

Summary of AMF community assembly

Aidee Guzman

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Description of study design

Study design: farm type x block

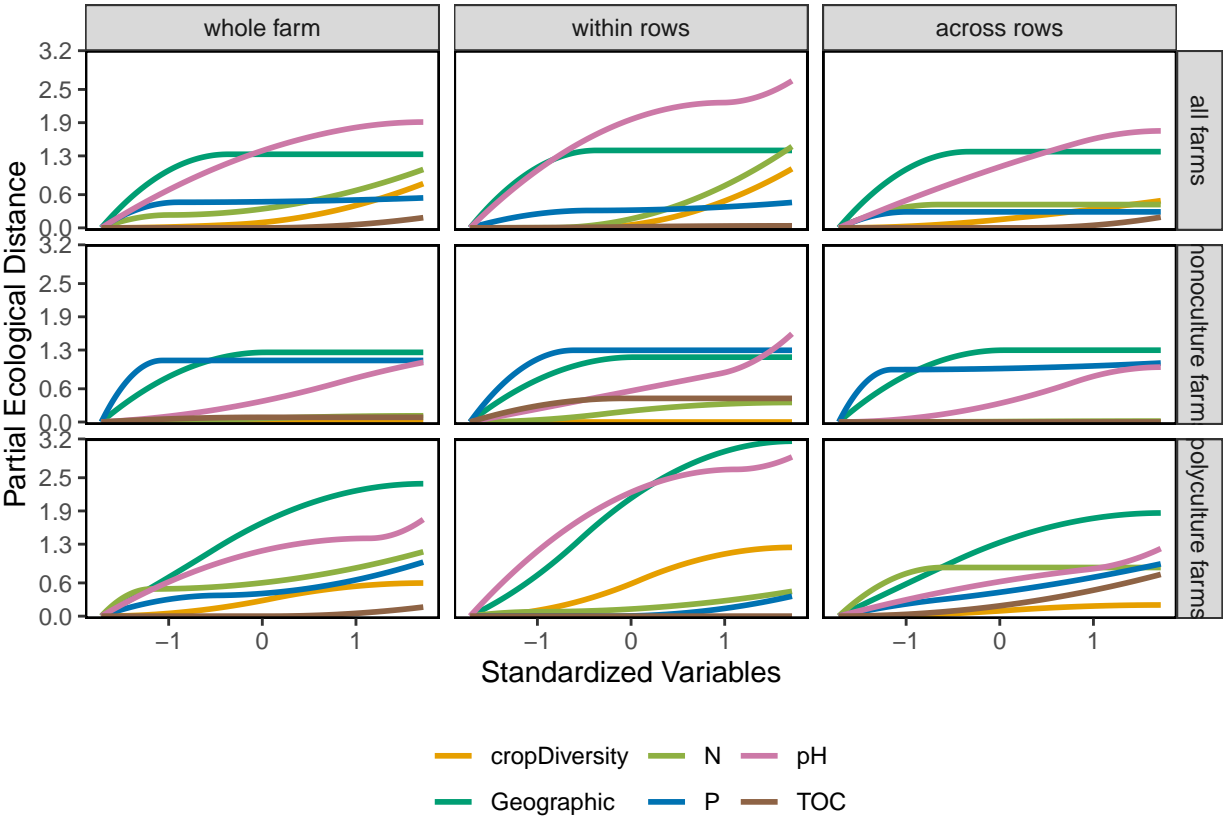
- 21 sites: 10 monoculture and 11 polyculture
- 2 transects per 2 blocks (within-rows vs across-rows) each site
- 10 = 2017 and 11 = 2018

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Drivers of AM compositional dissimilarity

Plots



All farms

Model output

Whole farm

##	Predictors	Coefficients
## 1	Geographic	1.329
## 2	pH	1.911
## 3	P	0.542
## 4	TOC	0.187
## 5	N	1.066
## 6	cropDiversity	0.810
## 7	Percent Deviance Explained	28.224
## 8	DIC	2109.035

Within-rows

##	Predictors	Coefficients
## 1	Geographic	1.399
## 2	pH	2.674
## 3	P	0.461
## 4	TOC	0.037
## 5	N	1.490
## 6	cropDiversity	1.083
## 7	Percent Deviance Explained	35.858
## 8	DIC	501.920

Across-rows

##	Predictors	Coefficients
## 1	Geographic	1.376
## 2	pH	1.751
## 3	P	0.289
## 4	TOC	0.197
## 5	N	0.421
## 6	cropDiversity	0.494
## 7	Percent Deviance Explained	25.233
## 8	DIC	511.525

Monoculture farms

Model output

Whole farm

##	Predictors	Coefficients
## 1	Geographic	1.258
## 2	pH	1.083
## 3	P	1.111
## 4	TOC	0.085
## 5	N	0.112
## 6	cropDiversity	0.000
## 7	Percent Deviance Explained	30.725
## 8	DIC	632.157

Within-rows

##	Predictors	Coefficients
## 1	Geographic	1.171
## 2	pH	1.620
## 3	P	1.296
## 4	TOC	0.427
## 5	N	0.353
## 6	cropDiversity	0.000
## 7	Percent Deviance Explained	34.585
## 8	DIC	155.935

Across-rows

##	Predictors	Coefficients
## 1	Geographic	1.297
## 2	pH	0.991
## 3	P	1.066
## 4	TOC	0.000
## 5	N	0.016
## 6	cropDiversity	0.000
## 7	Percent Deviance Explained	28.266
## 8	DIC	144.208

Polyculture farms

Model output

Whole farm

##	Predictors	Coefficients
## 1	Geographic	2.392
## 2	pH	1.767
## 3	P	0.985
## 4	TOC	0.167
## 5	N	1.170
## 6	cropDiversity	0.598
## 7	Percent Deviance Explained	36.804
## 8	DIC	539.177

Within-rows

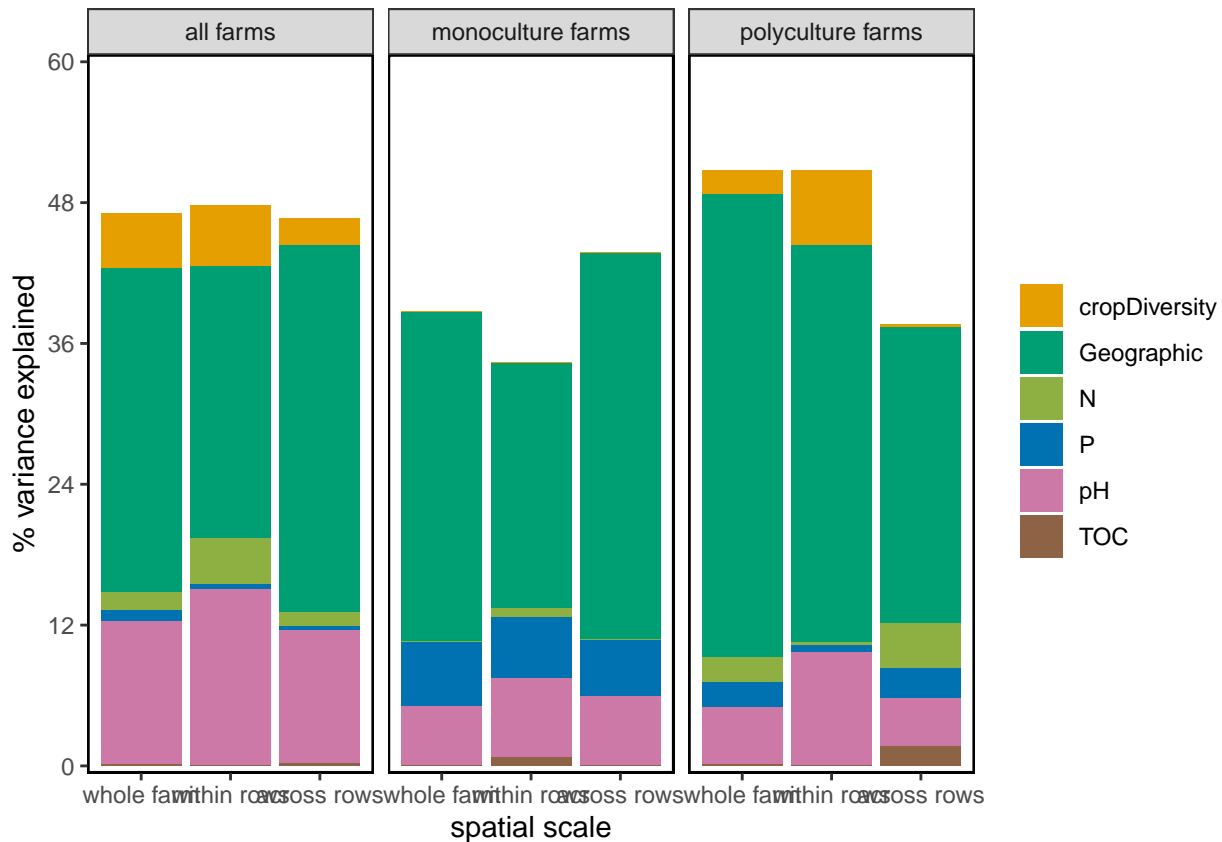
##	Predictors	Coefficients
## 1	Geographic	3.159
## 2	pH	2.889
## 3	P	0.366
## 4	TOC	0.000
## 5	N	0.453
## 6	cropDiversity	1.242
## 7	Percent Deviance Explained	49.548
## 8	DIC	121.025

