# Statistical analyses\_Github

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#### Required data sets

```
WD <- ".."
PATH_OUTPUT <- file.path(WD, "Outputs")
NLOG_VE <- read.csv(file.path(PATH_OUTPUT, "NLOG_VE.csv"), head = T)</pre>
NLOG_VE_sup_zero <- read.csv(file.path(PATH_OUTPUT, "NLOG_VE_sup_zero.csv"), head = T)
NLOG_VE_sup_zero_Moz <- read.csv(file.path(PATH_OUTPUT, "NLOG_VE_sup_zero_Moz.csv"), head = T)
NLOG_VE_sup_zero_North <- read.csv(file.path(PATH_OUTPUT, "NLOG_VE_sup_zero_North.csv"), head = T)</pre>
NLOG_VE_zero <- read.csv(file.path(PATH_OUTPUT, "NLOG_VE_zero.csv"), head = T)
NLOG VE zero Moz <- read.csv(file.path(PATH OUTPUT, "NLOG VE zero Moz.csv"), head = T)
NLOG_VE_zero_North <- read.csv(file.path(PATH_OUTPUT, "NLOG_VE_zero_North.csv"), head = T)</pre>
dfMN_epi<-read.csv(file.path(PATH_OUTPUT, "MN_epi_mean.csv"), header = T)</pre>
dfMN_u<-read.csv(file.path(PATH_OUTPUT, "MN_umeso_mean.csv"), header = T)</pre>
dfMN_mu<-read.csv(file.path(PATH_OUTPUT, "MN_mumeso_mean.csv"), header = T)</pre>
dfMN_ml<-read.csv(file.path(PATH_OUTPUT, "MN_mlmeso_mean.csv"), header = T)
dfMN_hml<-read.csv(file.path(PATH_OUTPUT, "MN_hmlmeso_mean.csv"), header = T)
df_eff <- read.csv(file.path(PATH_OUTPUT, "df_eff.csv"), head = T)</pre>
df_eff_new <- read.csv(file.path(PATH_OUTPUT, "df_eff_new_100.csv"), head = T)</pre>
df_eff_new_50 <- read.csv(file.path(PATH_OUTPUT, "df_eff_new_50.csv"), head = T)</pre>
df_eff_new_150 <- read.csv(file.path(PATH_OUTPUT, "df_eff_new_150.csv"), head = T)</pre>
# For the Sensitivity analysis (Oi,d > 10)
NLOG_VE_zero_Moz_10 <- NLOG_VE_zero_Moz[NLOG_VE_zero_Moz$NumOBS>=10,]
NLOG VE zero North 10 <- NLOG VE zero North[NLOG VE zero North$NumOBS>=10,]
NLOG_VE_sup_zero_Moz_10 <- NLOG_VE_sup_zero_Moz[NLOG_VE_sup_zero_Moz$NumOBS>=10,]
NLOG_VE_sup_zero_North_10 <- NLOG_VE_sup_zero_North[NLOG_VE_sup_zero_North$NumOBS>=10,]
```

## I: Differences between zones (Wilcoxon tests)

Here, no plots have been made but boxplots could be a good representation

```
wilcox.test(NLOG_VE_sup_zero_Moz$chlamean, NLOG_VE_sup_zero_North$chlamean)
##
##
   Wilcoxon rank sum test with continuity correction
## data: NLOG_VE_sup_zero_Moz$chlamean and NLOG_VE_sup_zero_North$chlamean
## W = 4797, p-value = 0.1827
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(NLOG_VE_sup_zero_Moz$slamean, NLOG_VE_sup_zero_North$slamean)
##
## Wilcoxon rank sum test with continuity correction
## data: NLOG VE sup zero Moz$slamean and NLOG VE sup zero North$slamean
## W = 3780, p-value = 0.3781
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(NLOG_VE_sup_zero_Moz$SSCImean, NLOG_VE_sup_zero_North$SSCImean)
##
##
   Wilcoxon rank sum test with continuity correction
## data: NLOG_VE_sup_zero_Moz$SSCImean and NLOG_VE_sup_zero_North$SSCImean
## W = 6005, p-value = 7.358e-05
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(NLOG_VE_sup_zero_Moz$FSLEmean, NLOG_VE_sup_zero_North$FSLEmean)
##
   Wilcoxon rank sum test with continuity correction
## data: NLOG_VE_sup_zero_Moz$FSLEmean and NLOG_VE_sup_zero_North$FSLEmean
## W = 257, p-value < 2.2e-16
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(NLOG_VE_sup_zero_Moz$MNmean, NLOG_VE_sup_zero_North$MNmean)
##
##
  Wilcoxon rank sum test with continuity correction
## data: NLOG_VE_sup_zero_Moz$MNmean and NLOG_VE_sup_zero_North$MNmean
## W = 5833, p-value = 0.0003311
## alternative hypothesis: true location shift is not equal to 0
```

### II: Checking the normal distribution (Shapiro tests)

H0 : the data follows a normal distribution

```
shapiro.test(NLOG_VE$NLOG_stand)
##
##
   Shapiro-Wilk normality test
## data: NLOG_VE$NLOG_stand
## W = 0.46653, p-value < 2.2e-16
shapiro.test(NLOG_VE$chlamean)
##
## Shapiro-Wilk normality test
##
## data: NLOG_VE$chlamean
## W = 0.75869, p-value < 2.2e-16
shapiro.test(NLOG_VE$sstmean)
##
## Shapiro-Wilk normality test
## data: NLOG_VE$sstmean
## W = 0.96988, p-value = 1.837e-12
shapiro.test(NLOG_VE$slamean)
##
## Shapiro-Wilk normality test
##
## data: NLOG_VE$slamean
## W = 0.99529, p-value = 0.00873
shapiro.test(NLOG_VE$SSCImean)
##
## Shapiro-Wilk normality test
##
## data: NLOG_VE$SSCImean
## W = 0.89761, p-value < 2.2e-16
shapiro.test(NLOG_VE$FSLEmean)
##
## Shapiro-Wilk normality test
## data: NLOG_VE$FSLEmean
## W = 0.91901, p-value < 2.2e-16
```

```
shapiro.test(NLOG_VE$MNmean)
##
##
   Shapiro-Wilk normality test
##
## data: NLOG VE$MNmean
## W = 0.92226, p-value < 2.2e-16
III: Correlations between Micronecton types (Kendall tests)
Related to Figure A2 and Table A1
df_list<-list(dfMN_epi,dfMN_u,dfMN_mu,dfMN_ml,dfMN_hml)</pre>
dftot<-Reduce(function(x, y) merge(x, y, by=c("lat_grid", "lon_grid", "year", "month")), df_list)</pre>
cor.test(dftot$micronec_epi,dftot$micronec_umeso, method = "kendall")
##
## Kendall's rank correlation tau
##
## data: dftot$micronec_epi and dftot$micronec_umeso
## z = 74.856, p-value < 2.2e-16
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
##
       tau
## 0.294818
cor.test(dftot$micronec_epi,dftot$micronec_mumeso, method = "kendall")
##
## Kendall's rank correlation tau
## data: dftot$micronec_epi and dftot$micronec_mumeso
## z = 147.2, p-value < 2.2e-16
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
##
         tau
## 0.5797324
cor.test(dftot$micronec_epi,dftot$micronec_mlmeso, method = "kendall")
##
##
   Kendall's rank correlation tau
## data: dftot$micronec_epi and dftot$micronec_mlmeso
## z = 66.482, p-value < 2.2e-16
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
       tau
## 0.261836
```

```
cor.test(dftot$micronec_epi,dftot$micronec_hmlmeso, method = "kendall")
##
   Kendall's rank correlation tau
##
## data: dftot$micronec_epi and dftot$micronec_hmlmeso
## z = 114.15, p-value < 2.2e-16
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
##
         tan
## 0.4495617
cor.test(dftot$micronec_umeso,dftot$micronec_mumeso, method = "kendall")
##
## Kendall's rank correlation tau
## data: dftot$micronec_umeso and dftot$micronec_mumeso
## z = 110.18, p-value < 2.2e-16
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
##
         tau
## 0.4339509
cor.test(dftot$micronec_umeso,dftot$micronec_mlmeso, method = "kendall")
##
## Kendall's rank correlation tau
## data: dftot$micronec_umeso and dftot$micronec_mlmeso
## z = 174.21, p-value < 2.2e-16
\#\# alternative hypothesis: true tau is not equal to 0
## sample estimates:
##
         tau
## 0.6861061
cor.test(dftot$micronec_umeso,dftot$micronec_hmlmeso, method = "kendall")
##
  Kendall's rank correlation tau
##
## data: dftot$micronec_umeso and dftot$micronec_hmlmeso
## z = 55.544, p-value < 2.2e-16
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
##
         tau
## 0.2187593
cor.test(dftot$micronec mumeso,dftot$micronec mlmeso, method = "kendall")
```

```
##
## Kendall's rank correlation tau
##
## data: dftot$micronec_mumeso and dftot$micronec_mlmeso
## z = 100.46, p-value < 2.2e-16
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
##
         tan
## 0.3956415
cor.test(dftot$micronec mumeso,dftot$micronec hmlmeso, method = "kendall")
##
   Kendall's rank correlation tau
##
##
## data: dftot$micronec_mumeso and dftot$micronec_hmlmeso
## z = 160.17, p-value < 2.2e-16
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
##
        tau
## 0.630834
cor.test(dftot$micronec_mlmeso,dftot$micronec_hmlmeso, method = "kendall")
##
   Kendall's rank correlation tau
## data: dftot$micronec_mlmeso and dftot$micronec_hmlmeso
## z = 74.65, p-value < 2.2e-16
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
##
         tau
## 0.2940077
```

