

# Macroinvertebrate metacommunity in Mediterranean mountain ponds (South Spain)

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## Small water bodies

- High production
- High habitat heterogeneity
- High biological diversity
- High environmental value (ecosystems services)

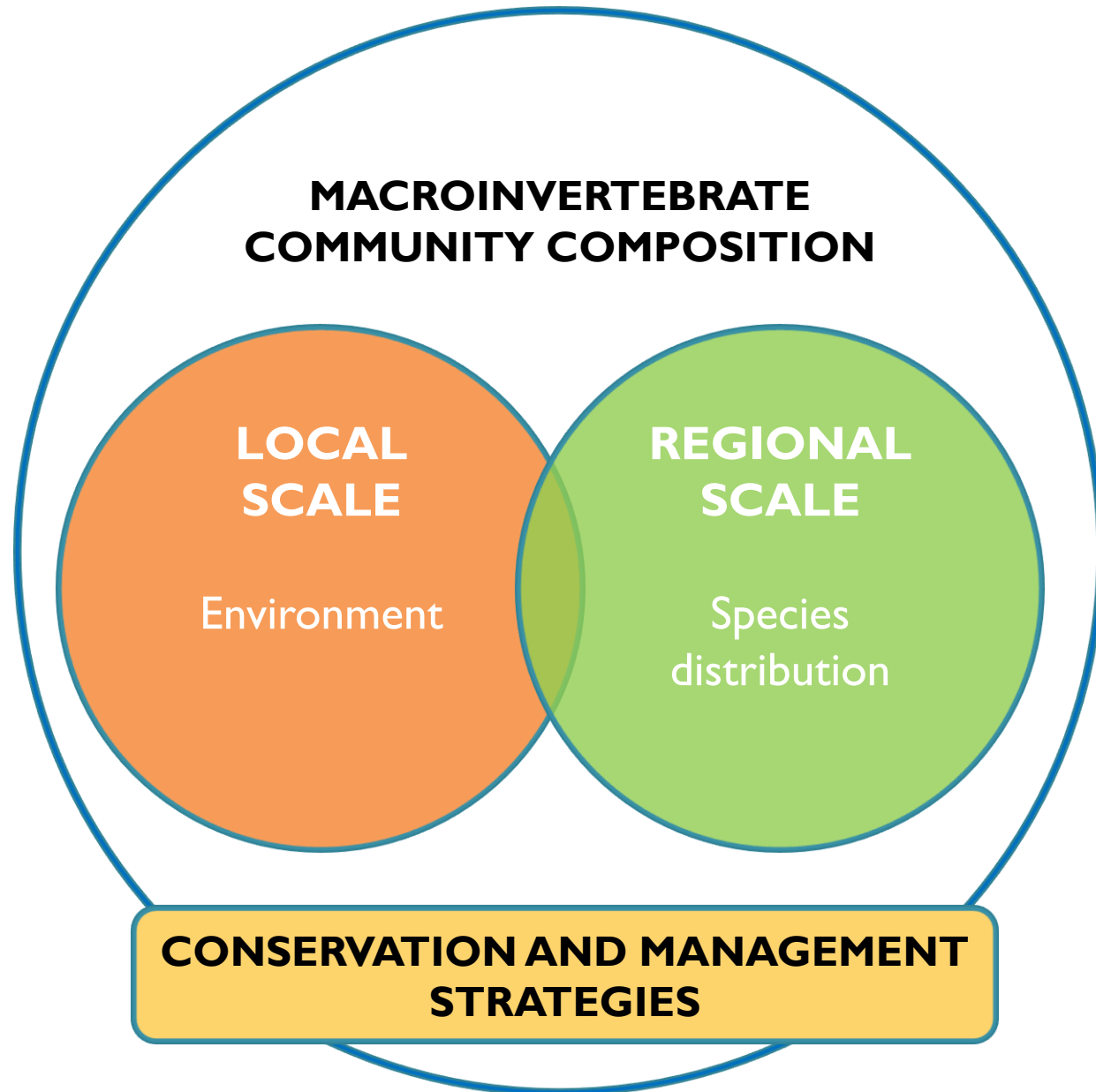


Isolated habitats inside a matrix of land



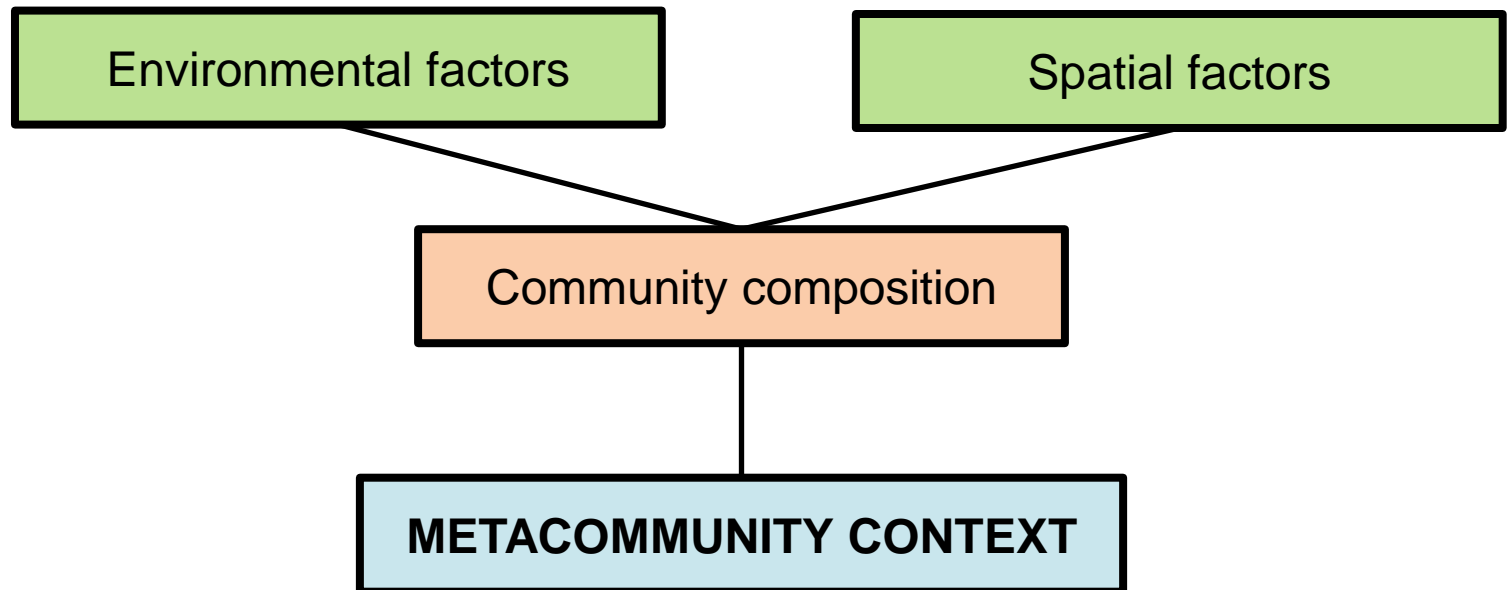
Human pressures → Degradation

**MEDITERRANEAN MOUNTAIN PONDS**



## HYPOTHESIS

Macroinvertebrate metacommunity of Mediterranean mountain ponds is a suitable element with implications in diversity maintenance and a key element to put forward wetlands conservation actions.



## STUDY AREA

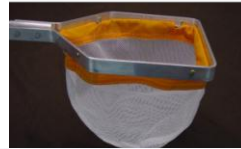
23 Mountain ponds Baetic Mountains  
Altitude: 387-2068 m.a.s.l  
Mediterranean climate: hydroperiod



Studied mountain ponds in the Baetic Mountains

## Macroinvertebrates samples

Hand net (250  $\mu\text{m}$ )



Spring 2017. Most ponds were dried

Sampling time:

>50 m<sup>2</sup>: 3 minutes

<50m<sup>2</sup> 30s per each 10m<sup>2</sup>



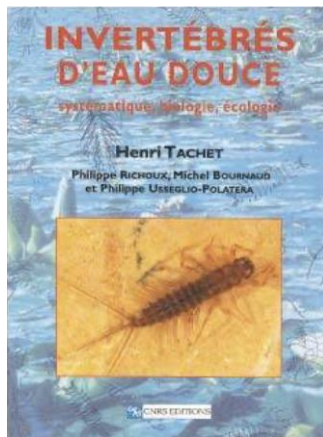
Sediments



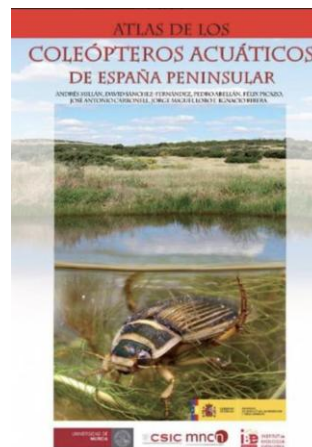
Open water



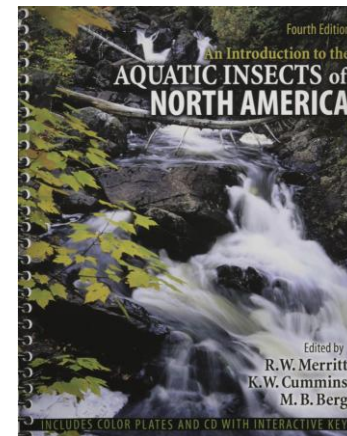
Submerged-emerged vegetation



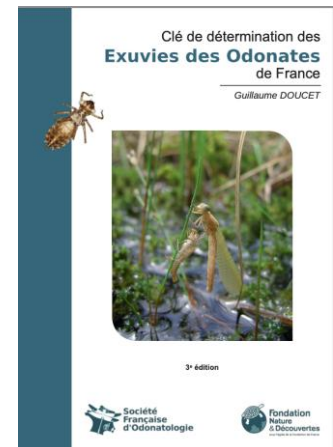
Tachet *et al.*, 2010



Millán-Sánchez *et al.*, 2014



Merritt *et al.*, 2008



Doucet, 2016

## Environmental variables

Water temperature and conductivity



Turbidity



Depth and  
pond size

Altitude

Substrate  
typology

Isolation

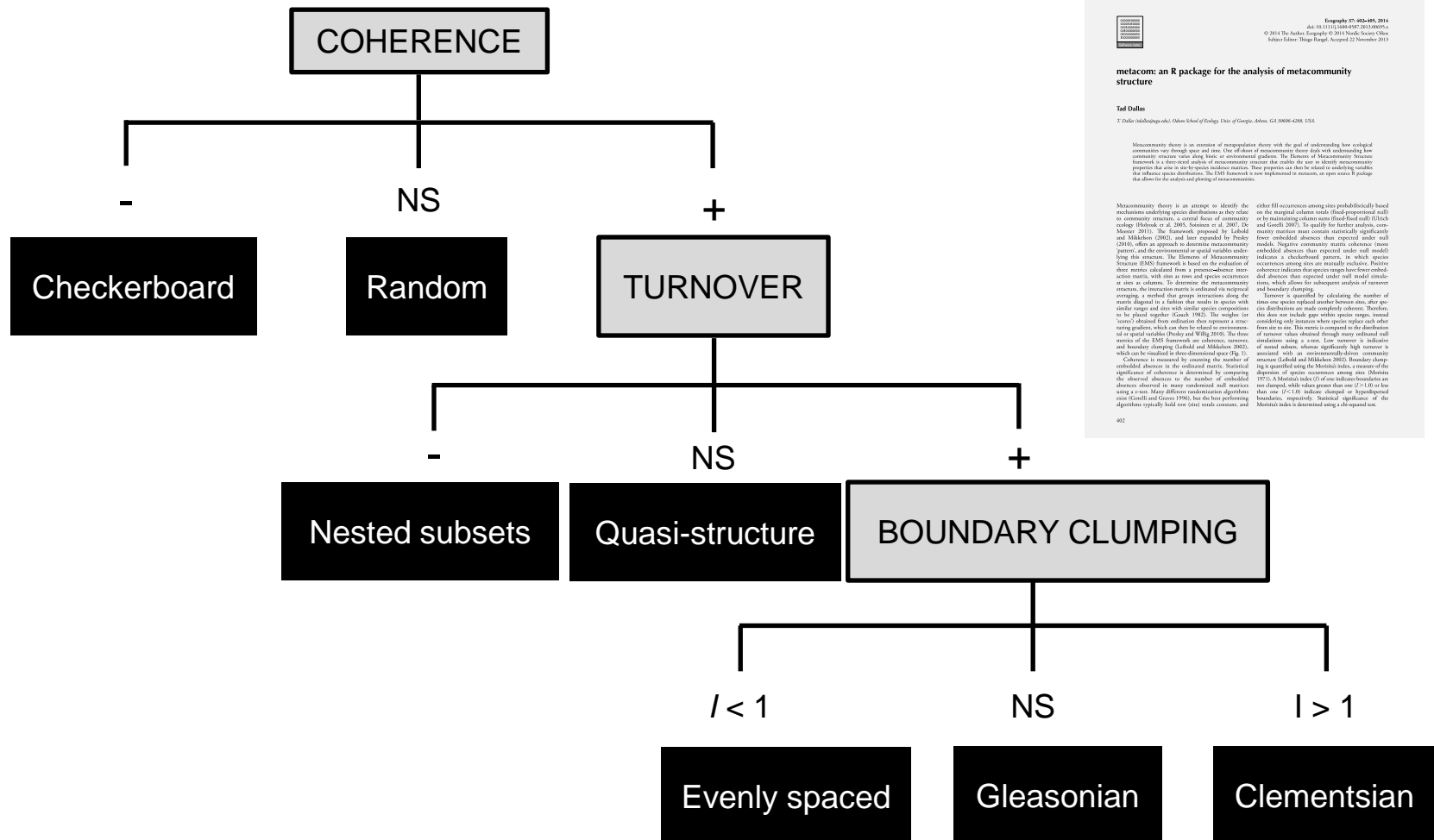
Hydroperiod

Nutrients: TP and TN



## Data analysis

## Elements of Metacommunity Structure EMS



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© 2014 The Authors. Ecography © 2014 Nordic Society Oikos  
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### metacom: an R package for the analysis of metacommunity structure

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Metacommunity theory is an extension of metacommunity theory with the goal of understanding how ecological communities vary through space and time. One of those of metacommunity theory deals with understanding how community structure varies along time or environmental gradients. The Elements of Metacommunity Structure framework is a structured analysis of metacommunity structure that enables the user to identify metacommunity properties like evenness, species incidence patterns. This program can then be used to understand whether the observed species distribution, by the EMS framework is more implemented in metacommunity structure. This package allows for the analysis and plotting of metacommunities.

Metacommunity theory is an attempt to identify the mechanisms underlying species distributions as they relate to community structure, a central focus of community ecology (Hubbell et al. 2005; Senterre et al. 2007; De Meester 2011). The framework proposed by Leibold and Mikšinskas (2002), and later expanded by Forber (2010), offers an approach to describe metacommunity patterns, and the environmental or spatial variables underlying this structure. The Elements of Metacommunity Structure (EMS) framework is based on the evaluation of three metrics calculated from a species-by-site incidence matrix, with data on time and space occurrence as well as on dispersal. To describe the metacommunity structure, the incidence matrix is evaluated on regional averaging a method that groups iterations along the matrix diagonal in a fashion that results in species with similar ranges and sites with similar species compositions to be placed together (Stamps 1982). The weights (or 'scores') obtained from calculations then represent a community gradient, which can then be related to environmental or spatial variables (Forber and Willig 2010). The three metrics of the EMS framework are coherence, turnover, and boundary clumping (Leibold and Mikšinskas 2002), which can be visualized in three-dimensional space (Fig. 1).