Across-rows

##	Predictors	Coefficients
## 1	Geographic	1.863
## 2	pH	1.233
## 3	P	0.950
## 4	TOC	0.761
## 5	N	0.878
## 6	cropDiversity	0.202
## 7	Percent Deviance Explained	33.035
## 8	DIC	138.996

Variance importance

Plots



All farms

##	dataset	location	predictor	variance	pvalue
## 1	all farms	whole farm	Geographic	27.657	0.0
## 2	all farms	whole farm	pH	12.180	0.0
## 3	all farms	whole farm	P	0.990	0.0
## 4	all farms	whole farm	TOC	0.156	0.4
## 5	all farms	whole farm	N	1.474	0.1
## 6	all farms	whole farm	cropDiversity	4.612	0.0
## 7	all farms	within rows	Geographic	23.228	0.0
## 8	all farms	within rows	pH	15.097	0.0
## 9	all farms	within rows	P	0.399	0.0
## 10	all farms	within rows	TOC	0.008	0.8
## 11	all farms	within rows	N	3.888	0.0
## 12	all farms	within rows	cropDiversity	5.127	0.0
## 13	all farms	across rows	Geographic	31.317	0.0
## 14	all farms	across rows	pH	11.356	0.0
## 15	all farms	across rows	P	0.410	0.5
## 16	all farms	across rows	TOC	0.210	0.3
## 17	all farms	across rows	N	1.122	0.4
## 18	all farms	across rows	cropDiversity	2.279	0.0

Monoculture farms

##	dataset	location	predictor	variance	pvalue
## 1	monoculture farms	whole farm	Geographic	28.082	0.0
## 2	monoculture farms	whole farm	pH	5.094	0.0

## 3	monoculture farms	whole farm	P	5.406	0.0
## 4	monoculture farms	whole farm	TOC	0.030	0.6
## 5	monoculture farms	whole farm	N	0.076	0.5
## 6	monoculture farms	whole farm	cropDiversity	0.000	1.0
## 7	monoculture farms	within rows	Geographic	20.919	0.0
## 8	monoculture farms	within rows	pH	6.793	0.0
## 9	monoculture farms	within rows	P	5.229	0.0
## 10	monoculture farms	within rows	TOC	0.722	0.4
## 11	monoculture farms	within rows	N	0.706	0.4
## 12	monoculture farms	within rows	cropDiversity	0.000	1.0
## 13	monoculture farms	across rows	Geographic	33.025	0.0
## 14	monoculture farms	across rows	pH	6.014	0.0
## 15	monoculture farms	across rows	P	4.734	0.0
## 16	monoculture farms	across rows	TOC	0.000	1.0
## 17	monoculture farms	across rows	N	0.002	0.9
## 18	monoculture farms	across rows	cropDiversity	0.000	1.0

Polyculture farms

##	dataset	location	predictor	variance	pvalue
## 1	polyculture farms	whole farm	Geographic	39.497	0.0
## 2	polyculture farms	whole farm	pH	4.883	0.0
## 3	polyculture farms	whole farm	P	2.162	0.0
## 4	polyculture farms	whole farm	TOC	0.133	0.5
## 5	polyculture farms	whole farm	N	2.048	0.0
## 6	polyculture farms	whole farm	cropDiversity	2.023	0.0
## 7	polyculture farms	within rows	Geographic	33.828	0.0
## 8	polyculture farms	within rows	pH	9.718	0.0
## 9	polyculture farms	within rows	P	0.584	0.1
## 10	polyculture farms	within rows	TOC	0.000	1.0
## 11	polyculture farms	within rows	N	0.244	0.3
## 12	polyculture farms	within rows	cropDiversity	6.370	0.0
## 13	polyculture farms	across rows	Geographic	25.263	0.0
## 14	polyculture farms	across rows	pH	4.152	0.0
## 15	polyculture farms	across rows	P	2.591	0.1
## 16	polyculture farms	across rows	TOC	1.681	0.1
## 17	polyculture farms	across rows	N	3.752	0.1
## 18	polyculture farms	across rows	cropDiversity	0.248	0.6

Mantel tests

Composition

##	Data	Species v Crop Diversity	Species v Geography	Species v pH
## 1	All	0.206 ***	0.218 ***	0.323 ***
## 2	All across	0.192 ***	0.233 ***	0.279 ***
## 3	All within	0.22 ***	0.207 ***	0.365 ***
## 4	Mono	NA NA	0.244 ***	0.369 ***
## 5	Mono across	NA NA	0.25 ***	0.317 ***
## 6	Mono within	NA NA	0.209 ***	0.423 ***
## 7	Poly	0.01	0.541 ***	0.289 ***
## 8	Poly across	-0.013	0.464 ***	0.22 **
## 9	Poly within	0.016	0.6 ***	0.346 ***
##	Species v P	Species v N	Species v TOC	
## 1	0.139 ***	0.019	0.012	
## 2	0.084 **	-0.013	-0.023	
## 3	0.202 ***	0.028	0.076 *	
## 4	0.172 ***	0.102 **	0.01	
## 5	0.123 *	0.176 ***	-0.051	
## 6	0.202 ***	0.049	0.137 **	
## 7	0.218 ***	0.133 ***	0.096 **	
## 8	0.209 ***	0.096 *	0.18 **	
## 9	0.203 ***	0.087 *	0.022	

Turnover

##	Data	Species v Crop Diversity	Species v Geography	Species v pH
## 1	All	0.205 ***	0.217 ***	0.275 ***
## 2	All across	0.205 ***	0.223 ***	0.227 ***
## 3	All within	0.208 ***	0.213 ***	0.324 ***
## 4	Mono	NA NA	0.239 ***	0.276 ***
## 5	Mono across	NA NA	0.228 ***	0.215 ***
## 6	Mono within	NA NA	0.225 ***	0.351 ***
## 7	Poly	0.017	0.473 ***	0.276 ***
## 8	Poly across	0.013	0.414 ***	0.214 ***
## 9	Poly within	0.005	0.495 ***	0.327 ***
##	Species v P	Species v N	Species v TOC	
## 1	0.131 ***	0.019	-0.012	
## 2	0.085 **	-0.013	-0.031	
## 3	0.19 ***	0.028	0.025	
## 4	0.161 ***	0.102 **	-0.002	
## 5	0.09 *	0.176 ***	-0.044	
## 6	0.223 ***	0.049	0.079	
## 7	0.175 ***	0.133 ***	0.06 *	
## 8	0.182 ***	0.096 *	0.158 ***	
## 9	0.153 ***	0.087 *	-0.016	

Nestedness

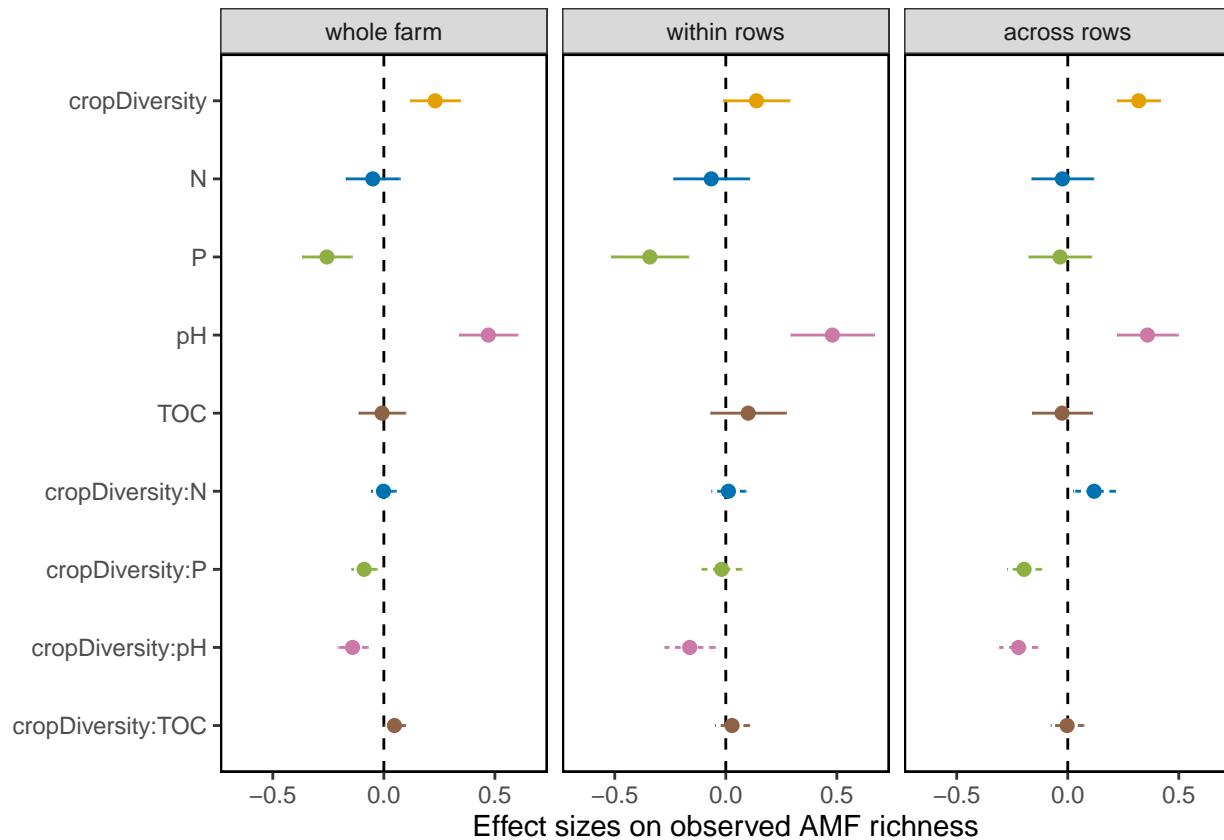
##	Data	Species v Crop Diversity	Species v Geography	Species v pH
## 1	All	-0.151	-0.167	-0.22
## 2	All across	-0.162	-0.182	-0.178
## 3	All within	-0.146	-0.158	-0.255
## 4	Mono	NA NA	-0.133	-0.184
## 5	Mono across	NA NA	-0.162	-0.141
## 6	Mono within	NA NA	-0.088	-0.235
## 7	Poly	0.011	-0.372	-0.216
## 8	Poly across	0.006	-0.32	-0.152

