# Biodiversity\_Assignment

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Setting up our packages

### Question

How does percent-cover of upright and crust seaweeds affect intertidal invertebrate diversity?

## Hypothesis

If an area has a higher percent cover of upright seaweeds, then intertidal diversity will be higher because they provide microhabitats and are recruiters.

# Data Collection and Data Type

At Scott's Bay, we placed two transects parallel to the shoreline with one wave exposed and the other wave-sheltered. On each transect, we surveyed three locations using 3-4 quadrats per location to record the following: percent cover of upright algae, percent cover of crustose algae, and the species abundance of invertebrates (down to the lowest taxa possible).

### **Analysis Summary**

We calculated the species richness and diversity (using the Shannon-Wiener index) of quadrates compared to percent cover of upright algae and crustose algae. We then performed a simple linear regression to see if percent cover of crustose or upright algae has a significant effect on diversity and richness.

### Results

#### **Calculations**

Seaweed percentages

```
seaweeds = perc %>%
    group_by(vertical_transect, quadrat) %>%
mutate(unique_ID = as.numeric( paste0(vertical_transect, ".", quadrat))) %>%
dplyr::select(c(unique_ID, quadrat, per_upright, per_crust)) %>%
arrange(unique_ID)
```

## Adding missing grouping variables: 'vertical\_transect'

```
seaweeds_clean = data.frame(seaweeds[,-c(1,3)])
# setting up our dataframe so that we can use matrify to arrange
#our data in long form and then calculate diversity with vegan
cleaned_biodiversity = biodiversity %>% # cleaning data to use matrify
   drop_na(n) %>%
   group_by(vertical_transect, quadrat) %>%
  mutate(unique_ID = as.numeric( pasteO(vertical_transect, ".", quadrat))) %>%
  dplyr::select(unique_ID, species, n)
## Adding missing grouping variables: 'vertical_transect', 'quadrat'
cleaned biodiversity = data.frame(cleaned biodiversity[,-c(1:2)]) # removing extra columns
# putting it into a matrix format
biodiversity_matrix <- matrify(cleaned_biodiversity) # using matrify
Calculating diversity indices and linking it to seaweed cover
ShannonIndex <- data.frame(</pre>
  diversity(biodiversity_matrix, index = "shannon"))
head(ShannonIndex)
       diversity.biodiversity_matrix..index....shannon..
##
## 1.1
                                                0.7595474
## 1.2
                                                0.6869616
## 1.3
                                                0.7529116
## 1.4
                                                0.0000000
## 2.1
                                                1.0690737
## 2.2
                                                0.8184438
ShannonIndex = ShannonIndex %>%
  rename("ShannonIndex" = diversity.biodiversity matrix..index....shannon...)
richness <- biodiversity %>% # calculating richness per quadrat
  drop_na(n) %>%
  group by (vertical transect, quadrat) %>%
  summarize(rich = length(unique(species)))    %>%
  mutate(unique_ID = as.numeric( pasteO(vertical_transect, ".", quadrat)))
## 'summarise()' has grouped output by 'vertical_transect'. You can override using the '.groups' argume
Data = cbind(ShannonIndex, seaweeds_clean, richness[,3]) # comining all our data
Now, we plot the relationship between diversity and seaweed cover using ggplot and geom_point
# visualizing upright percent cover and diversity
```

figure1<-ggplot(data = Data, aes(x=per\_upright, y =ShannonIndex))+</pre>

geom\_point()+

```
geom_smooth(method=lm, se=FALSE)+
  labs(x="Percent cover of upright algae (%)", y="Shannon-Wiener index")
# visualizing crust percent cover and diversity
figure2<-ggplot(data = Data, aes(x=per_crust, y =ShannonIndex))+</pre>
  geom_point()+
  geom_smooth(method=lm, se=FALSE)+
  labs(x="Percent cover of crustose algae (%)", y="Shannon-Wiener index")
# visualizing upright percent cover and richness
figure3<-ggplot(data = Data, aes(x=per_upright, y =rich))+</pre>
  geom_point()+
  geom_smooth(method=lm, se=FALSE)+
  labs(x="Percent cover of upright algae", y="Richness")
# visualizing crust percent cover and richness
figure4 <- ggplot(data = Data, aes(x=per_crust, y =rich))+</pre>
  geom_point()+
  geom_smooth(method=lm, se=FALSE)+
  labs(x="Percent cover of crustose algae", y="Richness")
figure1
```

