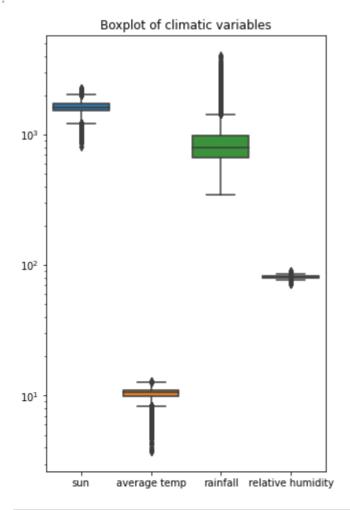
```
In []:
         import pandas as pd
         from sklearn.linear model import LinearRegression, LogisticRegression
         from sklearn.model selection import train test split, GridSearchCV
         from sklearn import metrics
         from sklearn import tree
         from sklearn.ensemble import ExtraTreesRegressor
         from sklearn.neighbors import KNeighborsClassifier
         from sklearn.preprocessing import StandardScaler, MinMaxScaler, RobustScale
         import matplotlib.pyplot as plt
         import seaborn as sns
         import numpy as np
         from imblearn import under sampling
In [ ]:
         #loading intermediate dataset ukbms and haduk, created in HADUK intermedia
         df = pd.read csv('BioD year site latlong.csv')
         df.drop(columns=df.columns[0], inplace = True)
In []:
         df1 climate = df.drop(columns=['Simpsons Index','SPECIES RICHNESS','TOTAL S
         df1 climate.describe()
         df2_clim = df.drop(columns=['Site Number', 'YEAR', 'Easting','Northing','la
         fig,ax = plt.subplots(figsize=(5,8))
         bxplot = sns.boxplot(data=df1 climate)
         bxplot.set yscale('log')
         plt.title('Boxplot of climatic variables')
```

## Out[]: Text(0.5, 1.0, 'Boxplot of climatic variables')

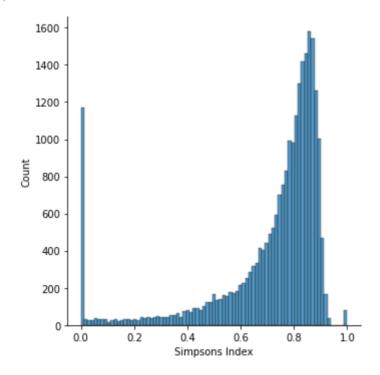


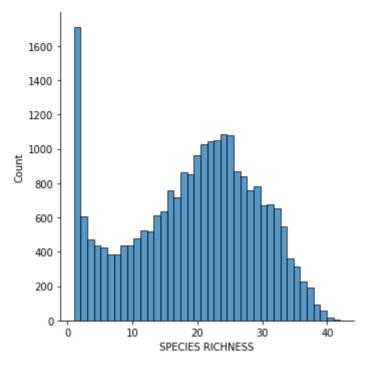
()::±	- 1	=
uu L	- 1	

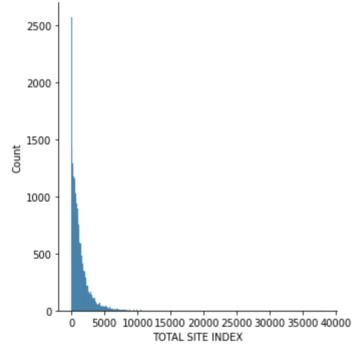
rel hun	rainfall	average temp	sun	TOTAL SITE INDEX	SPECIES RICHNESS	
24558.00	24558.000000	24558.000000	24558.000000	24558.000000	24558.000000	count
81.34	865.006929	10.396600	1613.894576	1242.418682	19.234425	mean
1.86	303.500379	0.968519	174.682971	1444.173563	9.719643	std
72.32	346.315933	3.743084	806.658881	2.000000	1.000000	min
80.3′	668.173773	9.918850	1519.217458	332.000000	12.000000	25%
81.41	801.808547	10.573544	1616.506798	839.000000	21.000000	50%
82.44	976.247667	11.021346	1722.647021	1639.000000	27.000000	75%
90.01	4071.312595	12.839887	2244.008016	38246.000000	42.000000	max