## 9 Poly within		0.037	-0.38	-0.25
##	Species v P	Species v N	Species v TOC	
## 1	-0.113	0.019	0.032	
## 2	-0.072	-0.013	0.043	
## 3	-0.16	0.028	0.006	
## 4	-0.109	0.102 **	0.012	
## 5	-0.054	0.176 **	0.031	
## 6	-0.144	0.049	-0.039	
## 7	-0.124	0.133 ***	-0.021	
## 8	-0.106	0.096 *	-0.097	
## 9	-0.104	0.087 *	0.041	

Drivers of AMF alpha diversity

Richness ~ crop diversity

Plots



Whole farm

```
## $model
## glmer(formula = observed ~ cropDiversity * scale(pH) + scale(P) *
##   cropDiversity + scale(TOC) * cropDiversity + scale(N) * cropDiversity +
##   (1 | FarmKey:Year), data = all_wa$df, family = MASS::negative.binomial(theta = 8.65005438353719),
##   nAGQ = 1, na.action = na.fail)
##
## $AIC
## [1] 1415.019
##
## $summary
##   response      covariate Estimate   SE    P sig
## 1 observed      cropDiversity   0.231 0.112 0.039  *
## 2 observed      scale(pH)       0.470 0.132 0.000 ***
## 3 observed      scale(P)       -0.256 0.112 0.022  *
## 4 observed      scale(TOC)      -0.008 0.105 0.938  ns
## 5 observed      scale(N)       -0.050 0.121 0.681  ns
## 6 observed  cropDiversity:scale(pH) -0.141 0.068 0.038  *
## 7 observed  cropDiversity:scale(P) -0.089 0.056 0.116  ns
## 8 observed  cropDiversity:scale(TOC)  0.048 0.048 0.324  ns
## 9 observed  cropDiversity:scale(N)  -0.002 0.055 0.976  ns
##
## $Anova
```

##	response	covariate	Chisq	P	sig
## 1	observed	cropDiversity	4.260	0.039	*
## 2	observed	scale(pH)	12.713	0.000	***
## 3	observed	scale(P)	5.274	0.022	*
## 4	observed	scale(TOC)	0.006	0.938	ns
## 5	observed	scale(N)	0.169	0.681	ns
## 6	observed	cropDiversity:scale(pH)	4.288	0.038	*
## 7	observed	cropDiversity:scale(P)	2.476	0.116	ns
## 8	observed	cropDiversity:scale(TOC)	0.975	0.324	ns
## 9	observed	cropDiversity:scale(N)	0.001	0.976	ns

Within-rows

```
## $model
## glmer(formula = observed ~ cropDiversity * scale(pH) + scale(P) *
##   cropDiversity + scale(TOC) * cropDiversity + scale(N) * cropDiversity +
##   (1 | FarmKey:Year), data = all_w$df, family = MASS::negative.binomial(theta = 19.7818601106867),
##   nAGQ = 1, na.action = na.fail)
##
## $AIC
## [1] 717.0261
##
## $summary
```

##	response	covariate	Estimate	SE	P	sig
## 1	observed	cropDiversity	0.138	0.149	0.353	ns
## 2	observed	scale(pH)	0.480	0.187	0.010	**
## 3	observed	scale(P)	-0.342	0.174	0.049	*
## 4	observed	scale(TOC)	0.101	0.170	0.552	ns
## 5	observed	scale(N)	-0.066	0.170	0.699	ns
## 6	observed	cropDiversity:scale(pH)	-0.162	0.113	0.150	ns
## 7	observed	cropDiversity:scale(P)	-0.019	0.089	0.833	ns
## 8	observed	cropDiversity:scale(TOC)	0.028	0.077	0.715	ns
## 9	observed	cropDiversity:scale(N)	0.012	0.077	0.878	ns

```
##
## $Anova
```

##	response	covariate	Chisq	P	sig
## 1	observed	cropDiversity	0.864	0.353	ns
## 2	observed	scale(pH)	6.558	0.010	**
## 3	observed	scale(P)	3.879	0.049	*
## 4	observed	scale(TOC)	0.354	0.552	ns
## 5	observed	scale(N)	0.149	0.699	ns
## 6	observed	cropDiversity:scale(pH)	2.073	0.150	ns
## 7	observed	cropDiversity:scale(P)	0.044	0.833	ns
## 8	observed	cropDiversity:scale(TOC)	0.133	0.715	ns
## 9	observed	cropDiversity:scale(N)	0.024	0.878	ns

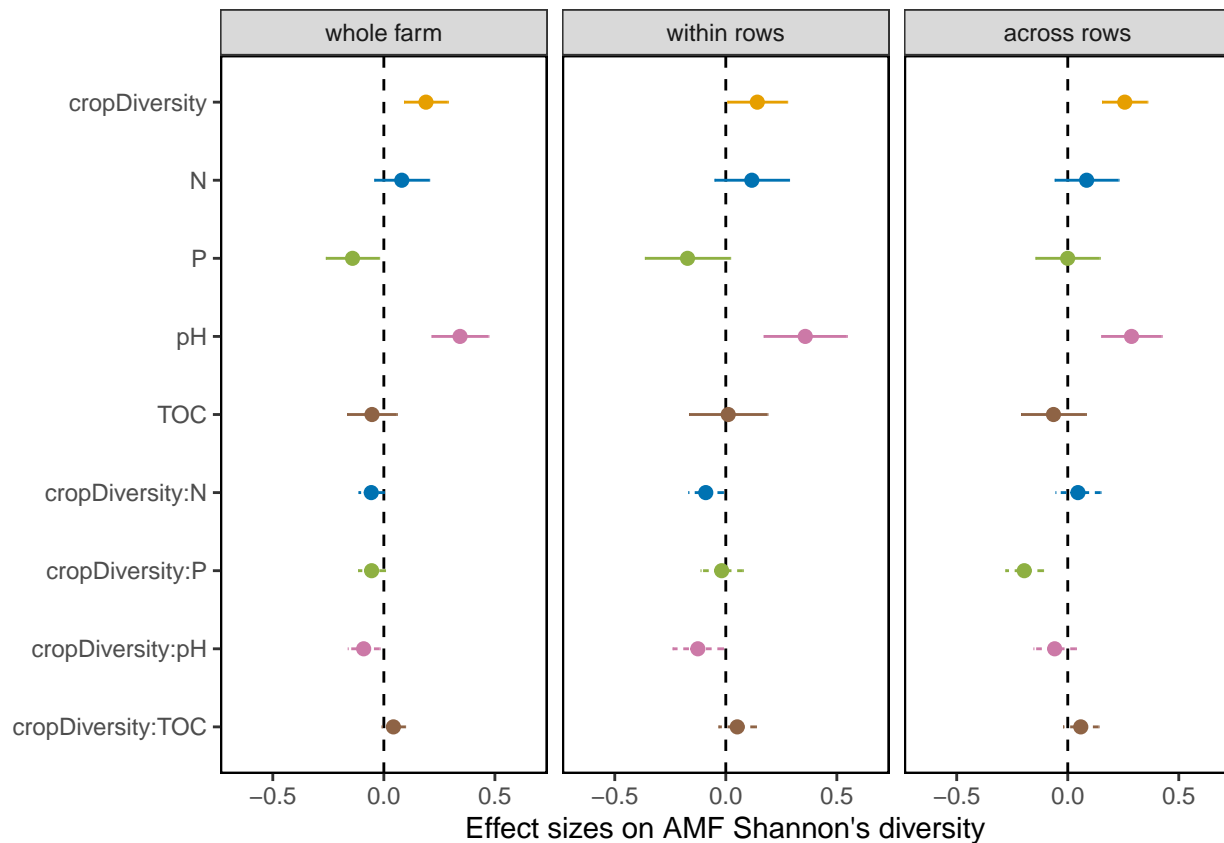
Across-rows

```
## $model
## glmer(formula = observed ~ cropDiversity * scale(pH) + scale(P) *
##   cropDiversity + scale(TOC) * cropDiversity + scale(N) * cropDiversity +
##   (1 | FarmKey:Year), data = all_a$df, family = MASS::negative.binomial(theta = 6.07236915915643),
##   nAGQ = 1, na.action = na.fail)
##
## $AIC
## [1] 729.5035
##
## $summary
```

```
## response      covariate Estimate SE P sig
## 1 observed    cropDiversity  0.319 0.096 0.001 ***
## 2 observed      scale(pH)    0.359 0.137 0.009 **
## 3 observed      scale(P)     -0.035 0.141 0.806 ns
## 4 observed      scale(TOC)    -0.026 0.135 0.849 ns
## 5 observed      scale(N)     -0.024 0.139 0.863 ns
## 6 observed    cropDiversity:scale(pH) -0.222 0.086 0.010 **
## 7 observed    cropDiversity:scale(P)  -0.197 0.077 0.011 *
## 8 observed    cropDiversity:scale(TOC) -0.003 0.074 0.968 ns
## 9 observed    cropDiversity:scale(N)   0.119 0.094 0.208 ns
##
## $Anova
## response      covariate Chisq P sig
## 1 observed    cropDiversity 10.968 0.001 ***
## 2 observed      scale(pH)    6.894 0.009 **
## 3 observed      scale(P)     0.060 0.806 ns
## 4 observed      scale(TOC)    0.036 0.849 ns
## 5 observed      scale(N)     0.030 0.863 ns
## 6 observed    cropDiversity:scale(pH) 6.630 0.010 **
## 7 observed    cropDiversity:scale(P)  6.522 0.011 *
## 8 observed    cropDiversity:scale(TOC) 0.002 0.968 ns
## 9 observed    cropDiversity:scale(N)   1.583 0.208 ns
```