## IV: Correlations between environmental variables (Kendall tests)

Related to Figure A3 and Table A2

```
cor.test(NLOG_VE$chlamean, NLOG_VE$sstmean, method = "kendall")

##

## Kendall's rank correlation tau

##

## data: NLOG_VE$chlamean and NLOG_VE$sstmean

## z = -27.801, p-value < 2.2e-16

## alternative hypothesis: true tau is not equal to 0

## sample estimates:

## tau

## -0.6281774</pre>
```

```
cor.test(NLOG_VE$chlamean, NLOG_VE$slamean, method = "kendall")
##
## Kendall's rank correlation tau
## data: NLOG_VE$chlamean and NLOG_VE$slamean
## z = -8.7996, p-value < 2.2e-16
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
##
          tau
## -0.1988312
cor.test(NLOG_VE$chlamean, NLOG_VE$SSCImean, method = "kendall")
## Kendall's rank correlation tau
## data: NLOG_VE$chlamean and NLOG_VE$SSCImean
## z = 3.8585, p-value = 0.0001141
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
##
## 0.08718457
cor.test(NLOG_VE$chlamean, NLOG_VE$FSLEmean, method = "kendall")
##
  Kendall's rank correlation tau
##
## data: NLOG_VE$chlamean and NLOG_VE$FSLEmean
## z = -0.76553, p-value = 0.444
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
           tau
## -0.01729746
cor.test(NLOG_VE$chlamean, NLOG_VE$MNmean, method = "kendall")
##
## Kendall's rank correlation tau
## data: NLOG_VE$chlamean and NLOG_VE$MNmean
## z = 4.046, p-value = 5.211e-05
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
##
          tau
## 0.09142047
cor.test(NLOG_VE$sstmean, NLOG_VE$slamean, method = "kendall")
```

```
##
## Kendall's rank correlation tau
##
## data: NLOG_VE$sstmean and NLOG_VE$slamean
## z = 9.4087, p-value < 2.2e-16
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
         tau
## 0.2125956
cor.test(NLOG_VE$sstmean, NLOG_VE$SSCImean, method = "kendall")
##
## Kendall's rank correlation tau
## data: NLOG_VE$sstmean and NLOG_VE$SSCImean
## z = -3.4449, p-value = 0.0005712
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
##
       tau
## -0.07784
cor.test(NLOG_VE$sstmean, NLOG_VE$FSLEmean, method = "kendall")
##
## Kendall's rank correlation tau
## data: NLOG_VE$sstmean and NLOG_VE$FSLEmean
## z = -2.3974, p-value = 0.01651
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
##
## -0.05417031
cor.test(NLOG_VE$sstmean, NLOG_VE$MNmean, method = "kendall")
##
  Kendall's rank correlation tau
## data: NLOG_VE$sstmean and NLOG_VE$MNmean
## z = -0.26774, p-value = 0.7889
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
           tau
## -0.006049795
cor.test(NLOG_VE$slamean, NLOG_VE$SSCImean, method = "kendall")
   Kendall's rank correlation tau
##
```

```
## data: NLOG_VE$slamean and NLOG_VE$SSCImean
## z = 3.5526, p-value = 0.0003815
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
          tau
## 0.08027271
cor.test(NLOG_VE$slamean, NLOG_VE$FSLEmean, method = "kendall")
##
  Kendall's rank correlation tau
##
## data: NLOG_VE$slamean and NLOG_VE$FSLEmean
## z = -1.8802, p-value = 0.06008
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
##
## -0.04248498
cor.test(NLOG_VE$slamean, NLOG_VE$MNmean, method = "kendall")
##
## Kendall's rank correlation tau
## data: NLOG_VE$slamean and NLOG_VE$MNmean
## z = -0.97376, p-value = 0.3302
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
           tau
## -0.02200265
cor.test(NLOG_VE$SSCImean, NLOG_VE$MNmean, method = "kendall")
##
##
   Kendall's rank correlation tau
## data: NLOG_VE$SSCImean and NLOG_VE$MNmean
## z = -4.9464, p-value = 7.56e-07
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
##
          tan
## -0.1117664
cor.test(NLOG_VE$FSLEmean, NLOG_VE$SSCImean, method = "kendall")
## Kendall's rank correlation tau
## data: NLOG_VE$FSLEmean and NLOG_VE$SSCImean
## z = 2.0827, p-value = 0.03728
## alternative hypothesis: true tau is not equal to 0
```

```
## sample estimates:
##
         tan
## 0.04705885
cor.test(NLOG_VE$FSLEmean, NLOG_VE$MNmean, method = "kendall")
##
## Kendall's rank correlation tau
## data: NLOG_VE$FSLEmean and NLOG_VE$MNmean
## z = -12.246, p-value < 2.2e-16
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
         tau
## -0.2766939
V: Differences between the environmental variables according to
NLOG = 0 or NLOG > 0 (Wilcoxon tests)
Related to Figure 3
wilcox.test(NLOG_VE_zero_Moz$chlamean,NLOG_VE_sup_zero_Moz$chlamean)
##
##
   Wilcoxon rank sum test
## data: NLOG_VE_zero_Moz$chlamean and NLOG_VE_sup_zero_Moz$chlamean
## W = 101, p-value = 0.4413
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(NLOG_VE_zero_North$chlamean, NLOG_VE_sup_zero_North$chlamean)
##
##
   Wilcoxon rank sum test with continuity correction
## data: NLOG_VE_zero_North$chlamean and NLOG_VE_sup_zero_North$chlamean
## W = 84443, p-value = 0.01223
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(NLOG_VE_zero_Moz$slamean,NLOG_VE_sup_zero_Moz$slamean)
##
## Wilcoxon rank sum test
## data: NLOG_VE_zero_Moz$slamean and NLOG_VE_sup_zero_Moz$slamean
## W = 96, p-value = 0.3458
## alternative hypothesis: true location shift is not equal to 0
```

```
wilcox.test(NLOG_VE_zero_North$slamean,NLOG_VE_sup_zero_North$slamean)
##
## Wilcoxon rank sum test with continuity correction
## data: NLOG_VE_zero_North$slamean and NLOG_VE_sup_zero_North$slamean
## W = 79708, p-value = 0.2924
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(NLOG_VE_zero_Moz$SSCImean,NLOG_VE_sup_zero_Moz$SSCImean)
##
##
  Wilcoxon rank sum test
##
## data: NLOG_VE_zero_Moz$SSCImean and NLOG_VE_sup_zero_Moz$SSCImean
## W = 76, p-value = 0.09919
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(NLOG_VE_zero_North$SSCImean,NLOG_VE_sup_zero_North$SSCImean)
## Wilcoxon rank sum test with continuity correction
## data: NLOG_VE_zero_North$SSCImean and NLOG_VE_sup_zero_North$SSCImean
## W = 84334, p-value = 0.01344
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(NLOG_VE_zero_Moz$FSLEmean,NLOG_VE_sup_zero_Moz$FSLEmean)
##
## Wilcoxon rank sum test
##
## data: NLOG_VE_zero_Moz$FSLEmean and NLOG_VE_sup_zero_Moz$FSLEmean
## W = 170, p-value = 0.1148
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(NLOG_VE_zero_North$FSLEmean,NLOG_VE_sup_zero_North$FSLEmean)
##
  Wilcoxon rank sum test with continuity correction
## data: NLOG_VE_zero_North$FSLEmean and NLOG_VE_sup_zero_North$FSLEmean
## W = 72834, p-value = 0.2913
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(NLOG_VE_zero_Moz$MNmean,NLOG_VE_sup_zero_Moz$MNmean)
##
```