## 'geom\_smooth()' using formula 'y ~ x'

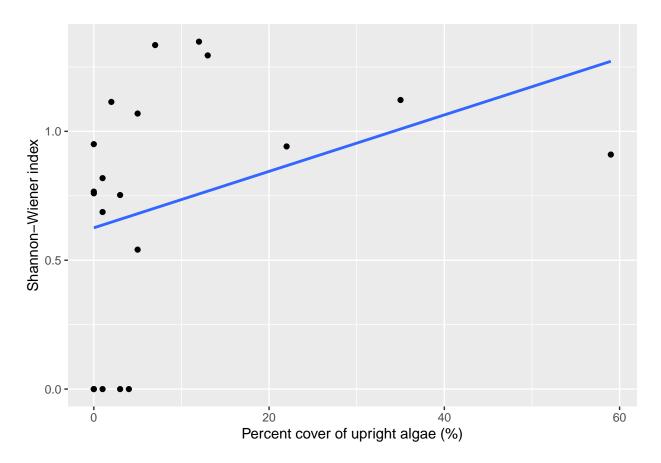


Figure 1. The impact of percent cover of upright algae on the diversity of marine invertebrates (as calculated by the Shannon-Wiener Index)

## figure2

## 'geom\_smooth()' using formula 'y ~ x'

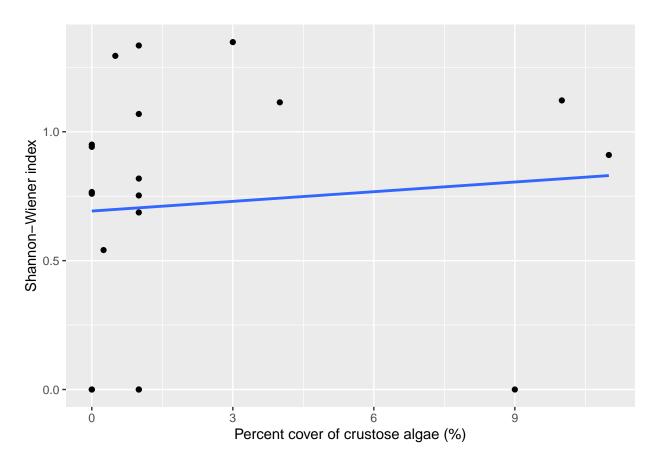


Figure 2. The impact of percent cover of crustose algae on the diversity of marine invertebrates (as calculated by the Shannon-Wiener Index)

# figure3

## 'geom\_smooth()' using formula 'y ~ x'

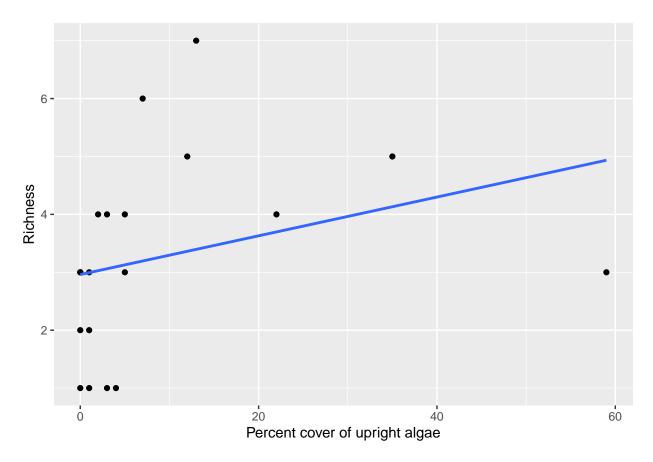


Figure 3. The impact of percent cover of crustose algae on the richness of marine invertebrates (as calculated by the Shannon-Wiener Index)

# figure4

## 'geom\_smooth()' using formula 'y ~ x'

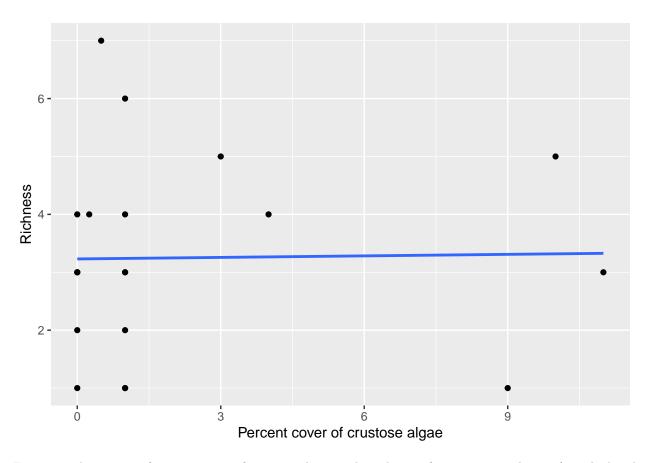


Figure 4. The impact of percent cover of crustose algae on the richness of marine invertebrates (as calculated by the Shannon-Wiener Index)