Shannon ~ crop diversity

Plots



Whole farm

```
## $model
```

```
## lmer(formula = shannon ~ cropDiversity * scale(pH) + scale(P) *
##      cropDiversity + scale(TOC) * cropDiversity + scale(N) * cropDiversity +
##      (1 | FarmKey:Year), data = all_wa$df, na.action = na.exclude)
##
## $AIC
## [1] 427.6784
##
## $summary
##      response      covariate Estimate    SE      P sig
## 1 shannon      cropDiversity    0.190 0.099 0.070  ns
## 2 shannon      scale(pH)        0.343 0.128 0.009  **
## 3 shannon      scale(P)        -0.141 0.120 0.241  ns
## 4 shannon      scale(TOC)       -0.053 0.112 0.633  ns
## 5 shannon      scale(N)         0.081 0.123 0.514  ns
## 6 shannon cropDiversity:scale(pH) -0.091 0.072 0.209  ns
## 7 shannon cropDiversity:scale(P)  -0.055 0.060 0.362  ns
## 8 shannon cropDiversity:scale(TOC)  0.043 0.053 0.416  ns
## 9 shannon cropDiversity:scale(N)  -0.057 0.057 0.319  ns
##
## $Anova
##      response      covariate Chisq      P sig
## 1 shannon      cropDiversity 3.707 0.054  ns
## 2 shannon      scale(pH)    7.203 0.007  **
## 3 shannon      scale(P)     1.383 0.240  ns
## 4 shannon      scale(TOC)    0.228 0.633  ns
## 5 shannon      scale(N)      0.428 0.513  ns
## 6 shannon cropDiversity:scale(pH) 1.598 0.206  ns
## 7 shannon cropDiversity:scale(P)  0.834 0.361  ns
## 8 shannon cropDiversity:scale(TOC) 0.664 0.415  ns
## 9 shannon cropDiversity:scale(N)  0.997 0.318  ns
```

Within-rows

```
## $model
## lmer(formula = shannon ~ cropDiversity * scale(pH) + scale(P) *
##      cropDiversity + scale(TOC) * cropDiversity + scale(N) * cropDiversity +
##      (1 | FarmKey:Year), data = all_w$df, na.action = na.exclude)
##
## $AIC
## [1] 224.9064
##
## $summary
##      response      covariate Estimate    SE      P sig
## 1 shannon      cropDiversity    0.142 0.135 0.307  ns
## 2 shannon      scale(pH)        0.358 0.187 0.061  ns
## 3 shannon      scale(P)        -0.172 0.192 0.372  ns
## 4 shannon      scale(TOC)       0.011 0.177 0.951  ns
## 5 shannon      scale(N)         0.117 0.168 0.486  ns
## 6 shannon cropDiversity:scale(pH) -0.126 0.113 0.273  ns
## 7 shannon cropDiversity:scale(P)  -0.019 0.095 0.845  ns
## 8 shannon cropDiversity:scale(TOC)  0.051 0.084 0.543  ns
## 9 shannon cropDiversity:scale(N)  -0.090 0.078 0.251  ns
##
## $Anova
##      response      covariate Chisq      P sig
## 1 shannon      cropDiversity 1.111 0.292  ns
## 2 shannon      scale(pH)     3.660 0.056  ns
```

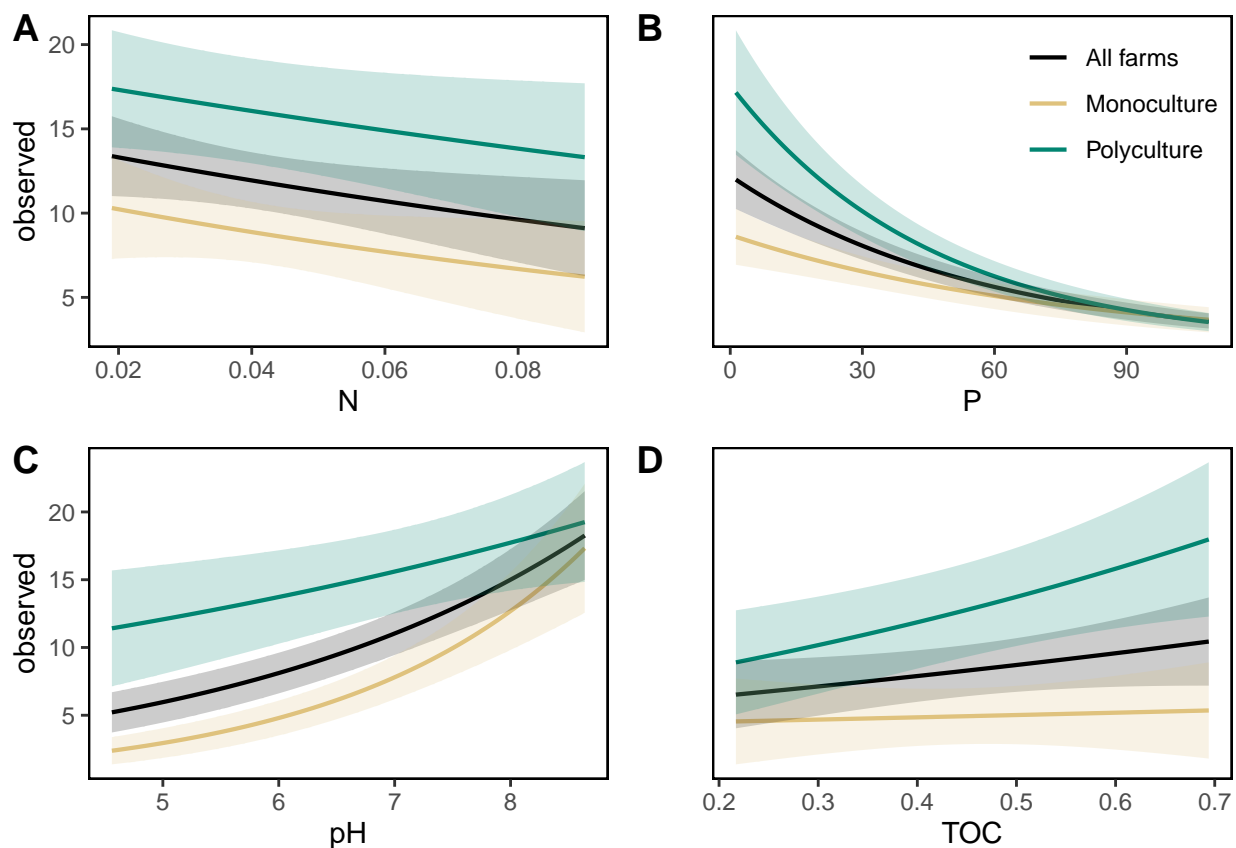
```
## 3 shannon          scale(P) 0.806 0.369 ns
## 4 shannon          scale(TOC) 0.004 0.951 ns
## 5 shannon          scale(N) 0.488 0.485 ns
## 6 shannon cropDiversity:scale(pH) 1.235 0.267 ns
## 7 shannon cropDiversity:scale(P) 0.038 0.845 ns
## 8 shannon cropDiversity:scale(TOC) 0.373 0.542 ns
## 9 shannon cropDiversity:scale(N) 1.334 0.248 ns
```

