## SpatCourse\_diversity

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## load packages

```
#install.packages('hillR')
library(hillR)

library(vegan)
library(tidyverse)
library(FD)

#install.packages('MeanRarity')
library(MeanRarity)
```

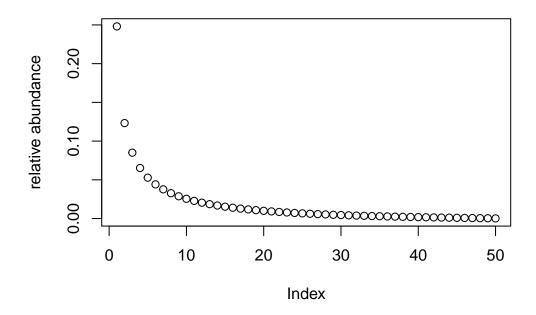
## Diversity measures

We explore how the classic diversity measures relate to the modern hill-numbers based metrics, which reflect the effective number of species in the community

We first simulate a community with known SAD (species abundance distribution), and calculate taxon richness, and classic Shannon, and Simpson diversity. Then we remove half of the taxa (randomly) and re-quantify diversity metrics, using both classic and hill-numbers based diversity metrics.

```
# using the 'MeanRarity::fit_SAD' function we simulate community with 50 spp
#and known abundance distribution
test.community1=data.frame(t(fit_SAD(rich=50, simpson = 10)$rel_abundances))
colnames(test.community1)=paste("sp",1:50)
plot(t(test.community1), ylab='relative abundance', main='test community1')
```

## test community1



let's calculate classic Shannon and Simpson diversity

```
diversity(test.community1, 'shannon')
```

## [1] 2.908134

```
diversity(test.community1, 'simpson')
```

## [1] 0.8999999

Now, lets remove 50% of species, randomly [like a large disturbance] Then calculate classic diversity for this halved community of 25 species

```
set.seed(1)
test.community2=sample(test.community1, 25); test.community2=sort(test.community2, decreasing = T)
diversity(test.community2, 'shannon')
```

## [1] 2.220554

```
diversity(test.community2, 'simpson')
```

## [1] 0.8123673

While the richness has declined by 50%, Shannon and Simpson diversity showed only modest decline; Let's try the same using the *effective number of species* hill-based diversity Hill-Shannon

```
renyi(test.community1, scale=1, hill=T) # this would be Shannon (q=1)
##
          1
## 18.32258
## attr(,"class")
## [1] "renyi"
                  "numeric"
renyi(test.community2, scale=1, hill=T)
##
          1
## 9.212432
## attr(,"class")
## [1] "renyi"
                  "numeric"
So hill-Shannon declines by \sim 50\% (from c.18 to c.9), mirroring the decline in species number;
Let' see Hill-Simpson.
renyi(test.community1, scale=2, hill=T) # this would be Simpson (q=2)
## 9.99999
## attr(,"class")
## [1] "renyi"
                  "numeric"
renyi(test.community2, scale=2, hill=T)
##
          2
## 5.329563
## attr(,"class")
## [1] "renyi"
                  "numeric"
```

The Hill-based conversion linearises the relation between taxonomic richness and diversity, making them more comparable.

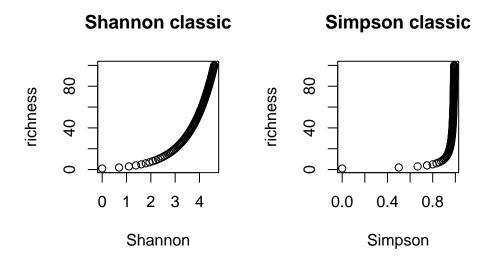
Hill-Simpson also declined by  $\sim 50\%$  (from c.10 to c.5).

We can visualise this simulating a community with 100 species with equal abundance = 1 ind. Call it 'comm1'

```
comm1=data.frame(t(rep(1,100))); colnames(comm1)=paste("sp",1:100)# simulated community of 100 spp
# calculate shannon classsic and hill-shannon for each species in the community
shannon.test=data.frame(matrix(ncol=5, nrow=100))# matrix to hold the values
shannon.test[,1]=seq(1,100,1)
for(i in 1:100) shannon.test[i,2]= diversity(comm1[1:i], 'shannon')
for(i in 1:100) shannon.test[i,3]= diversity(comm1[1:i], 'simpson')
for(i in 1:100) shannon.test[i,4]= renyi(comm1[1:i], scale=1, hill=T)
for(i in 1:100) shannon.test[i,5]= renyi(comm1[1:i], scale=2, hill=T)
names(shannon.test)=c ('richness', 'Shannon', 'Simpson', 'Hill_Shannon', 'Hill_Simpson')
```

