

# Sunset Crater Rock Lichen Community Composition

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## **1 Questions for Co-Authors**

1. Were all measurements a part of the same quadrat?
2. I.E., are lichen/bryophyte measurements absolute or a proportion of available habitat?
3. Rikke and Cameron, how were the lichen/bryophyte measurements done?
4. Richard, how were the plant and light measurements done?

## **2 Things to Do**

- Check Resistance vs Susceptible in barplots
- Check on bird preference
- Compile trait data from database

- Check on co-occurrence
- Do indicator species sems or summed species sems
- multiple resgresion for variables on community

### 3 Do rock lichen communities respond to phenotypic variation in a foundation species?

#### 3.1 Load and Pre-Process Data

```
> key <- read.csv('/Users/Aeolus/Documents/Active_Projects/Sunset_Crater_Lichens/data/
> x <- read.csv('/Users/Aeolus/Documents/Active_Projects/Sunset_Crater_Lichens/data/
>                                     #remove dead
> x <- x[x$Live.Dead == 1,]
> com <- x[,((1:ncol(x))[colnames(x) == 'Acacon']):((1:ncol(x))[colnames(x) == 'Xane]
> env <- x[,1:12]
>                                     #remove N and S light
> env <- env[,-10:-11]
>                                     #flip R and
> ## env$Moth[env$Moth==0] <- 2
> ## env$Moth[env$Moth==1] <- 0
> ## env$Moth[env$Moth==2] <- 1
>                                     #fix colnames
> colnames(env) <- sub('\\.\\.\\.',' ',colnames(env))
> colnames(env)
```

```

[1] "Tree.pairs"      "Moth"            "Live.Dead"      "Litter"
[5] "Big.rock"        "Small.rock"      "Shrubs"         "Grass"
[9] "Branches"        "Light.average"

>

```

### 3.2 Analysis of Moth Affects on "Env" Variables

```

> shapiro.test(fit.env$residuals)

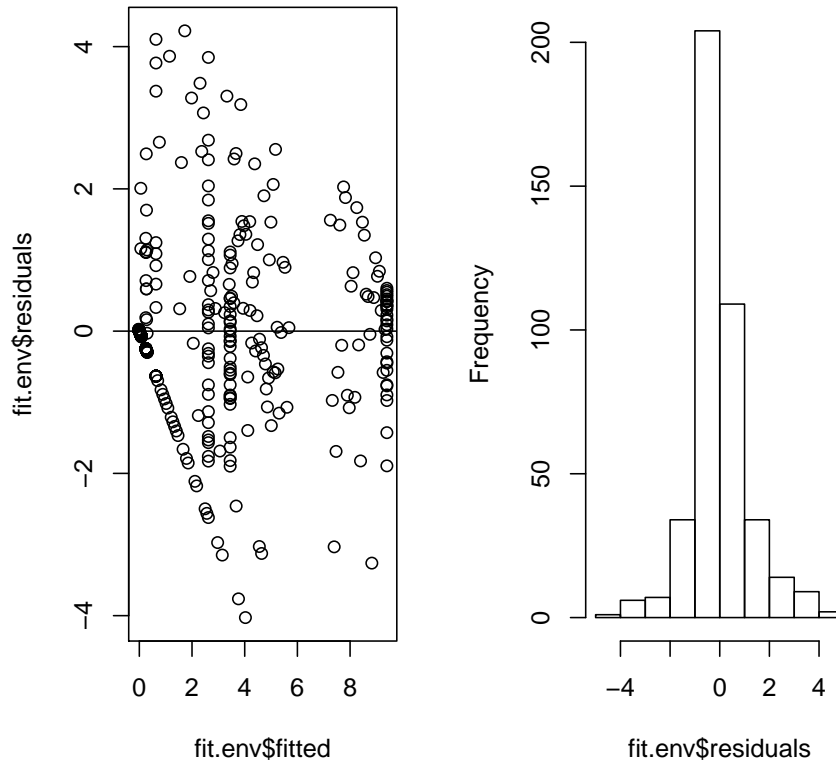
      Shapiro-Wilk normality test

data:  fit.env$residuals
W = 0.9019, p-value = 8.339e-16

> par(mfrow=c(1,2))
> plot(fit.env$residuals~fit.env$fitted)
> abline(h=0)
> hist(fit.env$residuals)
>

```

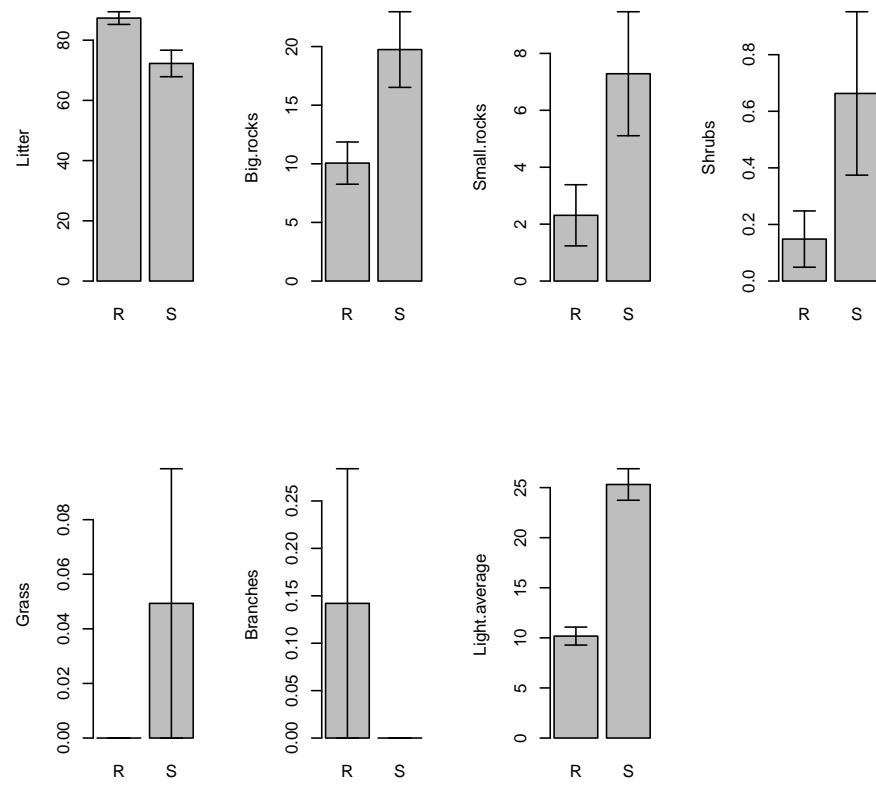
Histogram of fit.env\$residuals:



```
> library(gplots)
> par(mfrow=c(2,4))
> for (i in 4:ncol(env)){
+   mu <- tapply(env[,i],env$Moth,mean)
+   se <- tapply(env[,i],env$Moth,function(x)sd(x)/sqrt(length(x)))
+   x.names <- unique(env$Moth)
+   x.names[x.names==1] <- 'R'
+   x.names[x.names==0] <- 'S'
+   barplot2(mu,plot.ci=TRUE,ci.l=mu-se,ci.u=mu+se,ylab=colnames(env)[i],names=x.names)
```

