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When one phenotype is not enough

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1 **When one phenotype is not enough – divergent evolutionary trajectories govern venom**
2 **variation in a widespread rattlesnake species**

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Abstract

Understanding the origin and maintenance of phenotypic variation, particularly across a continuous spatial distribution, represents a key challenge in evolutionary biology. For this, animal venoms represent ideal study systems: they are complex, variable, yet easily quantifiable molecular phenotypes with a clear function. Rattlesnakes display tremendous variation in their venom composition, mostly through strongly dichotomous venom strategies, which may even coexist within single species. Here, through dense, widespread population-level sampling of the Mojave rattlesnake, *Crotalus scutulatus*, we show that genomic structural variation at multiple loci underlies extreme geographic variation in venom composition, which is maintained despite extensive gene flow. Unexpectedly, neither diet composition nor neutral population structure explain venom variation. Instead, venom divergence is strongly correlated with environmental conditions. Individual toxin genes correlate with distinct environmental factors, suggesting that different selective pressures can act on individual loci independently of their co-expression patterns or genomic proximity. Our results challenge common assumptions about diet composition as the key selective driver of snake venom evolution and emphasize how the interplay between genomic architecture and local-scale spatial heterogeneity in selective pressures may facilitate the retention of adaptive functional polymorphisms across a continuous space.

Introduction

The origin and genetic basis of phenotypic variation, and its retention in a population in the face of both random and deterministic forces, are pivotal questions for our understanding of evolutionary adaptations. Functional polymorphisms typically segregate in spatially isolated populations [1,2] and/or discrete ecological conditions [3-5]. In contrast, it is much more challenging to dissect the evolutionary processes involved in adaptive geographic variation across a continuous spatial distribution [6]. As a result, relatively few studies have comprehensively examined the relationship between genomic architecture, the resulting phenotypic variation and the ecological pressures maintaining that variation in continuously distributed organisms [2].

Animal venoms ~~are potent cocktails of bioactive molecules and~~ represent exemplar models for investigating the genetic basis of phenotypic variation [7]. Genes encoding for venom toxins are uniquely expressed in distinct, specialized glands, and their final product can be easily detected and quantified. This sidesteps the problem of pleiotropy in the genes involved in adaptive polygenic traits, which often obscures the phenotypic effects of individual genetic variants [8,9].

Rattlesnakes (*Crotalus*) produce highly complex and diverse venoms, with tens to hundreds of individual components. ~~Rattlesnake~~ These venoms display a puzzling phenotypic dichotomy, with two largely mutually exclusive “strategies”: type “A” venoms are highly lethal and characterized by heterodimeric, presynaptic β -neurotoxic phospholipases A₂ (PLA₂), ~~e.g. Mojave toxin (MTX),~~ whereas type “B” venoms are less toxic and ~~lack MTX, but are~~ rich in snake venom metalloproteinases (SVMPs) with haemorrhagic and proteolytic activity [10]. The distribution of these phenotypes across the phylogeny of rattlesnakes is highly irregular: both types occur within most major clades, and even ~~between populations of~~ within some individual species (Figure S1) [10].

Multiple studies have explored the drivers ~~underlying of~~ intraspecific variation in venom composition and found evidence for the effect of natural selection for the optimisation of venom to diet [7,11-13]. Even subtle differences, involving only a few low-expression toxins, appear to

75 ~~reflect natural selection for diet~~ have selectively significant consequences [14]. This suggests that
76 the much starker intraspecific variation in species with both venom A and ~~venom~~-B populations
77 would likely have very powerful selective consequences, and thus predicts a strong effect of diet-
78 related factors as drivers of this variation [15-17].

79 Whilst ~~Identifying~~ selective drivers has been a significant research focus, the role of neutral
80 factors, such as past population fragmentation [18] or current gene flow, has received less
81 attention. ~~While~~ ~~e~~Evolutionary theory traditionally emphasizes the role of gene flow in either
82 facilitating the transfer of selectively favourable alleles, or reducing the potential for local
83 adaptation through genotypic homogenization [19]; nonetheless, the relative importance of gene
84 flow and selection on venom have rarely been compared directly. Although ~~Recent~~ recent studies
85 [20, 21] have ~~variously~~ identified selection and inter-population genetic distances as better
86 predictors of venom composition, ~~but those~~ involved subtler differentiation than the ~~Venom~~ A/B
87 dichotomy.

