1 2	Abundance and Distribution of <i>Heterelmis cf. glabra</i> (Coleoptera: Elmidae) within Dolan Falls Preserve and the Devils River State Natural Area, Texas, USA
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20 Abstract

The Devils River watershed in south-central Texas has baseflows entirely attributable to 21 22 groundwater primarily sourced from the Edwards-Trinity and Edwards Aquifers. The largest 23 known populations of a species of riffle beetle, *Heterelmis cf. glabra*, are located in springs 24 associated with the upper Devils River. The focus of this study was to 1) determine site-level 25 abundances of H. cf. glabra using open system N-mixture models, 2) test mesohabitat 26 associations of members in the riffle beetle family Elmidae and, 3) measure and examine abiotic 27 and habitat associations for adult and larval beetles within the study area. We sampled 32 spring 28 sources to determine occupancy and abundance of adult and larval riffle beetles (Elmidae) within 29 the study area. Spring sources were mapped and categorized by type (orifice, upwelling, group of 30 springs, or seep). Basic water chemistry and flow rate categorization were also performed at each 31 site. Model results suggest that rainfall, flow and site are important for detection of *H. cf. glabra*.

Based on our results, regular monitoring of these 32 sites using these methods, is recommended to conduct hypotheses tests on covariates influencing abundance. Such baseline information will be important in measuring impacts to this and other spring-associated species as the habitats of this region are impacted by natural or anthropogenic phenomena.

36 Key Words: springs, Elmidae, abundance, karst, N-mixture, Heterelmis

37 Introduction

38 Groundwater extraction in the Permian Basin of West Texas has increased as industrial 39 pumping for gas and oil have increased. In addition, the region has been identified as a potential 40 source for water export to more highly populated regions of Texas [1]. The depletion of 41 groundwater in certain areas may in turn cause disruptions in flow or changes to historically 42 stable temperatures that endemic spring-adapted species have adapted to over the course of 43 geologic history. Many spring-adapted species are known to be associated with stenothermal 44 groundwater habitats of the Edwards Plateau of Central and West Texas [2-6]. Although springs may fluctuate in regards to flows over geological periods of time, direct correlations between 45 46 flows and anthropogenic pressures are observable in nearly real time in unconfined aquifers such 47 as the Edwards-Trinity aguifer that feed the Devils River [1]. This in turn may alter the 48 characteristics of the spring system under which these species naturally persist. Recent modeling 49 suggests that the impact of groundwater withdrawals on Devils River discharge is proportional to 50 the amount of water pumped [1]. Subsequently, significant groundwater pumping has the 51 potential to extinguish or shift the location of spring discharge points. During times of drought or 52 disturbance, some spring-adapted species are able to retreat into the aquifer for temporary refuge 53 [7] or live deeper within the aquifer permanently [8]. Other spring adapted-organisms, such as 54 Heterelmis comalensis and H. glabra, have life history patterns requiring surface components,

which makes them more susceptible to changes in springflow that alter the surface habitatscondition.

57 *Heterelmis cf. glabra* represents a potentially undescribed species of riffle beetle, is 58 known to exist in large permanent springs in Terrell, Val Verde, Kerr, Hays, Bell, and Tom 59 Green counties [9,10, unpublished data]. [9,10]. Despite recent applied research evaluating 60 tolerance to elevated temperatures and reduced dissolved oxygen for a population of H. cf. 61 glabra [11, 12], little information is available on the distribution and abundance of this species 62 within its known range or its habitat associations. Gathering and analyzing reliable data on 63 population size and distribution is a central theme in ecological research [13] and is essential for 64 management of endemic populations [14]. 65 The Comal Springs riffle beetle (*Heterelmis comalensis*; [15]) is a USA federally 66 endangered species [16] hypothesized to be similar to *H. cf. glabra* both phylogenetically [9, 10] 67 and ecologically (i.e., spring obligate). The major threats to *H. comalensis* are reduction in water 68 quality and quantity due to drought and development associated with an increased need for 69 groundwater resources as a result of accelerated population growth in the area [17, 18]. Similar 70 to *H. cf. glabra*, *H. comalensis* inhabits areas near and within spring sources [19, 8] and are often 71 found associated with woody debris or roots where they feed on biofilm produced as these 72 substrates decay [20, 8, 18, 21]. Heterelmis comalensis are thought to move through interstitial 73 alluvium within spring sources, making collections difficult as much of this habitat is not 74 accessible by traditional sampling techniques. Subsequently, a method was developed using 75 cotton lures placed in spring sources for monitoring *H. comalensis* [22]. Over several weeks, 76 biofilms on which riffle beetles feed grow on the cotton material. This provides a consistent 77 method of collection for this endangered species. Using this method, hundreds of larvae and

adults can be collected and returned to the habitat unharmed [8, 23]. Populations of *Heterelmis cf. glabra* in large perennial springs of the Edwards Plateau occupy ecologically similar habitat
as *H. comalensis* and are readily collected using the cotton lure methodology.

81 The life history characteristics of these riffle beetles provide statistical and study design 82 challenges for population monitoring. Adult *Heterelmis* beetles probably live about a year (San 83 Marcos Aquatic Resources Center - unpublished captive propagation data) and are small (~ 2 84 mm) creating issues with mark recapture studies [23]. Certain interstitial spring-adapted species 85 most likely occupy areas within spring sources not accessible to sampling gear such as a Hess 86 sampler or kick net producing low count data not or with many zeros. The use of count data for a 87 level of abundance without taking into account the organisms not detected can be misleading, by 88 invoking a suspect relationship between the count index and true abundance [24]. To rectify this 89 discrepancy, advances in monitoring techniques can allow for estimation of abundance using 90 count data and covariates that partition the distribution of the target organism spatially while 91 accounting for imperfect detection [25, 26]. These models are called N-mixture models and are a 92 class of state space models which assume the system is observed imperfectly [27]. These models 93 can be used on open [26] or closed systems [25].