### Linear Regression

Using regression to examine the relationships between seaweed % cover and shannon weiner diversity We used poission distribution because it's more appropriate for data that's never negative

```
# simple linear regression for upright and diversity
upright_diversity = lm(ShannonIndex ~ per_upright, data = Data)
summary(upright_diversity) # not significant
```

```
##
## Call:
## lm(formula = ShannonIndex ~ per_upright, data = Data)
##
## Residuals:
##
                1Q Median
                                 3Q
                                        Max
   -0.6695 -0.4279 0.1035
                            0.3406 0.6323
##
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) 0.625711
                           0.120190
                                      5.206 5.96e-05 ***
## per_upright 0.010950
                          0.007155
                                      1.530
                                               0.143
## ---
```

```
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4608 on 18 degrees of freedom
## Multiple R-squared: 0.1151, Adjusted R-squared: 0.06596
## F-statistic: 2.342 on 1 and 18 DF, p-value: 0.1433
# simple linear regression for crust and diversity
crust_diversity = lm(ShannonIndex ~ per_crust, data = Data)
summary(crust_diversity) # not significant
##
## Call:
## lm(formula = ShannonIndex ~ per_crust, data = Data)
## Residuals:
##
       Min
                 1Q
                     Median
                                   3Q
## -0.80504 -0.28904 0.07679 0.31905 0.62975
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 0.69243
                          0.13030
                                    5.314 4.73e-05 ***
## per_crust
               0.01251
                          0.03187
                                    0.393
                                             0.699
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 0.4877 on 18 degrees of freedom
## Multiple R-squared: 0.008489, Adjusted R-squared:
## F-statistic: 0.1541 on 1 and 18 DF, p-value: 0.6993
# simple linear regression for upright and richness
upright_rich = lm(rich ~ per_upright, data = Data)
summary(upright_rich) # not significant
##
## lm(formula = rich ~ per_upright, data = Data)
##
## Residuals:
      Min
               1Q Median
                               3Q
                                      Max
## -2.0944 -1.2292 0.0394 0.8889 3.6044
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 2.96055
                          0.43086
                                  6.871 1.99e-06 ***
## per_upright 0.03346
                          0.02565
                                    1.305
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 1.652 on 18 degrees of freedom
## Multiple R-squared: 0.08638,
                                  Adjusted R-squared: 0.03562
## F-statistic: 1.702 on 1 and 18 DF, p-value: 0.2085
```

```
# simple linear regression for crust and richness
crust_rich = lm(rich ~ per_crust, data = Data)
summary(crust_rich) # not significant
##
## Call:
## lm(formula = rich ~ per_crust, data = Data)
##
## Residuals:
##
      Min
               1Q Median
                               3Q
                                      Max
## -2.3096 -1.2325 -0.2303 0.7681 3.7653
##
## Coefficients:
##
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 3.230294
                         0.461596
                                     6.998 1.56e-06 ***
              0.008807
                          0.112902
                                     0.078
## per_crust
                                             0.939
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 1.728 on 18 degrees of freedom
## Multiple R-squared: 0.0003379, Adjusted R-squared: -0.0552
## F-statistic: 0.006085 on 1 and 18 DF, p-value: 0.9387
# we have no significant results (teehee) but that's ok! science
```

Table 1: Table 1. Linear regression results of the relationship between algae morphology and both Shannon-Wiener Index and Richness

| Analysis             | Algae.Type | P.Value | Adjusted.R.squared |
|----------------------|------------|---------|--------------------|
| Shannon-Wiener Index | upright    | 0.1433  | 0.06596            |
| Shannon-Wiener Index | crustose   | 0.6993  | -0.0466            |
| Richness             | upright    | 0.2085  | 0.03562            |
| Richness             | crustose   | 0.9387  | -0.0552            |

### Conclusions

As none of the p-values are less than 0.05, there is no significant relationship between algae type and both diversity and richness.