## Wilcoxon rank sum test

```
##
## data: NLOG_VE_zero_Moz$MNmean and NLOG_VE_sup_zero_Moz$MNmean
## W = 111, p-value = 0.6697
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(NLOG_VE_zero_North$MNmean, NLOG_VE_sup_zero_North$MNmean)
## Wilcoxon rank sum test with continuity correction
## data: NLOG_VE_zero_North$MNmean and NLOG_VE_sup_zero_North$MNmean
## W = 80846, p-value = 0.1609
## alternative hypothesis: true location shift is not equal to 0
Related to Table A3 (Sensitivity Analysis)
wilcox.test(NLOG_VE_zero_Moz_10$chlamean,NLOG_VE_sup_zero_Moz_10$chlamean)
##
## Wilcoxon rank sum test
## data: NLOG_VE_zero_Moz_10$chlamean and NLOG_VE_sup_zero_Moz_10$chlamean
## W = 12, p-value = 0.8167
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(NLOG_VE_zero_North_10$chlamean, NLOG_VE_sup_zero_North_10$chlamean)
##
## Wilcoxon rank sum test with continuity correction
## data: NLOG_VE_zero_North_10$chlamean and NLOG_VE_sup_zero_North_10$chlamean
## W = 20943, p-value = 0.01625
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(NLOG_VE_zero_Moz_10$slamean,NLOG_VE_sup_zero_Moz_10$slamean)
##
## Wilcoxon rank sum test
## data: NLOG_VE_zero_Moz_10$slamean and NLOG_VE_sup_zero_Moz_10$slamean
## W = 16, p-value = 0.8167
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(NLOG_VE_zero_North_10$slamean, NLOG_VE_sup_zero_North_10$slamean)
##
  Wilcoxon rank sum test with continuity correction
##
## data: NLOG_VE_zero_North_10$slamean and NLOG_VE_sup_zero_North_10$slamean
## W = 19292, p-value = 0.3733
## alternative hypothesis: true location shift is not equal to 0
```

```
##
##
  Wilcoxon rank sum test
## data: NLOG_VE_zero_Moz_10$SSCImean and NLOG_VE_sup_zero_Moz_10$SSCImean
## W = 6, p-value = 0.2667
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(NLOG_VE_zero_North_10$SSCImean, NLOG_VE_sup_zero_North_10$SSCImean)
##
##
   Wilcoxon rank sum test with continuity correction
## data: NLOG_VE_zero_North_10$SSCImean and NLOG_VE_sup_zero_North_10$SSCImean
## W = 19895, p-value = 0.1491
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(NLOG_VE_zero_Moz_10$FSLEmean,NLOG_VE_sup_zero_Moz_10$FSLEmean)
##
## Wilcoxon rank sum test
## data: NLOG_VE_zero_Moz_10$FSLEmean and NLOG_VE_sup_zero_Moz_10$FSLEmean
## W = 7, p-value = 0.3333
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(NLOG_VE_zero_North_10$FSLEmean, NLOG_VE_sup_zero_North_10$FSLEmean)
##
## Wilcoxon rank sum test with continuity correction
## data: NLOG_VE_zero_North_10$FSLEmean and NLOG_VE_sup_zero_North_10$FSLEmean
## W = 19614, p-value = 0.2359
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(NLOG_VE_zero_Moz_10$MNmean,NLOG_VE_sup_zero_Moz_10$MNmean)
##
##
   Wilcoxon rank sum test
## data: NLOG_VE_zero_Moz_10$MNmean and NLOG_VE_sup_zero_Moz_10$MNmean
## W = 15, p-value = 0.9333
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(NLOG_VE_zero_North_10$MNmean,NLOG_VE_sup_zero_North_10$MNmean)
##
## Wilcoxon rank sum test with continuity correction
## data: NLOG_VE_zero_North_10$MNmean and NLOG_VE_sup_zero_North_10$MNmean
## W = 18653, p-value = 0.7606
\#\# alternative hypothesis: true location shift is not equal to 0
```

wilcox.test(NLOG\_VE\_zero\_Moz\_10\$SSCImean, NLOG\_VE\_sup\_zero\_Moz\_10\$SSCImean)

### VI: Correlations between NLOG and VE (Kendall tests)

Related to Figure 4 and Table A4 cor.test(NLOG\_VE\_sup\_zero\_Moz\$NLOG\_stand, NLOG\_VE\_sup\_zero\_Moz\$chlamean, method = "kendall") ## ## Kendall's rank correlation tau ## ## data: NLOG\_VE\_sup\_zero\_Moz\$NLOG\_stand and NLOG\_VE\_sup\_zero\_Moz\$chlamean ## z = 0.13607, p-value = 0.8918 ## alternative hypothesis: true tau is not equal to 0 ## sample estimates: tau ## 0.01729755 cor.test(NLOG\_VE\_sup\_zero\_North\$NLOG\_stand, NLOG\_VE\_sup\_zero\_North\$chlamean, method = "kendall") Kendall's rank correlation tau ## ## data: NLOG\_VE\_sup\_zero\_North\$NLOG\_stand and NLOG\_VE\_sup\_zero\_North\$chlamean ## z = -1.4962, p-value = 0.1346 ## alternative hypothesis: true tau is not equal to 0 ## sample estimates: tau ## -0.06210949 cor.test(NLOG\_VE\_sup\_zero\_Moz\$NLOG\_stand, NLOG\_VE\_sup\_zero\_Moz\$slamean, method = "kendall") ## ## Kendall's rank correlation tau ## ## data: NLOG\_VE\_sup\_zero\_Moz\$NLOG\_stand and NLOG\_VE\_sup\_zero\_Moz\$slamean ## z = 2.4152, p-value = 0.01573 ## alternative hypothesis: true tau is not equal to 0 ## sample estimates: tau ## 0.3070315 cor.test(NLOG\_VE\_sup\_zero\_North\$NLOG\_stand, NLOG\_VE\_sup\_zero\_North\$slamean, method = "kendall") ## Kendall's rank correlation tau ## data: NLOG\_VE\_sup\_zero\_North\$NLOG\_stand and NLOG\_VE\_sup\_zero\_North\$slamean ## z = -2.2274, p-value = 0.02592 ## alternative hypothesis: true tau is not equal to 0 ## sample estimates:

## -0.09246481

```
cor.test(NLOG_VE_sup_zero_Moz$NLOG_stand, NLOG_VE_sup_zero_Moz$SSCImean, method = "kendall")
##
   Kendall's rank correlation tau
##
## data: NLOG_VE_sup_zero_Moz$NLOG_stand and NLOG_VE_sup_zero_Moz$SSCImean
## z = 0.10205, p-value = 0.9187
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
##
         tan
## 0.01297316
cor.test(NLOG_VE_sup_zero_North$NLOG_stand, NLOG_VE_sup_zero_North$SSCImean, method = "kendall")
##
##
  Kendall's rank correlation tau
##
## data: NLOG_VE_sup_zero_North$NLOG_stand and NLOG_VE_sup_zero_North$SSCImean
## z = -1.1883, p-value = 0.2347
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
           tau
## -0.04932883
cor.test(NLOG_VE_sup_zero_Moz$NLOG_stand, NLOG_VE_sup_zero_Moz$FSLEmean, method = "kendall")
## Kendall's rank correlation tau
## data: NLOG_VE_sup_zero_Moz$NLOG_stand and NLOG_VE_sup_zero_Moz$FSLEmean
## z = 0.95249, p-value = 0.3408
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
##
         tau
## 0.1210829
cor.test(NLOG VE sup zero North$NLOG stand, NLOG VE sup zero North$FSLEmean, method = "kendall")
##
## Kendall's rank correlation tau
## data: NLOG_VE_sup_zero_North$NLOG_stand and NLOG_VE_sup_zero_North$FSLEmean
## z = 0.3727, p-value = 0.7094
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
          tau
## 0.01547132
cor.test(NLOG_VE_sup_zero_Moz$NLOG_stand, NLOG_VE_sup_zero_Moz$MNmean, method = "kendall")
```

```
##
## Kendall's rank correlation tau
## data: NLOG_VE_sup_zero_Moz$NLOG_stand and NLOG_VE_sup_zero_Moz$MNmean
## z = -0.23812, p-value = 0.8118
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
## -0.03027071
cor.test(NLOG_VE_sup_zero_North$NLOG_stand, NLOG_VE_sup_zero_North$MNmean, method = "kendall")
##
## Kendall's rank correlation tau
## data: NLOG_VE_sup_zero_North$NLOG_stand and NLOG_VE_sup_zero_North$MNmean
## z = 1.3017, p-value = 0.193
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
         tau
## 0.0540375
Related to Table A5 (Sensitivity Analysis)
cor.test(NLOG_VE_sup_zero_Moz_10$NLOG_stand, NLOG_VE_sup_zero_Moz_10$chlamean, method = "kendall")
##
## Kendall's rank correlation tau
## data: NLOG_VE_sup_zero_Moz_10$NLOG_stand and NLOG_VE_sup_zero_Moz_10$chlamean
## T = 45, p-value = 1
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
           tau
## -0.01098901
cor.test(NLOG VE sup zero North 10$NLOG stand, NLOG VE sup zero North 10$chlamean, method = "kendall")
##
## Kendall's rank correlation tau
##
## data: NLOG_VE_sup_zero_North_10$NLOG_stand and NLOG_VE_sup_zero_North_10$chlamean
## z = -0.81672, p-value = 0.4141
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
           tau
## -0.04405533
cor.test(NLOG_VE_sup_zero_Moz_10$NLOG_stand, NLOG_VE_sup_zero_Moz_10$slamean, method = "kendall")
```

```
##
## Kendall's rank correlation tau
##
## data: NLOG_VE_sup_zero_Moz_10$NLOG_stand and NLOG_VE_sup_zero_Moz_10$slamean
## T = 60, p-value = 0.1268
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
         tau
## 0.3186813
cor.test(NLOG_VE_sup_zero_North_10$NLOG_stand, NLOG_VE_sup_zero_North_10$slamean, method = "kendall")
## Kendall's rank correlation tau
## data: NLOG_VE_sup_zero_North_10$NLOG_stand and NLOG_VE_sup_zero_North_10$slamean
## z = -2.5196, p-value = 0.01175
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
         tau
## -0.1359157
cor.test(NLOG_VE_sup_zero_Moz_10$NLOG_stand, NLOG_VE_sup_zero_Moz_10$SSCImean, method = "kendall")
##
## Kendall's rank correlation tau
## data: NLOG_VE_sup_zero_Moz_10$NLOG_stand and NLOG_VE_sup_zero_Moz_10$SSCImean
## T = 48, p-value = 0.8299
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
##
          tau
## 0.05494505
cor.test(NLOG VE sup zero North 10$NLOG stand, NLOG VE sup zero North 10$SSCImean, method = "kendall")
##
## Kendall's rank correlation tau
## data: NLOG VE sup zero North 10$NLOG stand and NLOG VE sup zero North 10$SSCImean
## z = -1.4104, p-value = 0.1584
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
           tau
## -0.07608109
cor.test(NLOG_VE_sup_zero_Moz_10$NLOG_stand, NLOG_VE_sup_zero_Moz_10$FSLEmean, method = "kendall")
##
   Kendall's rank correlation tau
##
```

```
## data: NLOG_VE_sup_zero_Moz_10$NLOG_stand and NLOG_VE_sup_zero_Moz_10$FSLEmean
## T = 50, p-value = 0.6672
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
         tau
## 0.0989011
cor.test(NLOG_VE_sup_zero_North_10$NLOG_stand, NLOG_VE_sup_zero_North_10$FSLEmean, method = "kendall")
##
##
   Kendall's rank correlation tau
##
## data: NLOG_VE_sup_zero_North_10$NLOG_stand and NLOG_VE_sup_zero_North_10$FSLEmean
## z = -0.14917, p-value = 0.8814
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
##
            tau
## -0.008046273
cor.test(NLOG_VE_sup_zero_Moz_10$NLOG_stand, NLOG_VE_sup_zero_Moz_10$MNmean, method = "kendall")
##
##
   Kendall's rank correlation tau
##
## data: NLOG_VE_sup_zero_Moz_10$NLOG_stand and NLOG_VE_sup_zero_Moz_10$MNmean
## T = 45, p-value = 1
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
## -0.01098901
cor.test(NLOG_VE_sup_zero_North_10$NLOG_stand, NLOG_VE_sup_zero_North_10$MNmean, method = "kendall")
##
## Kendall's rank correlation tau
## data: NLOG_VE_sup_zero_North_10$NLOG_stand and NLOG_VE_sup_zero_North_10$MNmean
## z = 0.93782, p-value = 0.3483
## alternative hypothesis: true tau is not equal to 0
## sample estimates:
## 0.05058795
```

#### VII: Construction of models

Related to Figure 5, Figure A1, Figure A4 and Table 3

```
NLOG_VE_sup_zero_North$chlacr <- scale(NLOG_VE_sup_zero_North$chlamean)
NLOG_VE_sup_zero_North$slacr <- scale(NLOG_VE_sup_zero_North$slamean)
NLOG_VE_sup_zero_North$SSCIcr <- scale(NLOG_VE_sup_zero_North$SSCImean)</pre>
```

```
NLOG_VE_sup_zero_North$FSLEcr <- scale(NLOG_VE_sup_zero_North$FSLEmean)
NLOG_VE_sup_zero_North$MNcr <- scale(NLOG_VE_sup_zero_North$MNmean)
NLOG_VE_sup_zero_North$logNLOG <- log(NLOG_VE_sup_zero_North$NLOG_stand)

NLOG_VE_sup_zero_Moz$chlacr <- scale(NLOG_VE_sup_zero_Moz$chlamean)
NLOG_VE_sup_zero_Moz$slacr <- scale(NLOG_VE_sup_zero_Moz$slamean)
NLOG_VE_sup_zero_Moz$SSCIcr <- scale(NLOG_VE_sup_zero_Moz$SSCImean)
NLOG_VE_sup_zero_Moz$FSLEcr <- scale(NLOG_VE_sup_zero_Moz$FSLEmean)
NLOG_VE_sup_zero_Moz$MNcr <- scale(NLOG_VE_sup_zero_Moz$MNmean)
NLOG_VE_sup_zero_Moz$logNLOG <- log(NLOG_VE_sup_zero_Moz$NLOG_stand)
```

#### Linear model for the >10°S zone (North)

## <none>

```
LM_North_chla <- lm(logNLOG ~ chlacr + slacr + SSCIcr + FSLEcr + MNcr, data = NLOG_VE_sup_zero_North)
summary(LM_North_chla)
##
## Call:
## lm(formula = logNLOG ~ chlacr + slacr + SSCIcr + FSLEcr + MNcr,
      data = NLOG_VE_sup_zero_North)
## Residuals:
       Min
                 1Q
                    Median
                                         Max
## -1.63432 -0.42382 0.01529 0.41113 1.39258
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -1.945243 0.038111 -51.042 < 2e-16 ***
                        0.041093 -2.111 0.03569 *
## chlacr
             -0.086757
## slacr
             ## SSCIcr
              -0.057175
                         0.039946 -1.431 0.15352
## FSLEcr
              -0.035422
                         0.041567 -0.852 0.39490
## MNcr
              -0.004671
                         0.041538 -0.112 0.91054
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 0.6262 on 264 degrees of freedom
## Multiple R-squared: 0.05425,
                                 Adjusted R-squared:
## F-statistic: 3.029 on 5 and 264 DF, p-value: 0.01121
stepAIC(LM_North_chla)
## Start: AIC=-246.82
## logNLOG ~ chlacr + slacr + SSCIcr + FSLEcr + MNcr
##
           Df Sum of Sq
                          RSS
## - MNcr
            1
                0.0050 103.53 -248.80
## - FSLEcr 1
                 0.2848 103.81 -248.07
```