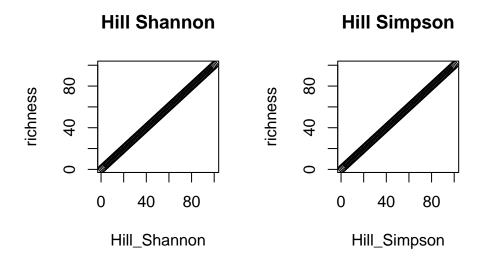
Plot classic Shannon & Simpson vs Richness.

```
par(mfrow=c(1,2))
plot(richness~Shannon, shannon.test, main='Shannon classic')
plot(richness~Simpson, shannon.test, main='Simpson classic')
```



Plot Hill-Shannon & Hill-Simpson vs Richness.

```
par(mfrow=c(1,2))
plot(richness~Hill_Shannon, shannon.test, main='Hill Shannon')
plot(richness~Hill_Simpson, shannon.test, main='Hill Simpson')
```



The same approach can be used for quantifying functional diversity

This can be done using the 'hillR' package

First, verify that diversity from the 'hillR' package are the same as those calculated with vegan::renyi

```
renyi(test.community1, scale=1, hill=T)
##
          1
## 18.32258
## attr(,"class")
## [1] "renyi"
                 "numeric"
hill_taxa(test.community1, q=1)
## [1] 18.32258
```

Yes! GOOD; Now lets see how it works on dummy communities with functional traits as well.

We can generate some fake trait values for the comm1.

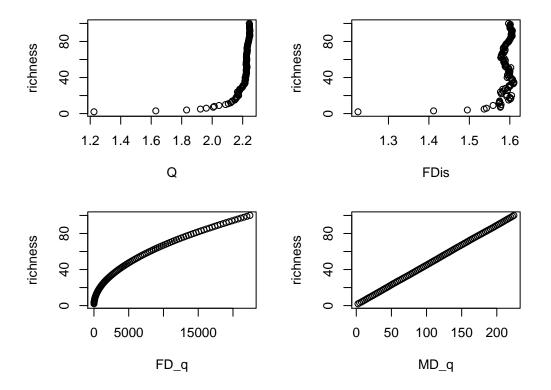
```
set.seed(1)
comm1.trt=data.frame(trt1=rpois(100, 500), trt2=sample(1:100, 100), trt3=rgamma(100, 2,1))
row.names(comm1.trt)= colnames(comm1) # need to provide the same sp names on the trait data
```

Calculate the range of functional diversities provided by hillR, over each richness value

```
# calculate some different functional diversity metrics; use q=1 to express it as hill-shannon
FD.test=data.frame(matrix(ncol=6, nrow=100))# matrix to hold the values
FD.test[,1] = seq(1,100,1)
for(i in 2:100) FD.test[i,2] = hill_func(comm1[1:i], comm1.trt[c(1:i),], q=1)[1]# Q
for(i in 2:100) FD.test[i,3] = hill_func(comm1[1:i], comm1.trt[c(1:i),], q=1)[2] # FDis
for(i in 2:100) FD.test[i,4]= hill_func(comm1[1:i], comm1.trt[c(1:i),], q=1)[3] # D_q
for(i in 2:100) FD.test[i,5] = hill_func(comm1[1:i], comm1.trt[c(1:i),], q=1)[4] # MD_q
for(i in 2:100) FD.test[i,6] = hill_func(comm1[1:i], comm1.trt[c(1:i),], q=1)[5] # FD_q
names(FD.test)=c ('richness', 'Q', 'FDis', 'D_q', 'MD_q', 'FD_q')
```

Let's plot the relation between number of taxa and functional diversities

```
par(mfrow=c(2,2), mar = c(4, 4, 2, 2))
plot(richness~Q, data=FD.test)
plot(richness~FDis, data=FD.test)
plot(richness~FD_q, FD.test)
plot(richness~MD_q, FD.test)
```



Here the results of course depends on the trait values of the species.

However, different patterns emerge looking at classic functional diversity measure (Q, FDis) relative to hill-based diversity measures (D\_q: functional hill number, the effective number of equally abundant and functionally equally distinct species; FD\_q: total functional diversity, the effective total functional distance between species of the assemblage). Refer to Chiu & Chao (2014) PLoS ONE for details on these metrics.