

Ecography

ECOG-02683

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Supplementary material

1 **Appendix 1:** COI data set.

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3 COI data set for the 3028 specimens. MDSs: species hypotheses (SHs) delimited using
4 morphology; TH: SHs delimited with the 16% COI divergence threshold; PTP (and their
5 support): SHs delimited with the Poisson tree processes model; bPTP (and their support): SHs
6 delimited with the Bayesian implementation of the Poisson tree processes model; Long.:
7 longitude of the locality, in decimal degrees; Lat.: latitude of the locality, in decimal degrees;
8 Dataset: “COI tree only” for sequences used only for species delimitation and “COI tree &
9 Europe” for sequences used both for species delimitation and macroecological analysis.
10 Range data: sequences used for the analysis of range size pattern.

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12 The table is too long to be embedded as Appendix in the main document but it is available at
13 the following address:

14 <https://drive.google.com/open?id=0BxluaQPA8MokbURaMzBjQ25mdmM>

15 and this table will be available in Dryad if the paper is accepted.

1 **Appendix 2:** Supplementary tables and figures.

2 **Table A1:** Taxonomic comparison among the four sets of species hypotheses (SHs) generated using morphology (Morph.), a COI divergence
3 threshold (TH), the Poisson tree processes model (PTP) and its Bayesian implementation (bPTP). Taxonomic inflation is defined as the number
4 of SHs in set y divided by the number of SHs in set x. See text and Figure 2 for definitions of match, lump, split and reshuffling.

Dataset x		Dataset y		Taxonomic inflation	Shift from x to y				Shift from y to x			
Type	Number of SHs	Type	Number of SHs		Match	Lump	Split	Reshuffling	Match	Lump	Split	Reshuffling
Morph.	263	TH	519	1.97	134	1	348	36	134	69	21	39
Morph.	263	PTP	646	2.46	131	1	492	22	131	95	10	27
Morph.	263	bPTP	650	2.47	133	1	496	20	133	97	9	24
TH	519	PTP	646	1.24	428	1	214	3	428	83	3	5
TH	519	bPTP	650	1.25	425	2	221	2	425	86	4	4
PTP	646	bPTP	650	1.01	597	13	40	0	597	18	31	0

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7 **Table A2.** Independent and shared contributions of historical climate variability (H),
8 productive energy (E) and spatial heterogeneity (S) to variation in species richness of obligate
9 groundwater crustaceans in Europe. . Ex_Var: total proportion of variance explained by the
10 models (%). In abbreviations, colons denote shared variance between mechanisms. Variance
11 partitioning is shown for the four different sets of species hypotheses delimited using
12 morphology (Morph.), a COI divergence threshold (TH), the Poisson tree processes model
13 (PTP) and its Bayesian implementation (bPTP). GLM.nb: generalized linear model with
14 negative binomial error; OLS/GLS: ordinary/generalized least squares models.