88 The Mojave rattlesnake (*Crotalus scutulatus*), ~~found across the southwestern USA and Mexico~~,
89 represents an ideal system to study the causes and mechanisms underlying variation in this
90 remarkable molecular phenotype. Four highly distinct phylogeographic lineages have been
91 identified across its wide range in southwestern USA and Mexico [17, 22]. Here, we focus on
92 the Mojave-Sonoran clade, ranging from California to south-western New Mexico, which in itself
93 represents a microcosm of the phenomenon of extreme intraspecific venom variation within a
94 single population [22]: most individuals ~~from most of its range~~ secrete type A venoms;
95 characterised by the neurotoxic Mojave toxin (MTX), whereas snakes from central Arizona secrete
96 type B venoms; ~~intermediate~~ Intermediate A+B venoms containing both SVMPs and MTX are
97 found at the contact zones between the two venom types [23,24]. Additional toxins belonging to
98 different gene families, such as other ~~PLA₂~~, myotoxin (MYO) and C-type lectins (CTL), also show
99 geographic variation in their expression [16,24]. We therefore used the Mojave-Sonoran clade of
100 *C. scutulatus* to investigate the causes and mechanisms generating and maintaining
101 polymorphisms across a widespread and continuously distributed species. We performed densely
102 sampled population-level analysis of the genomic basis of venom variation, investigated
103 population structure and diet, and then used in-depth environmental association analysis (EEA)
104 and climate reconstruction to disentangle the dynamics between genotype, phenotype and
105 environment.

107 **Material and Methods**

108 **Approach.** Initially, ~~We~~ ~~we~~ used in-depth proteomic analysis, genome sequencing and venom
109 gland transcriptomics of two field-caught adults of *C. scutulatus* from venom type A and B areas
110 (Figure 1) to identify ~~the~~ major toxins ~~in *C. scutulatus*~~, and to design primers to test for the presence
111 of specific toxin genes in additional specimens. We then mapped phenotype onto genotype by
112 comparing proteomic and genomic presence/absence of toxins across a larger sample, and, after
113 establishing a strict linkage, ~~extending~~ extended this to additional specimens at genomic level only.
114 We then correlated the venom profiles with new, densely sampled population genetic data,
115 geographic variation in diet, and ~~a number of~~ physical, climatic and vegetational parameters to
116 understand the drivers of venom variation.

117 **Sample collection.** ~~We collected venom and blood or tissue samples from field-caught specimens~~
118 (Figure 1). ~~Two field-caught adult snakes from venom type A and B areas were chosen as~~

119 ~~representatives of the two types for in-depth proteomic, venom-gland transcriptomic and whole~~
120 ~~genome sequencing. These data were then used to design toxin-gene specific primers (see below).~~

121 **Draft whole-genome sequencing.** For each representative individual we sequenced two genomic
122 libraries on an Illumina HiSeq2500. ~~high~~High-quality reads ~~were~~ assembled *de novo* using the
123 CLC Genomics Workbench platform v6.5, and contigs combined into scaffolds using SSPACE
124 Standard 3.0 [25]. Scaffolds containing putative toxin genes were identified by mapping all toxin
125 transcripts to genome assemblies using the GMAP software [26].

126 **Venom-gland transcriptomics.** ~~Venom gland~~ cDNA libraries ~~from the venom glands~~ of the two
127 representatives were sequenced on an Illumina HiSeq2500 and high-quality reads assembled *de*
128 *novo* using Trinity 2.0.4 [27]. We identified all possible toxin transcripts with blastx searches
129 against the NCBI nonredundant (nr) protein sequences [28], UniProtKB [29] and a custom
130 database containing only toxin protein sequences. Homologous toxin transcripts were identified
131 by reciprocal blast analysis and considered homologous if the coding sequences were 99%
132 identical, with minimum 70% sequence coverage. ~~Absence of toxins due to failure of Trinity to~~
133 ~~recover venom transcripts was verified by reciprocal mapping of reads against either~~
134 ~~transcriptomes and investigation of the proteome (see below).~~

135 **Venom proteomics.** To link venom proteins to their corresponding transcripts we analysed the
136 venoms of the two representative snakes by RP-HPLC and obtained molecular masses and peptide
137 sequences [30]. All sequences were blasted against the NCBI non-redundant database and the
138 venom-gland transcriptome assemblies using tblastn adjusted for short sequences. RP-HPLC
139 venom profiles of 50 additional specimens from different geographic areas were then examined to
140 identify the most highly expressed and variable toxins, and to test whether variation in venom
141 composition is caused by genome-level differences (see below).