103.53 -246.82

## - SSCIcr 1 0.8034 104.33 -246.73

```
## - chlacr 1
              1.7480 105.28 -244.30
## - slacr 1
              3.6212 107.15 -239.53
## Step: AIC=-248.8
## logNLOG ~ chlacr + slacr + SSCIcr + FSLEcr
           Df Sum of Sq RSS
                0.2927 103.83 -250.04
## - FSLEcr 1
## <none>
                       103.53 -248.80
## - SSCIcr 1
                0.8065 104.34 -248.71
## - chlacr 1 1.7463 105.28 -246.29
## - slacr 1
              3.6165 107.15 -241.53
## Step: AIC=-250.04
## logNLOG ~ chlacr + slacr + SSCIcr
##
                         RSS
##
           Df Sum of Sq
## <none>
                       103.83 -250.04
## - SSCIcr 1
               1.0037 104.83 -249.44
## - chlacr 1
               1.5971 105.42 -247.92
## - slacr 1 3.3747 107.20 -243.41
##
## Call:
## lm(formula = logNLOG ~ chlacr + slacr + SSCIcr, data = NLOG_VE_sup_zero_North)
## Coefficients:
                   chlacr
## (Intercept)
                                slacr
                                           SSCIcr
     -1.94524
                 -0.08227
                             -0.11915
                                         -0.06202
LM2_North_chla <- lm(logNLOG ~ chlacr + slacr + SSCIcr + FSLEcr, data = NLOG_VE_sup_zero_North)
summary(LM2 North chla)
##
## Call:
## lm(formula = logNLOG ~ chlacr + slacr + SSCIcr + FSLEcr, data = NLOG VE sup zero North)
## Residuals:
               1Q Median
##
       Min
                                 3Q
## -1.63680 -0.42294 0.01666 0.40919 1.38722
## Coefficients:
             Estimate Std. Error t value Pr(>|t|)
## chlacr
             -0.08671
                         0.04101 -2.114 0.03543 *
## slacr
             -0.12511
                         0.04112 -3.042 0.00258 **
## SSCIcr
             -0.05638
                         0.03924 -1.437 0.15196
             -0.03387
## FSLEcr
                         0.03913 -0.866 0.38750
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.6251 on 265 degrees of freedom
## Multiple R-squared: 0.05421, Adjusted R-squared: 0.03993
## F-statistic: 3.797 on 4 and 265 DF, p-value: 0.00508
```

```
summary(LM3_North_chla)
##
## Call:
## lm(formula = logNLOG ~ chlacr + slacr + SSCIcr, data = NLOG_VE_sup_zero_North)
## Residuals:
       Min
                 10
                     Median
                                   3Q
## -1.65378 -0.43897 0.02645 0.41915 1.41580
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
                         0.03802 -51.162 < 2e-16 ***
## (Intercept) -1.94524
             -0.08227
## chlacr
                          0.04067 -2.023 0.04410 *
## slacr
              -0.11915
                          0.04052 -2.940 0.00357 **
## SSCIcr
             -0.06202
                          0.03868 -1.604 0.11000
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 0.6248 on 266 degrees of freedom
## Multiple R-squared: 0.05153, Adjusted R-squared: 0.04084
## F-statistic: 4.818 on 3 and 266 DF, p-value: 0.002772
LM4_North_chla <- lm(logNLOG ~ chlacr + slacr, data = NLOG_VE_sup_zero_North)
summary(LM4 North chla)
##
## Call:
## lm(formula = logNLOG ~ chlacr + slacr, data = NLOG_VE_sup_zero_North)
## Residuals:
                 1Q
                     Median
                                   3Q
## -1.61352 -0.44007 0.01007 0.42305 1.48592
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -1.94524
                         0.03813 -51.012 < 2e-16 ***
## chlacr
              -0.09226
                          0.04031 -2.289 0.02289 *
                          0.04031 -3.160 0.00176 **
              -0.12740
## slacr
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 0.6266 on 267 degrees of freedom
## Multiple R-squared: 0.04236,
                                  Adjusted R-squared: 0.03519
## F-statistic: 5.906 on 2 and 267 DF, p-value: 0.003092
```

LM3\_North\_chla <- lm(logNLOG ~ chlacr + slacr + SSCIcr, data = NLOG\_VE\_sup\_zero\_North)