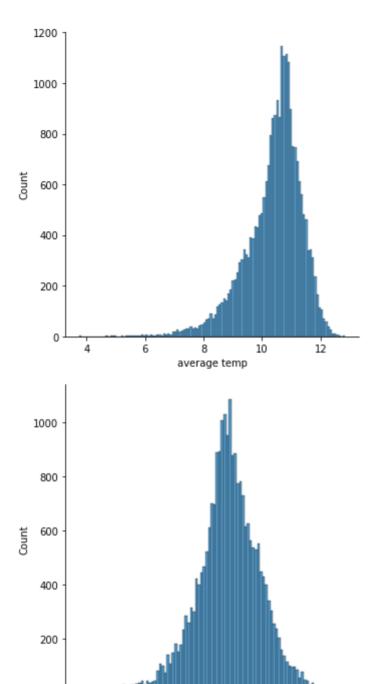
```
In []:
    sns.displot(df['SPECIES RICHNESS'])
    sns.displot(df['TOTAL SITE INDEX'])
    sns.displot(df['average temp'])
    sns.displot(df['sun'])
    sns.displot(df['rainfall'])
    sns.displot(df['relative humidity'])
```

Out[]: <seaborn.axisgrid.FacetGrid at 0x7fc28e0528b0>



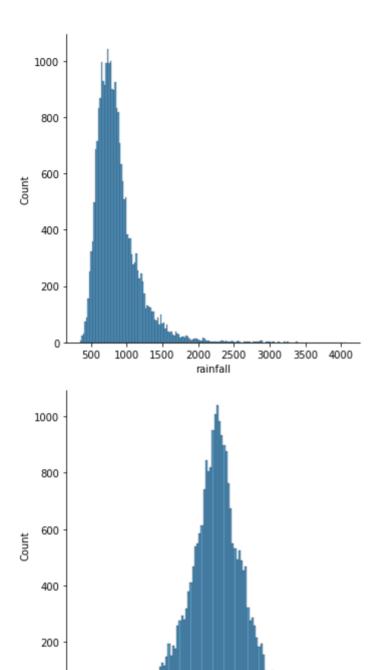






0 1

sun



0

72.5

75.0

77.5

80.0

Multiple Linear Regression, carried out on climatic variables (x) and species richness (y):

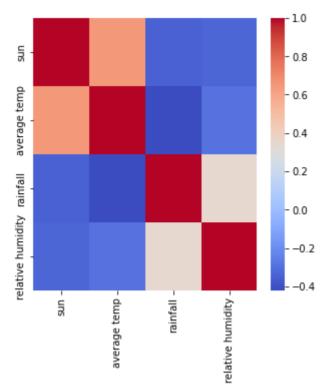
87.5

90.0

85.0

82.5 relative humidity

```
In [ ]:
         x = df[['sun','average temp','rainfall','relative humidity']]
         y= df[['SPECIES RICHNESS']]
         x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.3, random_
In [ ]:
         corr = df1_climate.corr()
         plt.figure(figsize=(5,5))
         sns.heatmap(corr, cmap='coolwarm')
        <AxesSubplot:>
Out[]:
```



```
In [ ]:
         x train.drop(columns=['sun'],inplace=True)
         x test.drop(columns=['sun'],inplace=True)
In []:
         #training multiple linear regression model
         model = LinearRegression()
         model.fit(x_train, y_train)
         print(model.coef )
         print(model.intercept_)
        [[ 0.96703309 -0.00211092 -0.45044449]]
        [47.625016]
In [ ]:
         preds SR = model.predict(x test)
In []:
         #plotting residuals
         residuals = y_test-preds_SR
         plt.scatter(preds_SR,residuals,5)
         plt.ylabel('Residuals')
         plt.xlabel('Predicted values')
         plt.title('Residuals plot')
         plt.show()
```

```
Residuals plot

20 - 10 - -10 - -20 - -10 12 14 16 18 20 22 24 26 Predicted values
```

```
In [ ]:
         mse = metrics.mean_squared_error(y_test,preds_SR)
         r2 = metrics.r2 score(y test,preds SR)
         print(mse)
         print(np.sqrt(mse))
         print(r2)
        91.9585298731896
        9.589501023160151
        0.039946953528546936
In [ ]:
         #Checking diversity
         x = df[['average temp', 'rainfall', 'relative humidity']]
         y= df[['Simpsons Index']]
         x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.3, random_
         model1 = LinearRegression()
         model1.fit(x_train, y_train)
         print(model1.coef )
         print(model1.intercept )
         preds Div = model1.predict(x test)
         mse = metrics.mean squared error(y test,preds Div)
         r2 = metrics.r2 score(y test,preds Div)
         print('mse: ' + str(mse))
         print('rmse: ' + str(np.sqrt(mse)))
         print('r2: ' + str(r2))
        [[ 9.50157468e-03 -7.94726070e-05 -8.92072927e-03]]
        [1.40639564]
        mse: 0.04964239506285567
        rmse: 0.22280573390928626
        r2: 0.03478576404689471
       Decision tree classifier:
```