142 **Toxin genotyping.** ~~We selected toxins that were always unambiguously scorable as either absent~~
143 ~~or highly expressed in the proteome, we and~~ designed gene-specific primer pairs based on our
144 genomic scaffolds using the Primer-BLAST tool [31]. Amplification specificity was checked
145 against our two transcriptomes and the NCBI nucleotide database. Twelve toxin genes belonging
146 to five families were selected for further investigation (see electronic supplementary material,
147 Table S3), in addition to the acidic (MTXa) and basic (MTXb) subunit genes of Mojave toxin [32].
148 Up to 163 individuals were screened for toxin gene presence, PCR products were checked on 1.5%
149 agarose gel, and a subset were sequenced to verify consistency of primer specificity. Sequences
150 were blasted against the NCBI nucleotide (nt) and whole-genome shotgun contigs (wgs) databases.
151 Pairwise Pearson correlation coefficients were calculated to test for linkage between toxin genes.

152 Given the absolute ~~correlation-link~~ between presence/absence of toxins in the proteome and the
153 corresponding coding genes (see below), we expanded our sampling ~~by using genotype~~
154 ~~information from genotyping additional additional~~ individuals without proteomic information
155 (e.g., road killed specimens) to assess toxin gene distributions.

156 **Venom fingerprinting.** Proteomic techniques allow detailed characterisation of individual venom
157 components, ~~but~~ do not allow for large-scale, standardised comparisons of overall variation and
158 diversity [30]. To increase our sampling and standardise our phenotype comparisons, we analysed
159 the same 50 venoms (see above) and 48 ~~additional~~ samples by on-chip electrophoresis [30]. All
160 samples were from adult snakes. The binary matrix of protein peak presence/absence was used to
161 calculate Shannon diversity index and pairwise Bray-Curtis dissimilarity matrices for subsequent
162 analyses.

163 **Population genetic analysis.** After preliminary analyses, we genotyped 290 specimens at 13
164 microsatellite loci (Table S5) (see electronic supplementary materials for details). Population
165 structure was determined using the spatial Bayesian clustering algorithm in TESS 2.3.1 [33].
166 Partitioning of genetic variation within and across subpopulations as inferred by TESS was
167 examined using analysis of molecular genetic variance (AMOVA) in GenAlex [34]. To test
168 whether spatial genetic patterns and population structure are the results of recent genetic
169 bottlenecks, heterozygosity excess and deficit were tested using the software BOTTLENECK
170 v1.2.02 [35] and Genepop [36].

171 Isolation by distance (IBD) was tested between pairs of individuals in GenAlex. A pairwise genetic
172 distance matrix was then estimated based on the proportion of shared alleles (*Dps*) [37] between
173 localities and used in a Mantel test against Euclidean geographic distances.

174 **Inference of past distributions.** To test whether current variation in venom composition could be
175 the result of past range fragmentation due to climatic changes, we performed niche modelling
176 using the program M_{AX}Ent [38]. Georeferenced occurrence localities of the Mojave-Sonoran
177 clade of *C. scutulatus* were gathered from the VertNet (<http://vertnet.org>) and Global Biodiversity
178 Information Facility (www.gbif.org) databases ~~and verified for possible mislabelled coordinates~~
179 ~~before analysis~~. Current climatic data were obtained from the WorldClim 1.4 database
180 (<http://www.worldclim.org>) at 30 sec resolution [39]. To avoid collinearity, highly correlated
181 variables (Pearson's coefficient $|r| \geq 0.8$) were pruned based on a pairwise correlation matrix,
182 leaving a total of 13 climatic variables (Table S10 and S11). Past climatic data for the Last Glacial
183 Maximum (LGM) were obtained from simulations with Global Climate Models (GCMs) estimated
184 by the Community Climate System Models (CCSM), and data from the Last Interglacial (LIG)
185 were obtained from [40]. All models were run with default regularization and 10 replicates
186 subsampled, using 20% of the points for test and 80% for training each replicate. We generated
187 ecological niche models for the species as well as for each individual toxin gene, and used present-
188 day climate envelopes for prediction-inference of past scenarios/distributions.

189 **Statistical analysis workflow.** All statistical analyses were performed in R version 3.4.2 [41]
190 using two approaches. First, we grouped individuals into discrete localities delineated by sampling
191 gaps and valley/mountain ridge systems. Individuals falling between localities were excluded.
192 Although this approach has the drawback of removing samples collected between localities, it can
193 exploit population-based association approaches, such as testing for relationships between venom
194 phenotype and diet composition. We ran Mantel and partial Mantel tests (controlling for
195 geographic distance) in the *vegan* 2.4-4 package [42] using the following response distance
196 matrices: i) venom phenotype: mean pairwise Bray-Curtis dissimilarities between localities
197 calculated from on-chip fingerprinting binary matrix; ii) venom genotype: pairwise Bray-Curtis
198 dissimilarity matrices based on toxin gene frequencies (one per gene).