#### Linear model for the <10°S zone (Moz)

```
LM_Moz_chla <- lm(logNLOG ~ chlacr + slacr + SSCIcr + FSLEcr + MNcr, data = NLOG_VE_sup_zero_Moz)
summary(LM_Moz_chla)
##
## Call:
## lm(formula = logNLOG ~ chlacr + slacr + SSCIcr + FSLEcr + MNcr,
      data = NLOG_VE_sup_zero_Moz)
##
## Residuals:
##
      Min
               1Q Median
                               3Q
## -1.4715 -0.6300 0.1197 0.4765 1.5835
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
##
                                            4e-05 ***
## (Intercept) -0.69572 0.13991 -4.973
## chlacr
              0.10054
                          0.25507 0.394
                                          0.6968
## slacr
              0.43936
                                  2.534
                          0.17342
                                          0.0179 *
## SSCIcr
              -0.30436
                          0.20373 - 1.494
                                           0.1477
              -0.03387
## FSLEcr
                          0.17002 -0.199
                                           0.8437
## MNcr
              -0.32920
                          0.27349 - 1.204
                                           0.2400
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.779 on 25 degrees of freedom
## Multiple R-squared: 0.2416, Adjusted R-squared: 0.08992
## F-statistic: 1.593 on 5 and 25 DF, p-value: 0.1986
stepAIC(LM_Moz_chla)
## Start: AIC=-10.15
## logNLOG ~ chlacr + slacr + SSCIcr + FSLEcr + MNcr
##
           Df Sum of Sq
                          RSS
                                   AIC
## - FSLEcr 1
                0.0241 15.195 -12.1041
## - chlacr 1
                 0.0943 15.265 -11.9612
## - MNcr 1 0.8792 16.050 -10.4068
                        15.171 -10.1533
## <none>
## - SSCIcr 1
                1.3544 16.525 -9.5023
## - slacr 1
                 3.8952 19.066 -5.0687
##
## Step: AIC=-12.1
## logNLOG ~ chlacr + slacr + SSCIcr + MNcr
##
##
           Df Sum of Sq
                                    AIC
                           RSS
## - chlacr 1
               0.0778 15.273 -13.9459
                 0.8617 16.056 -12.3942
## - MNcr
            1
## <none>
                        15.195 -12.1041
## - SSCIcr 1
                 1.5405 16.735 -11.1105
## - slacr
                 4.3740 19.569 -6.2616
            1
##
## Step: AIC=-13.95
## logNLOG ~ slacr + SSCIcr + MNcr
```

```
##
            Df Sum of Sq
##
                            RSS
                                     ATC
## <none>
                         15.273 -13.9459
## - MNcr
                  1.2855 16.558 -13.4407
             1
## - SSCIcr 1
                  1.5186 16.791 -13.0072
## - slacr
                  4.3054 19.578 -8.2472
             1
##
## Call:
## lm(formula = logNLOG ~ slacr + SSCIcr + MNcr, data = NLOG_VE_sup_zero_Moz)
## Coefficients:
## (Intercept)
                      slacr
                                  SSCIcr
                                                 MNcr
##
       -0.6957
                     0.4203
                                 -0.2827
                                              -0.2418
LM2_Moz_chla <- lm(logNLOG ~ chlacr + slacr + SSCIcr + MNcr, data = NLOG_VE_sup_zero_Moz)
summary(LM2_Moz_chla)
##
## Call:
## lm(formula = logNLOG ~ chlacr + slacr + SSCIcr + MNcr, data = NLOG_VE_sup_zero_Moz)
##
## Residuals:
##
       Min
                1Q Median
                                3Q
                                       Max
## -1.5012 -0.6008 0.1150 0.4633 1.5958
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) -0.69572
                           0.13730 -5.067 2.82e-05 ***
## chlacr
                0.08885
                           0.24360
                                     0.365
                                             0.7182
## slacr
                0.42531
                           0.15546
                                     2.736
                                             0.0111 *
## SSCIcr
               -0.28492
                           0.17549
                                    -1.624
                                             0.1165
## MNcr
               -0.31563
                           0.25994 - 1.214
                                             0.2356
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 0.7645 on 26 degrees of freedom
## Multiple R-squared: 0.2404, Adjusted R-squared: 0.1235
## F-statistic: 2.057 on 4 and 26 DF, p-value: 0.1156
LM3_Moz_chla <- lm(logNLOG ~ slacr + SSCIcr + MNcr, data = NLOG_VE_sup_zero_Moz)
summary(LM3_Moz_chla)
##
## lm(formula = logNLOG ~ slacr + SSCIcr + MNcr, data = NLOG_VE_sup_zero_Moz)
##
## Residuals:
        Min
                  1Q
                      Median
                                            Max
                                    3Q
## -1.47876 -0.59818 0.09487 0.49096 1.61276
##
## Coefficients:
```

```
Estimate Std. Error t value Pr(>|t|)
## (Intercept) -0.6957
                           0.1351 -5.150 2.03e-05 ***
## slacr
                0.4203
                           0.1524
                                   2.759
                                            0.0103 *
## SSCIcr
               -0.2827
                           0.1725 -1.639
                                            0.1129
## MNcr
               -0.2418
                           0.1604 -1.507
                                            0.1433
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.7521 on 27 degrees of freedom
## Multiple R-squared: 0.2365, Adjusted R-squared: 0.1517
## F-statistic: 2.788 on 3 and 27 DF, p-value: 0.05976
LM4_Moz_chla <- lm(logNLOG ~ slacr + SSCIcr, data = NLOG_VE_sup_zero_Moz)
summary(LM4_Moz_chla)
##
## Call:
## lm(formula = logNLOG ~ slacr + SSCIcr, data = NLOG_VE_sup_zero_Moz)
##
## Residuals:
##
      Min
               1Q Median
                               3Q
                                      Max
## -1.4896 -0.6056 0.1438 0.4021 1.6381
## Coefficients:
##
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -0.6957
                           0.1381 -5.037 2.51e-05 ***
                           0.1511
                                    2.412
## slacr
                0.3644
                                            0.0227 *
## SSCIcr
               -0.1484
                           0.1511 -0.982
                                            0.3343
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 0.769 on 28 degrees of freedom
## Multiple R-squared: 0.1722, Adjusted R-squared: 0.1131
## F-statistic: 2.913 on 2 and 28 DF, p-value: 0.07089
LM5_Moz_chla <- lm(logNLOG ~ slacr, data = NLOG_VE_sup_zero_Moz)</pre>
summary(LM5_Moz_chla)
##
## Call:
## lm(formula = logNLOG ~ slacr, data = NLOG_VE_sup_zero_Moz)
##
## Residuals:
               1Q Median
                               3Q
## -1.7310 -0.5754 0.1378 0.4193 1.5177
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
                           0.1380 -5.040 2.27e-05 ***
## (Intercept) -0.6957
## slacr
                0.3096
                           0.1403
                                    2.206
                                          0.0354 *
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
##
```

```
## Residual standard error: 0.7685 on 29 degrees of freedom
## Multiple R-squared: 0.1437, Adjusted R-squared: 0.1142
## F-statistic: 4.867 on 1 and 29 DF, p-value: 0.03545
Related to Table A6 (Sensitivity analysis)
NLOG_VE_sup_zero_North_10$chlacr <- scale(NLOG_VE_sup_zero_North_10$chlamean)
NLOG VE sup zero North 10$slacr <- scale(NLOG VE sup zero North 10$slamean)
NLOG_VE_sup_zero_North_10$SSCIcr <- scale(NLOG_VE_sup_zero_North_10$SSCImean)
NLOG_VE_sup_zero_North_10$FSLEcr <- scale(NLOG_VE_sup_zero_North_10$FSLEmean)
NLOG_VE_sup_zero_North_10$MNcr <- scale(NLOG_VE_sup_zero_North_10$MNmean)
NLOG_VE_sup_zero_North_10$logNLOG <- log(NLOG_VE_sup_zero_North_10$NLOG_stand)
LM_North_chla_10 <- lm(logNLOG ~ chlacr + slacr + SSCIcr + FSLEcr + MNcr, data = NLOG_VE_sup_zero_North
summary(LM_North_chla_10)
##
## Call:
## lm(formula = logNLOG ~ chlacr + slacr + SSCIcr + FSLEcr + MNcr,
##
       data = NLOG_VE_sup_zero_North_10)
##
## Residuals:
##
       Min
                  1Q
                     Median
                                    3Q
## -1.45771 -0.50415 -0.04823 0.38847 1.59035
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
                          0.05152 -41.691 < 2e-16 ***
## (Intercept) -2.14786
## chlacr
              -0.08549
                           0.05676 -1.506 0.13413
## slacr
              -0.17669
                          0.05862 -3.014 0.00302 **
## SSCIcr
              -0.05316
                          0.05481 -0.970 0.33365
## FSLEcr
              -0.06643
                          0.05860 -1.134 0.25867
                          0.05691 -0.266 0.79079
## MNcr
              -0.01512
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
##
## Residual standard error: 0.6517 on 154 degrees of freedom
                                    Adjusted R-squared: 0.04506
## Multiple R-squared: 0.07509,
## F-statistic: 2.501 on 5 and 154 DF, p-value: 0.03296
LM2_North_chla_10 <- lm(logNLOG ~ chlacr + slacr + SSCIcr + FSLEcr, data = NLOG_VE_sup_zero_North_10)
summary(LM2 North chla 10)
##
## Call:
## lm(formula = logNLOG ~ chlacr + slacr + SSCIcr + FSLEcr, data = NLOG_VE_sup_zero_North_10)
##
## Residuals:
                1Q Median
                                3Q
      Min
                                       Max
## -1.4666 -0.5017 -0.0445 0.3888 1.5717
##
## Coefficients:
```

```
Estimate Std. Error t value Pr(>|t|)
## chlacr
           -0.08457
                      0.05649 -1.497 0.13641
                      0.05836 -3.013 0.00302 **
## slacr
            -0.17586
## SSCIcr
            -0.05080
                     0.05392 -0.942 0.34765
            -0.06084
                     0.05453 -1.116 0.26622
## FSLEcr
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.6497 on 155 degrees of freedom
## Multiple R-squared: 0.07467, Adjusted R-squared:
## F-statistic: 3.127 on 4 and 155 DF, p-value: 0.01659
LM3_North_chla_10 <- lm(logNLOG ~ chlacr + slacr + FSLEcr, data = NLOG_VE_sup_zero_North_10)
summary(LM3_North_chla_10)
##
## Call:
## lm(formula = logNLOG ~ chlacr + slacr + FSLEcr, data = NLOG_VE_sup_zero_North_10)
## Residuals:
     Min
             1Q Median
                           30
                                 Max
## -1.4358 -0.4828 -0.0568 0.4069 1.6088
##
## Coefficients:
##
            Estimate Std. Error t value Pr(>|t|)
## chlacr
            -0.09442
                       0.05549 -1.702 0.09083 .
## slacr
            -0.18890
                       0.05668 -3.333 0.00107 **
## FSLEcr
            -0.07177
                      0.05326 -1.347 0.17979
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.6495 on 156 degrees of freedom
## Multiple R-squared: 0.06937, Adjusted R-squared: 0.05147
## F-statistic: 3.876 on 3 and 156 DF, p-value: 0.01044
LM4 North chla 10 <- lm(logNLOG ~ chlacr + slacr, data = NLOG VE sup zero North 10)
summary(LM4_North_chla_10)
##
## lm(formula = logNLOG ~ chlacr + slacr, data = NLOG_VE_sup_zero_North_10)
##
## Residuals:
               1Q Median
                              3Q
## -1.46401 -0.46228 -0.07624 0.41669 1.56878
## Coefficients:
            Estimate Std. Error t value Pr(>|t|)
## chlacr
           -0.08360
                      0.05505 -1.519 0.13088
            ## slacr
```

```
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.6512 on 157 degrees of freedom
## Multiple R-squared: 0.05854,
                                   Adjusted R-squared:
## F-statistic: 4.881 on 2 and 157 DF, p-value: 0.00878
LM5_North_chla_10 <- lm(logNLOG ~ slacr, data = NLOG_VE_sup_zero_North_10)</pre>
summary(LM5_North_chla_10)
##
## Call:
## lm(formula = logNLOG ~ slacr, data = NLOG_VE_sup_zero_North_10)
## Residuals:
                 1Q Median
##
## -1.48800 -0.49092 -0.09125 0.46222 1.62824
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
                          0.05169 -41.552 < 2e-16 ***
## (Intercept) -2.14786
                          0.05185 -2.719 0.00727 **
## slacr
              -0.14101
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 0.6538 on 158 degrees of freedom
## Multiple R-squared: 0.04471,
                                   Adjusted R-squared: 0.03866
## F-statistic: 7.395 on 1 and 158 DF, p-value: 0.007273
NLOG_VE_sup_zero_Moz_10$chlacr <- scale(NLOG_VE_sup_zero_Moz_10$chlamean)
NLOG_VE_sup_zero_Moz_10$slacr <- scale(NLOG_VE_sup_zero_Moz_10$slamean)
NLOG_VE_sup_zero_Moz_10$SSCIcr <- scale(NLOG_VE_sup_zero_Moz_10$SSCImean)
NLOG_VE_sup_zero_Moz_10$FSLEcr <- scale(NLOG_VE_sup_zero_Moz_10$FSLEmean)
NLOG_VE_sup_zero_Moz_10$MNcr <- scale(NLOG_VE_sup_zero_Moz_10$MNmean)
NLOG_VE_sup_zero_Moz_10$logNLOG <- log(NLOG_VE_sup_zero_Moz_10$NLOG_stand)
LM_Moz_chla_10 <- lm(logNLOG ~ chlacr + slacr + SSCIcr + FSLEcr + MNcr, data = NLOG_VE_sup_zero_Moz_10)
summary(LM Moz chla 10)
##
## Call:
## lm(formula = logNLOG ~ chlacr + slacr + SSCIcr + FSLEcr + MNcr,
##
      data = NLOG_VE_sup_zero_Moz_10)
##
## Residuals:
                     Median
                                   ЗQ
                 1Q
## -1.12851 -0.41877 0.04339 0.35694 1.21036
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -0.44312
                          0.21967 -2.017
                                            0.0784 .
## chlacr
                          0.27460
                                   0.599
              0.16459
                                            0.5655
               0.29006
                                   1.085 0.3096
## slacr
                          0.26737
```

```
## SSCIcr
              -0.20297
                          0.29208 -0.695
                                            0.5068
## FSLEcr
              0.02013
                          0.32786
                                   0.061
                                            0.9525
              -0.34025
## MNcr
                          0.39769 -0.856 0.4171
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.8219 on 8 degrees of freedom
## Multiple R-squared: 0.188, Adjusted R-squared: -0.3195
## F-statistic: 0.3705 on 5 and 8 DF, p-value: 0.8555
LM2_Moz_chla_10 <- lm(logNLOG ~ chlacr + slacr + SSCIcr + MNcr, data = NLOG_VE_sup_zero_Moz_10)
summary(LM2_Moz_chla_10)
##
## Call:
## lm(formula = logNLOG ~ chlacr + slacr + SSCIcr + MNcr, data = NLOG_VE_sup_zero_Moz_10)
##
## Residuals:
                     Median
##
       Min
                 1Q
                                   3Q
                                           Max
## -1.13496 -0.40338 0.04165 0.35748 1.20849
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -0.4431
                           0.2072 -2.139
                                            0.0611 .
## chlacr
                0.1634
                           0.2583
                                   0.633
                                            0.5428
## slacr
                0.2949
                           0.2409
                                    1.224
                                            0.2519
## SSCIcr
               -0.2050
                           0.2737 - 0.749
                                            0.4731
## MNcr
               -0.3287
                           0.3307 -0.994
                                          0.3462
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 0.7751 on 9 degrees of freedom
## Multiple R-squared: 0.1876, Adjusted R-squared: -0.1734
## F-statistic: 0.5196 on 4 and 9 DF, p-value: 0.724
LM3_Moz_chla_10 <- lm(logNLOG ~ slacr + SSCIcr + MNcr, data = NLOG_VE_sup_zero_Moz_10)
summary(LM3 Moz chla 10)
##
## Call:
## lm(formula = logNLOG ~ slacr + SSCIcr + MNcr, data = NLOG_VE_sup_zero_Moz_10)
##
## Residuals:
                 1Q
                      Median
                                   3Q
## -1.17001 -0.37633 0.08566 0.47975 1.00725
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
                           0.2008 -2.206
## (Intercept) -0.4431
                                            0.0519 .
## slacr
                0.3006
                           0.2334
                                    1.288
                                            0.2268
## SSCIcr
               -0.1832
                           0.2633 -0.696
                                            0.5025
## MNcr
               -0.2302
                           0.2828 -0.814
                                            0.4347
## ---
```

```
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.7515 on 10 degrees of freedom
## Multiple R-squared: 0.1515, Adjusted R-squared: -0.103
## F-statistic: 0.5952 on 3 and 10 DF, p-value: 0.6323
LM4_Moz_chla_10 <- lm(logNLOG ~ slacr + MNcr, data = NLOG_VE_sup_zero_Moz_10)
summary(LM4_Moz_chla_10)
##
## Call:
## lm(formula = logNLOG ~ slacr + MNcr, data = NLOG_VE_sup_zero_Moz_10)
##
## Residuals:
##
      Min
               1Q Median
                               3Q
                                      Max
## -1.2103 -0.2840 0.1214 0.5201 1.0904
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) -0.4431
                           0.1961 - 2.260
                                            0.0451 *
## slacr
                0.2548
                           0.2186 1.166
                                            0.2684
## MNcr
               -0.1100
                           0.2186 -0.503
                                            0.6249
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 0.7337 on 11 degrees of freedom
## Multiple R-squared: 0.1104, Adjusted R-squared: -0.0513
## F-statistic: 0.6828 on 2 and 11 DF, p-value: 0.5254
LM5_Moz_chla_10 <- lm(logNLOG ~ slacr, data = NLOG_VE_sup_zero_Moz_10)</pre>
summary(LM5_Moz_chla_10)
##
## Call:
## lm(formula = logNLOG ~ slacr, data = NLOG_VE_sup_zero_Moz_10)
## Residuals:
##
                 1Q
                     Median
## -1.09691 -0.18624 0.09067 0.51312 1.00997
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -0.4431
                           0.1899 -2.334
                                            0.0378 *
                0.2146
                                   1.089
                                            0.2974
## slacr
                           0.1970
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.7105 on 12 degrees of freedom
## Multiple R-squared: 0.08998,
                                   Adjusted R-squared: 0.01414
## F-statistic: 1.186 on 1 and 12 DF, p-value: 0.2974
```