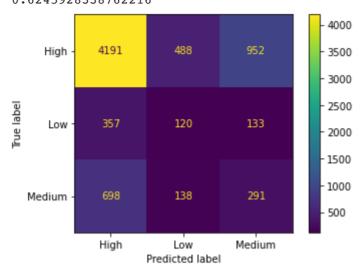
Decision tree classifier.

```
In []:
    #multiclass classifcation decision tree classifier
    x_c = df[['sun','average temp','rainfall','relative humidity']]
    y_c = df[['Diversity']]
    x_ctrain,x_ctest,y_ctrain,y_ctest = train_test_split(x_c,y_c,test_size=0.3,
    dtree_model = tree.DecisionTreeClassifier(class_weight='balanced')
    dtree_model.fit(x_ctrain,y_ctrain)
    dtree_preds = dtree_model.predict(x_ctest)

cm = metrics.confusion_matrix(y_ctest,dtree_preds)
    acc = metrics.accuracy_score(y_ctest, dtree_preds)
```

```
metrics.ConfusionMatrixDisplay.from_predictions(y_ctest,dtree_preds)
print(metrics.fl_score(y_ctest,dtree_preds, average='weighted'))
print(acc)
```

```
0.6391638834523053
0.6245928338762216
```



```
In [ ]:
         #hyperparameter tuning
         parameters={"splitter":["best", "random"],
                     "max_depth" : [1,3,5,7,9,11],
                    "min_samples_leaf":[1,3,5,7,9],
                    "min weight fraction leaf":[0,0.1,0.2,0.3,0.4],
                    "max features":[1,"log2","sqrt",None],
                    "max leaf nodes":[None, 10, 30, 50, 70, 90] }
In [ ]:
         custom f1 = metrics.make scorer(
             metrics.fl score, greater is better=True, average="weighted", labels=["]
           )
In []:
         #d tree classifier tunin
         ctree_tuning = tree.DecisionTreeClassifier(class_weight='balanced')
         c_tuning_model=GridSearchCV(ctree_tuning,param_grid=parameters,scoring=cust
In [ ]:
         c tuning model.fit(x ctrain,y ctrain)
Out[]:
                      GridSearchCV
         ▶ estimator: DecisionTreeClassifier
               ▶ DecisionTreeClassifier
```

```
In []: print(c_tuning_model.best_params_)
    print(c_tuning_model.best_score_)

{'max_depth': 3, 'max_features': 'sqrt', 'max_leaf_nodes': 30, 'min_samples_
    leaf': 9, 'min_weight_fraction_leaf': 0.1, 'splitter': 'best'}
    0.23916131013041977
```

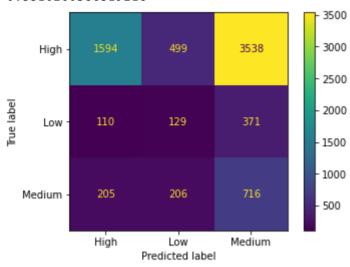
```
In [ ]: tuned_ctree = tree.DecisionTreeClassifier(class_weight='balanced', max_depth
```

```
tuned_ctree.fit(x_ctrain,y_ctrain)
tuned_ctree_preds = tuned_ctree.predict(x_ctest)

acc = metrics.accuracy_score(y_ctest, tuned_ctree_preds)

metrics.ConfusionMatrixDisplay.from_predictions(y_ctest,tuned_ctree_preds)
print(acc)
```

## 0.3310260586319218

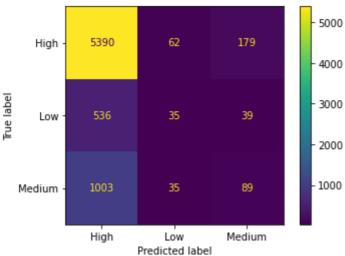


## KNN CLASSIFICATION - SCALED X VARIABLES:

```
In [ ]:
         #scaled knn
         scaler = StandardScaler()
         df scale = df.copy()
         features = [['sun','average temp','rainfall','relative humidity']]
         for feature in features:
             df_scale[feature] = scaler.fit_transform(df_scale[feature])
         x scaled = df scale[['sun','average temp','rainfall','relative humidity']]
         y s = df scale[['Diversity']]
         x_strain,x_stest,y_strain,y_stest = train_test_split(x_scaled,y_s,test_size
         knn scaled model = KNeighborsClassifier()
         knn_scaled_model.fit(x_strain,y_strain)
         knn scaled preds = knn scaled model.predict(x stest)
         acc_scaled = metrics.accuracy_score(y_stest,knn_scaled_preds)
         metrics.ConfusionMatrixDisplay.from predictions(y stest,knn scaled preds)
         print(metrics.fl_score(y_stest,knn_scaled_preds, average = 'weighted'))
         print(acc_scaled)
```

/Library/Frameworks/Python.framework/Versions/3.9/lib/python3.9/site-package s/sklearn/neighbors/\_classification.py:207: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n\_samples,), for example using ravel().