199 Secondly, we used an individual-based approach, including all samples, to allow better detection
200 of association along gradients. For the venom phenotype, we analysed patterns of variation using
201 non-metric multidimensional scaling (NMDS) based on a pairwise Bray-Curtis distance matrix
202 and used the individual scores on the first two axes as response variables in regression models. For
203 the venom genotype, presence or absence of each toxin gene were used as response variables in
204 logistic regression models using the *glm* (generalized linear model) function with binomial
205 (link="logit") error distribution.

206 False discovery rates for all p-values of multiple comparison analyses were corrected using the
207 method of Benjamini & Hochberg [43]. One locality (“Gila”), where we were unable to collect
208 venoms, was only included in the genotype analysis.

209 **Venom variation and current gene flow.** Multiple approaches were used to test whether variation
210 in venom composition reflects current patterns of gene flow and neutral genetic structure. First,
211 we used AMOVA in GenAlex to estimate numbers of migrants and compare molecular variance
212 between (i) the three major venom types (i.e. A, B, A+B), and (ii) sampling localities. Secondly,
213 we ran partial Mantel tests between venom and genetic (*Dps*) distance matrices based on localities.
214 Finally, we tested for correlations between individual-level venom variation and neutral genetic
215 structure using the admixture proportions estimated by TESS as the explanatory variables.

216 **Venom variation and diet.** To test whether geographic variation in venom phenotypes and
217 distribution of toxin genes is associated with differences in diet composition, we recorded stomach
218 and gut contents from 463 preserved, geo-referenced specimens from several-museum collections.
219 All prey items were either mammals or reptiles, except for three amphibians, two arthropods and
220 one bird, which were excluded from further analyses. Altogether, 445 items were identified to
221 class level, 327 to family, 249 to genus, and 192 to species level.

222 For each taxonomic level we calculated the “frequency occurrence”, defined as the number of
223 samples in which a food item occurs expressed as a frequency of the total number of samples with
224 identifiable prey [44], the most commonly used method for diet analysis [45]. For each locality,
225 we used the frequency occurrence to calculate two measures of dietary composition: i) diet niche
226 overlap, ranging from 0 (no overlap) to 1 (complete overlap), describes diet composition similarity
227 between localities and corresponds to the pairwise Bray-Curtis dissimilarity index; ii) niche width
228 (Shannon diversity index), describes the diet diversity within a locality, with values near 0
229 indicating a narrow niche and values near 1 a broad niche. Both metrics were calculated with prey
230 identified to class, family, genus and species level. Pairwise distance matrices based on these
231 metrics were used for Mantel tests. Additionally, we tested for correlation between venom
232 diversity and niche width, and between frequencies of individual prey species and toxin genes in
233 order to identify potential key species involved in predator-prey arm races.

234 **Environmental association analysis (EAA).** To test whether the observed variation in venom
235 phenotype and toxin gene distributions were associated with spatial heterogeneity, and to identify
236 environmental factors potentially contributing to local adaptation and genetic variation, we
237 performed EAA.

238 In addition to the WorldClim data (see above), we used the high resolution digital elevation model
239 (DEM) raster (<http://asterweb.jpl.nasa.gov>) to produce additional topographic variables including
240 slope, solar radiation, aspect and topographic position index (TPI) using the Spatial Analyst
241 toolbox in ArcMap 10.3 (ESRI®). Land cover data describing North American ecological areas
242 (level III “ecoregions”) were obtained from the US EPA (<https://www.epa.gov/ecoresearch/ecoregions-north-america>), and vegetation data from the Gap Analysis Project
243 (<https://gapanalysis.usgs.gov/gaplandcover/data/download/>).

245 Patterns of environmental heterogeneity across the study areas were examined using Principal
246 Component Analysis (PCA), and the significance of differences between localities were tested
247 with pairwise t-tests.

248 For climatic and topographic variables, Euclidean distance matrices were calculated based on the
249 average values within each locality, whereas for categorical variables (ecoregion and vegetation)
250 distance matrices were generated based on the proportion of each factor level within localities.
251 Prior to Mantel test analysis the BIOENV procedure [46] in the *vegan* package was used to reduce
252 the climatic variables contributing to the final distance matrix. This function calculates Euclidean
253 distances for all possible subsets of scaled climatic variables and finds the maximum Spearman
254 (rank) correlation with the response distance matrix.