The focus of this study was to determine site-level abundances of *H. cf. glabra* within the study area. This was accomplished by testing a series of models based on hypotheses associated with the detection of the beetle. Other objectives of the study include testing spring associated affinities of members in the riffle beetle family Elmidae present within the system. Measured abiotic associations and basic habitat associations for adult and larval beetles were examined.

99 Methods

100 Study Area

101	The Devils River watershed is in south-central Texas and is one of two principal Texas
102	tributaries to the Rio Grande. The Devils River is primarily sourced by the Edwards-Trinity
103	Aquifer, with baseflows entirely attributable to groundwater [1]. The largest known populations
104	of H. cf. glabra are located in Finegan, Blue, and Dolan springs. These springs issue into the
105	upper Devils River portion of The Nature Conservancy's Dolan Falls Preserve (DFP) and Texas
106	Parks and Wildlife Department's Devils River State Natural Area (DRSNA) property (Fig 1).
107	
108	Fig 1. Map of study area in Val Verde County Texas. Mapped springs on The Nature
109	Conservancy's Dolan Falls Preserve and Texas Parks and Wildlife Department's Devils River
110	State Natural Area. Sites were selected randomly from these available springs.
111	
112	The spring complexes in the study area issue from Cretaceous Edwards Limestone [28]
113	along the east bank of the Devils River and Dolan Creek. Finegan Springs comprises (discharge
114	= 99-760 L/s from 5 measurements during 1928-1971; [29]) 44 mapped springs and seeps along
115	a stretch of around 333 m at the base of a bluff flowing over chert bedrock forming small
116	rheocrene streams that merge into larger pools and streams emptying into the Devils River as far
117	as 25 m from spring sources. Blue Springs is a small group of spring conduits and gravel seeps
118	with most (5 springs/seeps) flowing into a short (ca. 5 m) cobble/gravel rheocrene and four
119	marginal seeps that empty into a backwater pool (ca. 90 x 30 m) that connects directly to the
120	Devils River ca. 800 m downstream of Finegan Springs. Dolan Springs comprises (discharge =
121	34-510 L/s from 7 measurements during 1928-1970; [29]) 48 springs or seeps along a stretch of
122	762 m at the base of a bluff and flow over limestone bedrock forming small rheocrene streams
123	and shallow pools that empty into Dolan Creek as far as 30 m from spring sources. The

124	confluence of Dolan Creek with the Devils River is ca. 500 m downstream of Dolan Creek from
125	the stretch of Dolan Springs and is ca. 1 km downstream of the Devils River from Blue Springs
126	(Fig 1). Water quality issuing from 12 spring sites (seven from Finegan; two from Blue, and two
127	from Dolan springs) in February 2010 (average temperature \approx 22 °C; conductivity \approx 504 $\mu S/cm;$
128	pH \approx 7.1; dissolved oxygen \approx 7.9) were similar to those measured during this study from 64
129	spring sites in February 2016 (average temperature \approx 22 °C; conductivity \approx 494 $\mu S/cm;$ pH \approx
130	7.2; dissolved oxygen \approx 7.9). These springs are habitat for several rare endemic stygobiontic
131	species including insects, crustaceans, and salamanders [30, 2].

132 Data collection and N-mixture model

133 Individual spring sources were mapped during the week of January 12, 2016. Data 134 collected during the mapping event consisted of basic water chemistry (temperature, dissolved 135 oxygen, pH, conductivity, and total dissolved solids) and a categorical designation of flow from 136 one to five (five being the highest). Springs were identified and categorized as orifice, 137 upwellings, group of springs, and seeps. The designation, "groups of springs", was used when 138 the springs were too close in proximity to each other to allow the Global Positioning System 139 (GPS) unit to distinguish between the individual orifices accurately. Different types of springs 140 and their placement within the system (above or below the waterline) may have effects on the 141 types of invertebrate communities present. For this study, all of the mapped locations had sites 142 above the waterline. Therefore, sites were randomly selected from two groups within the 143 mapping data (groups of springs/orifice and seeps). Fourteen sites were selected from Dolan, 14 144 sites from Finegan and four sites from Blue springs. Although seeps were the second most 145 abundant spring type available, most consisted of a thin layer of water moving over bedrock 146 which is not conducive for the cotton lure sampling method as it requires water depths of at least

147 2 cm. Within the 32 sites, six seeps were selected randomly from the data set although not all
148 were used for previously mentioned reasons. Sites are, at a minimum, a meter apart or separated
149 by terrestrial environment to maintain independence.

150 To examine abundance of *H. cf. glabra* within the study area, lures were deployed in 151 February, May, August and November of 2016. Each event consisted of burying a folded cotton 152 cloth encased in a metal cage in the substrate of the spring source outflow (Fig 2). The cotton 153 cloth lure is standardized in size (15 cm x 15 cm) and folded the same for each cage. The cage is 154 used to hold the shape of the lure over time and prevent potential anoxia by becoming squeezed [23]. Thirty two lures were set in springs for each of the four events. Each lure represents a 155 156 sampling site and were left in the substrate for 28 to 35 days to allow for biofilm growth. After 157 that time, the lures were removed and adult and larval riffle beetles (Elmidae) found on each lure 158 were counted and recorded. All riffle beetles were carefully returned to the site of capture 159 following each event. For each subsequent event, this process was repeated at the same sites. 160

Fig 2. Cotton lure. Cotton lure and cage used to collect riffle beetles to determine site level
population estimates from springs in Val Verde County TX.

