

1 **Multiple performance peaks for scale-biting in an adaptive radiation of pupfishes**

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3 Anson Tan^{1,2}, Michelle St. John³, Dylan Chau^{1,2}, Chloe Clair^{1,2}, HoWan Chan⁴, Roi Holzman^{5,6},

4 Christopher H. Martin^{1,2*}

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6 ¹Department of Integrative Biology, University of California, Berkeley

7 ²Museum of Vertebrate Zoology, University of California, Berkeley

8 ³Department of Biology, University of Oklahoma

9 ⁴Department of Biology, Rice University

10 ⁵School of Zoology, Tel Aviv University, Eilat, Israel

11 ⁶Inter-University Institute for Marine Sciences, Eilat, Israel

12

13 *Corresponding author: Christopher Martin, chmartin@berkeley.edu, 3101 Valley Life Sciences

14 Building, University of California, Berkeley 94720

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22 **Keywords:** adaptive radiation, functional morphology, ecomorphology, bite kinematics,

23 lepidophagy, trophic specialization, performance landscape

24 **Abstract**

25 The physical interactions between organisms and their environment ultimately shape their rate of
26 speciation and adaptive radiation, but the contributions of biomechanics to evolutionary
27 divergence are frequently overlooked. Here we investigated an adaptive radiation of *Cyprinodon*
28 pupfishes to measure the relationship between feeding kinematics and performance during
29 adaptation to a novel trophic niche, lepidophagy, in which a predator removes only the scales,
30 mucus, and sometimes tissue from their prey using scraping and biting attacks. We used high-
31 speed video to film scale-biting strikes on gelatin cubes by scale-eater, molluscivore, generalist,
32 and hybrid pupfishes and subsequently measured the dimensions of each bite. We then trained
33 the SLEAP machine-learning animal tracking model to measure kinematic landmarks and
34 automatically scored over 100,000 frames from 227 recorded strikes. Scale-eaters exhibited
35 increased peak gape and greater bite length; however, substantial within-individual kinematic
36 variation resulted in poor discrimination of strikes by species or strike type. Nonetheless, a
37 complex performance landscape with two distinct peaks best predicted gel-biting performance,
38 corresponding to a significant nonlinear interaction between peak gape and peak jaw protrusion
39 in which scale-eaters and their hybrids occupied a second performance peak requiring larger
40 peak gape and greater jaw protrusion. A bite performance valley separating scale-eaters from
41 other species may have contributed to their rapid evolution and is consistent with multiple
42 estimates of a multi-peak fitness landscape in the wild. We thus present an efficient deep-
43 learning automated pipeline for kinematic analyses of feeding strikes and a new biomechanical
44 model for understanding the performance and rapid evolution of a rare trophic niche.

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46

47 **Introduction**

48 Most biodiversity evolved during adaptive radiation, the process by which a single lineage
49 colonizes myriad ecological niches and rapidly diversifies (Gillespie et al., 2020; Martin and
50 Richards, 2019; Simpson, 1944; Stroud and Losos, 2016). However, this process of rapid
51 adaptation is also constrained by physical laws governing interactions between the organism and
52 the essential tasks needed for survival and reproduction within its environment, known as the
53 emerging field of ecomechanics (Higham et al., 2016; Higham et al., 2021; Perevolotsky et al.,
54 2020). Understanding how functional traits interact in complex ways to achieve these essential
55 life history tasks is fundamental for connecting population genomics to phenotypes to
56 performance to fitness to adaptive radiation within new environments (Arnold, 1983; Martin et
57 al., 2019). For example, estimating the performance landscape for a functional task or tasks can
58 lead to deep insights about patterns of morphospace occupation and rates of morphological
59 diversification across diverse taxa (Armbruster, 1990; Benkman, 1993; Figueirido et al., 2013;
60 Holzman et al., 2022; Keren et al., 2018; Olsson et al., 2020; Raup, 1966; Schultz et al., 2021;
61 Stayton, 2019; Tseng and Flynn, 2018). However, few systems have been investigated across
62 these levels of biological organization.