255 In the individual-level analysis, univariate regression models were generated for all variables in
256 order to identify the strength, direction and nature of the relationships between each environmental
257 factor and venom variation/toxin gene presence. ~~To test to what extent climatic variables could~~
258 ~~predict the distribution of individual toxin genes, we~~ We also generated climatic niche models for
259 the individual genes using the WorldClim data, and used MaxEnt to generate predicted distribution
260 maps for each.

261 **Gradient analysis.** To investigate local environmental patterns at the interface between the two
262 main venom types, we performed a gradient analysis to test associations between phenotypic or
263 genetic variation and environmental factors along a continuous cline. We identified two suitable
264 venom B – venom A transects, one running west (“Maricopa”) and the other south (“Sasabe”) from
265 the core of the venom B area (Figure 3b). We intensively sampled~~s~~ these two transects and tested
266 for presence of MTX and SVMP genes. Trends along the transects were analysed for each climatic
267 variable ~~and correlated with toxin gene presence/absence~~.

268

269 Results and Discussion

270 **Venom variation is due to structural genomic variation.** High-throughput genome sequencing
271 of *C. scutulatus* generated ~~a total of~~ 652865 contigs for the venom type A representative individual
272 and 597176 for the type B, with sequencing coverage of approximately 8x (Table S1). RNA-Seq
273 of the venom glands generated 37162 contigs for the venom A and 56627 for the type B (Table
274 S2). We identified a total of 96 unique toxin transcripts in the venom A transcriptome and 115 in
275 the venom B. Both venom gland transcriptomes and proteomes showed marked differences, with
276 several toxins highly expressed in either one or the other venom (Figure S2 and S3), including ~~;~~
277 ~~Those included toxins belonging to the~~ SVMPs, PLA₂s, serine proteases (SVSPs), C-type lectins
278 (CTLs) and myotoxin (MYO).

279 Comparison of the proteomic profiles and genotypes of 50 specimens confirmed that the presence
280 or absence of 14 differentially expressed toxins in the proteome was invariably associated with the
281 presence or absence of the corresponding coding genes (Figure S4). This was previously
282 documented for MTX, other PLA₂s and SVMPs [16,32], and is here confirmed for CTLs and
283 MYO. Based on this strict phenotype-genotype link, we analysed the spatial distribution of toxin
284 genes in a larger sample to identify gene complexes and linkage patterns (Figure 2a, Table S4). In
285 both main venom types, some genes appeared tightly linked, whereas others varied independently.
286 In the core venom B area there were two main genotypes, both characterized by the presence of
287 SVMPs, PLA₂s (gA1, gB1 and gK) and CTL-B7, but differing in the presence of myotoxin
288 (MyoB). Much greater diversity was observed across the venom A genotypes; ~~;~~ which all ~~;~~ all
289 characterized by the tightly linked neurotoxic MTXa and MTXb, the absence of SVMPs, PLA₂gK
290 and gB1, but ~~showing great variation~~ varied in the occurrence of PLA₂gA1, MyoB and CTL-B7,
291 each with unique spatial distribution patterns. While MTXa and MTXb, as well as PLA₂gK and

292 gB1, remained linked in all specimens, other linkages between gene complexes were disrupted
293 across the contact zone between venom types, where mixed (A+B) genotypes and multiple
294 different gene combinations occur. Interestingly, the intergrade zones also produced three
295 individuals lacking both neurotoxic MTX ~~nor~~ and SVMP genes (type O), suggesting that mating
296 between mixed genotypes can not only disrupt adaptive genomic linkages, ~~and~~ but even lead to
297 the complete loss of multiple key components. This raises the question how these different
298 genomic variants persist in the species, and what determines the distribution of venom phenotypes.

299 **Venom variation is not associated with population genetic structure.** Our climatic niche
300 modelling suggests a past range fragmentation into western, Sonoran (AZW), and eastern,
301 Madrean (AZE), refuges (Figure 2b). ~~Unlike a previous analysis of *C. scutulatus* population~~
302 ~~structure [22], both~~ TESS and sPCA detected a genetic discontinuity with extensive admixture
303 corresponding to the boundaries between the Sonoran and Madrean ecoregions (Figure 2b),
304 ~~therefore~~ reflecting predicted Pleistocene vicariance and consistent with postglacial range
305 expansion. No evidence of recent bottlenecks (Table S6) or further subpopulation structuring
306 (Figure S6 and S7) was detected. Our results contrast with previous inferences of panmixia within
307 the Mojave-Sonoran clade of *C. scutulatus*, based on analyses of mtDNA, or RADseq data from
308 much smaller samples [22,47].