Across-rows

```
## $model
## lmer(formula = shannon ~ cropDiversity * scale(pH) + scale(P) *
##      cropDiversity + scale(TOC) * cropDiversity + scale(N) * cropDiversity +
##      (1 | FarmKey:Year), data = all_a$df, na.action = na.exclude)
##
## $AIC
## [1] 249.4365
##
## $summary
##      response          covariate Estimate    SE      P sig
## 1 shannon      cropDiversity    0.257 0.102 0.024  *
## 2 shannon          scale(pH)    0.287 0.136 0.046  *
## 3 shannon          scale(P)     0.000 0.145 0.998 ns
## 4 shannon      scale(TOC)   -0.064 0.146 0.662 ns
## 5 shannon          scale(N)     0.085 0.144 0.560 ns
## 6 shannon cropDiversity:scale(pH) -0.059 0.096 0.541 ns
## 7 shannon cropDiversity:scale(P) -0.196 0.085 0.026  *
## 8 shannon cropDiversity:scale(TOC)  0.059 0.080 0.467 ns
## 9 shannon cropDiversity:scale(N)   0.046 0.102 0.655 ns
##
## $Anova
##      response          covariate Chisq      P sig
## 1 shannon      cropDiversity 6.345 0.012  *
## 2 shannon          scale(pH) 4.459 0.035  *
## 3 shannon          scale(P) 0.000 0.998 ns
## 4 shannon      scale(TOC) 0.193 0.660 ns
## 5 shannon          scale(N) 0.351 0.554 ns
## 6 shannon cropDiversity:scale(pH) 0.379 0.538 ns
## 7 shannon cropDiversity:scale(P) 5.349 0.021  *
## 8 shannon cropDiversity:scale(TOC) 0.537 0.464 ns
## 9 shannon cropDiversity:scale(N) 0.205 0.651 ns
```

Richness ~ farm type

Plots



Whole farm

```
## $model
## glmer(formula = observed ~ FarmType * scale(pH) + scale(P) *
##   FarmType + scale(TOC) * FarmType + scale(N) * FarmType +
##   (1 | FarmKey:Year), data = all_wa$df, family = MASS::negative.binomial(theta = 8.62627505625531),
##   nAGQ = 1, na.action = na.fail)
##
## $AIC
## [1] 1414.883
##
## $summary
##   response      covariate Estimate    SE    P sig
## 1 observed      FarmTypePolyculture  0.584 0.280 0.037  *
## 2 observed      scale(pH)           0.482 0.134 0.000 ***
## 3 observed      scale(P)            -0.278 0.115 0.016  *
## 4 observed      scale(TOC)           0.017 0.109 0.879  ns
## 5 observed      scale(N)            -0.090 0.126 0.473  ns
## 6 observed      FarmTypePolyculture:scale(pH) -0.355 0.171 0.038  *
## 7 observed      FarmTypePolyculture:scale(P) -0.182 0.150 0.225  ns
## 8 observed      FarmTypePolyculture:scale(TOC)  0.088 0.132 0.506  ns
## 9 observed      FarmTypePolyculture:scale(N)   0.043 0.139 0.759  ns
##
## $Anova
##   response      covariate  Chisq    P sig
## 1 observed      FarmType    4.364 0.037  *
## 2 observed      scale(pH)  12.867 0.000 ***
## 3 observed      scale(P)    5.842 0.016  *
## 4 observed      scale(TOC)  0.023 0.879  ns
## 5 observed      scale(N)    0.515 0.473  ns
```



```
## 6 observed FarmType:scale(pH) 4.298 0.038 *
## 7 observed FarmType:scale(P) 1.469 0.225 ns
## 8 observed FarmType:scale(TOC) 0.441 0.506 ns
## 9 observed FarmType:scale(N) 0.094 0.759 ns
```

Within-rows

```
## $model
## glmer(formula = observed ~ FarmType * scale(pH) + scale(P) *
## FarmType + scale(TOC) * FarmType + scale(N) * FarmType +
## (1 | FarmKey:Year), data = all_w$df, family = MASS::negative.binomial(theta = 19.901483453616),
## nAGQ = 1, na.action = na.fail)
##
## $AIC
## [1] 716.1875
##
## $summary
## response covariate Estimate SE P sig
## 1 observed FarmTypePolyculture 0.465 0.362 0.199 ns
## 2 observed scale(pH) 0.490 0.191 0.010 **
## 3 observed scale(P) -0.354 0.179 0.047 *
## 4 observed scale(TOC) 0.143 0.179 0.426 ns
## 5 observed scale(N) -0.107 0.177 0.545 ns
## 6 observed FarmTypePolyculture:scale(pH) -0.452 0.289 0.118 ns
## 7 observed FarmTypePolyculture:scale(P) -0.022 0.234 0.924 ns
## 8 observed FarmTypePolyculture:scale(TOC) 0.010 0.201 0.961 ns
## 9 observed FarmTypePolyculture:scale(N) 0.077 0.189 0.685 ns
##
## $Anova
## response covariate Chisq P sig
## 1 observed FarmType 1.651 0.199 ns
## 2 observed scale(pH) 6.611 0.010 **
## 3 observed scale(P) 3.929 0.047 *
## 4 observed scale(TOC) 0.635 0.426 ns
## 5 observed scale(N) 0.367 0.545 ns
## 6 observed FarmType:scale(pH) 2.445 0.118 ns
## 7 observed FarmType:scale(P) 0.009 0.924 ns
## 8 observed FarmType:scale(TOC) 0.002 0.961 ns
## 9 observed FarmType:scale(N) 0.164 0.685 ns
```

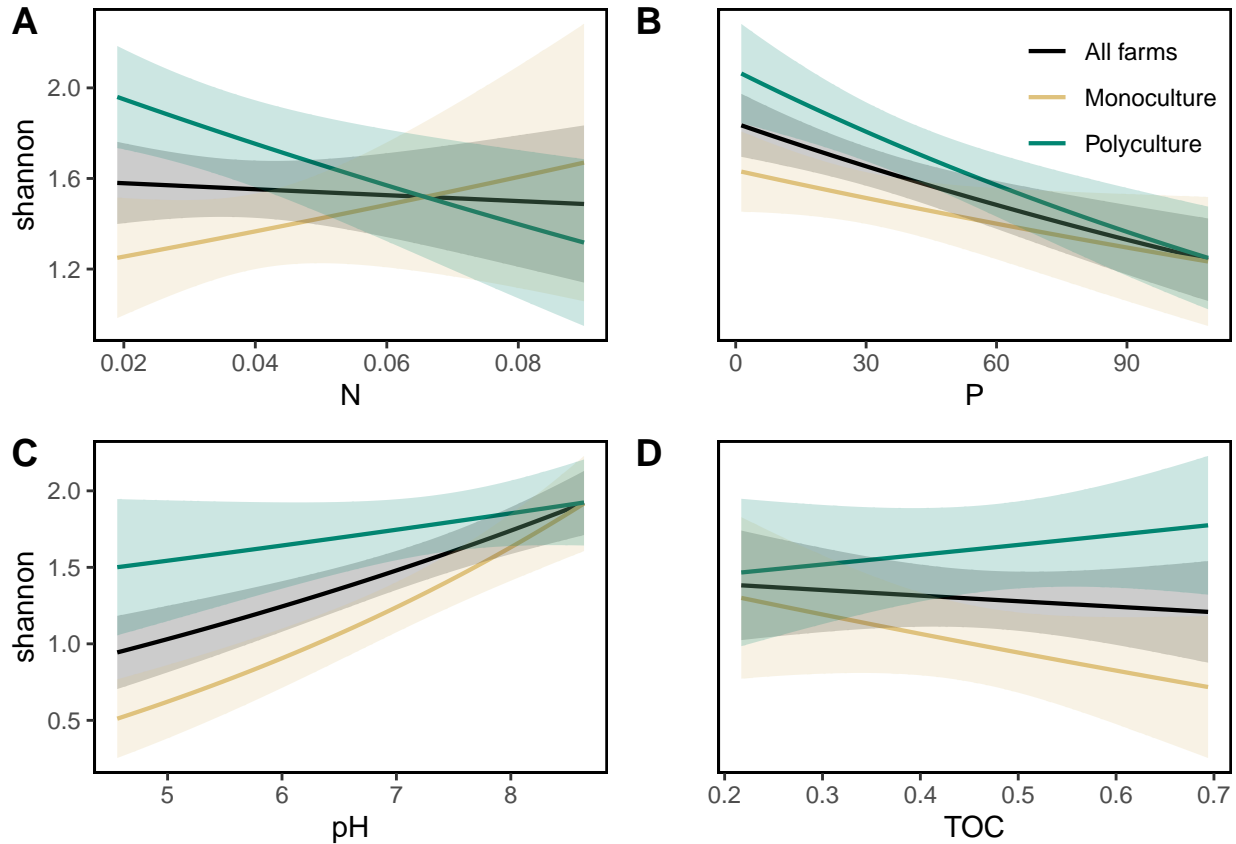
Across-rows

```
## $model
## glmer(formula = observed ~ FarmType * scale(pH) + scale(P) *
## FarmType + scale(TOC) * FarmType + scale(N) * FarmType +
## (1 | FarmKey:Year), data = all_a$df, family = MASS::negative.binomial(theta = 5.93197226682143),
## nAGQ = 1, na.action = na.fail)
##
## $AIC
## [1] 728.933
##
## $summary
## response covariate Estimate SE P sig
## 1 observed FarmTypePolyculture 0.739 0.225 0.001 ***
## 2 observed scale(pH) 0.357 0.137 0.009 **
## 3 observed scale(P) -0.010 0.143 0.946 ns
## 4 observed scale(TOC) -0.045 0.139 0.748 ns
## 5 observed scale(N) -0.014 0.138 0.922 ns
```

```
## 6 observed FarmTypePolyculture:scale(pH) -0.584 0.215 0.007 **
## 7 observed FarmTypePolyculture:scale(P) -0.482 0.199 0.015 *
## 8 observed FarmTypePolyculture:scale(TOC) -0.009 0.193 0.961 ns
## 9 observed FarmTypePolyculture:scale(N) 0.289 0.243 0.234 ns
##
## $Anova
## response covariate Chisq P sig
## 1 observed FarmType 10.779 0.001 ***
## 2 observed scale(pH) 6.758 0.009 **
## 3 observed scale(P) 0.005 0.946 ns
## 4 observed scale(TOC) 0.103 0.748 ns
## 5 observed scale(N) 0.010 0.922 ns
## 6 observed FarmType:scale(pH) 7.357 0.007 **
## 7 observed FarmType:scale(P) 5.867 0.015 *
## 8 observed FarmType:scale(TOC) 0.002 0.961 ns
## 9 observed FarmType:scale(N) 1.413 0.234 ns
```