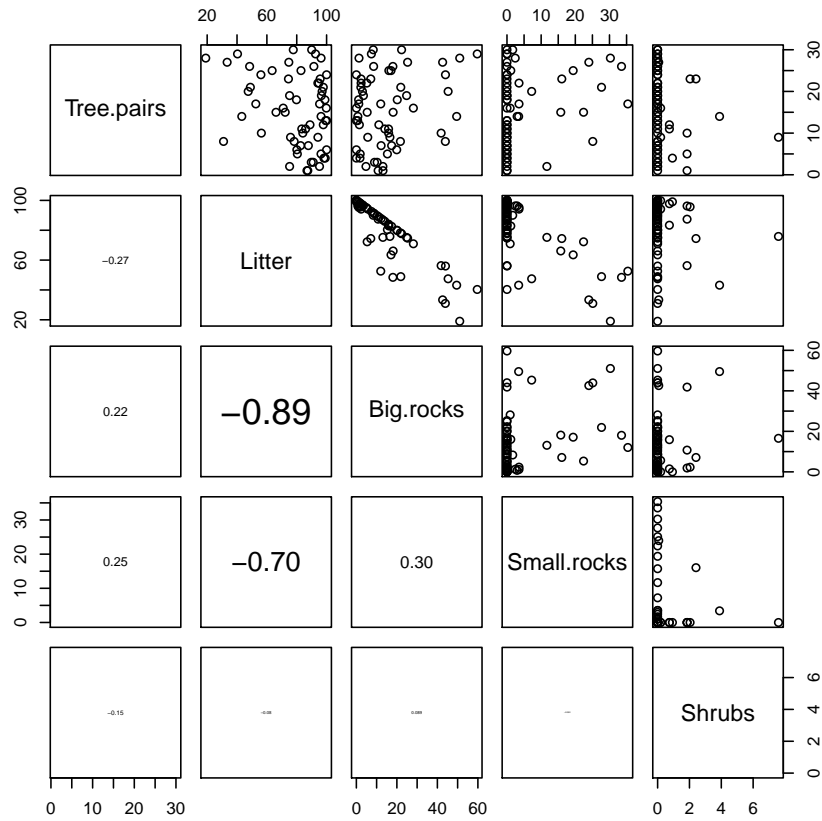
```
+ }
```

```
>
```



```
> pairs(env[,c(-2:-3,-8:-11)],lower.panel=panel.cor)
```

```
>
```



## 4 Community Response to Moth

### 4.1 Sampling

```
> library(vegan)
> spac <- specaccum(com)
> plot(spac,xlab='Number of Trees',ylab='Number of Lichen Species')
>
```

## 4.2 Abundance, Richness and Diversity

```
>                                     #abundance
> A <- apply(com,1,sum)
> A.fit <- glm(log(A+1)~Moth:Tree.pairs,data=env)
>                                     #richness
> R <- apply(com,1,function(x) length(x[x!=0]))
> R.fit <- glm(R~Moth:Tree.pairs,data=env,family='poisson')
>                                     #Shannon's diversity
> H <- apply(com,1,diversity)
> H.fit <- glm(H~Moth:Tree.pairs,data=env)
>
> library(xtable)
> xtable(summary(A.fit))
```

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	0.6388	0.1172	5.45	0.0000
Moth:Tree.pairs	0.0182	0.0093	1.95	0.0555

```
> xtable(summary(R.fit))
```

```
> xtable(summary(H.fit))
```

```
>
```

	Estimate	Std. Error	z value	Pr(> z )
(Intercept)	1.3234	0.0832	15.91	0.0000
Moth:Tree.pairs	0.0209	0.0057	3.66	0.0002

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	0.7461	0.1074	6.94	0.0000
Moth:Tree.pairs	0.0205	0.0086	2.40	0.0197

### 4.3 Composition

```
>                                     #paired test of composition
> ds <- rep(min(com[com!=0]),nrow(com))
> com. <- cbind(com,ds)
> adonis(com.~env$Moth:env$Tree.pairs)
```

Call:

```
adonis(formula = com. ~ env$Moth:env$Tree.pairs)
```

Terms added sequentially (first to last)

	Df	SumsOfSqs	MeanSqs	F.Model	R2	Pr(>F)
env\$Moth:env\$Tree.pairs	1	0.8211	0.82114	2.5772	0.04254	0.037 *
Residuals	58	18.4799	0.31862		0.95746	
Total	59	19.3010			1.00000	

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1



```

> ##                                     #mvabund
> ## attach(env)
> ## mva.com <- com[,apply(com,2,sum)!=0]
> ##                                     #many glm with gaussian distribution
> ## mva.com <- apply(mva.com,2,sqrt)
> ## mva.com <- mvabund(mva.com)
> ## glm.fit <- manyglm(mva.com~Moth:Tree.pairs,family='gaussian')
> ## detach(env)
> ## anova(glm.fit)
>
>                                     #paired distance based
> d. <- as.matrix(vegdist(com.))
> pd <- 0
> for (i in 1:(nrow(d.)-1)){
+   pd[i] <- d.[(i+1),i]
+ }
> pd. <- pd[(1:length(pd)) %% 2 == 1]
> wilcox.test(pd.,exact=FALSE)

      Wilcoxon signed rank test with continuity correction

data:  pd.
V = 406, p-value = 4.003e-06
alternative hypothesis: true location is not equal to 0

> env.. <- env[(1:nrow(env)) %% 2 == 1,]
> library(tweedie)

```

```
> glm. <- glm(pd.~Light.average*Litter*Big.ocks,data=env...,family=tweedie(2))
>
```

```
> xtable(summary(glm.))
```

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	-0.8025	3.4980	-0.23	0.8207
Light.average	0.1093	0.1556	0.70	0.4897
Litter	0.0339	0.0395	0.86	0.4001
Big.ocks	0.0378	0.0733	0.51	0.6117
Light.average:Litter	-0.0016	0.0017	-0.91	0.3724
Light.average:Big.ocks	-0.0023	0.0031	-0.76	0.4529
Litter:Big.ocks	-0.0007	0.0011	-0.68	0.5009
Light.average:Litter:Big.ocks	0.0000	0.0000	0.95	0.3503

```
>
```

## 4.4 Indicator Species Analysis

```
> #indicator species
> ## library(labdsv)
> ## ind.spp <- indval(com,env$Moth)
> ## summary(ind.spp)
> ## detach(package:labdsv)
>
```

Indicator Species Analysis Results using Moth as the grouping factor

species cluster indicator.value probability

Canros 2 0.6397 0.006

Acasup 2 0.6295 0.002

Acacon 2 0.4769 0.001

Acaobp 2 0.4241 0.008

Phydub 2 0.4125 0.018

Calare 2 0.2966 0.036

```
> #NMDS
> library(ecodist)
> d <- vegdist(com.)
> if (any(ls()=='my.nmds')){ }else{my.nmds <- nmds(d,3,3,100)}

NULL

> env.nms <- env.[,-c(3,4,5,6)]
> par(mfrow=c(3,3))
> for (i in 1:3){
+   for (j in 1:3){
+     if (i!=j){
+       par(mar=c(5.1, 4.1, 4.1, 2.1)-0.75)
+       vectors <- envfit(nmds.min(my.nmds)[,c(i,j)]~env.nms)
```

```

+       plot(nmds.min(my.nmds)[,c(j,i)],pch=19,col=grey(c(0.75,0))[(env$Moth+1)],x1
+       plot(vectors,col='black')
+     }else{
+       par(mar=c(5.1, 4.1, 4.1, 2.1)-2)
+       plot(1,1,axes=FALSE,xlab='',ylab='',type='n')
+       text(1,1,labels=paste('X',i,sep=''),cex=10)
+     }
+   }
+ }

```

Minimum stress for given dimensionality: 0.1487555

r<sup>2</sup> for minimum stress configuration: 0.8666012

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Minimum stress for given dimensionality: 0.1487555

$r^2$  for minimum stress configuration: 0.8666012

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Minimum stress for given dimensionality: 0.1487555

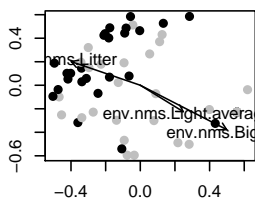
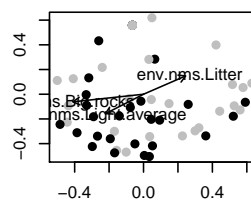
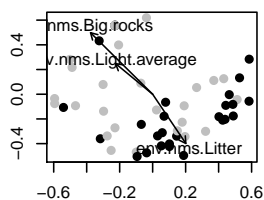
$r^2$  for minimum stress configuration: 0.8666012