Coherence is measured by counting the number of contiguously adjacent sites in the regional matrix. Statistical significance of coherence is determined by comparing the observed coherence to the number of contiguously adjacent sites in many randomized null matrices using a one-tailed different randomization algorithm (see Stenseth and Gross 1996), has the best performing algorithm typically build new 'null' matrix contents, and either fill occurrences among sites probabilistically based on the original coherence with fixed proportional nulls or by maintaining column sums fixed but nulls (Black and Green 2007). To qualify for further analysis, community matrices must contain statistically significantly fewer contiguously adjacent sites than expected under null models. Negative community matrix coherence (more contiguously adjacent sites than expected under null models) indicates a checkerboard pattern, in which species occurrences among sites are mutually exclusive. Positive coherence indicates that species ranges have been extended, which allows for the interpretation of turnover and boundary clumping.

Turnover is quantified by calculating the number of sites one species replaced another between sites, after species distribution is made completely coherent. Therefore, this does not include gaps within species ranges, instead considering only species where species replace each other from site to site. This metric is compared to the distribution of turnover values derived through many randomized null simulations using a one-tailed test. Low turnover is indicative of nested subsets, whereas significantly high turnover is associated with an intermediate-level community structure (Leibold and Mikšinskas 2002). Boundary clumping is a quantified using the null matrix, a measure of the dispersion of species occurrences among sites (Morris 1971). Mikšinskas index (CI) of null matrix boundaries are not changed, while values greater than one ( $I > 1$ ) or less than one ( $I < 1$ ) indicate clumped or aggregated boundaries, respectively. Statistical significance of the Mikšinskas index is determined using a chi-squared test.

## Data analysis

## RDA analysis and variation partitioning analysis

06 October 2017 | 10:12:47  
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COMMUNITY ECOLOGY • ORIGINAL RESEARCH

**Metacommunity ecology meets biogeography: effects of geographical region, spatial dynamics and environmental filtering on community structure in aquatic organisms**

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**Abstract** Metacommunity patterns and underlying processes in aquatic organisms have typically been studied within a drainage basin. We examined variation in the composition of six freshwater epifaunal groups across various drainage basins in Finland. We first modelled spatial structure within each drainage basin using Moran's eigenfunction maps. Second, we partitioned variation in community structure across three groups of predictor using canonical ordination: (1) local environmental variables, (2) spatial variables, and (3) density variables through-basin identity. Third, we examined turnover and nestedness components of multiplicative beta diversity, and tested the best fit partition of our density using the 'variance of metacommunity structure' analysis. Our results showed that basin identity and local environmental variables were significant predictors of community structure, whereas within-basin spatial effects were typically negligible. In half of the regional groups (diatoms, bryophytes, zooplankton), basin identity was a slightly better predictor of community structure than local environmental variables, whereas the opposite was true for the remaining three regional groups (insects, macrophytes, fish). Both pure basin and local environmental factors were, however, significant when accounting for the effects of the other predictor variable sets. All regional groups exhibited high levels of beta diversity, which was mostly attributable to the turnover component. Our results showed consistent Chao2-metacommunity structure patterns, suggesting that subgroups of species responded similarly to environmental factors or drainage basin limits. We conclude that aquatic communities across large scales are mostly determined by environmental and basin effects, which leads to high beta diversity and prevalence of Chao2-metacommunity types.