Shannon ~ farm type

Plots



Whole farm

```
## $model
## lmer(formula = log(shannon + 1) ~ FarmType * scale(pH) + scale(P) *
## FarmType + scale(TOC) * FarmType + scale(N) * FarmType +
## (1 | FarmKey:Year), data = all_wa$df, na.action = na.exclude)
##
## $AIC
## [1] 77.06143
```

```
##
## $summary
##           response                covariate Estimate    SE      P sig
## 1 log(shannon + 1)      FarmTypePolyculture    0.167 0.099 0.111  ns
## 2 log(shannon + 1)                scale(pH)    0.160 0.055 0.005  **
## 3 log(shannon + 1)                scale(P)   -0.051 0.053 0.333  ns
## 4 log(shannon + 1)                scale(TOC)  -0.035 0.050 0.488  ns
## 5 log(shannon + 1)                scale(N)    0.031 0.055 0.575  ns
## 6 log(shannon + 1) FarmTypePolyculture:scale(pH) -0.122 0.078 0.120  ns
## 7 log(shannon + 1) FarmTypePolyculture:scale(P)  -0.042 0.069 0.548  ns
## 8 log(shannon + 1) FarmTypePolyculture:scale(TOC)  0.051 0.062 0.420  ns
## 9 log(shannon + 1) FarmTypePolyculture:scale(N)  -0.075 0.063 0.240  ns
##
## $Anova
##           response                covariate Chisq      P sig
## 1 log(shannon + 1)      FarmType 2.838 0.092  ns
## 2 log(shannon + 1)                scale(pH) 8.499 0.004  **
## 3 log(shannon + 1)                scale(P) 0.944 0.331  ns
## 4 log(shannon + 1)                scale(TOC) 0.482 0.488  ns
## 5 log(shannon + 1)                scale(N) 0.315 0.574  ns
## 6 log(shannon + 1) FarmType:scale(pH) 2.467 0.116  ns
## 7 log(shannon + 1) FarmType:scale(P) 0.364 0.547  ns
## 8 log(shannon + 1) FarmType:scale(TOC) 0.653 0.419  ns
## 9 log(shannon + 1) FarmType:scale(N) 1.394 0.238  ns
```

Within-rows

```
## $model
## lmer(formula = log(shannon + 1) ~ FarmType * scale(pH) + scale(P) *
##       FarmType + scale(TOC) * FarmType + scale(N) * FarmType +
##       (1 | FarmKey:Year), data = all_w$df, na.action = na.exclude)
##
## $AIC
## [1] 59.52043
##
## $summary
##           response                covariate Estimate    SE      P sig
## 1 log(shannon + 1)      FarmTypePolyculture    0.172 0.136 0.226  ns
## 2 log(shannon + 1)                scale(pH)    0.147 0.083 0.082  ns
## 3 log(shannon + 1)                scale(P)   -0.048 0.088 0.586  ns
## 4 log(shannon + 1)                scale(TOC)  -0.026 0.084 0.755  ns
## 5 log(shannon + 1)                scale(N)    0.069 0.078 0.378  ns
## 6 log(shannon + 1) FarmTypePolyculture:scale(pH) -0.169 0.128 0.195  ns
## 7 log(shannon + 1) FarmTypePolyculture:scale(P)  -0.046 0.112 0.683  ns
## 8 log(shannon + 1) FarmTypePolyculture:scale(TOC)  0.057 0.100 0.573  ns
## 9 log(shannon + 1) FarmTypePolyculture:scale(N)  -0.127 0.089 0.159  ns
##
## $Anova
##           response                covariate Chisq      P sig
## 1 log(shannon + 1)      FarmType 1.596 0.207  ns
## 2 log(shannon + 1)                scale(pH) 3.169 0.075  ns
## 3 log(shannon + 1)                scale(P) 0.299 0.584  ns
## 4 log(shannon + 1)                scale(TOC) 0.098 0.754  ns
## 5 log(shannon + 1)                scale(N) 0.786 0.375  ns
## 6 log(shannon + 1) FarmType:scale(pH) 1.747 0.186  ns
## 7 log(shannon + 1) FarmType:scale(P) 0.168 0.682  ns
## 8 log(shannon + 1) FarmType:scale(TOC) 0.319 0.572  ns
```

```
## 9 log(shannon + 1) FarmType:scale(N) 2.015 0.156 ns
```

Across-rows

```
## $model
```

```
## lmer(formula = log(shannon + 1) ~ FarmType * scale(pH) + scale(P) *  
## FarmType + scale(TOC) * FarmType + scale(N) * FarmType +  
## (1 | FarmKey:Year), data = all_a$df, na.action = na.exclude)
```

```
##
```

```
## $AIC
```

```
## [1] 70.73671
```

```
##
```

```
## $summary
```

	response	covariate	Estimate	SE	P	sig
## 1	log(shannon + 1)	FarmTypePolyculture	0.173	0.095	0.088	ns
## 2	log(shannon + 1)	scale(pH)	0.142	0.054	0.017	*
## 3	log(shannon + 1)	scale(P)	0.003	0.058	0.953	ns
## 4	log(shannon + 1)	scale(TOC)	-0.061	0.061	0.326	ns
## 5	log(shannon + 1)	scale(N)	0.022	0.057	0.706	ns
## 6	log(shannon + 1)	FarmTypePolyculture:scale(pH)	-0.129	0.098	0.197	ns
## 7	log(shannon + 1)	FarmTypePolyculture:scale(P)	-0.140	0.089	0.125	ns
## 8	log(shannon + 1)	FarmTypePolyculture:scale(TOC)	0.081	0.086	0.352	ns
## 9	log(shannon + 1)	FarmTypePolyculture:scale(N)	0.047	0.106	0.665	ns

