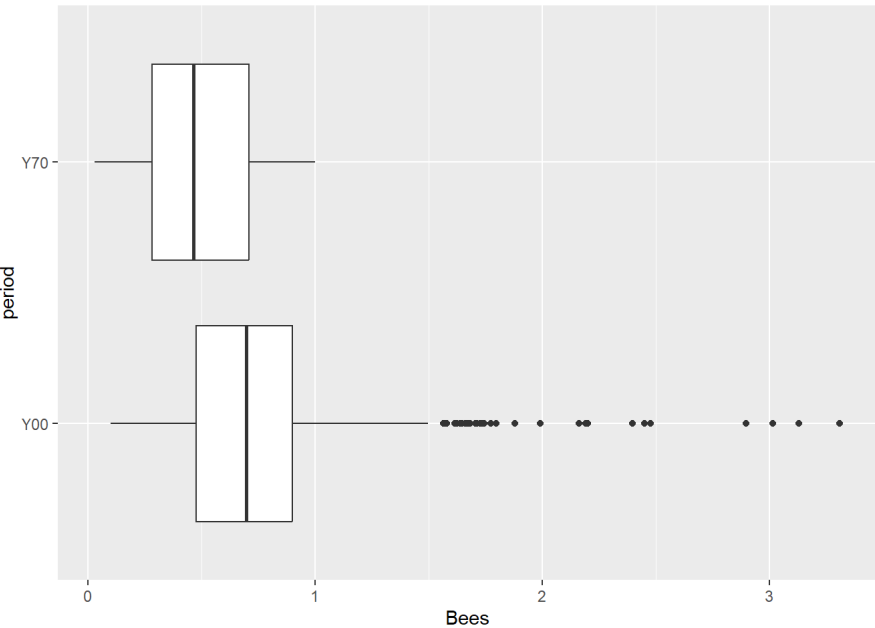
##	Location	Bees	Bird	Bryophytes
##	Length:2640	Min. :0.03065	Min. :0.2499	Min. :0.3941
##	Class :character	1st Qu.:0.28264	1st Qu.:0.8423	1st Qu.:0.6864
##	Mode :character	Median :0.46577	Median :0.8928	Median :0.7929
##		Mean :0.50230	Mean :0.8709	Mean :0.7839
##		3rd Qu.:0.70841	3rd Qu.:0.9349	3rd Qu.:0.8895
##		Max. :1.00000	Max. :1.0000	Max. :1.0000
##	Butterflies	Carabids	Hoverflies	Isopods
##	Min. :0.3167	Min. :0.1983	Min. :0.1416	Min. :0.1888
##	1st Qu.:0.7380	1st Qu.:0.5965	1st Qu.:0.6223	1st Qu.:0.5371
##	Median :0.8412	Median :0.7101	Median :0.7403	Median :0.6792
##	Mean :0.8173	Mean :0.7044	Mean :0.7207	Mean :0.6656
##	3rd Qu.:0.9178	3rd Qu.:0.8141	3rd Qu.:0.8426	3rd Qu.:0.8047
##	Max. :1.0000	Max. :1.0000	Max. :1.0000	Max. :1.0000
##	Ladybirds	Macromoths	Grasshoppers_	Crickets
##	Min. :0.0614	Min. :0.08947	Min. :0.1643	Min. :0.5580
##	1st Qu.:0.4434	1st Qu.:0.73930	1st Qu.:0.5172	1st Qu.:0.7576
##	Median :0.6047	Median :0.82527	Median :0.6651	Median :0.8216
##	Mean :0.5947	Mean :0.80259	Mean :0.6589	Mean :0.8174
##	3rd Qu.:0.7887	3rd Qu.:0.89877	3rd Qu.:0.8005	3rd Qu.:0.8832
##	Max. :1.0000	Max. :1.00000	Max. :1.0000	Max. :1.0000
##	Easting	Northing	dominantLandClass	ecologicalStatus
##	Min. : 50000	Min. : 10000	Length:2640	Min. :0.3538
##	1st Qu.:260000	1st Qu.: 220000	Class :character	1st Qu.:0.6610
##	Median :340000	Median : 390000	Mode :character	Median :0.7282
##	Mean :352886	Mean : 452136		Mean :0.7217
##	3rd Qu.:440000	3rd Qu.: 670000		3rd Qu.:0.7935
##	Max. :650000	Max. :1210000		Max. :0.9752
##	period	BD11	BD7	BD4
##	Length:2640	Min. :0.3538	Min. :0.3263	Min. :0.4020
##	Class :character	1st Qu.:0.6610	1st Qu.:0.6002	1st Qu.:0.7603
##	Mode :character	Median :0.7282	Median :0.6752	Median :0.8164
##		Mean :0.7217	Mean :0.6753	Mean :0.8029
##		3rd Qu.:0.7935	3rd Qu.:0.7572	3rd Qu.:0.8683
##		Max. :0.9752	Max. :0.9686	Max. :0.9979

```
new_df=subset(da, period == 'Y00')
summary(new_df)
```