**Keywords** Bryophytes • Diatoms • Fish • Invertebrates • Lakes • Macrophytes • Metacommunity • Streams

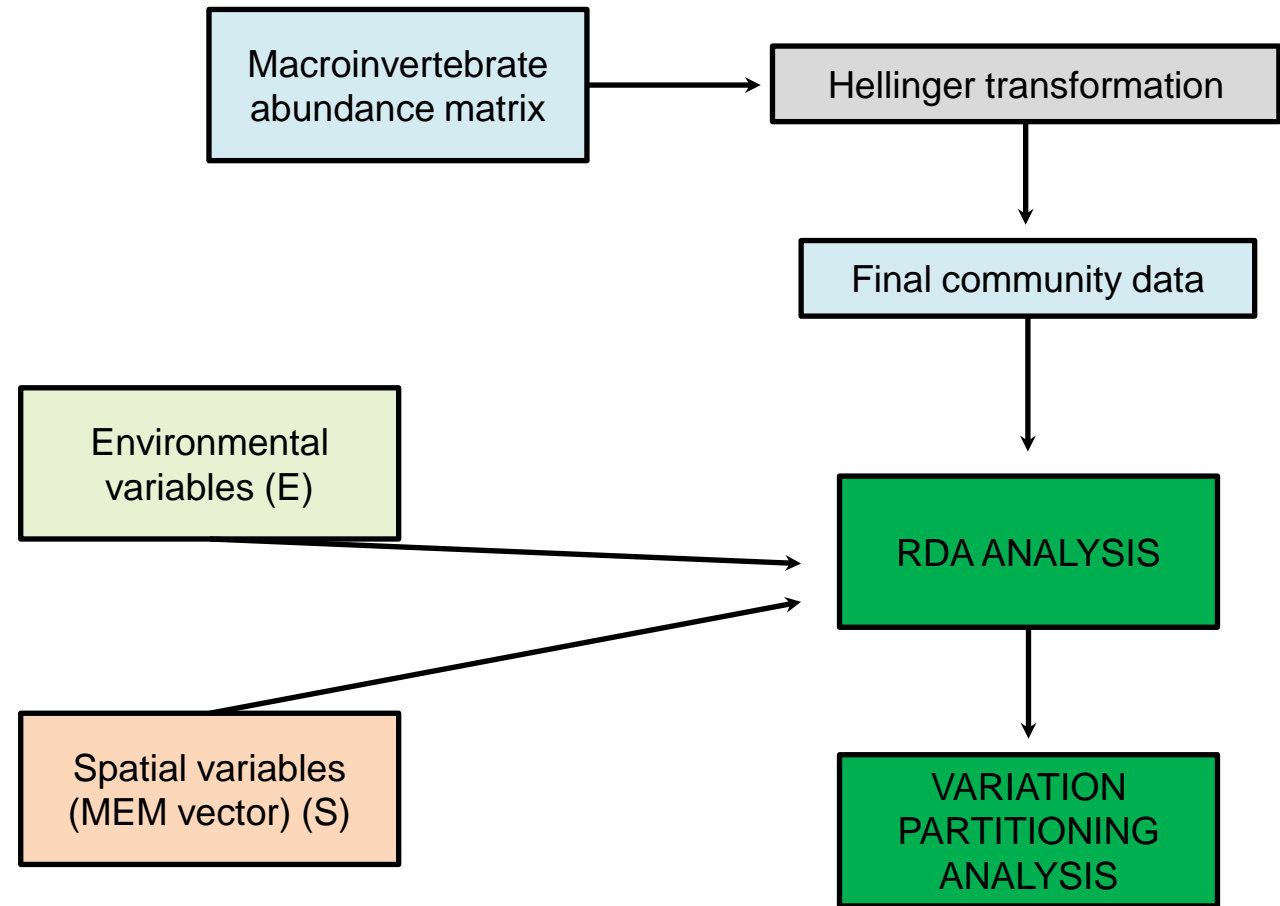
**Introduction**

Biogeography and community ecology are two disciplines that combine history, dispersal, biotic interaction and environmental filtering to determine the structure of biotic assemblages. However, a better understanding of the determinants of biotic assemblages might benefit from a closer conceptual articulation of these disciplines (Jenkins and