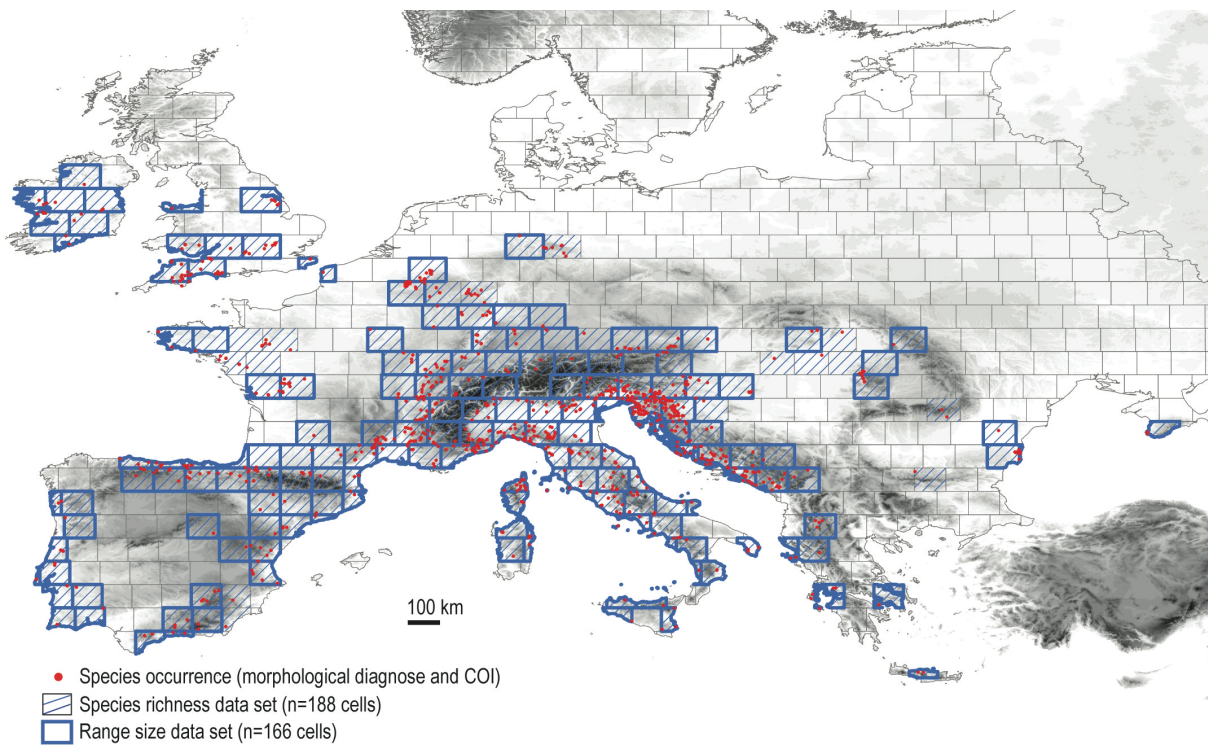
Dataset	Model type	Ex_Var	Pure components			Shared components			
			H	E	S	H:E	E:S	H:S	H:S:E
Morph.	GLM.nb	28.54	2.02	13.45	4.51	5.44	7.04	-1.53	-2.39
	OLS	16.61	0.35	8.34	4.06	2.53	3.82	-0.99	-1.51
	GLS	18.11	0.53	8.41	4.66	3.11	3.67	-0.73	-1.55
TH	GLM.nb	27.54	0.66	11.56	5.47	4.15	8.83	-0.49	-2.65
	OLS	16.46	-0.36	8.51	4.02	1.75	4.10	-0.15	-1.40
	GLS	18.33	0.21	8.85	4.69	1.78	4.11	0.01	-1.31
PTP	GLM.nb	26.58	0.57	10.53	5.70	3.96	8.96	-0.44	-2.68
	OLS	16.44	-0.72	8.34	4.53	1.29	4.27	0.15	-1.41
	GLS	18.39	-0.10	8.91	5.29	1.13	4.02	0.25	-1.11
bPTP	GLM.nb	26.80	0.58	10.72	5.54	3.99	9.04	-0.47	-2.60
	OLS	16.32	-0.67	8.34	4.21	1.31	4.31	0.20	-1.37
	GLS	18.31	0.00	9.06	4.94	0.97	3.97	0.31	-0.94

16 **Table A3:** Independent and shared contributions of historical climate variability (H), habitat
 17 area/heterogeneity (S) and climate seasonality (S) to variation in median range size of
 18 obligate groundwater crustaceans in Europe. Ex_Var: total proportion of variance explained
 19 by the models (%). In abbreviations, colons denote shared variance between mechanisms.
 20 Variance partitioning is shown for the four different sets of species hypotheses delimited
 21 using morphology (Morph.), a COI divergence threshold (TH), the Poisson tree processes
 22 model (PTP) and its Bayesian implementation (bPTP). OLS/GLS: ordinary/generalized least
 23 squares models.

Dataset	Model type	Ex_Var	Pure components			Shared components			
			H	S	C	H:S	S:C	H:C	H:S:C
Morph.	OLS	19.42	15.325	0.144	-0.506	0.314	0.114	3.409	0.622
	GLS	18.19	15.244	-0.870	-1.411	-0.407	-0.165	4.139	1.660
TH	OLS	33.37	16.167	1.340	1.870	1.083	-1.469	11.726	2.651
	GLS	33.95	24.733	1.880	2.850	-7.956	-1.995	6.328	8.110
PTP	OLS	40.14	20.781	1.884	1.207	0.582	-0.995	13.610	3.073
	GLS	41.70	31.314	2.733	2.077	-9.742	-1.246	8.272	8.295
bPTP	OLS	40.90	19.766	2.406	1.444	1.146	-1.085	13.862	3.354
	GLS	42.78	28.734	3.532	2.553	-7.566	-1.209	11.489	5.243

25 **Figure A1:** Map showing the location of sampling sites and cells used for analyzing the
26 determinants of species richness and median range size of groundwater crustaceans in Europe.
27 The cell area of 0.9°-latitude cells was kept constant (10,000 km²) all over the grid by
28 adjusting the longitudinal divisions between adjacent cells in each latitudinal band (see Sastre,
29 P. et al. 2009. A geoplatform for the accessibility to environmental cartography. – J.
30 Biogeogr. 36: 568).