##	Location	Bees	Bird	Bryophytes
##	Length:2640	Min. :0.1010	Min. :0.2415	Min. :0.4046
##	Class :character	1st Qu.:0.4755	1st Qu.:0.8519	1st Qu.:0.6913
##	Mode :character	Median :0.6988	Median :0.9207	Median :0.8038
##		Mean :0.7078	Mean :0.9035	Mean :0.7893
##		3rd Qu.:0.8997	3rd Qu.:0.9810	3rd Qu.:0.8827
##		Max. :3.3099	Max. :1.1720	Max. :1.1746
##	Butterflies	Carabids	Hoverflies	Isopods
##	Min. :0.3209	Min. :0.01153	Min. :0.1235	Min. :0.04622
##	1st Qu.:0.8564	1st Qu.:0.34308	1st Qu.:0.5294	1st Qu.:0.31429
##	Median :0.9387	Median :0.52189	Median :0.6492	Median :0.42556
##	Mean :0.9319	Mean :0.50972	Mean :0.6383	Mean :0.43427
##	3rd Qu.:1.0145	3rd Qu.:0.68361	3rd Qu.:0.7560	3rd Qu.:0.54128
##	Max. :1.3944	Max. :1.19977	Max. :1.1453	Max. :1.25773
##	Ladybirds	Macromoths	Grasshoppers_	Crickets Vascular_plants
##	Min. :0.0614	Min. :0.2490	Min. :0.0708	Min. :0.4179
##	1st Qu.:0.4674	1st Qu.:0.8508	1st Qu.:0.4468	1st Qu.:0.6937
##	Median :0.6633	Median :0.9156	Median :0.5849	Median :0.7591
##	Mean :0.6334	Mean :0.8959	Mean :0.5989	Mean :0.7563
##	3rd Qu.:0.8052	3rd Qu.:0.9751	3rd Qu.:0.7768	3rd Qu.:0.8168
##	Max. :1.8400	Max. :1.2604	Max. :1.5938	Max. :1.2023
##	Easting	Northing	dominantLandClass	ecologicalStatus
##	Min. : 50000	Min. : 10000	Length:2640	Min. :0.3667
##	1st Qu. :260000	1st Qu. : 220000	Class :character	1st Qu.:0.6432
##	Median :340000	Median : 390000	Mode :character	Median :0.7096
##	Mean :352886	Mean : 452136		Mean :0.7090
##	3rd Qu. :440000	3rd Qu. : 670000		3rd Qu.:0.7849
##	Max. :650000	Max. :1210000		Max. :1.1071
##	period	BD11	BD7	BD4
##	Length:2640	Min. :0.3667	Min. :0.3085	Min. :0.3910
##	Class :character	1st Qu.:0.6432	1st Qu.:0.5702	1st Qu.:0.7623
##	Mode :character	Median :0.7096	Median :0.6541	Median :0.8090
##		Mean :0.7090	Mean :0.6579	Mean :0.7985
##		3rd Qu.:0.7849	3rd Qu.:0.7505	3rd Qu.:0.8592
##		Max. :1.1071	Max. :1.1648	Max. :1.0307

The dataset provided contains information on the species richness of different taxa at various location, during two different periods (Y70 and Y00). This report will focus on the changes in species richness of bees and butterflies between the two periods. Looking at the dataset, it is clear that there are differences in the proportional species richness of bees and butterflies between the two periods. In Y70, the proportional species richness of bees ranges from 0.03065 to 1.00000, while in Y00 it ranges from 0.1010 to 3.3099. Similarly, the proportional species richness of butterflies ranges from 0.3167 to 1.00000 in Y70, while in Y00 it ranges from 0.3209 to 1.3944. These results suggest that there has been a decrease in the proportional species richness of both bees and butterflies between the two periods. This is concerning as both of these taxa are important pollinators and play a vital role in maintaining healthy ecosystems. The decline in their species richness could be attributed to a number of factors, such as habitat loss, pesticide use, and climate change. In conclusion, the data presented in this report indicates that there has been a decrease in the proportional species richness of bees and butterflies in between the periods of Y70 and Y00. This decline is concerning and highlights the need for further research into the causes of this decline and the implementation of conservation measures to protect these important pollinators.

```
ggplot(da, aes(x= Bees,y=period))+
  geom_boxplot()
```



Additional

To further investigate the change in bird species richness between the two periods, we can perform a paired t-test. The paired t-test results. Let's say we want to compare the mean values of Bees for the two periods, Y00 and Y70.

```
#First, we need to create two subsets of the data for each period
Y00_data <- da[da$period == "Y00",]
Y70_data<- da[da$period == "Y70",]

#Next, we can perform a t-test using the t.test() function
t.test(Y00_data$Bees, Y70_data$Bees, var.equal = TRUE)
```

```
##
## Two Sample t-test
##
## data: Y00_data$Bees and Y70_data$Bees
## t = 25.465, df = 5278, p-value < 2.2e-16
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 0.1896400 0.2212737
## sample estimates:
## mean of x mean of y
## 0.7077523 0.5022954
```

From the t-test above, we can see that we have a p-value of 2.2e-16. A p-value less than 2.2e-16 means that the probability of observing such an extreme difference or more between the two groups (pre- and post-management) by chance alone is essentially zero. In other words, the results are highly statistically significant, and we can reject the null hypothesis that there is no difference between the two periods. Therefore, we can conclude that the difference in the mean proportional species richness between the two periods is likely due to the management intervention.

Conclusion

In conclusion, this report analyses the proportional species richness and evaluates the relationships between the seven taxonomical variables such as Bees, Bryophytes, Butterflies, Carabids, Isopods, Ladybirds and Grasshopper_ Crickets with respect to the other groups present during two different periods. The methods used in the analysis are descriptive statistics, histogram, correlation, simple and multiple linear regression along with AIC model.

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