163

Since the abundance models use probability of detection in determining the organism abundance, covariates that may influence detection of *H. cf. glabra* were examined. Spring location was considered a site covariate and consisted of grouping sites on the Devils River (Finegan and Blue springs) as one site and Dolan Springs as the other. Flow was considered an ordinal site covariate and ranged from one to five (five having the most flow). Sites with a flow of one were considered seep sites, and ranged from wet rock to a small pooling area with organic

debris. Sites with flows above one would be considered more traditional springs with a runterminating at the creek or the river.

172 Most of the springs arise from the bottom of bluffs that run parallel to the creek or river. 173 During rain events, material is dislodged from higher elevations along the bluff and deposited 174 over the spring sites. This may cause lures to be covered by silt, reducing dissolved oxygen and 175 thereby decreasing the available area for food resources and beetle respiration thus influencing 176 the detection of the beetles after rain events. In addition, rain can cause changes in water chemistry at a local level potentially influencing the areas where riffle beetles would associate. 177 To account for this, rain events during the sampling period were considered as a sampling 178 179 covariate which has an associated binary score for each sampling event. A score of 1 was given 180 to the model for rain if the rainfall total for a day exceeded 2.54 cm, examined over the duration 181 while the lure was deployed. Rainfall totals were determined by checking the Del Rio 182 International Airport station accessed through Weather Underground (www.wunderground.com). 183 Capture data of adult, and not larval, H. cf. glabra from each event was used to populate 184 the model. The software package "unmarked" in R [31, 32] was used to analyze the data for 185 abundance using the function "pcountOpen" for fitting the open population models [26]. Open 186 models were run due to the relatively short life span of the beetles and the duration of the study. 187 Lost lures during an event were accommodated as per the unmarked manual within the model 188 framework. Model selection was aided using Akaike's information criterion corrected (AICc) for 189 small sample size [33]. To examine the fit of the model, goodness of fit test were conducted in 190 the R package "AICcmodavg" with 100 parametric bootstrap iterations. The final models were 191 selected based upon the AICc score and the goodness of fit tests.

Candidate abundance models were analyzed using two different abundance distributions: a Poisson distribution with K set at 200 and negative binomial (NB) models with K set at 400. Abundance can vary based on the selection of K (upper bounds of integration), especially when low detection probabilities are calculated [34]. Therefore, for negative binomial models, K was allowed to fluctuate with the model that had the lowest AICc score to examine when the population estimate levels off, as well as changes in p, and goodness of fit.

198 Univariate Relationships and Habitat Associations

199 Present within the system are species of at least nine different genera of Elmidae. Most of 200 these species are likely more riverine, however, our interest was in spring-adapted species. To 201 examine community structure of elmid beetles and determine spring-adapted species present, 202 lures were set in a longitudinal fashion from the origin of the spring to parts of the spring-run 203 more influenced by external temperature and mixing of the spring water with organic material. A 204 total of 17 other lures were set out during the course of the study (March, April, and November) 205 downstream of the spring origin within the spring-run, which we termed the transition zone. The 206 transition zone has subtle changes in water chemistry compared to the spring origin which may 207 restrict the distribution of spring-adapted or associated organisms. Lures were set in the same 208 fashion as the abundance model lures and left in place for the same amount of time. When lures 209 were collected, elmid adults and larvae were counted and returned to the site of capture.

To examine differences between the two elmid communities present at the spring origin and the transition zone, and to determine which elmids appear to be spring adapted, indicator species analysis [35] was conducted using the statistical package "labdsv" in R [36]. Count data from the abundance model was averaged (n = 32) from the four events and compared to the transition zone lure averages (n = 17). Other univariate relationships with water chemistry and

215 substrate were explored using Pearson correlation analysis. Variables that were correlated with 216 count data at ± 0.50 were examined using linear regression models. This was done by averaging 217 the count data from all events and testing against the data collected during the mapping event 218 (i.e. specific conductance, temperature, etc). To test for correlations with the substrate, the 219 designation of substrate types was expanded to run from 0 to 15 using a modified Wentworth 220 scale.

Results 221

222

N-mixture Abundance Model

223 The final data set consisted of 32 sites, 28 spring sites and four seep sites. Some seep 224 sites were disregarded due to the low flow and the collection method used. Specifically, most 225 seep sites were not conducive to the lure method and alternate spring sites were used instead. 226 Final sites used in the abundance models are presented in Table 1. A total of 3,122 adult *H. cf.* 227 glabra were counted from lures during the four events. The highest capture rate occurred in 228 November (n = 1,078), while the lowest capture rate (n = 602) was in August. Finegan Springs 229 consistently had higher capture rates than at Dolan Springs.

230

231 Table 1. Randomly selected spring sites in Val Verde County used to populate *Heterelmis*

232	c.f. glabra	abundance	models.
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Springs	Site	Туре	Flow	Substrate	Temperature	Conductivity	DO	pН	TDS
BH	8	group of springs	2	3	22.11	495	7.63	6.82	0.3175
BH	9	group of springs	2	3	22.12	490	7.97	7.06	0.3139
BH	1	orifice	3	4	22.13	495	7.72	6.91	0.3169
BH	2	orifice	4	4	21.98	495	7.36	7.60	0.3168
BH	5	orifice	2	3	22.03	537	7.60	7.13	0.3435
FS	84.0	orifice	2	4	22.34	498	8.14	7.14	0.3188