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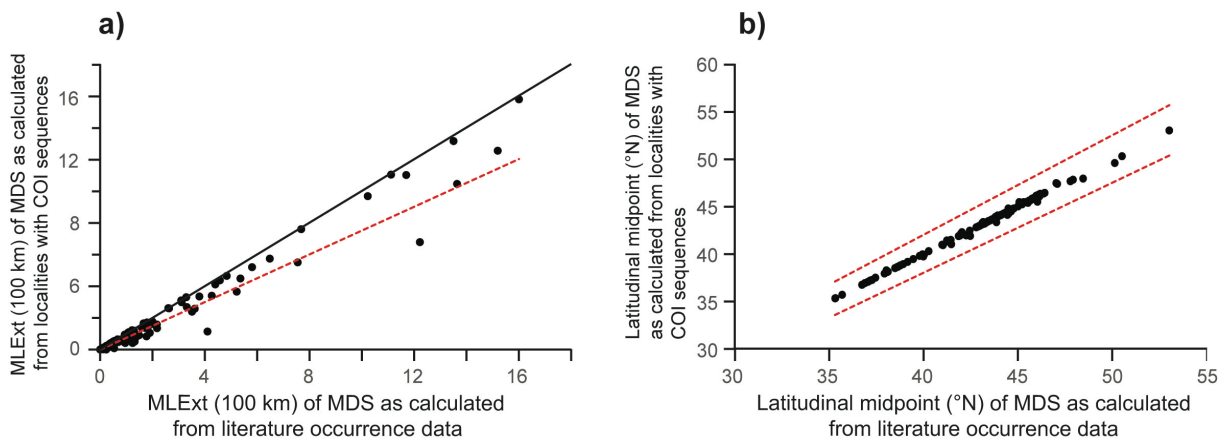


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35 **Figure A2:** Plots of maximum linear extent (MLExt) (a) and latitudinal midpoints (b) of
36 selected morphologically distinguishable species (MDSs, n=147) as measured from localities
37 with COI sequences against the MLExt and latitudinal midpoints of MDSs as measured from
38 occurrence data from the literature. The solid black line and broken red line in (a) represent
39 the 1:1 equivalence line and the 0.75 interval line (i.e. the MLExt in y is 0.75 time the MLExt
40 in x), respectively. Broken red lines in (b) represent the +/- 0.95 interval lines.