Minimum stress for given dimensionality: 0.1487555

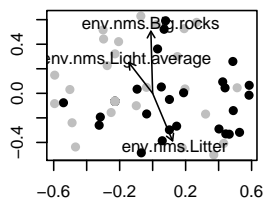
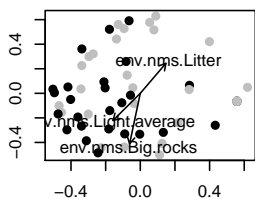
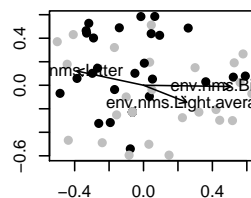
$r^2$  for minimum stress configuration: 0.8666012

>

X1



X2



X3

## 4.5 SEM

Model Chisquare = 5.0513 Df = 4 Pr(>Chisq) = 0.28207

Chisquare (null model) = 158.65 Df = 10

Goodness-of-fit index = 0.96765

Adjusted goodness-of-fit index = 0.87868

RMSEA index = 0.066744 90% CI: (NA, 0.21702)

Bentler-Bonnett NFI = 0.96816

Tucker-Lewis NNFI = 0.98232

Bentler CFI = 0.99293

SRMR = 0.039061

AIC = 27.051

AICc = 10.551

BIC = 50.089

CAIC = -15.326

### Normalized Residuals

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
-0.9340	0.0000	0.0000	-0.0639	0.0297	0.3070

### R-square for Endogenous Variables

Light.average	Litter	Abundance	Big.rocks
0.5583	0.1442	0.3360	0.7041

### Parameter Estimates

Estimate	Std Error	z value	Pr(> z )
----------	-----------	---------	----------

```

g.1.2  1.86246 0.215650      8.63652 5.7952e-18 Light.average <--- Moth
g.1.3 -0.96225 0.305240     -3.15244 1.6191e-03 Litter <--- Moth
b.2.5  0.26986 0.226461      1.19165 2.3340e-01 Abundance <--- Light.average
b.3.4 -1.32422 0.111764    -11.84832 2.1954e-32 Big.rocks <--- Litter
b.3.5  0.11105 0.397655      0.27926 7.8004e-01 Abundance <--- Litter
b.4.5  0.74057 0.248775      2.97687 2.9121e-03 Abundance <--- Big.rocks
e.1     0.25424 0.046809      5.43139 5.5917e-08 Moth <--> Moth
e.2     0.69757 0.128434      5.43139 5.5917e-08 Light.average <--> Light.average
e.3     1.39757 0.257314      5.43139 5.5917e-08 Litter <--> Litter
e.4     1.20348 0.221579      5.43139 5.5917e-08 Big.rocks <--> Big.rocks
e.5     4.39445 0.809084      5.43139 5.5917e-08 Abundance <--> Abundance

```

Iterations = 0

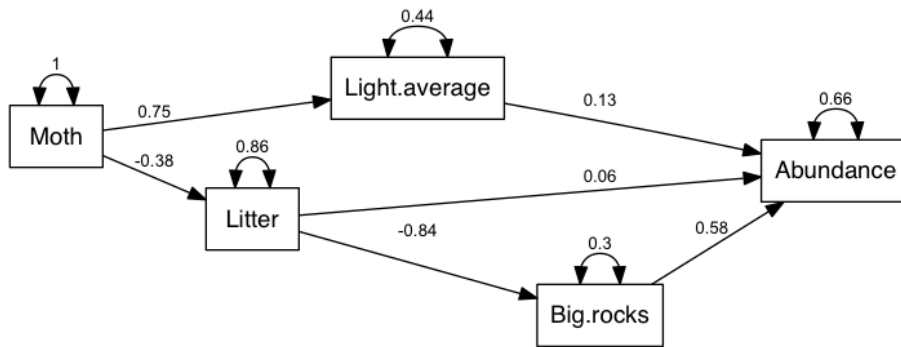
5 largest modification indices, A matrix:

Litter<-Light.average	Light.average<-Litter	Abundance<-Moth
2.494770	2.494770	1.254584
Big.rocks<-Abundance	Big.rocks<-Light.average	
1.137006	1.137006	

5 largest modification indices, P matrix:

Light.average<->Litter	Abundance<->Moth	Abundance<->Light.average
2.4947699	1.2545837	1.2545837
Abundance<->Litter	Big.rocks<->Moth	
1.2545837	0.8392215	

Running dot -Tpng -o semPathA.png semPathA.dot





Model Chisquare = 4.2927 Df = 4 Pr(>Chisq) = 0.36785  
 Chisquare (null model) = 198.77 Df = 10  
 Goodness-of-fit index = 0.9727  
 Adjusted goodness-of-fit index = 0.89762  
 RMSEA index = 0.035215 90% CI: (NA, 0.20261)  
 Bentler-Bonnett NFI = 0.9784  
 Tucker-Lewis NNFI = 0.99612  
 Bentler CFI = 0.99845  
 SRMR = 0.04022  
 AIC = 26.293  
 AICc = 9.7927  
 BIC = 49.33  
 CAIC = -16.085

#### Normalized Residuals

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
-0.934	-0.216	0.000	-0.141	0.000	0.252

#### R-square for Endogenous Variables

Light.average	Litter	Richness	Big.rocks
0.5583	0.1442	0.6740	0.7041

#### Parameter Estimates

	Estimate	Std Error	z value	Pr(> z )	
g.1.2	1.86246	0.215650	8.6365	5.7952e-18	Light.average <--- Moth
g.1.3	-0.96225	0.305240	-3.1524	1.6191e-03	Litter <--- Moth

```

b.2.5  0.83401 0.200015    4.1697 3.0497e-05 Richness <--- Light.average
b.3.4 -1.32422 0.111764   -11.8483 2.1954e-32 Big.ocks <--- Litter
b.3.5  1.17773 0.351217    3.3533 7.9862e-04 Richness <--- Litter
b.4.5  1.68066 0.219723    7.6490 2.0257e-14 Richness <--- Big.ocks
e.1    0.25424 0.046809    5.4314 5.5917e-08 Moth <--> Moth
e.2    0.69757 0.128434    5.4314 5.5917e-08 Light.average <--> Light.average
e.3    1.39757 0.257314    5.4314 5.5917e-08 Litter <--> Litter
e.4    1.20348 0.221579    5.4314 5.5917e-08 Big.ocks <--> Big.ocks
e.5    3.42801 0.631147    5.4314 5.5917e-08 Richness <--> Richness

```

Iterations = 0

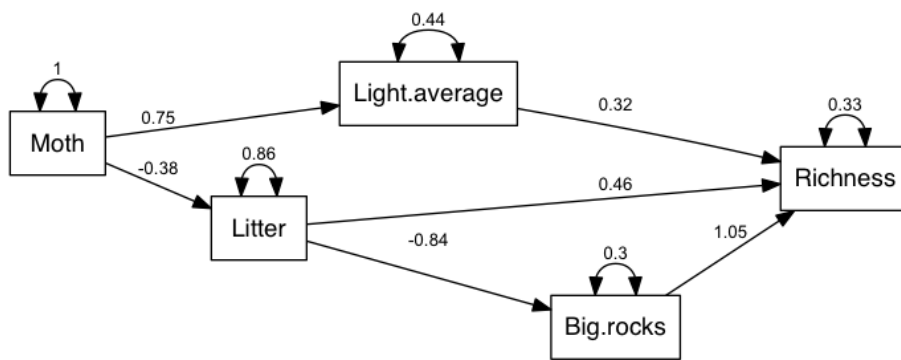
5 largest modification indices, A matrix:

Litter<-Light.average	Light.average<-Litter	Litter<-Richness
2.494770	2.494770	1.639637
Big.ocks<-Richness	Big.ocks<-Light.average	
1.137006	1.137006	

5 largest modification indices, P matrix:

Light.average<->Litter	Big.ocks<->Moth	Big.ocks<->Litter
2.4947699	0.8392215	0.8392215
Richness<->Moth	Richness<->Light.average	
0.4628788	0.4628788	

Running dot -Tpng -o semPathR.png semPathR.dot



Model Chisquare = 3.8579 Df = 4 Pr(>Chisq) = 0.42557  
 Chisquare (null model) = 183.64 Df = 10  
 Goodness-of-fit index = 0.97565  
 Adjusted goodness-of-fit index = 0.90868  
 RMSEA index = 0 90% CI: (NA, 0.19348)  
 Bentler-Bonnett NFI = 0.97899  
 Tucker-Lewis NNFI = 1.002  
 Bentler CFI = 1  
 SRMR = 0.042232  
 AIC = 25.858  
 AICc = 9.3579  
 BIC = 48.896  
 CAIC = -16.519

#### Normalized Residuals

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
-0.934	-0.271	0.000	-0.156	0.000	0.252

#### R-square for Endogenous Variables

Light.average	Litter	Diversity	Big.rocks
0.5583	0.1442	0.5808	0.7041

#### Parameter Estimates

	Estimate	Std Error	z value	Pr(> z )	
g.1.2	1.86246	0.215650	8.6365	5.7952e-18	Light.average <--- Moth
g.1.3	-0.96225	0.305240	-3.1524	1.6191e-03	Litter <--- Moth

```

b.2.5  0.16786 0.047902      3.5042 4.5794e-04 Diversity <--- Light.average
b.3.4 -1.32422 0.111764    -11.8483 2.1954e-32 Big.ocks <--- Litter
b.3.5  0.22462 0.084115      2.6704 7.5763e-03 Diversity <--- Litter
b.4.5  0.32511 0.052622      6.1782 6.4842e-10 Diversity <--- Big.ocks
e.1     0.25424 0.046809      5.4314 5.5917e-08 Moth <--> Moth
e.2     0.69757 0.128434      5.4314 5.5917e-08 Light.average <--> Light.average
e.3     1.39757 0.257314      5.4314 5.5917e-08 Litter <--> Litter
e.4     1.20348 0.221579      5.4314 5.5917e-08 Big.ocks <--> Big.ocks
e.5     0.19662 0.036201      5.4314 5.5917e-08 Diversity <--> Diversity

```

Iterations = 0

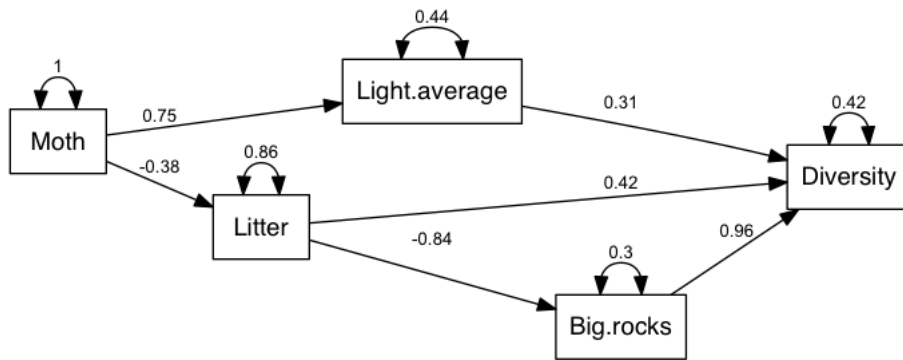
5 largest modification indices, A matrix:

Light.average<-Litter	Litter<-Light.average	Litter<-Diversity
2.494770	2.494770	2.420293
Big.ocks<-Light.average	Big.ocks<-Diversity	
1.137006	1.137006	

5 largest modification indices, P matrix:

Light.average<->Litter	Big.ocks<->Moth	Big.ocks<->Litter
2.49476987	0.83922155	0.83922155
Light.average<->Big.ocks	Diversity<->Litter	
0.34301774	0.00461226	

Running dot -Tpng -o semPathD.png semPathD.dot



Minimum stress for given dimensionality: 0.1487555

$r^2$  for minimum stress configuration: 0.8666012

Model Chisquare = 8.4011 Df = 9 Pr(>Chisq) = 0.49428

Chisquare (null model) = 203.09 Df = 21

Goodness-of-fit index = 0.95968

Adjusted goodness-of-fit index = 0.87455

RMSEA index = 0 90% CI: (NA, 0.13967)

Bentler-Bonnett NFI = 0.95863

Tucker-Lewis NNFI = 1.0077

Bentler CFI = 1

SRMR = 0.047869

AIC = 46.401

AICc = 27.401

BIC = 86.194

CAIC = -37.448

#### Normalized Residuals

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
-0.9340	-0.1790	0.0000	-0.0171	0.1080	0.8890

#### R-square for Endogenous Variables

Light.average	Litter	X1	X2	X3
0.5583	0.1442	0.4217	0.4097	0.0521
Big.rocks				
0.7041				

# Parameter Estimates

	Estimate	Std Error	z value	Pr(> z )	
g.1.2	1.862463	0.215650	8.63652	5.7952e-18	Light.average <--- Moth
g.1.3	-0.962252	0.305240	-3.15244	1.6191e-03	Litter <--- Moth
b.2.5	-0.080703	0.029471	-2.73840	6.1739e-03	X1 <--- Light.average
b.2.6	0.054260	0.027169	1.99716	4.5808e-02	X2 <--- Light.average
b.2.7	-0.026432	0.031000	-0.85264	3.9386e-01	X3 <--- Light.average
b.3.4	-1.324220	0.111764	-11.84832	2.1954e-32	Big.rocks <--- Litter
b.3.5	-0.160718	0.051750	-3.10569	1.8984e-03	X1 <--- Litter
b.3.6	0.067605	0.047707	1.41709	1.5646e-01	X2 <--- Litter
b.3.7	0.074578	0.054434	1.37007	1.7067e-01	X3 <--- Litter
b.4.5	-0.167030	0.032375	-5.15927	2.4791e-07	X1 <--- Big.rocks
b.4.6	0.124920	0.029846	4.18552	2.8451e-05	X2 <--- Big.rocks
b.4.7	0.035507	0.034054	1.04266	2.9711e-01	X3 <--- Big.rocks
e.1	0.254237	0.046809	5.43139	5.5917e-08	Moth <--> Moth
e.2	0.697573	0.128434	5.43139	5.5917e-08	Light.average <--> Light.average
e.3	1.397574	0.257314	5.43139	5.5917e-08	Litter <--> Litter
e.4	1.203481	0.221579	5.43139	5.5917e-08	Big.rocks <--> Big.rocks
e.5	0.074423	0.013702	5.43139	5.5917e-08	X1 <--> X1
e.6	0.063249	0.011645	5.43139	5.5917e-08	X2 <--> X2
e.7	0.082344	0.015161	5.43139	5.5917e-08	X3 <--> X3

Iterations = 0



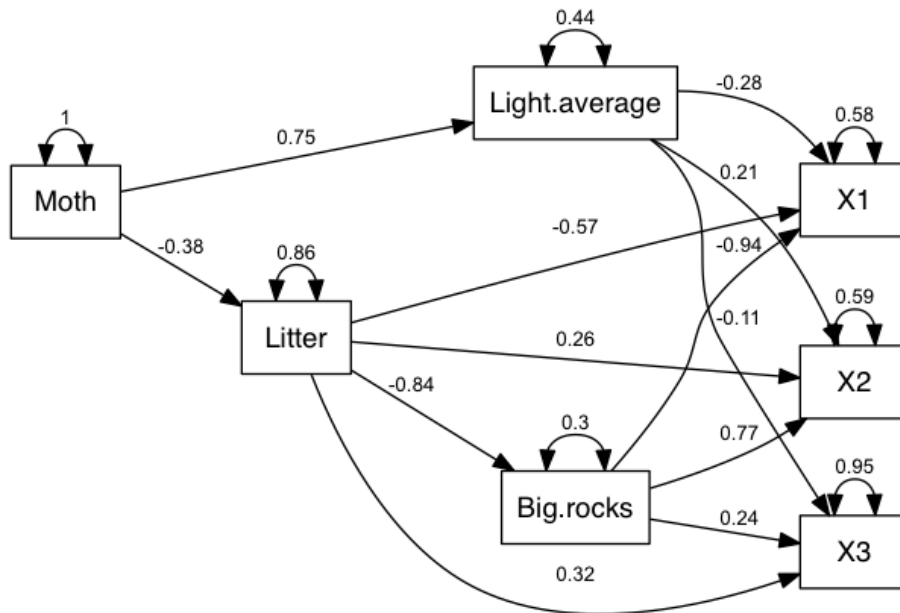
5 largest modification indices, A matrix:

Litter<-X1	Litter<-Light.average	Light.average<-Litter
3.438300	2.494770	2.494770
X1<-X2	X2<-X1	
2.245427	2.245427	

5 largest modification indices, P matrix:

Light.average<->Litter	X2<->X1	X2<->Moth
2.494770	2.245427	1.144694
X2<->Light.average	X2<->Litter	
1.144694	1.144694	

Running dot -Tpng -o semPath1.png semPath1.dot



Model Chisquare = 3.9132 Df = 4 Pr(>Chisq) = 0.41787

Chisquare (null model) = 166.87 Df = 10

Goodness-of-fit index = 0.97527  
 Adjusted goodness-of-fit index = 0.90726  
 RMSEA index = 0 90% CI: (NA, 0.19468)  
 Bentler-Bonnett NFI = 0.97655  
 Tucker-Lewis NNFI = 1.0014  
 Bentler CFI = 1  
 SRMR = 0.04181  
 AIC = 25.913  
 AICc = 9.4132  
 BIC = 48.951  
 CAIC = -16.464

#### Normalized Residuals

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
-0.9340	0.0000	0.0000	-0.0575	0.0736	0.4100

#### R-square for Endogenous Variables

Light.average	Litter	Acacon	Big.rocks
0.5583	0.1442	0.4175	0.7041

#### Parameter Estimates

	Estimate	Std Error	z value	Pr(> z )
g.1.2	1.8624634	0.2156498	8.63652	5.7952e-18
g.1.3	-0.9622521	0.3052403	-3.15244	1.6191e-03
b.2.5	0.0119617	0.0045513	2.62819	8.5840e-03
b.3.4	-1.3242197	0.1117643	-11.84832	2.1954e-32

```

b.3.5 -0.0186545 0.0079919 -2.33419 1.9586e-02
b.4.5 0.0025973 0.0049998 0.51949 6.0342e-01
e.1 0.2542373 0.0468089 5.43139 5.5917e-08
e.2 0.6975726 0.1284335 5.43139 5.5917e-08
e.3 1.3975744 0.2573143 5.43139 5.5917e-08
e.4 1.2034807 0.2215787 5.43139 5.5917e-08
e.5 0.0017750 0.0003268 5.43139 5.5917e-08

```

```

g.1.2 Light.average <--- Moth
g.1.3 Litter <--- Moth
b.2.5 Acacon <--- Light.average
b.3.4 Big.rocks <--- Litter
b.3.5 Acacon <--- Litter
b.4.5 Acacon <--- Big.rocks
e.1 Moth <--> Moth
e.2 Light.average <--> Light.average
e.3 Litter <--> Litter
e.4 Big.rocks <--> Big.rocks
e.5 Acacon <--> Acacon

```

Iterations = 0

5 largest modification indices, A matrix:

```

Litter<-Light.average    Light.average<-Litter Big.rocks<-Light.average
                2.4947699                2.4947699                1.1370060
Big.rocks<-Acacon Light.average<-Big.rocks

```

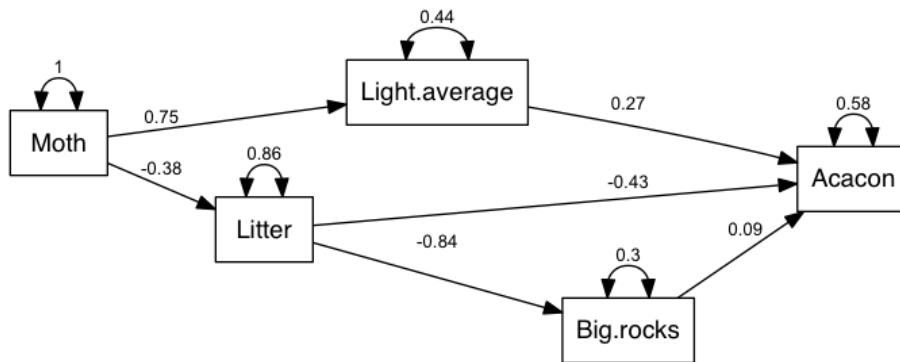
1.1370060

0.9165941

5 largest modification indices, P matrix:

Light.average $\leftrightarrow$ Litter	Big.rocks $\leftrightarrow$ Litter	Big.rocks $\leftrightarrow$ Moth
2.49476987	0.83922155	0.83922155
Light.average $\leftrightarrow$ Big.rocks	Acacon $\leftrightarrow$ Litter	
0.34301774	0.06310573	

Running dot -Tpng -o semPathAcacon.png semPathAcacon.dot



Model Chisquare = 3.9045 Df = 4 Pr(>Chisq) = 0.41908  
 Chisquare (null model) = 185.53 Df = 10  
 Goodness-of-fit index = 0.97533  
 Adjusted goodness-of-fit index = 0.90749  
 RMSEA index = 0 90% CI: (NA, 0.19449)  
 Bentler-Bonnett NFI = 0.97896  
 Tucker-Lewis NNFI = 1.0014  
 Bentler CFI = 1  
 SRMR = 0.04161  
 AIC = 25.905  
 AICc = 9.4045  
 BIC = 48.942  
 CAIC = -16.473

#### Normalized Residuals

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
-0.9340	-0.1620	0.0000	-0.0840	0.0431	0.3860

#### R-square for Endogenous Variables

Light.average	Litter	Acasup	Big.rocks
0.5583	0.1442	0.5794	0.7041

#### Parameter Estimates

	Estimate	Std Error	z value	Pr(> z )	
g.1.2	1.862463	0.2156498	8.6365	5.7952e-18	Light.average <--- Moth
g.1.3	-0.962252	0.3052403	-3.1524	1.6191e-03	Litter <--- Moth

```

b.2.5  0.036276 0.0156584  2.3167 2.0521e-02 Acasup <--- Light.average
b.3.4 -1.324220 0.1117643 -11.8483 2.1954e-32 Big.ocks <--- Litter
b.3.5 -0.056637 0.0274954 -2.0599 3.9411e-02 Acasup <--- Litter
b.4.5  0.042716 0.0172013  2.4833 1.3017e-02 Acasup <--- Big.ocks
e.1    0.254237 0.0468089  5.4314 5.5917e-08 Moth <--> Moth
e.2    0.697573 0.1284335  5.4314 5.5917e-08 Light.average <--> Light.average
e.3    1.397574 0.2573143  5.4314 5.5917e-08 Litter <--> Litter
e.4    1.203481 0.2215787  5.4314 5.5917e-08 Big.ocks <--> Big.ocks
e.5    0.021009 0.0038681  5.4314 5.5917e-08 Acasup <--> Acasup