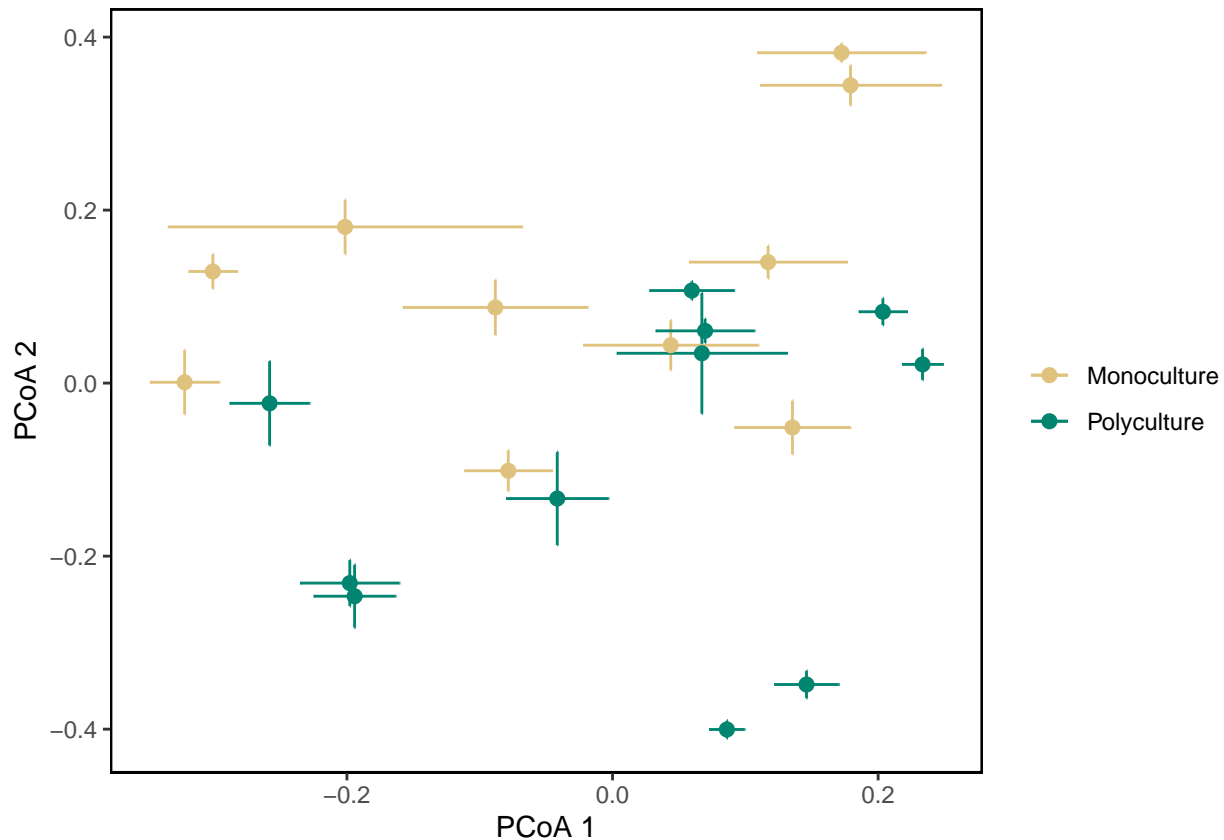
```
##
```

```
## $Anova
```

	response	covariate	Chisq	P	sig
## 1	log(shannon + 1)	FarmType	3.356	0.067	ns
## 2	log(shannon + 1)	scale(pH)	6.836	0.009	**
## 3	log(shannon + 1)	scale(P)	0.003	0.953	ns
## 4	log(shannon + 1)	scale(TOC)	0.984	0.321	ns
## 5	log(shannon + 1)	scale(N)	0.147	0.702	ns
## 6	log(shannon + 1)	FarmType:scale(pH)	1.714	0.191	ns
## 7	log(shannon + 1)	FarmType:scale(P)	2.482	0.115	ns
## 8	log(shannon + 1)	FarmType:scale(TOC)	0.887	0.346	ns
## 9	log(shannon + 1)	FarmType:scale(N)	0.192	0.661	ns

AMF compositional differences between farm type

Plots

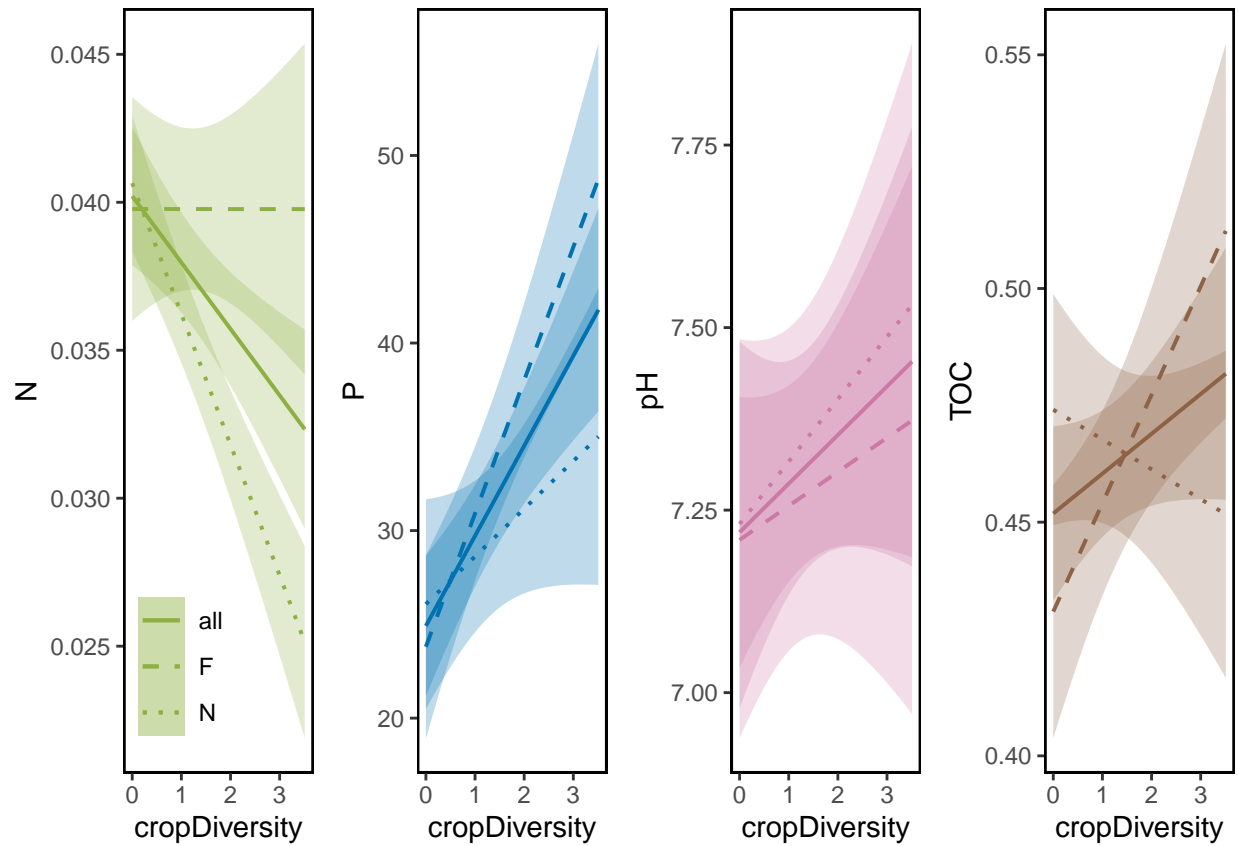


Model output

```
##
## Call:
## adonis(formula = finalFormula, data = permanovaDF, strata = permanovaDF$strata)
##
## Blocks: strata
## Permutation: free
## Number of permutations: 999
##
## Terms added sequentially (first to last)
##
##              Df SumsOfSqs MeanSqs F.Model    R2 Pr(>F)
## cropDiversity  1     5.268   5.2681 16.0011 0.06285 0.009 **
## P              1     2.253   2.2532  6.8437 0.02688 0.001 ***
## N              1     1.935   1.9353  5.8782 0.02309 0.651
## pH             1     4.092   4.0923 12.4297 0.04882 0.008 **
## TOC            1     1.138   1.1379  3.4564 0.01358 0.418
## Residuals     210    69.139   0.3292      0.82479
## Total         215    83.826
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Edaphic variables ~ crop diversity

Plots

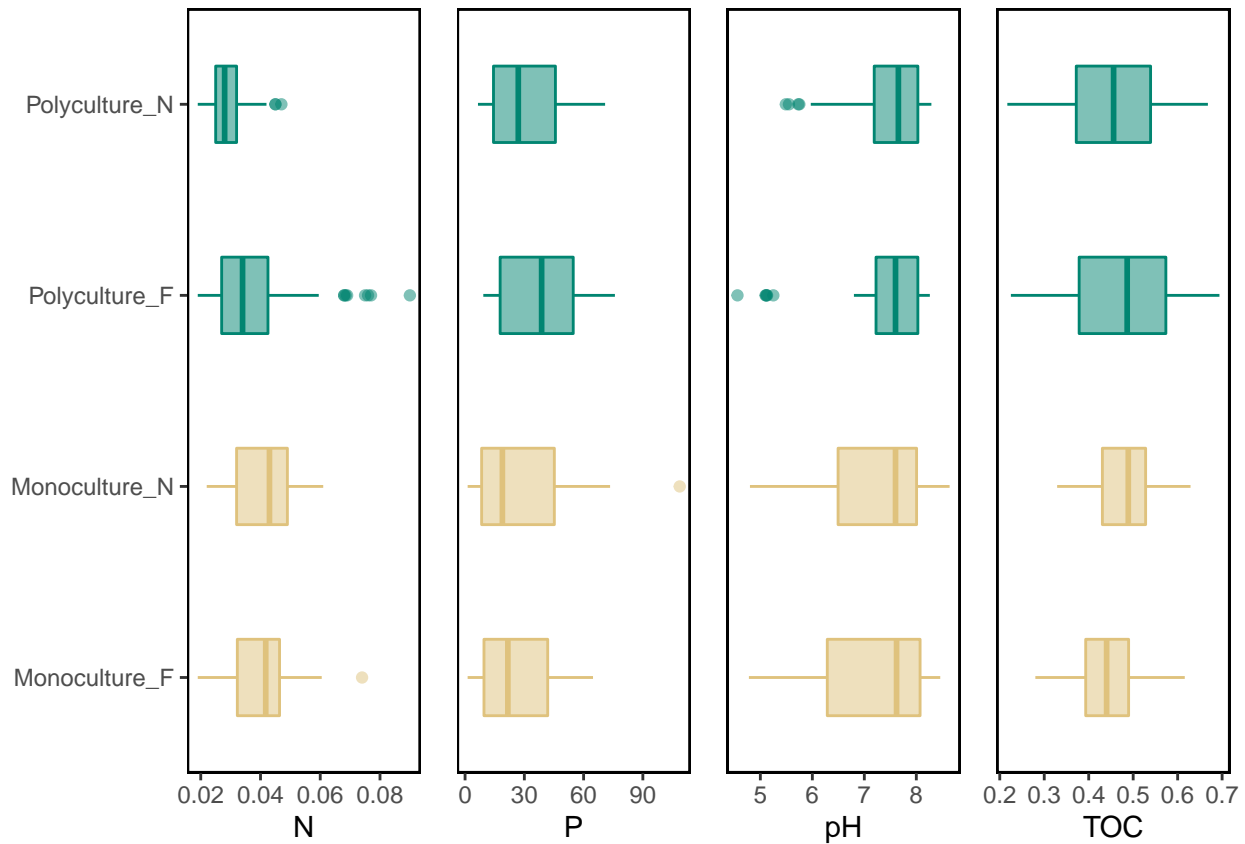


Model output