FS	94.7	Seep	1	2	21.81	498	8.68	7.37	0.3194
FS	105.0	orifice	2	3	22.45	497	8.09	7.34	0.3180
FS	110.0	orifice	2	3	22.43	489	8.24	7.22	0.3134
FS	151.5	orifice	3	3	22.48	497	8.13	7.16	0.3183
FS	175.7	orifice	4	2	22.49	498	8.12	7.20	0.3185
FS	184.3	orifice	4	4	22.48	498	7.88	7.16	0.3185
FS	191.0	orifice	3	6	22.48	497	8.03	7.19	0.3178
FS	212.0	Seep	1	6	NA	NA	NA	NA	NA
FS	280.5	group of springs	3	2	22.50	498	8.04	7.19	0.3185
FS	285.0	group of springs	3	6	22.50	497	7.96	7.20	0.3186
FS	310.4	group of springs	3	4	22.49	498	7.67	7.28	0.3186
FS	319.0	orifice	3	6	22.49	497	8.10	7.20	0.3179
FS	291.5	orifice	5	6	22.49	498	8.13	7.22	0.3188
DC	43.0	group of springs	3	4	22.29	477	7.27	7.09	0.3053
DC	120.0	Seep	1	6	NA	NA	NA	NA	NA
DC	135.3	orifice	3	1	21.86	485	6.75	7.15	0.3191
DC	252.7	orifice	4	3	22.56	507	8.08	7.05	0.3322
DC	261.5	group of springs	2	3	22.49	497	6.73	7.24	0.3301
DC	264.5	orifice	3	3	22.42	485	7.53	7.16	0.3107
DC	267.5	Seep	1	1	NA	NA	NA	NA	NA
DC	278.6	orifice	3	3	22.55	486	7.77	7.14	0.3114
DC	279.7	orifice	3	3	22.48	494	7.55	7.21	0.3250
DC	641.0	orifice	3	3	22.54	479	8.16	7.27	0.3069
DC	644.4	orifice	3	3	22.53	479	8.05	7.13	0.3071
DC	650.5	group of springs	2	6	NA	NA	NA	NA	NA
DC	658.2	group of springs	3	6	21.70	488	8.12	7.45	0.3169
233	DU –	Blue Hole: FS - Fin	agan Spr	ingg: DC - D	alan Craak: NA -	- too challow to	ampla		

234

233 BH = Blue Hole; FS = Finegan Springs; DC = Dolan Creek; NA = too shallow to sample.

235 The top ranking abundance models for both the Poisson and NB models contained the

236 additive effects of rain, site and flow within the detection parameter (global models). The global

negative binomial model had the lowest AIC_c value with an overall AIC_c weight of 0.84 237

238 compared to the other 12 models. There was a change in AIC_c of 3 compared to the following

239 model and a separation of AIC_c of 424 to the closest Poisson model (Table 2). All NB models

had lower AIC_c scores than the Poisson. The global Poisson model scored 143 AIC_c points lower 240

- than the null negative binomial model. Site and flow had positive relationships within the models
- 242 with rain having a negative effect on the detection of riffle beetles.

Table 2. Results from negative binomial (NB) and Poisson (P) open season N-mixture abundance models for *Heterelmis c f glabra*

Model	# of Parameters	Δ AICc	AICcWt	Cum.Wt	LL	Distribution
$\lambda(.) p(Flow+Rain+Site)$	8	0	0.84	0.84	-1004.11	NB
$\lambda(.) p(Flow+Rain)$	7	3.38	0.16	1	-1007.6	NB
$\lambda(.) p(Flow+Site)$	7	132.45	0	1	-1072.13	NB
$\lambda(.) p(\text{Rain})$	6	143.95	0	1	-1079.54	NB
$\lambda(.) p(Flow)$	6	145.89	0	1	-1080.51	NB
$\lambda(.) p(.)$	5	279.44	0	1	-1148.81	NB
$\lambda(.) p(Site)$	6	281.39	0	1	-1148.26	NB
$\lambda(.) p(Flow+Rain+Site)$	7	424.09	0	1	-1217.96	Р
$\lambda(.) p(\text{Flow}+\text{Rain})$	6	487.43	0	1	-1251.28	Р
$\lambda(.) p(Flow+Site)$	6	569.90	0	1	-1292.51	Р
$\lambda(.) p(Rain+Site)$	6	799.78	0	1	-1407.46	Р
$\lambda(.) p(\text{Rain})$	5	836.67	0	1	-1427.42	Р
$\lambda(.) p(.)$	4	1032.48	0	1	-1526.74	Р

245 NB = negative binomial; P = Poisson; Cum WT = cumulative weight; LL = log-likelihood.

246 Model selection was based upon the AICc score initially. Subsequently, the preliminary

selection of the global NB at a K of 400 was modeled with a goodness of fit and evaluated at

248 different levels of K. The levels of K ranged from the default value of 157 to the selected value

of 600 (Fig 3). The abundance parameter estimates of K at 500 and 600 became inseparable.

250 However, the probabilities of detection decreased as K level increased. Goodness of fit tests at

251 500 K showed the mathematical fit of the model (p = 0.08) with a c-hat (\hat{c}) of 1.68 (Table 3).

Goodness of fit tests for the 600 K model showed a lack of fit (p = 0.01) and had a higher \hat{c}

253 (1.91) than the 500 K model. Therefore, the global NB model at a K of 500 was selected as the

appropriate model for this data set.

Fig 3. Results from the selected N-mixture model. Negative binomial mixture of global
 models showing range of site level estimates with varying values of K.

Κ	AICc	COE	•
	AICC	GOF	Ĉ
ite) 600	2030.13	0.01	1.91
ite) 500	2030.14	0.08	1.68
ite) 400	2030.48	0.03	1.89
ite) 300	2032.62	NA	NA
ite) 200	2040.48	NA	NA
ite) 157	2052.59	0.02*	2.1*
	ite) 500 ite) 400 ite) 300	$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	ite)6002030.130.01ite)5002030.140.08ite)4002030.480.03ite)3002032.62NAite)2002040.48NA

258 Table 3. Selected global abundance model and the relationship with K.

259 K = values for the upper bound of integration.

260 *Asterisks indicate warnings associated with the analysis

261	The abundance parameter estimates taken from the global NB model at 500 K for
262	combined spring sites sampled increased over the year from 2,609 (+690/-874) to 2,946 (+451/-
263	479). Overall probabilities of detection in the NB 500 model ranged from 0.07 to 0.80 on
264	average (Table 4). Site 291 at Finegan Springs had the highest probability of detection for the
265	data set (0.856). Site 120 at Dolan Springs had the lowest probability of detection (0.045). The
266	results suggest that as the flow increases the probability of detecting a beetle increases. The
267	probability of detection is the lowest for the events (April and August) with rainfall during the
268	deployment.