```

Iterations = 0

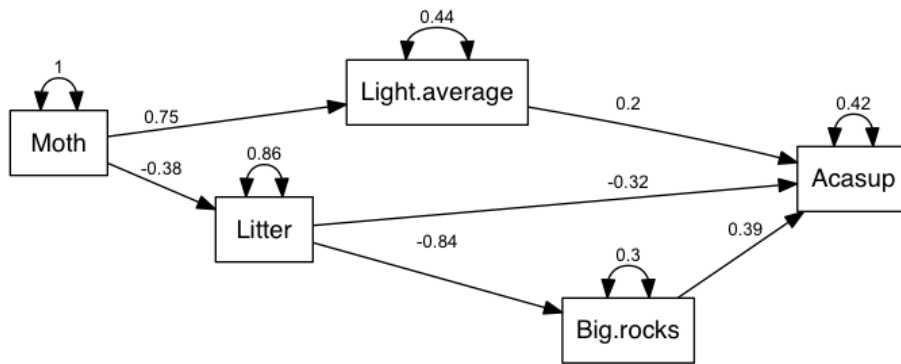
5 largest modification indices, A matrix:

Light.average<-Litter	Litter<-Light.average	Litter<-Acasup
2.494770	2.494770	1.982069
Light.average<-Acasup	Big.ocks<-Light.average	
1.397081	1.137006	

5 largest modification indices, P matrix:

Light.average<->Litter	Big.ocks<->Moth	Big.ocks<->Litter
2.49476987	0.83922155	0.83922155
Light.average<->Big.ocks	Acasup<->Litter	
0.34301774	0.05389307	

Running dot -Tpng -o semPathAcasup.png semPathAcasup.dot



Model Chisquare = 3.8536 Df = 3 Pr(>Chisq) = 0.27772  
 Chisquare (null model) = 142.14 Df = 10  
 Goodness-of-fit index = 0.97568  
 Adjusted goodness-of-fit index = 0.87839  
 RMSEA index = 0.069444 90% CI: (NA, 0.24056)  
 Bentler-Bonnett NFI = 0.97289  
 Tucker-Lewis NNFI = 0.97847  
 Bentler CFI = 0.99354  
 SRMR = 0.042811  
 AIC = 27.854  
 AICc = 10.492  
 BIC = 52.986  
 CAIC = -11.429

#### Normalized Residuals

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
-0.9340	-0.0505	0.0000	-0.0853	0.0000	0.5020

#### R-square for Endogenous Variables

Light.average	Litter	Acaobp	Big.rocks
0.5583	0.1442	0.1465	0.7041

#### Parameter Estimates

	Estimate	Std Error	z value	Pr(> z )	
g.1.2	1.862463	0.215650	8.63652	5.7952e-18	Light.average <--- Moth
g.1.3	-0.962252	0.305240	-3.15244	1.6191e-03	Litter <--- Moth



```

g.1.5  1.056837 0.384638    2.74761 6.0030e-03 Acaobp <--- Moth
b.2.5  -0.427636 0.148880   -2.87236 4.0742e-03 Acaobp <--- Light.average
b.3.4  -1.324220 0.111764  -11.84832 2.1954e-32 Big.rocks <--- Litter
b.3.5   0.034415 0.183282    0.18777 8.5106e-01 Acaobp <--- Litter
b.4.5   0.056794 0.113347    0.50106 6.1633e-01 Acaobp <--- Big.rocks
e.1     0.254237 0.046809    5.43139 5.5917e-08 Moth <--> Moth
e.2     0.697573 0.128434    5.43139 5.5917e-08 Light.average <--> Light.average
e.3     1.397574 0.257314    5.43139 5.5917e-08 Litter <--> Litter
e.4     1.203481 0.221579    5.43139 5.5917e-08 Big.rocks <--> Big.rocks
e.5     0.912248 0.167958    5.43139 5.5917e-08 Acaobp <--> Acaobp

```

Iterations = 0

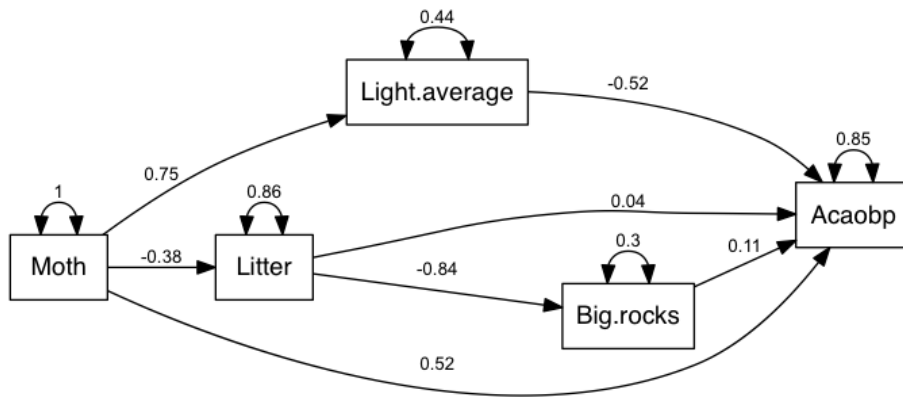
5 largest modification indices, A matrix:

Light.average<-Litter	Litter<-Light.average	Litter<-Acaobp
2.4947699	2.4947699	2.2967054
Big.rocks<-Light.average	Light.average<-Big.rocks	
1.1370060	0.9165941	

5 largest modification indices, P matrix:

Light.average<->Litter	Big.rocks<->Moth	Big.rocks<->Litter
2.4947699	0.8392215	0.8392215
Light.average<->Big.rocks	<NA>	
0.3430177	NA	

Running dot -Tpng -o semPathAcaobp.png semPathAcaobp.dot



Model Chisquare = 3.8661 Df = 4 Pr(>Chisq) = 0.42443  
 Chisquare (null model) = 191.38 Df = 10  
 Goodness-of-fit index = 0.97559  
 Adjusted goodness-of-fit index = 0.90847  
 RMSEA index = 0 90% CI: (NA, 0.19366)  
 Bentler-Bonnett NFI = 0.9798  
 Tucker-Lewis NNFI = 1.0018  
 Bentler CFI = 1  
 SRMR = 0.040632  
 AIC = 25.866  
 AICc = 9.3661  
 BIC = 48.904  
 CAIC = -16.511

#### Normalized Residuals

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
-0.9340	-0.2020	0.0000	-0.1080	0.0802	0.2520

#### R-square for Endogenous Variables

Light.average	Litter	Canros	Big.rocks
0.5583	0.1442	0.6211	0.7041

#### Parameter Estimates

	Estimate	Std Error	z value	Pr(> z )
g.1.2	1.8624634	0.215650	8.63652	5.7952e-18
g.1.3	-0.9622521	0.305240	-3.15244	1.6191e-03

```

b.2.5  0.1295623 0.028641    4.52371 6.0765e-06
b.3.4 -1.3242197 0.111764   -11.84832 2.1954e-32
b.3.5 -0.0083869 0.050292   -0.16676 8.6756e-01
b.4.5  0.1249703 0.031463    3.97200 7.1271e-05
e.1    0.2542373 0.046809    5.43139 5.5917e-08
e.2    0.6975726 0.128434    5.43139 5.5917e-08
e.3    1.3975744 0.257314    5.43139 5.5917e-08
e.4    1.2034807 0.221579    5.43139 5.5917e-08
e.5    0.0702887 0.012941    5.43139 5.5917e-08

```

```

g.1.2 Light.average <--- Moth
g.1.3 Litter <--- Moth
b.2.5 Canros <--- Light.average
b.3.4 Big.rockes <--- Litter
b.3.5 Canros <--- Litter
b.4.5 Canros <--- Big.rockes
e.1  Moth <--> Moth
e.2  Light.average <--> Light.average
e.3  Litter <--> Litter
e.4  Big.rockes <--> Big.rockes
e.5  Canros <--> Canros