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# RESULTS

	Conductivity (mS/cm)	Turbidity (FTU) (FTU)	Total nitrogen	Total phosphorous	Elevation (m.a.s.l)	Hydroperiod	Size	Deep	Subst	Isolation
Laguna del Picacho del Algibe	0.21	6.94	2.01	147.23	538	T	2026.4	1.5	S	0.82
Charca de los Llanosde Líbar	0.58	3.58	0.32	62.73	970	P	380.65	3	L	0.75
Laguna de Coripe	0.62	45.33	4.85	391.87	406	T	17908.9	2	T	0.86
Laguna de Caja	0.86	137	2.31	263.55	729	T	59161.2	3	T	0.57
Laguna del Hondonero	0.33	3.9	0.57	19.95	1162	T	913.68	0.75	L	0.06
Charca Loma León	0.94	3.76	0.37	16.3	387	T	551.17	2.5	T	<b>0.67</b>
Laguna Chica	10.4	1.93	1.82	25.69	794	P	67113.5	5	T	0.04
Charco del Nevazo Largo	0.15	2.36	0.84	22.56	1347	T	2336.88	1	L	0.16
Laguna del Rico	0.34	8.17	3.08	2428.83	896	P	5752.28	2.5	L	0.65
Charca de Juan Ramos	0.29	3.51	2.57	80.46	1317	P	17.34	0.4	L	0.91
Balsa del Almiar	0.11	5.46	0.48	53.34	1775	P	1197.69	1.5	S	0.78
Manantial de la Cuerda del Alguacil	0.03	1.63	0.45	77.59	2068	T	22.74	0.2	S	0.85
Balsa Sabinar	0.12	23.27	1.01	110.71	1830	T	4787.22	2.5	L	0.54
Balsa Barjalí	0.13	61	1.55	175.4	1713	T	4084.82	1.5	L	0.42
Balsa Calabrial	0.28	67	0.52	45.51	1340	P	1513.04	2.5	L	0.69
Charca Filabres	0.18	72	8.82	712.67	1050	T	180.06	0.5	S	0.88
Charca de Balax	0.07	3.73	1.2	37.17	1892	P	51.81	0.6	S	0.68
Balsa Pocico	1.56	2.6	0.73	13.17	1275	P	283.71	1.25	L	0.74
Charca de la Franciscuela	0.31	1.54	0.26	12.13	1340	P	265.15	2	L	0.87
Laguna de Castril	0.28	>1000	28.54	3310.38	1967	T	959.03	0.5	L	0.9
Charco de la Tiná de las Cruces	0.42	41.27	1.94	104.46	1661	P	229.71	2	L	0.82
Laguna de Orcera	0.35	8.09	4.02	110.45	1264	T	4669.39	2	L	0.82
Laguna de Siles	0.25	6.5	0.54	14.21	1288	T	4406.9	2.75	L	0.85

Conductivity: 0.07 - 10.4 mS/cm

Turbidity: 1.54 - 1000 FTU

Total Nitrogen: 0.26 - 28.54 mg/L

Total Phosphorus: 12.12 - 3310.38 µg/L

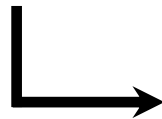
Substrate: limestone (60.8%) - siliceous (17.4%)