269	Table 4. Probabilities of detection by site along with averages calculated from the selected
270	negative binomial global model.

Spring	Site	March	May	August	November	Average
Blue	8	0.241	0.134	0.134	0.241	0.187
Blue	9	0.241	0.134	0.134	0.241	0.187
Blue	1	0.457	0.291	0.291	0.457	0.374
Blue	2	0.691	0.522	0.522	0.691	0.607
Blue	5	0.241	0.134	0.134	0.241	0.187
Finegan	84.0	0.241	0.134	0.134	0.241	0.187
Finegan	94.7	0.107	0.055	0.055	0.107	0.081
Finegan	105.0	0.241	0.134	0.134	0.241	0.187
Finegan	110.0	0.241	0.134	0.134	0.241	0.187
Finegan	151.5	0.457	0.291	0.291	0.457	0.374
Finegan	175.7	0.691	0.522	0.522	0.691	0.607
Finegan	184.3	0.691	0.522	0.522	0.691	0.607
Finegan	191.0	0.457	0.291	0.291	0.457	0.374
Finegan	212.0	0.107	0.055	0.055	0.107	0.081

Finegan	280.5	0.457	0.291	0.291	0.457	0.374
Finegan	285.0	0.457	0.291	0.291	0.457	0.374
Finegan	310.4	0.457	0.291	0.291	0.457	0.374
Finegan	319	0.457	0.291	0.291	0.457	0.374
Finegan	291.5	0.856	0.744	0.744	0.856	0.800
Dolan	43.0	0.403	0.248	0.248	0.403	0.326
Dolan	120.0	0.087	0.045	0.045	0.087	0.066
Dolan	135.3	0.403	0.248	0.248	0.403	0.326
Dolan	252.7	0.643	0.467	0.467	0.643	0.555
Dolan	261.5	0.203	0.110	0.110	0.203	0.156
Dolan	264.5	0.403	0.248	0.248	0.403	0.326
Dolan	267.5	0.087	0.045	0.045	0.087	0.066
Dolan	278.6	0.403	0.248	0.248	0.403	0.326
Dolan	279.7	0.403	0.248	0.248	0.403	0.326
Dolan	641.0	0.403	0.248	0.248	0.403	0.326
Dolan	644.4	0.403	0.248	0.248	0.403	0.326
Dolan	650.5	0.203	0.110	0.110	0.203	0.156
Dolan	658.2	0.403	0.248	0.248	0.403	0.326

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273 Univariate Relationships and Habitat Associations

274 Indicator analysis on the Elmidae communities collected on transition lures and spring

275 origin lures showed that *H. cf. glabra* adults and larvae are associated with spring sources (p =

276 0.001; Table 5). The associations of *Microcylloepus* sp. adults and larvae with spring sites were

277 not significant although were shown to associate with the transition zone more than the spring

278 origin. *Phanocerus* larvae were significantly associated with the spring origin sites (p = 0.001).

279 Other genera were not detected within the transition zone or the spring area.

Table 5. Indicator species analysis results showing treatment, relevant indicator value, and
 significance.

	Treatment	Indicator Value	p-value
Adult Heterelmis cf. glabra	Spring	0.85	0.001
Larval Heterelmis cf. glabra	Spring	0.81	0.001
Adult Microcylloepus pusillus	Transition	0.28	0.953
Larval Microcylloepus pusillus	Transition	0.58	0.167

	Larval Phanocerus clavicornis	Spring	0.71	0.001
202	Significance set at $\alpha < 0.05$			

282 Significance set at $\alpha \le 0.05$.

283 Only three abiotic variables (temperature, flow, and substrate) had significant correlations 284 with either adult or larval H. cf. glabra average count data. The first relationship was between 285 the measured temperature at the time of the mapping and the average count data of adult H. cf. 286 glabra (Fig 4). As the temperature increases there is a significant increase in the presence of adult H. cf. glabra ($F_{1,26} = 10.14$, $r^2 = 0.28$, p = 0.003). Heterelmis cf. glabra also exhibited a 287 288 significant relationship with flow (Fig 4). As flow increased the average number of adult H. cf. 289 glabra collected increased ($F_{1,30} = 16.64$, $r^2 = 0.35$, p = 0.003). A negative relationship was 290 observed with *H. cf. glabra* larvae and substrate (Fig 4). As substrate size increased the average 291 count of *H. cf. glabra* larvae decreased ($F_{1,30} = 9.39$, $r^2 = 0.23$, p = 0.004), suggesting potential 292 habitat partitioning between adult and larval H. cf. glabra as the adult correlation was positive 293 although not significant. 294 Figure 4. Univariate relationships with measured abiotic parameters of adult and larval 295 Heterelmis cf. glabra collected in Val Verde, TX. Presented are relationships for temperature,

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296

298 **Discussion**

flow and substrate. Significance set at $\alpha \le 0.05$

The results from the models suggest that rainfall, flow and the site were important to detections of *H. cf. glabra* within this area of its distribution. Individually, the covariates had different levels of fit to the models. Flow and site both had significant positive relationships highlighting differences between the site level estimates of abundance at different flows and between the sites. Rainfall did have a negative relationship with the probability of detection,

which was hypothesized to have such an effect due to the sedimentation issues and local changesin water chemistry during and post rainfall.