```

```

Iterations = 0

```

5 largest modification indices, A matrix:

```

Light.average<-Litter    Litter<-Light.average    Litter<-Canros

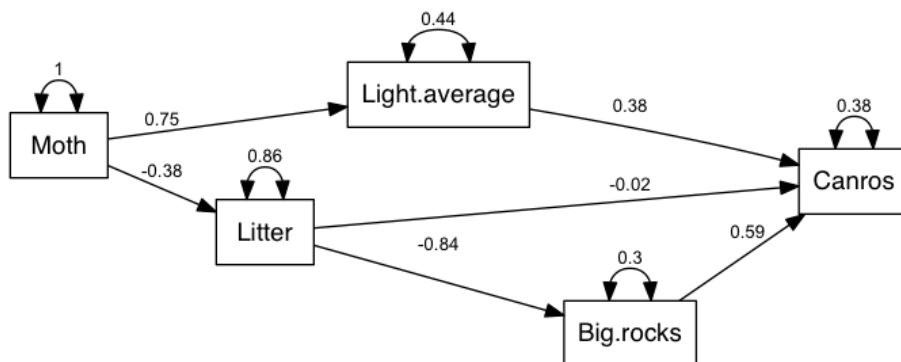
```

	2.494770	2.494770	2.299203
Big.rocks<-Light.average		Big.rocks<-Canros	
	1.137006		1.137006

5 largest modification indices, P matrix:

Light.average<->Litter	Big.rocks<->Moth	Big.rocks<->Litter
2.49476987	0.83922155	0.83922155
Light.average<->Big.rocks	Canros<->Moth	
0.34301774	0.01323831	

Running dot -Tpng -o semPathCanros.png semPathCanros.dot



Model Chisquare = 5.0461 Df = 4 Pr(>Chisq) = 0.2826  
 Chisquare (null model) = 138.66 Df = 10  
 Goodness-of-fit index = 0.96768  
 Adjusted goodness-of-fit index = 0.87881  
 RMSEA index = 0.066579 90% CI: (NA, 0.21693)  
 Bentler-Bonnett NFI = 0.96361  
 Tucker-Lewis NNFI = 0.97967  
 Bentler CFI = 0.99187  
 SRMR = 0.042323  
 AIC = 27.046  
 AICc = 10.546  
 BIC = 50.084  
 CAIC = -15.331

#### Normalized Residuals

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
-0.9340	-0.1910	0.0000	-0.0819	0.0000	0.5010

#### R-square for Endogenous Variables

Light.average	Litter	Calare	Big.rocks
0.5583	0.1442	0.0825	0.7041

#### Parameter Estimates

	Estimate	Std Error	z value	Pr(> z )
g.1.2	1.8624634	0.21564981	8.6365	5.7952e-18
g.1.3	-0.9622521	0.30524028	-3.1524	1.6191e-03

```

b.2.5  0.0117248 0.00774787  1.5133 1.3021e-01
b.3.4 -1.3242197 0.11176430 -11.8483 2.1954e-32
b.3.5  0.0202336 0.01360492  1.4872 1.3696e-01
b.4.5  0.0145328 0.00851130  1.7075 8.7735e-02
e.1    0.2542373 0.04680888  5.4314 5.5917e-08
e.2    0.6975726 0.12843352  5.4314 5.5917e-08
e.3    1.3975744 0.25731430  5.4314 5.5917e-08
e.4    1.2034807 0.22157875  5.4314 5.5917e-08
e.5    0.0051438 0.00094705  5.4314 5.5917e-08

```

```

g.1.2 Light.average <--- Moth
g.1.3 Litter <--- Moth
b.2.5 Calare <--- Light.average
b.3.4 Big.rock <--- Litter
b.3.5 Calare <--- Litter
b.4.5 Calare <--- Big.rock
e.1  Moth <--> Moth
e.2  Light.average <--> Light.average
e.3  Litter <--> Litter
e.4  Big.rock <--> Big.rock
e.5  Calare <--> Calare

```

```

Iterations = 0

```

5 largest modification indices, A matrix:

```

Light.average<-Litter Litter<-Light.average Light.average<-Calare

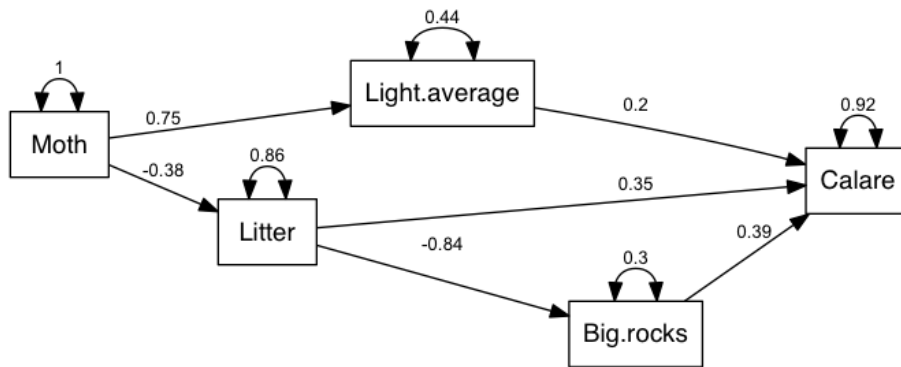
```

2.494770	2.494770	1.624254
Calare<-Moth	Big.rocks<-Calare	
1.249195	1.137006	

5 largest modification indices, P matrix:

Light.average<->Litter	Calare<->Litter	Calare<->Light.average
2.4947699	1.2491953	1.2491953
Calare<->Moth	Big.rocks<->Moth	
1.2491953	0.8392215	

Running dot -Tpng -o semPathCalare.png semPathCalare.dot





Model Chisquare = 4.5 Df = 4 Pr(>Chisq) = 0.34255  
 Chisquare (null model) = 146.05 Df = 10  
 Goodness-of-fit index = 0.97131  
 Adjusted goodness-of-fit index = 0.8924  
 RMSEA index = 0.046028 90% CI: (NA, 0.20672)  
 Bentler-Bonnett NFI = 0.96919  
 Tucker-Lewis NNFI = 0.99081  
 Bentler CFI = 0.99633  
 SRMR = 0.039249  
 AIC = 26.5  
 AICc = 10  
 BIC = 49.538  
 CAIC = -15.877

#### Normalized Residuals

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
-0.934	-0.189	0.000	-0.105	0.000	0.252

#### R-square for Endogenous Variables

Light.average	Litter	Phydub	Big.rocks
0.5583	0.1442	0.1979	0.7041

#### Parameter Estimates

	Estimate	Std Error	z value	Pr(> z )	
g.1.2	1.862463	0.2156498	8.6365	5.7952e-18	Light.average <--- Moth
g.1.3	-0.962252	0.3052403	-3.1524	1.6191e-03	Litter <--- Moth

```

b.2.5  0.038712 0.0204195    1.8959 5.7979e-02 Phydub <--- Light.average
b.3.4 -1.324220 0.1117643 -11.8483 2.1954e-32 Big.ocks <--- Litter
b.3.5  0.059666 0.0358558    1.6640 9.6103e-02 Phydub <--- Litter
b.4.5  0.062696 0.0224315    2.7950 5.1903e-03 Phydub <--- Big.ocks
e.1    0.254237 0.0468089    5.4314 5.5917e-08 Moth <--> Moth
e.2    0.697573 0.1284335    5.4314 5.5917e-08 Light.average <--> Light.average
e.3    1.397574 0.2573143    5.4314 5.5917e-08 Litter <--> Litter
e.4    1.203481 0.2215787    5.4314 5.5917e-08 Big.ocks <--> Big.ocks
e.5    0.035728 0.0065781    5.4314 5.5917e-08 Phydub <--> Phydub

```

Iterations = 0

5 largest modification indices, A matrix:

```

Litter<-Light.average    Light.average<-Litter    Big.ocks<-Phydub
                2.4947699                2.4947699                1.1370060
Big.ocks<-Light.average  Light.average<-Big.ocks
                1.1370060                0.9165941

