Data repository readme for the manuscript [Accounting for forest community type gives] Novel insights into how the mean and heterogeneity of abiotic conditions together shape forb species richness patterns in the Allegheny plateau ecoregion.

All files last updated June 20, 2019

Folders:

• original abiotic factor predictors measured in each subplot (light, 13 soil chemistry variables, as well as litter, organic layer, and restrictive layer depths), from which we calculated 29 plot-level means and coefficients of variation. These three measurements (raw subplot values α_j , means calculated from raw subplot values $\bar{\alpha}_j$, and coefficients of variation calculated from raw subplot values for each plot $cv(\alpha_j)$) were transformed (indicated by a t in the output file name) and normalized (indicated by a z in the output file name) for comparison.

o input

- SPRING 2015 soil1.csv and SPRING 2015 soil2.csv
- SPRING 2015 light.csv
- SPRING 2015 litter_rd.csv
- SPRING 2015 abiotic predictors.R
- SUMMER 2015 soil1.csv and SUMMER 2015 soil2.csv
- SUMMER 2015 light.csv
- SUMMER 2015 litter rd.csv
- SUMMER 2015 abiotic predictors.R

o output

- spalpha.ab.t.z.csv (n=114 of 17 variables)
- spplot.ab.t.z.csv (n=29 of 34 variables)
- summalpha.ab.t.z.csv (n=114 of 17 variables)
- summplot.ab.t.z.csv (n=29 of 34 variables)
- **original community data** measured in each subplot (% cover of each species in each subplot) and in each plot (% cover of each species in each plot). Cover data was converted to presence-absence data to obtain subplot- and plot-level species counts (alpha and gamma species richness, γ_S and α_S).

o input

- SPRING 2015 seasonal designations.csv
- SPRING 2015 response.csv
- SUMMER 2015 seasonal designations.csv
- SUMMER 2015 response.csv

o output

- spalpha.patterns.forbs.csv and spplot.patterns.forbs.csv
- summalpha.patterns.forbs.csv and summplot.patterns.forbs.csv
- categorical data contained information on which land holding plots were located, and what tree species dominated the canopy.

land holdings:

- Brecksville Reservation (BR)
- Hinckley Reservation (HI)
- Mill Stream Run Reservation (MS)

- North Chagrin Reservation (NC)
- Rocky River Reservation (RR)
- West Creek Reservation (WC)

forest community types:

- Beech-Maple (BM) (Beech and/or Sugar Maple canopies, or Beech-Red
 Oak canopies)
- Floodplain (FP) (Sycamore, Cottonwood, and/or Black Walnut canopies)
- Mixed (M) (Tulip, Cherry, and/or Mixed canopy containing species from multiple other categories)
- Oak (OAK) (Red, White, Black, Scarlet, and/or Chinkapin Oak canopies)

analyses

modeling

- in which we dredge and compare models of species richness at the plotand subplot-levels using different abiotic measurements and different model structures (Table 1).
- \circ varpartitioning.R partitions contributions to plot-level species richness, γ_S that come from mean abiotic predictors compared to the heterogeneity in abiotic predictors
 - EXCLUDING trends along abiotic factors that change with forest type:
 - $\bar{\alpha}_{\mathit{Ca}}$, $cv(\alpha_{\mathit{P}})$, and $cv(\alpha_{\mathit{K}})$ in spring
 - $\bar{\alpha}_{Ca}$, $\bar{\alpha}_{P}$, $\bar{\alpha}_{K}$, $cv(\alpha_{pH})$, and $cv(\alpha_{N})$ in summer

- INCLUDING trends along pairs of abiotic factors that changed with forest type:
 - $\bar{\alpha}_{Ca}$ and $cv(\alpha_P)$, $\bar{\alpha}_{Ca}$ and $cv(\alpha_K)$, $\bar{\alpha}_{rd}$ and $cv(\alpha_P)$, $\bar{\alpha}_{rd}$ and $cv(\alpha_R)$, $\bar{\alpha}_N$ and $cv(\alpha_R)$, $\bar{\alpha}_N$ and $cv(\alpha_R)$ in each forest type in spring
 - $\bar{\alpha}_{Ca}$ and $cv(\alpha_{pH})$, $\bar{\alpha}_{Ca}$ and $cv(\alpha_N)$, $\bar{\alpha}_{Ca}$ and $cv(\alpha_{CEC})$, $\bar{\alpha}_P$ and $cv(\alpha_{pH})$, $\bar{\alpha}_P$ and $cv(\alpha_N)$, $\bar{\alpha}_P$ and $cv(\alpha_{CEC})$, $\bar{\alpha}_K$ and $cv(\alpha_{pH})$, $\bar{\alpha}_K$ and $cv(\alpha_N)$, $\bar{\alpha}_K$ and $cv(\alpha_{CEC})$ in each forest type in summer

Table 1: Location of the code for each type of full model compared in our analysis.

	abiotic predictor variables used in the full model			
model	$\overline{\alpha}_{j}$	$cv(\alpha_j)$	$\overline{\alpha}_i \& cv(\alpha_i)$	α_{j}
structure				
used				
A: linear	SPRING2015(1).R	SPRING2015(2).R	SPRING2015(1&2).R	
model	and	and	and	n/a
	SUMMER2015(1).R	SUMMER2015(2).R	SUMMER2015(1&2).R	
B: null				
linear				
mixed-				
effects	SPRING2015(1).R	SPRING2015(2).R	SPRING2015(1&2).R	SPRING2015(4).R
model with	and	and	and	and
grouping	SUMMER2015(1).R	SUMMER2015(2).R	SUMMER2015(1&2).R	SUMMER2015(4).R
structure				
but no				
predictors				
C1: linear				
mixed-	SPRING2015(1).R	SPRING2015(2).R	SPRING2015(1&2).R	SPRING2015(4).R
effects	and	and	and	and
model with	SUMMER2015(1).R	SUMMER2015(2).R	SUMMER2015(1&2).R	SUMMER2015(4).R
variable	SOMMENZOIS(I).N	SUMMERZUIS(Z).R	SUMMERZUIS(IQZ).N	301VIIVIENZU13(4).N
intercepts				
C2: linear				
mixed-	SPRING2015(1).R	SPRING2015(2).R	SPRING2015(1&2).R	SPRING2015(4).R
effects	and	and	and	and
model with	SUMMER2015(1).R	SUMMER2015(2).R	SUMMER2015(1&2).R	SUMMER2015(4).R
variable	301411412013(1).11	301411412013(2).11	30111112112013(102).11	301411412013(1).11
slopes				
D1: best				
linear				
mixed-	,			
effect(s)	SPRING2015(1).R	SPRING2015(2).R	SPRING2015(1&2).R	SPRING2015(4).R
model(s)	and	and	and	and
from C1	SUMMER2015(1).R	SUMMER2015(2).R	SUMMER2015(1&2).R	SUMMER2015(4).R
with				
variable				
slopes D2: best				
linear				
mixed-				
effect(s)	SPRING2015(1).R	SPRING2015(2).R	SPRING2015(1&2).R	SPRING2015(4).R
model(s)	and	and	and	and
from C2	SUMMER2015(1).R	SUMMER2015(2).R	SUMMER2015(1&2).R	SUMMER2015(4).R
with	JOIVIIVILIZUIJ(I).N	JOIVIIVILIZUIJ(Z).N	JOIVIIVILINZUIJ(IQZ).N	JOIVIIVILIZUIJ(4).N
variable				
intercepts				
intercepts				