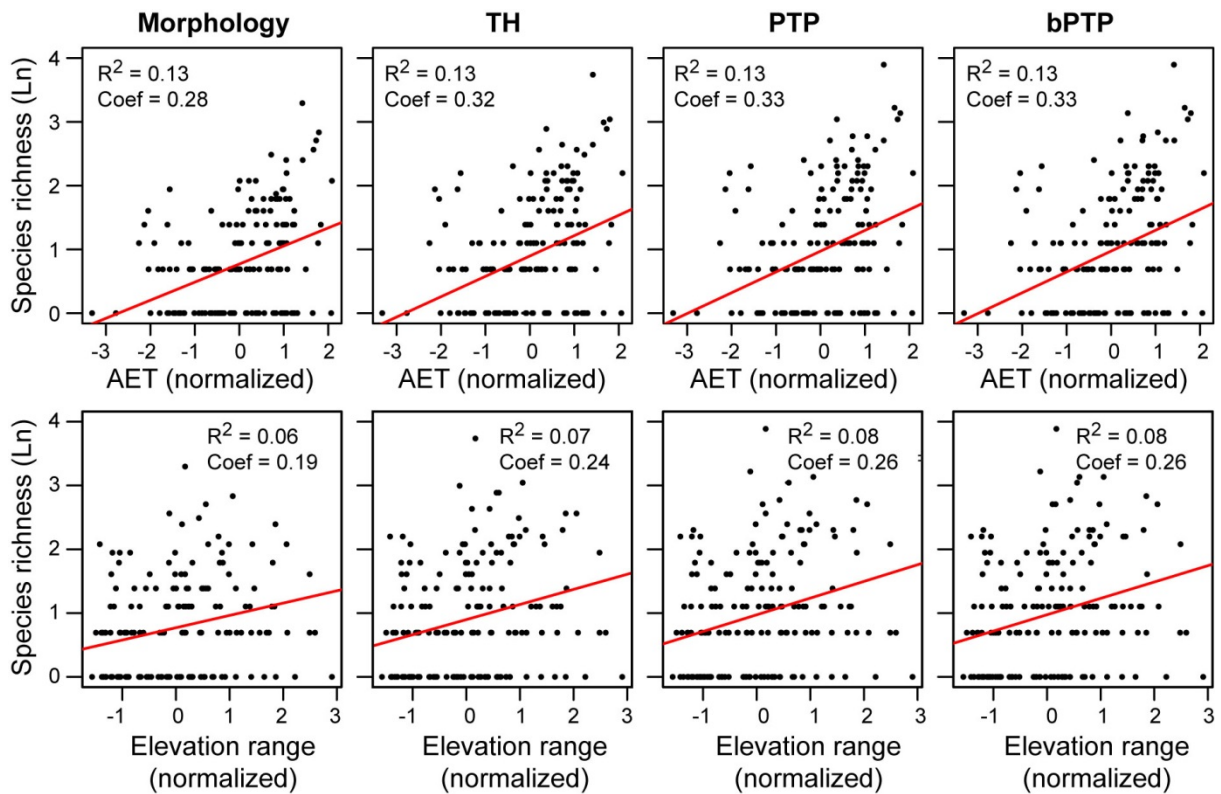
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44 **Figure A3:** Pairwise relationships between species richness (log-transformed) and actual
 45 evapotranspiration (AET) and elevation range (n=188 cells). The red line represents the fit of
 46 an ordinary least squares model. Morphology: species hypotheses (SHs) delimited using
 47 morphology; TH: SHs delimited using a COI divergent threshold; PTP: SHs delimited using
 48 the Poisson tree processes; bPTP: SHs delimited using the Bayesian implementation of the
 49 Poisson tree processes.

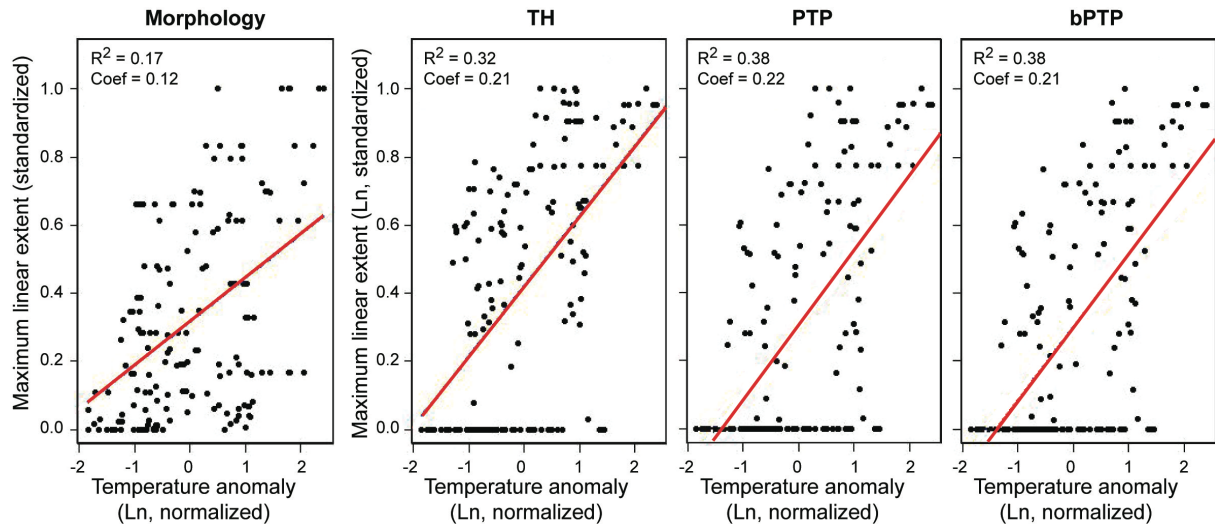
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52 **Figure A4:** Pairwise relationships between maximum linear extent and temperature anomaly
53 (n=166 cells). The red line represents the fit of an ordinary least squares model. Morphology:
54 species hypotheses (SHs) delimited using morphology; TH: SHs delimited using a COI
55 divergent threshold; PTP: SHs delimited using the Poisson tree processes; bPTP: SHs
56 delimited using the Bayesian implementation of the Poisson tree processes.