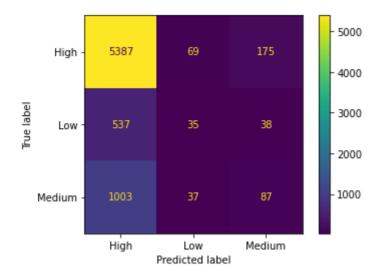
```
return self._fit(X, y)
0.682738313321154
0.748371335504886
```



return self.\_fit(X, y)

0.7476927252985885 0.6820484131213636

```
In [ ]:
         metrics.fl score(y stest,knn scaled preds, average='weighted')
        0.682738313321154
Out[ ]:
In [ ]:
         #knn hyperparameter tuning
         n neighbors = [5,10,15,20,25,30]
         leaf size= list(range(1,25))
         p=[1,2]
         knn_hyperparameters = dict(n_neighbors=n_neighbors,leaf_size=leaf_size,p=p)
         knn tuned = KNeighborsClassifier()
         knn_tuned_model = GridSearchCV(knn_tuned,knn_hyperparameters, scoring=custo
In []:
         knn_tuned_model.fit(x_strain,y_strain.values.ravel())
         print('best leaf size:', knn tuned model.best estimator .get params()['leaf
         print('best p:', knn_tuned_model.best_estimator_.get_params()['p'])
         print('best n neighbors:', knn tuned model.best estimator .get params()['n
        best leaf size: 1
        best p: 2
        best n_neighbors: 5
In []:
         tuned knn model = KNeighborsClassifier(leaf size=1,p=2,n neighbors=5)
         tuned knn model.fit(x strain,y strain)
         tuned_knn_preds = tuned_knn_model.predict(x_stest)
         acc tuned knn = metrics.accuracy score(y stest,tuned knn preds)
         metrics.ConfusionMatrixDisplay.from predictions(y stest, tuned knn preds)
         print(acc tuned knn)
         print(metrics.fl_score(y_stest,tuned_knn_preds, average='weighted'))
        /Library/Frameworks/Python.framework/Versions/3.9/lib/python3.9/site-package
        s/sklearn/neighbors/_classification.py:207: DataConversionWarning: A column-
        vector y was passed when a 1d array was expected. Please change the shape of
        y to (n_samples,), for example using ravel().
```



Condensed Neareset Neighbor Undersamling:

```
In [ ]: undersample = under_sampling.CondensedNearestNeighbour()
    x_under,y_under = undersample.fit_resample(x_strain,y_strain)

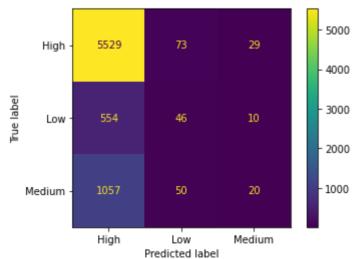
In [ ]: cnn_knn = KNeighborsClassifier(leaf_size=1,p=2,n_neighbors=30)
    cnn_knn.fit(x_under,y_under)
    cnn_knn_preds = cnn_knn.predict(x_stest)

acc_knn_cnn = metrics.accuracy_score(y_stest,cnn_knn_preds)
    metrics.ConfusionMatrixDisplay.from_predictions(y_stest,cnn_knn_preds)
    print(metrics.fl_score(y_stest,cnn_knn_preds, average = 'weighted'))
    print(acc_knn_cnn)
```

/Library/Frameworks/Python.framework/Versions/3.9/lib/python3.9/site-package s/sklearn/neighbors/\_classification.py:207: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n\_samples,), for example using ravel().

```
return self._fit(X, y)
0.6766766834716903
```

0.7593648208469055



```
In []:
    rus_undersample = under_sampling.RandomUnderSampler(random_state=44,replace
    x_rus,y_rus = undersample.fit_resample(x_strain,y_strain)
```

```
In []: rus_knn = KNeighborsClassifier(leaf_size=1,p=2,n_neighbors=5)
    rus_knn.fit(x_rus,y_rus)
    rus_knn_preds = rus_knn.predict(x_stest)

acc_knn_rus = metrics.accuracy_score(y_stest,rus_knn_preds)
    metrics.ConfusionMatrixDisplay.from_predictions(y_stest,rus_knn_preds)
    print(metrics.fl_score(y_stest,rus_knn_preds, average = 'weighted'))
    print(acc_knn_rus)
```

/Library/Frameworks/Python.framework/Versions/3.9/lib/python3.9/site-package s/sklearn/neighbors/\_classification.py:207: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n\_samples,), for example using ravel().

return self.\_fit(X, y)

0.6587949710559801

0.6980184581976113