63 Scale-biting, or lepidophagy, is a fascinating ecological niche that has evolved at least 20
64 times in fishes and within a diverse range of environments, from rift lakes (Hori, 1993; Raffini et
65 al., 2017; Stewart and Albertson, 2010; Takahashi et al., 2007) to tropical streams and rivers
66 (Evans et al., 2017; Gosavi et al., 2018; Gosavi et al., 2019; Grubh et al., 2004) to the
67 mesopelagic zone (Nakae and Sasaki, 2002) and across ontogenetic stages (Davis et al., 2011;
68 MacLeod, 2020; Novakowski et al., 2004; Szelistowski, 1989) in a diverse range of fish groups,
69 including cichlidiformes, characiformes, siluriformes, and sharks (Kolmann et al., 2018; Martin

70 and Wainwright, 2013c; Sazima, 1984). This trophic niche provides an excellent setting for
71 investigating the functional traits underpinning performance and fitness because successful scale-
72 eating appears to require both intensive prey capture behaviors and efficient energy use per strike
73 (Sazima, 1984). In contrast to piscivory, all scale-eaters are generally smaller than their larger
74 prey and numerous strikes must be performed efficiently due to the low energy payoff in calories
75 per strike, resulting in an excellent laboratory and field model for observing repeated prey
76 capture behaviors. Scale-eaters rarely or never consume whole prey (Gosavi et al., 2018; Kovac
77 et al., 2019; Martin and Wainwright, 2013c; Sazima, 1984; Sazima, 1986).

78 The scale-biting pupfish, *Cyprinodon desquamator*, is the youngest lepidophage so far
79 discovered; this species evolved within the past 10-15 kya on San Salvador Island, Bahamas
80 (Martin and Wainwright, 2013a; Martin et al., 2019). Scales comprise approximately 50% of its
81 diet in addition to macroalgae and microinvertebrates in several hypersaline lakes where it is
82 endemic (Martin and Wainwright, 2011; Martin and Wainwright, 2013c; McLean and Lonzarich,
83 2017). All other extant scale-eating lineages likely evolved at least 1 Mya (Koblmüller et al.,
84 2007), except for the extinct Lake Victorian lepidophagous cichlid *Haplochromis welcommei*
85 (Greenwood, 1965). In several interior hypersaline lakes on San Salvador Island (SSI), *C.*
86 *desquamator* occurs in sympatry at low frequencies in the same benthic macroalgae-dominated
87 habitats as *C. variegatus*, a widespread generalist species, *C. brontotheroides*, an endemic oral-
88 shelling molluscivore species, and *C. sp.* ‘wide-mouth’, a newly discovered intermediate scale-
89 eating ecomorph which is not included in this study (Hernandez et al., 2018; Martin and
90 Feinstein, 2014; Richards and Martin, 2022; St John et al., 2020).

91 The SSI adaptive radiation exhibits one of fastest rates of craniofacial diversification
92 among any measured vertebrate group, up to 1,000 times faster than generalist populations on

93 neighboring Bahamian islands for oral jaw length, due to rapid adaptation to the novel trophic
94 niche of lepidophagy (Martin, 2016b; Martin and Wainwright, 2011). Previous field experiments
95 measuring the growth and survival of F2 and F5 hybrids among all three species placed in field
96 enclosures within two hypersaline lakes on San Salvador Island estimated a two-peak fitness
97 landscape and a large fitness valley isolating the hybrids with greatest phenotypic similarity to
98 scale-eaters (Martin and Wainwright, 2013b; Patton et al., 2022). This two-peak landscape was
99 stable over multiple years, lake environments, and phenotype-frequency manipulations,
100 suggesting that biophysical constraints on the interaction between pupfish craniofacial traits and
101 their bite performance on scales may ultimately shape the adaptive landscape in this system,
102 rather than frequency-dependent competition (e.g. (Hori, 1993; Martin, 2012)) or seasonal
103 resource abundance (Grant and Grant, 2002), which would predict a dynamically changing
104 landscape (Martin, 2016a; Martin and Gould, 2020). Furthermore, our genomic and
105 developmental genetic analyses suggest that introgression of genetic variation for traits related to
106 feeding behavior triggered adaptive radiation on SSI and led to the reassembly of Caribbean-
107 wide genetic variation into a specialized scale-eater and molluscivore on a single island over
108 several thousand years (McGirr and Martin, 2017; McGirr and Martin, 2018; Palominos et al.,
109 2023; Richards and Martin, 2017; Richards et al., 2021).