309 ~~In the absence of a correlation between venom types and neutral genetic clusters~~ Since the two
310 genetic clusters did not predict the distribution of venom types (Figure 2a, Table S8), we further
311 assessed the relationship between venom composition and genetic structure by grouping the
312 samples geographically into localities (Figure 1b) and calculating venom distance matrices and
313 toxin gene frequencies. Overall genetic differentiation was weak, including between venom A and
314 B localities ($F_{st} = 0.003-0.05$), with high levels of gene flow ($N_m = 8-75$). Analysis of genetic
315 variation showed evidence of deviation from Hardy-Weinberg equilibrium (HWE) and
316 heterozygosity deficit in the venom B and adjoining localities, suggesting strong selective regimes
317 (Table S7). AMOVA analysis grouping either by venom types or localities confirmed an absence
318 of finer substructure, with most of the variance arising from within individuals (Table S8). Partial
319 Mantel tests showed ~~a non-no significant weak~~ association between venom phenotype variation
320 and neutral genetic distance; similarly, individual toxin gene frequencies were not correlated with
321 gene flow (Table 1). While a significant pattern of isolation by distance (IBD) (Mantel $r^2=0.70$,
322 $p=0.006$), weak genetic structure ($F_{st}=0.02$) and heterozygosity deficit ($p=0.001$) are consistent
323 with population expansion following LGM, the complete absence of association between
324 phenotype and neutral genetic differentiation suggests that strong selective forces are driving the
325 distribution of venom types, rather than differentiation in allopatry followed by range expansion.

326 **Venom composition is not associated with diet spectrum.** Because adaptation to diet is generally
327 invoked as the foremost driver of venom evolution [11,14,16,17,47,48], we tested whether the
328 divergent phenotypes are associated with differences in local diet. Our diet data show that *C.*
329 *scutulatus* feeds primarily on small mammals, with the rodent families Heteromyidae and
330 Cricetidae alone constituting forming the bulk (78.8% of prey items overall) of the diet (Figure 1b
331 and S86b). ~~—~~Partial Mantel tests found no significant association between overall venom
332 composition and diet spectrum measured as niche overlap or niche width, irrespective of whether
333 the spectrum was resolved to class, family, genus or species level (Table 1).

334 Similarly, we found no significant pairwise relationships between individual toxin gene
335 frequencies and individual prey species; in particular, neither MTX nor SVMPs, the two main
336 players in the venom dichotomy, were linked to any specific prey. We also tested the hypothesis

337 ~~of an association between toxin diversity and diet niche width, with that-~~ more complex venoms
338 would allowing predation upon a more diverse array of prey [48,49]. Interestingly, we found the
339 opposite trend; ~~i.e.~~ localities with less diverse venoms had broader prey spectra, although this was
340 only weakly significant (Figure S86a). None of the frequencies of the individual toxin genes were
341 significantly correlated with either diet composition or niche width, except PLA_{2gA1}, an inhibitor
342 of ADP-induced platelet aggregation ([50]), which showed a strong association with climate and
343 ecoregion, and a lower weaker, but a-significant, correlation with diet composition at the family
344 level (Table 1). The functional significance of this is unclear, as this gene is widespread in the
345 genomes of both type A and type B rattlesnakes in general [16]. Whether this association is due to
346 direct selection for diet or a partial correlation between diet and climate or ecoregion is also
347 unclear.

348 Because the primary function of venom in snakes is prey acquisition [7], adaptation to specific diet
349 as the key selective driver of venom evolution has become the dominant paradigm in the study of
350 snake venom evolution. Since even subtle variation in venom composition can reflect selection for
351 local prey [12,14], we had hypothesized that the stark contrast in toxicity and mode of action
352 (neurotoxic vs. haemorrhagic) between A and B venoms in *C. scutulatus* would have a significant
353 impact on the snakes' foraging biology. Our results thus challenge the widespread assumption of
354 diet composition as the main determinant of the venom dichotomy in this or other rattlesnake
355 species [16,17] and its universality as a selective driver of snake venom evolution in general [7].