#### VIII: Sampling bias analysis (Wilcoxon tests)

```
Related to Figure B1 and Table B1
wilcox.test(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Moz",]$chlamean,
            df_eff_new[df_eff_new$Zone=="Moz",]$chlamean)
## Wilcoxon rank sum test with continuity correction
##
## data: df_eff[df_eff$threshold == "Fisheries" & df_eff$Zone == "Moz", and df_eff_new[df_eff_new$Zon
## W = 56620, p-value = 5.017e-08
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Above_10S",]$chlamean,
            df_eff_new[df_eff_new$Zone=="Above_10S",]$chlamean)
##
## Wilcoxon rank sum test with continuity correction
## data: df_eff[df_eff$threshold == "Fisheries" & df_eff$Zone == "Above_10S", and df_eff_new[df_eff_n
## W = 2557336, p-value = 7.101e-14
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Moz",]$slamean,
            df_eff_new[df_eff_new$Zone=="Moz",]$slamean)
##
## Wilcoxon rank sum test with continuity correction
## data: df_eff[df_eff$threshold == "Fisheries" & df_eff$Zone == "Moz", and df_eff_new[df_eff_new$Zon
## W = 25298, p-value = 0.0004887
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Above_10S",]$slamean,
            df_eff_new[df_eff_new$Zone=="Above_10S",]$slamean)
##
## Wilcoxon rank sum test with continuity correction
## data: df_eff[df_eff$threshold == "Fisheries" & df_eff$Zone == "Above_10S", and df_eff_new[df_eff_n
## W = 2137026, p-value = 0.1654
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Moz",]$SSCImean,
            df_eff_new[df_eff_new$Zone=="Moz",]$SSCImean)
## Wilcoxon rank sum test with continuity correction
## data: df_eff[df_eff$threshold == "Fisheries" & df_eff$Zone == "Moz", and df_eff_new[df_eff_new$Zon
## W = 43741, p-value = 0.0758
```

## alternative hypothesis: true location shift is not equal to 0

```
wilcox.test(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Above_10S",]$SSCImean,
            df_eff_new[df_eff_new$Zone=="Above_10S",]$SSCImean)
## Wilcoxon rank sum test with continuity correction
## data: df_eff[df_eff$threshold == "Fisheries" & df_eff$Zone == "Above_10S", and df_eff_new[df_eff_n
## W = 2288471, p-value = 0.07029
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(df eff[df eff$threshold=="Fisheries" & df eff$Zone=="Moz",]$FSLEmean,
            df_eff_new[df_eff_new$Zone=="Moz",]$FSLEmean)
##
## Wilcoxon rank sum test with continuity correction
## data: df_eff[df_eff$threshold == "Fisheries" & df_eff$Zone == "Moz", and df_eff_new[df_eff_new$Zon
## W = 16705, p-value = 2.869e-09
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Above_10S",]$FSLEmean,
            df_eff_new[df_eff_new$Zone=="Above_10S",]$FSLEmean)
##
## Wilcoxon rank sum test with continuity correction
##
## data: df_eff[df_eff$threshold == "Fisheries" & df_eff$Zone == "Above_10S", and df_eff_new[df_eff_n
## W = 2241418, p-value = 0.4141
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(df eff[df eff$threshold=="Fisheries" & df eff$Zone=="Moz",]$micronec epi,
            df_eff_new[df_eff_new$Zone=="Moz",]$micronec_epi)
##
## Wilcoxon rank sum test with continuity correction
## data: df_eff[df_eff$threshold == "Fisheries" & df_eff$Zone == "Moz", and df_eff_new[df_eff_new$Zon
## W = 48289, p-value = 0.002116
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Above_10S",]$micronec_epi,
            df_eff_new[df_eff_new$Zone=="Above_10S",]$micronec_epi)
##
## Wilcoxon rank sum test with continuity correction
## data: df eff[df eff$threshold == "Fisheries" & df eff$Zone == "Above 10S", and df eff new[df eff n
## W = 2297977, p-value = 0.04435
## alternative hypothesis: true location shift is not equal to 0
```

```
#### Medians ####
## CHLA
median(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Moz",]$chlamean)
## [1] 0.1257596
median(df_eff_new[df_eff_new$Zone=="Moz",]$chlamean)
## [1] 0.08913168
median(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Above_10S",]$chlamean)
## [1] 0.136388
median(df_eff_new[df_eff_new$Zone=="Above_10S",]$chlamean)
## [1] 0.1165741
##SLA
median(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Moz",]$slamean)
## [1] 0.05261875
median(df_eff_new[df_eff_new$Zone=="Moz",]$slamean)
## [1] 0.0916
median(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Above_10S",]$slamean)
## [1] 0.07872656
median(df_eff_new[df_eff_new$Zone=="Above_10S",]$slamean)
## [1] 0.08061172
## SSCI
median(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Moz",]$SSCImean)
## [1] 0.1988864
median(df_eff_new[df_eff_new$Zone=="Moz",]$SSCImean)
```