110 In our initial pilot study of scale-biting kinematics in this system, we found that scale-
111 eaters behaviorally reduced their realized peak gape during feeding strikes by reducing their jaw
112 opening angle (St. John et al., 2020). However, we were unable to robustly quantify the
113 performance landscape with limited strike and landmark data and did not measure F2 hybrids.
114 Here we used all new feeding videos filmed at substantially improved resolution and developed a
115 new automated machine-learning pipeline to quantify five landmarks on nearly every frame. We

116 sampled extensive phenotypic diversity within the radiation using both purebred species and
117 multiple F2 hybrid intercrosses and backcrosses to estimate the five-dimensional kinematic
118 performance landscape for scale-biting.

119

120 **Methods**

121 *Collection and Husbandry*

122 Using seine nets or hand nets, we collected molluscivore (*C. brontotheroides*) and scale-eating
123 (*C. desquamator*) pupfish from Crescent Pond, Little Lake, and Osprey Lake on San Salvador
124 Island, Bahamas, and generalist (*C. variegatus*) pupfish from Lake Cunningham on New
125 Providence Island, Bahamas and Fort Fisher estuary in North Carolina, United States in 2017 and
126 2018. Generalist pupfish from SSI could not be readily trained to feed on gelatin cubes during
127 the period of this study; however, generalists from neighboring islands exhibit essentially
128 identical craniofacial morphology and kinematics relative to SSI generalists (Martin, 2016b; St.
129 John et al., 2020) . Wild-caught and lab-reared fish were maintained in 40-80 l aquaria at
130 salinities of 2-8 ppt and 23-27°C and fed a diet of frozen bloodworms, frozen mysis shrimp, and
131 commercial pellet foods daily. This study used only second-generation through fourth generation
132 lab-reared pupfishes. All newly hatched fry were fed *Artemia* nauplii for approximately one
133 month. All SSI species can be readily crossed to produce viable and fertile hybrids (Martin et al.,
134 2017; St. John et al., 2021). F1 hybrid and F2 hybrid intercrosses were generated from
135 molluscivore x scale-eater crosses from both Osprey Lake and Crescent Pond, generalist x scale-
136 eater and generalist x molluscivore crosses from Little Lake, and generalist x *C. sp.*
137 ‘broadmouth’ F1 hybrids from Osprey Lake (Richards and Martin, 2022). Prior to filming,

138 pupfishes were fed exclusively Repashy Superfood gel diets for acclimation and training for at
139 least one week before filming.

140

141 *High-speed filming and measurement of gelatin bites*

142 We recorded pupfishes feeding on standardized gelatin cubes (dimensions: 1.5 cm x 1.5cm x 1.5
143 cm x 1.5 cm cube; Community Plus Omnivore Gel Premix, Repashy Specialty Pet Products).
144 Gels were prepared in batches of 50 at precisely a 4:1 water:mix ratio in silicone molds
145 following the manufacturer's instructions and allowed to set overnight at 4° C. Gels were stored
146 covered at 4° C for a maximum of two weeks before discarding. The gel cube retains its shape in
147 water and therefore allows precise measurements of the dimensions and area removed by each
148 bite.

149 Individuals were trained in group tanks to feed on gelatin cubes and then netted
150 individually for filming. We filmed all strikes at 1,100 fps using a full-color Phantom VEO 440S
151 (Vision Research Inc.) with a Canon EF-S 60 mm lens mounted on a standard tripod. Fish were
152 filmed individually in 7.5 liter bare glass tanks with a solid matte background at salinities of 2-3
153 ppt and 21-23°C. Illumination was provided by two dimmable full spectrum LED lights placed
154 on either side of the filming tank. Gelatin cubes were held with forceps with one edge facing in a
155 horizontal direction toward the fish (Fig. 1). Trained fish usually attacked the gel almost
156 immediately after placement in the filming tank. After each strike the cube was immediately
157 removed and inspected to confirm a bite mark; missed strikes were confirmed from the video
158 replay. New cubes were used for each feeding strike and never re-used. All videos were filmed in
159 lateral view.