The calculated estimates from the NB global model of site level abundance and total abundance seem realistic and ecologically plausible. Although the area from which the lure is sampling the beetles is not known, the sampling consistency at all sites provides reliable comparisons between sites. As mentioned previously, flow displayed a strong positive relationship to the probability of detecting beetles. The flow at each site may in part determine the area from which these beetles were drawn. Therefore, the greater the flow the more potential microhabitat from which to draw beetles to the lure.

313 When abundance parameter estimates from model runs are compared to raw data, the 314 sites with lower raw counts seem to have higher predicted abundance than in the calculated data 315 for sites with higher raw counts. For example, seep sites, ranked with a flow of 'one', have the 316 lowest probabilities of detection among all of the flow categories and higher estimates of 317 abundance than the count data for these sites. The model appears to be accounting for riffle 318 beetles potentially not present at the site due to the low probability of detection at these lower 319 flowing sites, suggesting a sampling issue with the seep sites. Therefore, one scenario would be 320 where the beetles are not being detected, although present, thereby inflating the abundance score 321 associated with these types of sites. Another possibility is that at these lower flowing spring sites, 322 the zeros in count data could be true zeros not modeled within the predicted data set. Either 323 scenario discussed above, is highlighting the need for a better estimate of flow than the 324 categorical type that was used for these models, or disregarding seep sites for these types of 325 models. Adding flow as a continuous variable, collected at deployment and pick up, may provide 326 more reliable site level abundances than using the ordinal flow designation. Monitoring the flow

at each site with a weir or other technique at the beginning and end of each event may provide
the subtle changes in flow data that could be used as a sampling covariate to explain fluctuations
in abundance over time.

330 The negative binomial models were ranked lower by AICc scores than any of the Poisson 331 models. This seems to be the trend for the NB models showing higher site level abundance when 332 compared to the corresponding Poisson models. In some cases, NB models in other studies have 333 produced approximately double the abundance of the normal territory mapping method [24; 334 Add]. Goodness of fit test on selected NB model results I this study showed acceptable fit as the 335 Poisson models were not in congruence [24]. Subsequently, a recent study [37] suggested the 336 estimates from K'ery et al. (2005; [24]) may be more reliable had the Poisson distribution been 337 selected and not the NB model. For this study, the Poisson models (global and null) both had 338 very high values (+15) of \hat{c} , indicating lack of fit within the model. In this case, based upon the 339 fitted levels of K, goodness of fit tests, and the AICc score the NB model at 500 K was selected. 340 There are a number of differences between our study organism and the organisms studied 341 in the available literature where n-mixture models were employed. For example, our site level 342 population estimates are consistently larger than many other reported estimates of lambda [38, 343 39]. Due to the small size of the beetle there is the potential to have large numbers of individuals 344 detected within or on the lure. Many studies have much shorter sampling periods (e.g., five to ten 345 minutes for bird surveys; [13, 14], however, with this study the lures were deployed for weeks as 346 it takes time for biofilms to grow on the lures attracting the beetles. Although N-mixture models 347 have not been used for many studies involving invertebrates, this approach is been useful with 348 this particular species.

349 Flow and temperature were both significant abiotic variables when compared to the 350 average adult H. cf. glabra count data. To determine if flow is the actual mechanism or if size of 351 the spring-run is creating larger abundances, size of the run should be considered a site covariate. 352 Greater spring flows presumably sustain more suitable habitat from which to draw beetles to the 353 lure. Temperature values were collected during the mapping event in January of 2016, therefore, 354 lower temperatures may show spring sites with more exposure to the environment or shallow 355 laminar flow which is more susceptible to temperature fluctuations at the surface or farther away 356 from the origin of the spring. Future efforts may consider incorporating temperature as a 357 covariate measured at the beginning and end of each sampling event. 358 Overall, N-mixture models have great potential as a monitoring tool for rare, small, and

difficult to collect interstitial species, such as riffle beetles. In order to determine trends within the population, regular monitoring of the beetles should be done with set monitoring locations at least three times a year for a number of years. In addition, sites should be added if possible to increase the sample size in order to conduct hypothesis tests on covariates that influence abundance, not only the probability of detection. After these baseline surveys are conducted, future surveys could be compared and changes in surface populations and available habitat of spring-adapted riffle beetles could be elucidated.

The Devils River has long been recognized as a least disturbed stream and has many unique species associated with its watershed. While anthropogenic stressors have been lacking in the area historically, advances in nontraditional oil and gas activity has created opportunity for industrial expansion in this region. Over 47,000 oil and gas wells were permitted in the Permian Basin region between 2011 and 2016, with water usage per well increasing 770% (to up to 42,500 m³ per well) during that same period [40]. Current commercial estimates predict

372 continued growth of production within the region over the next few years, suggesting the 373 demand for water in the region will increase. Groundwater dependent rivers and streams, such as 374 the Devils River, may experience decreases in springflow and thus overall streamflow due to 375 pumping activity. Modeling demonstrated that "production of groundwater in the [Devils River] 376 basin will result in a proportional reduction in the flow of the Devils River" with the impact 377 "most pronounced during low flow conditions" potentially impacting the ecology of the system 378 as spring discharge points in the river are extinguished [1]. While H. cf glabra currently has no 379 protected status, the need for such protections would grow if known populations are negatively 380 impacted by water development. Incorporating springflow metrics protective of spring habitats 381 into groundwater management is needed to reduce potential impacts to rare spring dependent 382 species, such as *H. cf glabra* while ensure sustainable water supplies.