356 **Spatial environmental heterogeneity predicts venom variation.** Spatial heterogeneity in
357 environmental variables is a key driver of genotypic and phenotypic polymorphism [49]. In the
358 absence of a strong venom-diet association, we performed EAA to understand whether differences
359 in other biotic and/or abiotic factors contribute to geographic variation of venom composition
360 [50,52, 47,53]. Overall venom variation was strongly associated with temperature (Table 1), and
361 the longitudinal climatic gradient characterizing the Sonoran desert (Figure S9 and S10) was
362 reflected in the differentiation across venom A profiles along the first NMDS axis (Figure 3a).
363 ~~Yet, the divergence~~ In contrast, the second NMDS axis, which broadly separates between A and
364 B venoms, showed weaker correlations ~~was not strongly correlated with any specific~~
365 ~~environmental variable~~ (Table S12). However, across a large, continuous distribution without
366 discrete physical barriers, large-scale analyses may fail to detect the effect of local ecotones and
367 short environmental clines of potential selective importance. We thus analysed local scale climatic
368 trends along two A-B transects and discovered the presence of sharp clines associated with venom
369 composition for several ~~ecological variables, especially those related to precipitation~~ (Figure
370 3b-g); In agreement with this and previous findings [47,53], logistic regression models revealed
371 significant associations of MTX and SVMPs with climatic variables, with venom B areas ~~being~~
372 characterized by larger diurnal thermal fluctuations, milder winters and less seasonal variation in
373 precipitation (Table S12).

374 ~~The Logistic regression models revealed that certain other~~ toxin genes, even though highly co-
375 expressed in some phenotypes, ~~correlate with different ecological variables~~ showed different
376 correlation patterns, suggesting that different selective forces orchestrate individual loci to create
377 complex, dynamic phenotypes (Table S12). Strikingly, genes located few kb apart, such as some
378 PLA_{2s} [15], also displayed independent associations, demonstrating that divergent selective
379 pressures can differentially affect parts of the same genomic region. Climatic niche modelling of
380 the distribution of individual toxin genes yielded different predictions even for neighbouring
381 genes, and ~~showed~~ the models proved to be accurate predictors of gene distribution (Figure S54),

382 emphasising the environment-genotype link. This interesting phenomenon deserves further
383 investigation, since genes coding for the same adaptive phenotype are generally brought closer
384 together by means of chromosomal rearrangements such as inversions or supergenes [5454].

385 **Genome, environment and the maintenance of geographic variation.** The emerging picture of
386 the mechanisms and drivers governing venom variation in *C. scutulatus* is thus one of ~~an~~-adaptive
387 polymorphism with gene flow, with the distribution of toxin genes shaped by directional natural
388 selection for local environmental factors other than diet spectrum or neutral gene flow. Margres et
389 al. [20] recently suggested that gene flow may be more likely to drive venom composition in
390 dietary generalists than in specialists; the lack of association between gene flow and venom
391 composition in the specialist mammal-feeder *C. scutulatus* is consistent with this, but the lack of
392 association between diet spectrum and venom suggests that other determinants are involved.

393 The precise nature and mechanism of selection, and especially the association of venom with
394 environmental parameters, remain unclear. ~~Analyses of venom composition in *Crotalus oreganus*
395 found either environmental parameters [50] or a combination of genetic distance and diet [21] to
396 predict venom variation.~~ It seems to us unlikely that climate by itself exerts strong selection on
397 venom composition. ~~In fact, The generally positive association between type B venoms and higher
398 winter temperatures runs contrary to the hypothesis that SVMPs are needed to assist digestion at
399 lower temperatures [10, 55].~~ However, climatic stability and seasonality may affect other factors,
400 for instance prey community composition and dynamics [52]. These, in turn, could influence snake
401 foraging strategies, and potentially also the exposure of snakes to predation, an understudied
402 source of selection on venom [5256]. In widely distributed species occupying diverse
403 environmental conditions, spatial heterogeneity could thus select for local fitness optima, resulting
404 in the maintenance of disparate, locally adaptive gene complexes.

405 While venom composition does not correlate with diet spectrum, the possibility of more subtle
406 diet-related selection deserves further study: predator-prey arms races, pitting resistance to venom
407 in prey against the snakes' venom, appear to be important drivers of venom evolution in at least
408 some cases [12]. While many desert rodents display resistance to type B venoms [5357], there are
409 virtually no corresponding data for type A venoms. Geographic variation in the prevalence of prey
410 resistance to different venom types, perhaps correlated with other environmental variables, could
411 conceivably act as a driver of venom composition in *C. scutulatus*. This could constitute a fruitful
412 focus for future research. Potential prey-specific toxicity in PLA₂gA1, the only diet-associated
413 toxin, may also repay further investigation.