## [1] 0.171358

```
median(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Above_10S",]$SSCImean)
## [1] 0.150017
median(df_eff_new[df_eff_new$Zone=="Above_10S",]$SSCImean)
## [1] 0.1404724
## FSLE
median(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Moz",]$FSLEmean)
## [1] -0.08792684
median(df_eff_new[df_eff_new$Zone=="Moz",]$FSLEmean)
## [1] -0.05425415
median(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Above_10S",]$FSLEmean)
## [1] -0.04729317
median(df_eff_new[df_eff_new$Zone=="Above_10S",]$FSLEmean)
## [1] -0.04796686
median(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Moz",]$micronec_epi)
## [1] 0.4471993
median(df_eff_new[df_eff_new$Zone=="Moz",]$micronec_epi)
## [1] 0.3700588
median(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Above_10S",]$micronec_epi)
## [1] 0.3779635
median(df_eff_new[df_eff_new$Zone=="Above_10S",]$micronec_epi)
## [1] 0.3676887
Related to Table B2 (Sensitivity analysis of the sampled size)
```

```
# Sampled size = 50
wilcox.test(df eff[df eff$threshold=="Fisheries" & df eff$Zone=="Moz",]$chlamean,
            df_eff_new_50[df_eff_new_50$Zone=="Moz",]$chlamean)
##
## Wilcoxon rank sum test with continuity correction
## data: df_eff[df_eff$threshold == "Fisheries" & df_eff$Zone == "Moz", and df_eff_new_50[df_eff_new_
## W = 29183, p-value = 1.17e-08
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Above_10S",]$chlamean,
            df_eff_new_50[df_eff_new_50$Zone=="Above_10S",]$chlamean)
##
## Wilcoxon rank sum test with continuity correction
## data: df_eff[df_eff$threshold == "Fisheries" & df_eff$Zone == "Above_10S", and df_eff_new_50[df_ef
## W = 1269572, p-value = 6.509e-12
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Moz",]$slamean,
            df_eff_new_50[df_eff_new_50$Zone=="Moz",]$slamean)
##
## Wilcoxon rank sum test with continuity correction
## data: df_eff[df_eff$threshold == "Fisheries" & df_eff$Zone == "Moz", and df_eff_new_50[df_eff_new_
## W = 12822, p-value = 0.0005888
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Above_10S",]$slamean,
            df_eff_new_50[df_eff_new_50$Zone=="Above_10S",]$slamean)
## Wilcoxon rank sum test with continuity correction
## data: df_eff[df_eff$threshold == "Fisheries" & df_eff$Zone == "Above_10S", and df_eff_new_50[df_ef
## W = 1048082, p-value = 0.05296
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Moz",]$SSCImean,
            df_eff_new_50[df_eff_new_50$Zone=="Moz",]$SSCImean)
## Wilcoxon rank sum test with continuity correction
## data: df_eff[df_eff$threshold == "Fisheries" & df_eff$Zone == "Moz", and df_eff_new_50[df_eff_new_
## W = 22457, p-value = 0.05165
## alternative hypothesis: true location shift is not equal to 0
```

```
wilcox.test(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Above_10S",]$SSCImean,
            df_eff_new_50[df_eff_new_50$Zone=="Above_10S",]$SSCImean)
## Wilcoxon rank sum test with continuity correction
## data: df_eff[df_eff$threshold == "Fisheries" & df_eff$Zone == "Above_10S", and df_eff_new_50[df_ef
## W = 1152155, p-value = 0.02772
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(df eff[df eff$threshold=="Fisheries" & df eff$Zone=="Moz",]$FSLEmean,
            df_eff_new_50[df_eff_new_50$Zone=="Moz",]$FSLEmean)
##
## Wilcoxon rank sum test with continuity correction
## data: df_eff[df_eff$threshold == "Fisheries" & df_eff$Zone == "Moz", and df_eff_new_50[df_eff_new_
## W = 8040, p-value = 1.006e-09
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Above_10S",]$FSLEmean,
            df_eff_new_50[df_eff_new_50$Zone=="Above_10S",]$FSLEmean)
##
## Wilcoxon rank sum test with continuity correction
##
## data: df_eff[df_eff$threshold == "Fisheries" & df_eff$Zone == "Above_10S", and df_eff_new_50[df_ef
## W = 1154138, p-value = 0.02261
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(df eff[df eff$threshold=="Fisheries" & df eff$Zone=="Moz",]$micronec epi,
            df_eff_new_50[df_eff_new_50$Zone=="Moz",]$micronec_epi)
##
## Wilcoxon rank sum test with continuity correction
## data: df_eff[df_eff$threshold == "Fisheries" & df_eff$Zone == "Moz", and df_eff_new_50[df_eff_new_
## W = 24796, p-value = 0.001142
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Above_10S",]$micronec_epi,
            df_eff_new_50[df_eff_new_50$Zone=="Above_10S",]$micronec_epi)
##
## Wilcoxon rank sum test with continuity correction
## data: df eff[df eff$threshold == "Fisheries" & df eff$Zone == "Above 10S", and df eff new 50[df ef
## W = 1118172, p-value = 0.395
## alternative hypothesis: true location shift is not equal to 0
```

```
# Sampled size = 150
wilcox.test(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Moz",]$chlamean,
            df_eff_new_150[df_eff_new_150$Zone=="Moz",]$chlamean)
##
## Wilcoxon rank sum test with continuity correction
## data: df_eff[df_eff$threshold == "Fisheries" & df_eff$Zone == "Moz", and df_eff_new_150[df_eff_new
## W = 86969, p-value = 3.262e-08
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Above_10S",]$chlamean,
            df_eff_new_150[df_eff_new_150$Zone=="Above_10S",]$chlamean)
##
## Wilcoxon rank sum test with continuity correction
## data: df_eff[df_eff$threshold == "Fisheries" & df_eff$Zone == "Above_10S", and df_eff_new_150[df_e
## W = 3787786, p-value = 1.64e-13
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Moz",]$slamean,
            df_eff_new_150[df_eff_new_150$Zone=="Moz",]$slamean)
##
## Wilcoxon rank sum test with continuity correction
## data: df_eff[df_eff$threshold == "Fisheries" & df_eff$Zone == "Moz", and df_eff_new_150[df_eff_new
## W = 39283, p-value = 0.0006875
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Above_10S",]$slamean,
            df_eff_new_150[df_eff_new_150$Zone=="Above_10S",]$slamean)
## Wilcoxon rank sum test with continuity correction
## data: df_eff[df_eff$threshold == "Fisheries" & df_eff$Zone == "Above_10S", and df_eff_new_150[df_e
## W = 3151606, p-value = 0.06359
\mbox{\tt \#\#} alternative hypothesis: true location shift is not equal to 0
wilcox.test(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Moz",]$SSCImean,
            df_eff_new_150[df_eff_new_150$Zone=="Moz",]$SSCImean)
## Wilcoxon rank sum test with continuity correction
## data: df_eff[df_eff$threshold == "Fisheries" & df_eff$Zone == "Moz", and df_eff_new_150[df_eff_new
## W = 68544, p-value = 0.03756
## alternative hypothesis: true location shift is not equal to 0
```

```
wilcox.test(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Above_10S",]$SSCImean,
            df_eff_new_150[df_eff_new_150$Zone=="Above_10S",]$SSCImean)
##
## Wilcoxon rank sum test with continuity correction
## data: df_eff[df_eff$threshold == "Fisheries" & df_eff$Zone == "Above_10S", and df_eff_new_150[df_e
## W = 3428013, p-value = 0.03114
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Moz",]$FSLEmean,
            df_eff_new_150[df_eff_new_150$Zone=="Moz",]$FSLEmean)
##
## Wilcoxon rank sum test with continuity correction
##
## data: df_eff[df_eff$threshold == "Fisheries" & df_eff$Zone == "Moz", and df_eff_new_150[df_eff_new
## W = 24954, p-value = 1.238e-09
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Above_10S",]$FSLEmean,
            df_eff_new_150[df_eff_new_150$Zone=="Above_10S",]$FSLEmean)
##
  Wilcoxon rank sum test with continuity correction
## data: df_eff[df_eff$threshold == "Fisheries" & df_eff$Zone == "Above_10S", and df_eff_new_150[df_e
## W = 3381546, p-value = 0.1386
## alternative hypothesis: true location shift is not equal to 0
wilcox.test(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Moz",]$micronec_epi,
            df_eff_new_150[df_eff_new_150$Zone=="Moz",]$micronec_epi)
##
## Wilcoxon rank sum test with continuity correction
## data: df_eff[df_eff$threshold == "Fisheries" & df_eff$Zone == "Moz", and df_eff_new_150[df_eff_new
## W = 74139, p-value = 0.00177
\mbox{\tt \#\#} alternative hypothesis: true location shift is not equal to 0
wilcox.test(df_eff[df_eff$threshold=="Fisheries" & df_eff$Zone=="Above_10S",]$micronec_epi,
            df_eff_new_150[df_eff_new_150$Zone=="Above_10S",]$micronec_epi)
## Wilcoxon rank sum test with continuity correction
## data: df eff[df eff$threshold == "Fisheries" & df eff$Zone == "Above 10S", and df eff new 150[df e
## W = 3388090, p-value = 0.115
## alternative hypothesis: true location shift is not equal to 0
```