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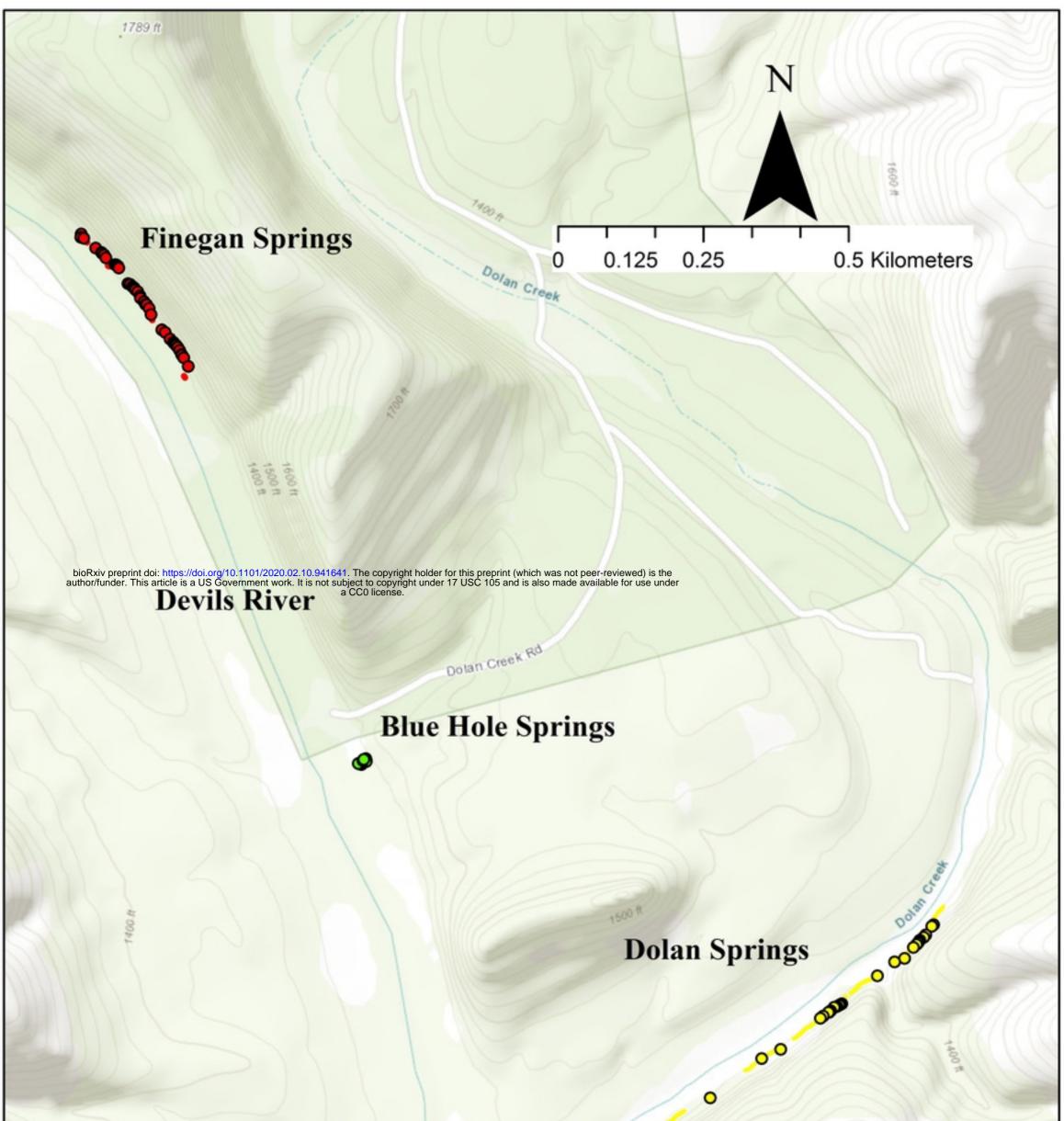
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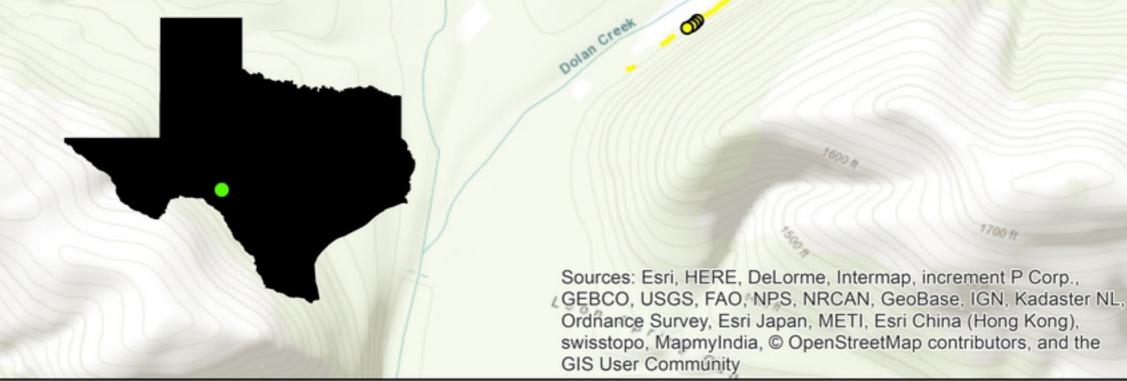