```
## $pH
##   response      covariate Estimate   SE    P sig
## 1 scale(pH)    cropDiversity  0.057 0.162 0.729  ns
## 2 scale(pH)      BlockN       0.051 0.066 0.438  ns
## 3 scale(pH) cropDiversity:BlockN 0.014 0.037 0.709  ns
##
## $P
##   response      covariate Estimate   SE    P sig
## 1 scale(P)      cropDiversity  0.356 0.148 0.026   *
## 2 scale(P)      BlockN        0.147 0.088 0.096  ns
## 3 scale(P) cropDiversity:BlockN -0.252 0.049 0.000 ***
##
## $TOC
##   response      covariate Estimate   SE    P sig
## 1 scale(TOC)    cropDiversity  0.239 0.146 0.118  ns
## 2 scale(TOC)    BlockN        0.469 0.117 0.000 ***
## 3 scale(TOC) cropDiversity:BlockN -0.322 0.065 0.000 ***
##
## $N
##   response      covariate Estimate   SE    P sig
## 1 scale(N)      cropDiversity  0.024 0.129 0.852  ns
## 2 scale(N)      BlockN        0.078 0.131 0.550  ns
## 3 scale(N) cropDiversity:BlockN -0.371 0.073 0.000 ***
```

Edaphic variables ~ farm type

Plots



Model output

```
## $pH
##   response                covariate Estimate   SE    P sig
## 1 scale(pH)   FarmTypePolyculture    0.233 0.408 0.575  ns
## 2 scale(pH)                BlockN     0.084 0.068 0.214  ns
## 3 scale(pH) FarmTypePolyculture:BlockN -0.030 0.094 0.747  ns
##
## $P
##   response                covariate Estimate   SE    P sig
## 1 scale(P)   FarmTypePolyculture    0.630 0.391 0.122  ns
## 2 scale(P)                BlockN     0.137 0.091 0.136  ns
## 3 scale(P) FarmTypePolyculture:BlockN -0.594 0.127 0.000 ***
##
## $TOC
##   response                covariate Estimate   SE    P sig
## 1 scale(TOC)   FarmTypePolyculture    0.354 0.373 0.353  ns
## 2 scale(TOC)                BlockN     0.409 0.122 0.001 ***
## 3 scale(TOC) FarmTypePolyculture:BlockN -0.668 0.171 0.000 ***
##
## $N
##   response                covariate Estimate   SE    P sig
## 1 scale(N)   FarmTypePolyculture   -0.096 0.320 0.766  ns
## 2 scale(N)                BlockN     0.051 0.136 0.705  ns
## 3 scale(N) FarmTypePolyculture:BlockN -0.854 0.189 0.000 ***
```

mean +/- SE

##	FarmType	Block	variable	mean	SE	min	max
## 1	Monoculture	-	cropDiversity	0.000	0.000	0.000	0.000
## 2	Polyculture	-	cropDiversity	2.444	0.050	1.242	3.511
## 3	Monoculture	-	N	0.041	0.001	0.019	0.074
## 4	Polyculture	-	N	0.034	0.001	0.019	0.090
## 5	Monoculture	-	observed	10.905	0.668	1.000	27.000
## 6	Polyculture	-	observed	19.099	1.443	1.000	68.000
## 7	Monoculture	-	P	27.171	2.100	1.283	108.662
## 8	Polyculture	-	P	34.519	1.867	6.527	75.843
## 9	Monoculture	-	pH	7.175	0.110	4.780	8.640
## 10	Polyculture	-	pH	7.425	0.078	4.560	8.290
## 11	Monoculture	-	shannon	1.483	0.063	0.000	2.804
## 12	Polyculture	-	shannon	1.874	0.075	0.000	3.551
## 13	Monoculture	-	TOC	0.463	0.007	0.280	0.629
## 14	Polyculture	-	TOC	0.462	0.011	0.217	0.694
## 15	-	F	cropDiversity	1.226	0.122	0.000	3.511
## 16	-	N	cropDiversity	1.286	0.124	0.000	3.511
## 17	-	F	N	0.040	0.001	0.019	0.090
## 18	-	N	N	0.035	0.001	0.019	0.061
## 19	-	F	observed	15.642	1.245	1.000	60.000
## 20	-	N	observed	14.579	1.172	1.000	68.000
## 21	-	F	P	32.516	1.986	1.283	75.843
## 22	-	N	P	29.348	2.027	1.297	108.662
## 23	-	F	pH	7.266	0.100	4.560	8.460
## 24	-	N	pH	7.341	0.091	4.800	8.640
## 25	-	F	shannon	1.714	0.075	0.000	3.377
## 26	-	N	shannon	1.652	0.068	0.000	3.551
## 27	-	F	TOC	0.459	0.010	0.225	0.694
## 28	-	N	TOC	0.466	0.009	0.217	0.668
## 29	Monoculture	F	cropDiversity	0.000	0.000	0.000	0.000
## 30	Monoculture	N	cropDiversity	0.000	0.000	0.000	0.000
## 31	Polyculture	F	cropDiversity	2.430	0.073	1.242	3.511
## 32	Polyculture	N	cropDiversity	2.457	0.068	1.242	3.511
## 33	Monoculture	F	N	0.041	0.002	0.019	0.074
## 34	Monoculture	N	N	0.041	0.001	0.022	0.061
## 35	Polyculture	F	N	0.039	0.002	0.019	0.090
## 36	Polyculture	N	N	0.029	0.001	0.019	0.047
## 37	Monoculture	F	observed	11.778	1.027	1.000	27.000
## 38	Monoculture	N	observed	9.980	0.831	1.000	21.000
## 39	Polyculture	F	observed	19.436	2.143	1.000	60.000
## 40	Polyculture	N	observed	18.768	1.954	1.000	68.000
## 41	Monoculture	F	P	26.157	2.577	1.283	64.761
## 42	Monoculture	N	P	28.244	3.374	1.297	108.662
## 43	Polyculture	F	P	38.759	2.790	9.226	75.843
## 44	Polyculture	N	P	30.353	2.381	6.527	70.880
## 45	Monoculture	F	pH	7.151	0.157	4.780	8.460
## 46	Monoculture	N	pH	7.201	0.157	4.800	8.640
## 47	Polyculture	F	pH	7.380	0.123	4.560	8.260
## 48	Polyculture	N	pH	7.469	0.099	5.490	8.290
## 49	Monoculture	F	shannon	1.503	0.093	0.000	2.804
## 50	Monoculture	N	shannon	1.461	0.085	0.000	2.465
## 51	Polyculture	F	shannon	1.921	0.111	0.000	3.377
## 52	Polyculture	N	shannon	1.827	0.100	0.000	3.551
## 53	Monoculture	F	TOC	0.444	0.010	0.280	0.616
## 54	Monoculture	N	TOC	0.483	0.010	0.329	0.629
## 55	Polyculture	F	TOC	0.474	0.018	0.225	0.694

## 56 Polyculture	N	TOC	0.451	0.014	0.217	0.668
-------------------	---	-----	-------	-------	-------	-------