```

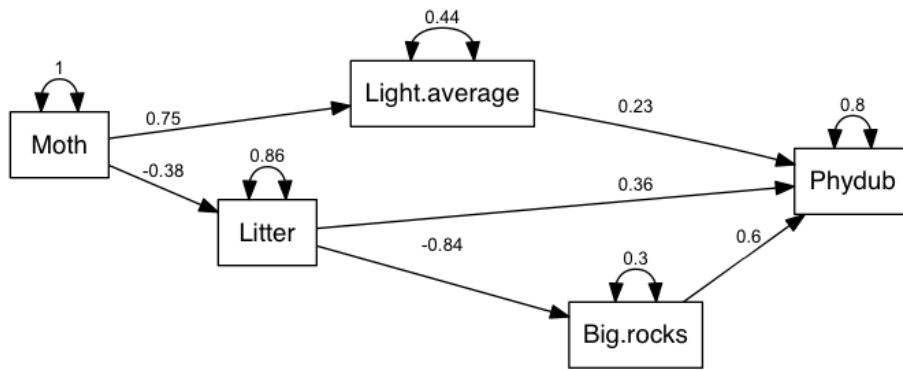
5 largest modification indices, P matrix:

```

Light.average<->Litter    Big.ocks<->Moth    Big.ocks<->Litter
                2.4947699                0.8392215                0.8392215
Phydub<->Light.average    Phydub<->Moth
                0.6802393                0.6802393

```

Running dot -Tpng -o semPathPhydub.png semPathPhydub.dot



```

>                                     #co-occurrence null models
> library(bipartite)
> source('~/.cor_nets/CorNets.R')
> library(audio)
>                                     #all
> x <- com
> if (any(ls()=='null')){}else{null <- nullSim(x,10000)}

NULL

> length(null)

[1] 10000

> if (any(ls()=='null.all')){}else{sim <- lapply(null,C.score);null.all <- unlist(sim)}

NULL

> p.all <- min(c(length(sim[sim>=C.score(x)])/length(sim),length(sim[sim<=C.score(x)]))
> ses.all <- (C.score(x) - mean(sim)) / sd(sim)
>                                     #moth susceptible
> x <- com[env$Moth==0,]
> if (any(ls()=='null')){}else{null <- nullSim(x,10000)}

NULL

> length(null)

[1] 10000

```

```
> if (any(ls()=='null.0')){}else{sim <- lapply(null,C.score);null.0 <- unlist(sim)}
```

NULL

```
> p.0 <- min(c(length(sim[sim>=C.score(x)])/length(sim),length(sim[sim<=C.score(x)]
```

```
> ses.0 <- (C.score(x) - mean(sim)) / sd(sim)
```

```
>                                     #moth resistant
```

```
> x <- com[env$Moth==1,]
```

```
> if (any(ls()=='null')){}else{null <- nullSim(x,10000)}
```

NULL

```
> length(null)
```

```
[1] 10000
```

```
> if (any(ls()=='null.1')){}else{sim <- lapply(null,C.score);null.1 <- unlist(sim)}
```

NULL

```
> p.1 <- min(c(length(sim[sim>=C.score(x)])/length(sim),length(sim[sim<=C.score(x)]
```

```
> ses.1 <- (C.score(x) - mean(sim)) / sd(sim)
```

```
>                                     #compare co-occurrence patterns
```

```
> cooc <- cbind(c(p.0,ses.0),c(p.1,ses.1),c(p.all,ses.all))
```

```
> colnames(cooc) <- c('S','R','ALL')
```

```
> rownames(cooc) <- c('p.min','SES')
```

```
> cooc
```

	S	R	ALL
p.min	0.171600	0.049900	0.2852000
SES	0.892948	-1.574635	-0.5934756

```
>
```

```
> library(xtable)
```

```
> xtable(cooc)
```

	S	R	ALL
p.min	0.17	0.05	0.29
SES	0.89	-1.57	-0.59

```
>
```

```
> #co-occurrence network
```

```
> source('~/.cor_nets/araujo_method/araujo_method.R')
```

```
> library(sna)
```

```
> x.net <- cbind(env$Moth,com)
```

```
> names(x.net)[1] <- 'moth'
```

```
> if (any(ls()=='net')){else{net <- araujoNet(x.net)$dp}
```

```
NULL
```

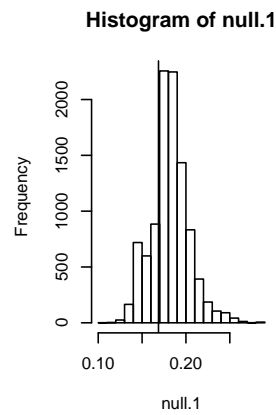
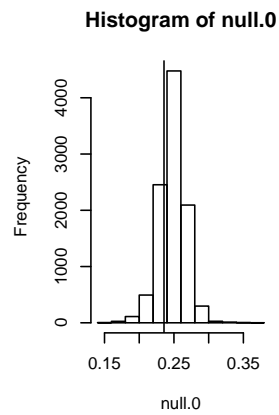
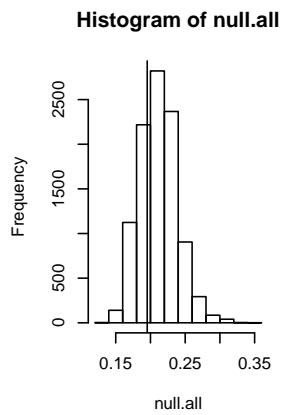
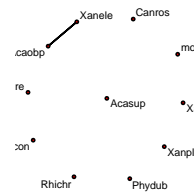
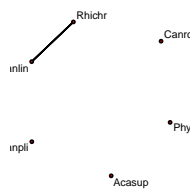
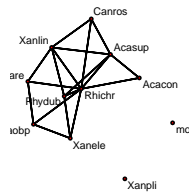
```
> if (any(ls()=='net.0')){else{net.0 <- araujoNet(x.net[env$Moth==0,])$dp}
```

```
NULL
```

```
> if (any(ls()=='net.1')){else{net.1 <- araujoNet(x.net[env$Moth==1,])$dp}
```

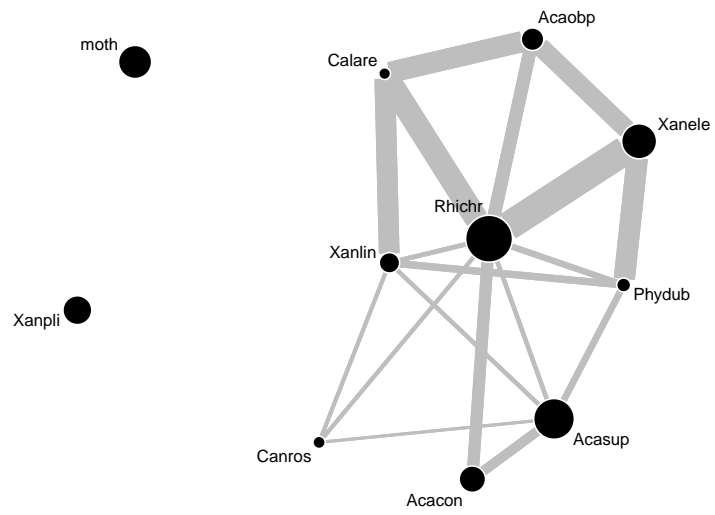
```
NULL
```

```
> par(mfrow=c(2,3))
> gplot(net,gmode='graph',displaylabels=TRUE,label.cex=0.65)
> gplot(net.0,gmode='graph',displaylabels=TRUE,label.cex=0.65)
> gplot(net.1,gmode='graph',displaylabels=TRUE,label.cex=0.65)
> hist(null.all)
> abline(v=C.score(com))
> hist(null.0)
> abline(v=C.score(com[env$Moth==0,]))
> hist(null.1)
> abline(v=C.score(com[env$Moth==1,]))
>
```



```
> e.col <- net
> e.col[net!=0] <- 'grey'
> v.cex <- apply(x.net,2,function(x) length(x[x!=0]))
> v.cex <- (v.cex / max(v.cex) + 1)^2
> gplot(net,gmode='graph',displaylabels=TRUE,label.cex=0.65,
+       edge.col=e.col,edge.lwd=(net+1)^10,vertex.cex=v.cex,
+       vertex.sides=50,vertex.col='black',vertex.border='white')
>
```





## References