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1 **Appendix 3:** Supplementary methods.

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3 *Predictors of species richness*

4 Spatial heterogeneity was represented by elevation range and habitat diversity. Elevation
5 range is a proxy for topographic heterogeneity (Leprieur et al. 2011) and was calculated as the
6 highest difference in elevation between any two locations in a cell using elevation data from
7 the pan-European river and catchment database (Vogt et al. 2007). Habitat diversity was
8 estimated for each cell using the Shannon's diversity index defined as $H' = -\sum p_i \times \ln p_i$,
9 where p_i represented the areal proportion of 12 groundwater habitat types. The areal
10 proportions of each habitat were computed from the European vector map of groundwater
11 habitats (Cornu et al. 2013). To represent climate/productivity, we used mean annual actual
12 evapotranspiration (AET) as an index of water-energy dynamics that subsumes ambient
13 energy (temperature) and water availability (precipitation) (O' Brien, 2006). Data on AET
14 were obtained from the 30 arc-second resolution world map released by Trabucco and Zomer
15 (2010). To test for the effect of history, we computed temperature and precipitation
16 anomalies, defined as the differences in mean annual temperature and annual precipitation
17 between the present and the Last Glacial Maximum (LGM) (Leprieur et al. 2011). Data for
18 the LGM were extracted from two global circulation models, CCSM and MIROC2 (Hijmans
19 et al. 2005), and values from both were averaged to account for variation between models.

20

21 *Predictors of median range size*

22 For range size, we used seven predictors to test for the role of three influential mechanisms –
23 habitat area/heterogeneity, climate seasonality and historical climate variability. Habitat
24 area/heterogeneity was represented by elevation range, AET, aquifer area, and climatic rarity.
25 Aquifer area was calculated using the European vector map of groundwater habitats as the

26 total area of aquifer available within a 1000-km radius around a cell (Cornu et al. 2013). To
27 compute climatic rarity, we extracted and averaged for each grid cell data on mean annual
28 temperature and all eight bioclimatic precipitation layers of the WorldClim dataset, except
29 precipitation seasonality (30 arcsec resolution; Hijmans et al. 2005). A normalized principal
30 components analysis (PCA) was then performed on grid-cell averaged data, log-transformed
31 when appropriate to approximate a normal distribution. The first two PCA axes accounted for
32 86.8% of the climatic variance. They were used to calculate the average Euclidian distance in
33 climatic space between each cell and all neighboring cells within a radius of 500 km. High
34 average distance values correspond to cells showing rare climates relative to their neighboring
35 cells (Morueta-Holme et al. 2013). For climate seasonality, we computed the per-cell-average
36 of precipitation seasonality from the WorldClim dataset (Hijmans et al. 2005) as a surrogate
37 of present intra-annual environmental variability because there is very little thermal
38 seasonality in groundwater.

39

40 *Statistical procedure to take into account spatial autocorrelation in the residuals of species*
41 *richness and median range size models*

42 We performed generalized least square (GLS) models in order to take into account spatial
43 autocorrelation in the residuals of the OLS models (Beale et al. 2010). Using the OLS model
44 including all predictors, we selected the best spatial variogram structure of the residuals and
45 tested for the presence of a nugget effect using restricted maximum-likelihood optimization
46 based on the AICc criterion (Zuur et al. 2009). The GLS model was fitted using maximum-
47 likelihood optimization with the best spatial residual structure. Then, variance partitioning
48 and multimodel inference were performed as described in the section “Determinants of spatial
49 variation in species richness and median range size” of the Materials and Methods. As a
50 measure of explained variance, we used pseudo- R^2 , defined as the squared Pearson correlation

51 coefficient between the fitted and observed values of species richness (Ballesteros-Mejia et al.
52 2013).

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75 York.

Appendix 4: Supplementary acknowledgments

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