Fig 1



Fig 2

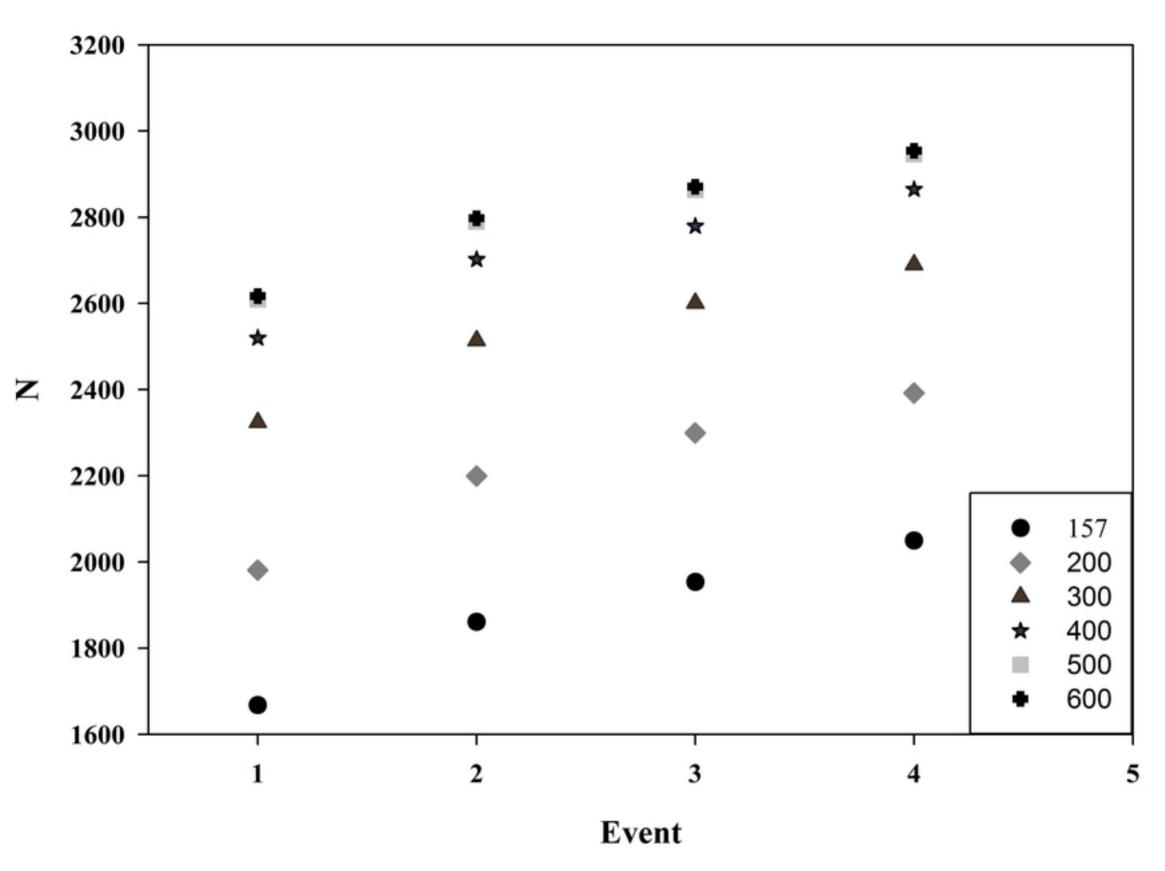
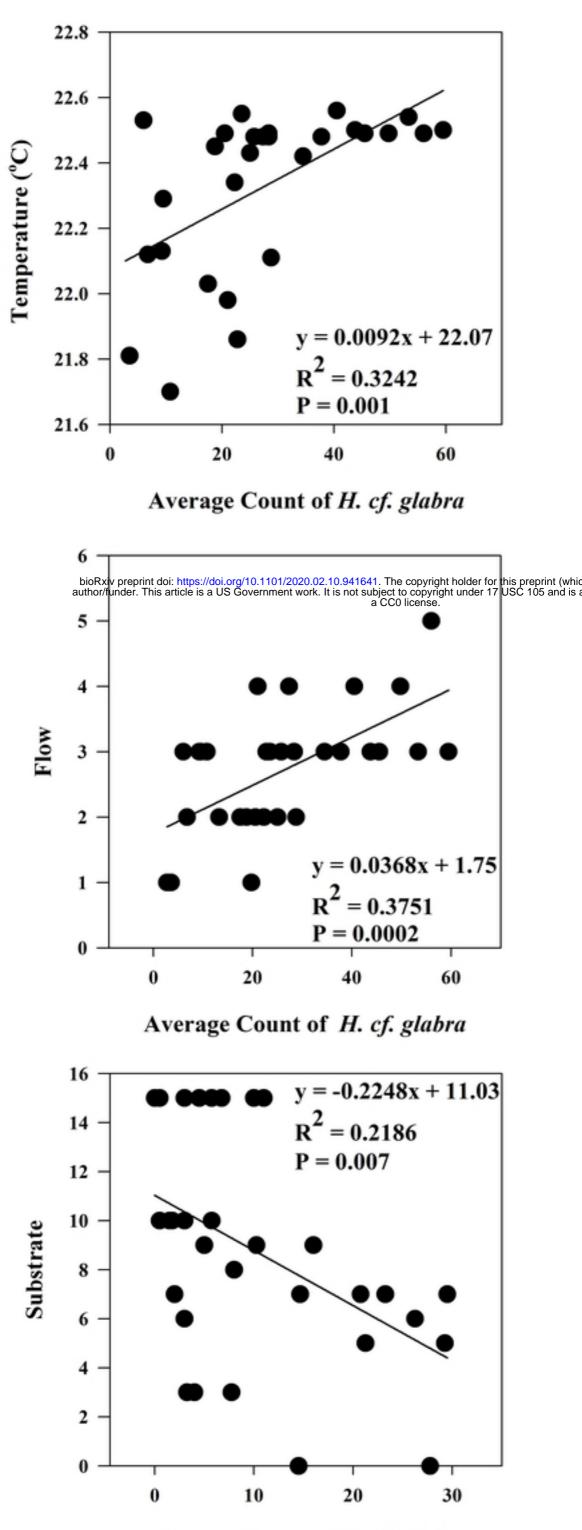


Fig 3



Average Count of H. cf. glabra

Fig 4