## Elements of Metacommunity Structure EMS

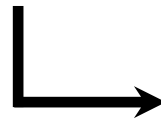
COHERENCE  
+

Embedded Absences= 512  
 $z = -6.70$   
 $p < 0.001$



TURNOVER  
+

Replaces= 5908  
 $z = -6.12$   
 $p < 0.001$



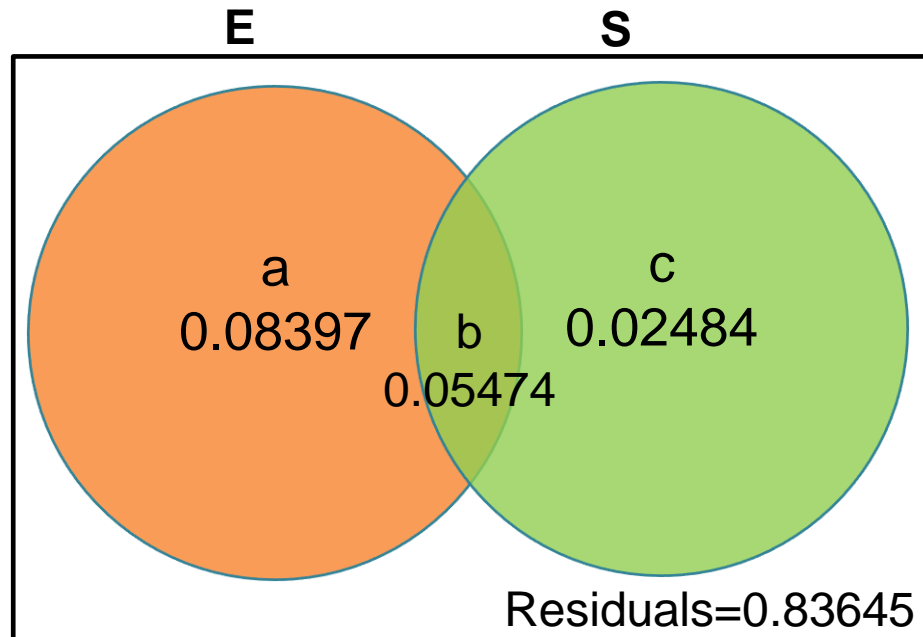
BOUNDARING  
CLUMPING  
+

Morisita's Index= 1.41  
 $p < 0.001$

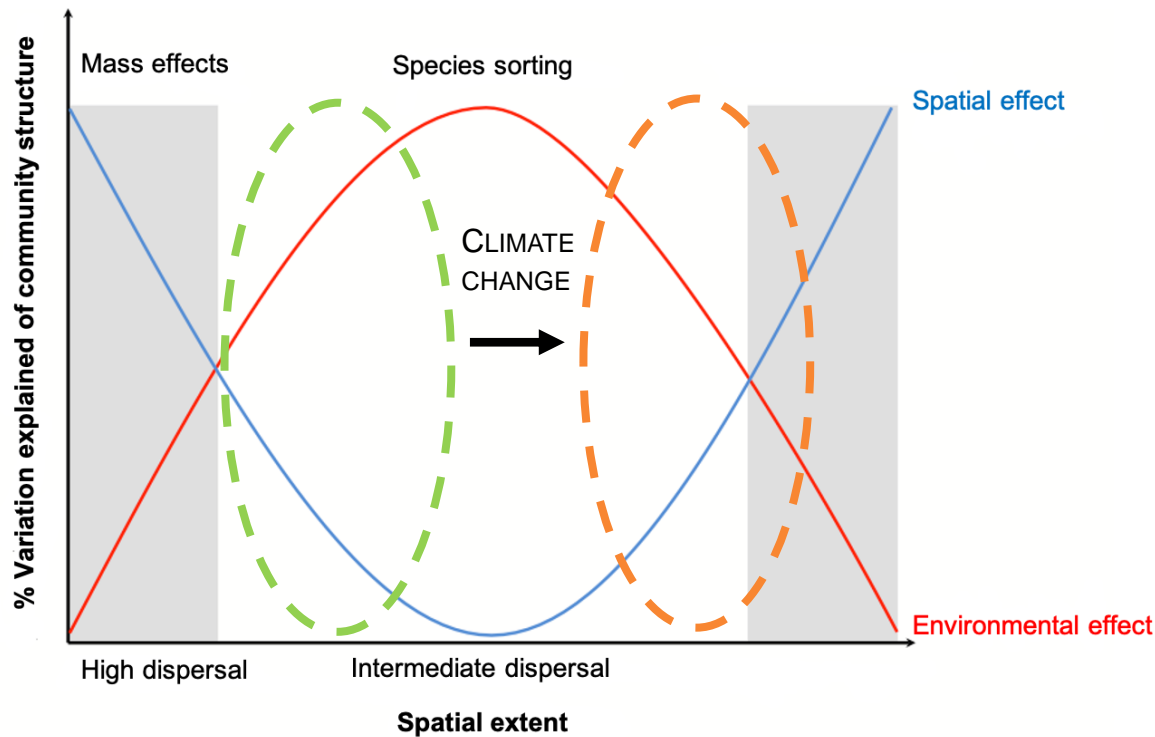
**CLEMENTSIAN PATTERN**

## RDA analysis and variation partitioning analysis

Fraction	df	Adj.R2	p	Variables in the model
$E (a \cup b)$	3	0.13871	0.001	Size, hydroperiod and altitude
$S (b \cup c)$	2	0.07958	0.001	MEM1 and MEM4
$E \cap S (b)$	5	0.05474	0.001	Size, hydroperiod, altitude, MEM1 and MEM4
$1-(E+S)$		0.83645		



The environmental variation in macroinvertebrate community composition is due by environmental variables



Modified figure from Heino *et al.*, 2015

Environmental variables-Species sorting

MEM vectors-Mass effects

THANKS