414 As in previous studies [47,53], ~~W~~we hypothesise that disruptive selection against intermediate
415 A+B phenotypes may ensure spatial segregation, thereby favouring persistence of gene complexes
416 and divergent phenotypes. The role of relatively subtle environmental changes in driving the
417 dramatic differences in venom composition in this species, coupled with selection against
418 intermediate phenotypes, suggests the existence of steep clines in the adaptive fitness landscape,
419 where one phenotype gains a selective advantage over the other across short geographic distances.
420 However, the proximate factors mediating the geographic variation is-in selection pressures
421 remain to be fully understood.

422 **Conclusions.** The unique genomic architecture of rattlesnake venom provides an important
423 addition to the catalogue of mechanisms underlying adaptive phenotypic variation, and establishes
424 a promising system for investigating the ecological and evolutionary implications of genomic
425 structural variation in non-model organisms. Together, our results emphasise the importance of

426 combining large-scale genotype, phenotype and ecological data in natural populations to uncover
427 the wide variety of mechanisms and drivers underlying phenotypic variation, and emphasise the
428 need to consider a multitude of factors as potential selective drivers of phenotypic variation.

429 **Data accessibility.** Raw [Illumina](#) sequences have been deposited in the European Nucleotide
430 Archive (ENA) under project accession PRJEB29193. RNA-seq accession numbers: venom type
431 A: ERS2793705 (right venom gland); ERS2793704 (left venom gland); type B: ERS2793703
432 (right venom gland). [Whole genome sequencing accession numbers](#): type A: ERS2793891 (300bp
433 insert) and ERS2793890 (600bp insert); type B: ERS2793893 (300bp insert) and ERS2793892
434 (600bp insert). [Toxin gene sequences are deposited in GenBank with accession numbers:](#)
435 [MG948948-MG949116](#). [Samples localities, microsatellite and diet data are found at](#)
436 [doi:10.5061/dryad.d21k432](#).

437 **Authors' contributions.** Conceptualization: WW, GZ; Formal analysis: GZ; Methodology: GZ,
438 JJC, MH; Investigation: all authors; Writing – original draft: GZ; review & editing: all authors.

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574

575 Legends:

576 Figure 1. Geographic variation in venom and diet of adult *C. scutulatus*. (a) Distribution of
577 samples for which the major venom types were identified based on toxin genotypes; stars represent
578 the sampling locations of the two representative individuals used for the genome-transcriptome-
579 proteome analyses. (b) Two-ring pie-charts showing the proportion of mammals and reptiles from
580 stomach contents (inner charts) and venom types (outer ring) for each locality.

581 **Figure 2. Toxin genotype and niche modelling.** (a) Presence-absence matrix of toxin genes and
582 admixture plot (TESS) with $K=2$.– (b) Niche models and sample distribution of the Mojave-
583 Sonoran clade of *Crotalus scutulatus* with individuals represented by proportion of genetic
584 clusters. Grey lines delineate ecoregion boundaries.

585 **Figure 3. Association between venom phenotypic variation, neutral genetic differentiation**
586 **and environment.** (a) Non-metric multidimensional scaling (NMDS) analysis of venom profiles
587 shows great overall variation. Variation along NMDS1 is strongly correlated with the marked east-
588 west environmental cline across Arizona (Table S12, Figure S9 and S10), whereas environmental
589 associations along NMDS2, broadly separating the A-B transition, are weaker because global-
590 scale variation hinders the detection of local-scale patterns. (b to g) Local-scale analysis along two
591 transects (b) reveals sharp clines in various temperature (c-e) and precipitation (f, g) variables (see
592 Table S11 for bioclimatic variable description) across the venom A-B transition zone.

593 **Table 1.** Environmental association analysis between localities. Correlation matrix of partial
594 Mantel tests (Spearman R partial correlation coefficients multiplied by 100) between overall
595 venom phenotype or individual toxin gene frequencies against environmental variables, with
596 geographic Euclidean distance matrix as covariate. Isolation by distance (IBD) is the null model.
597 Proportion of shared alleles (Dps) was used as index for neutral genetic differentiation. Variables
598 selected with the BIOENV procedure to generate climatic and topography distance matrix are
599 reported: BIO1-BIO9 correspond to measures related to temperature and BIO12-BIO19 to
600 precipitation. Values with $p < 0.05$ are in bold.