160 The length, width, and depth of each gelatin bite were measured using digital high-
161 precision calipers (Mitutoyo) under a stereomicroscope for nearly all strikes (Fig. S1). Strikes
162 were characterized as an edge, corner, scrape, or miss. Edge bites occurred along the edge but
163 not the corner of the gelatin cubes and comprised the majority of strikes. Corner bites were
164 removed from one of the corners, which may affect bite dimensions, so were distinguished from
165 edge strikes. Scrapes were defined by bites in which the jaws did not completely close around
166 the gel to remove a chunk of the material, but instead left two distinct indentations. Misses were
167 defined as strikes in which the oral jaws of the fish contacted the gel but did not leave any marks.
168 Strikes in which the jaws did not make visible contact with the gel were excluded. Most
169 individuals were filmed consecutively over one or two filming periods for up to sixteen strikes.
170 After each filming session for each individual, an image of a ruler was photographed in the
171 filming tank at the same distance as the gelatin cubes for calibration of videos.

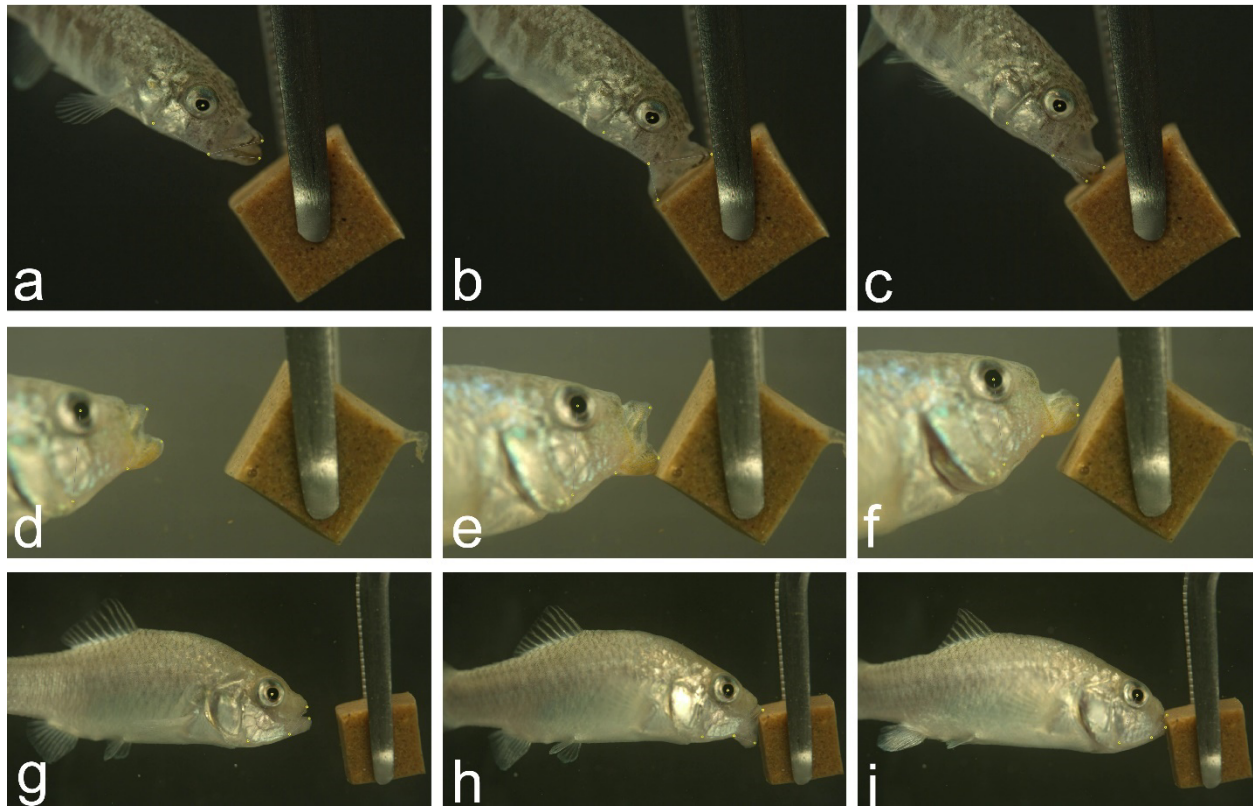
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173 ***Machine learning for quantifying kinematic landmarks on each frame***

174 We used the SLEAP (Social LEAP Estimates Animal Poses) analysis pipeline to automatically
175 detect and place morphometric landmarks on each frame (Pereira et al., 2020; Pereira et al.,
176 2022). This software supports data input of raw videos and then provides an interactive GUI to
177 create a labeled training dataset. Predictions from trained models can then be adjusted to enable
178 a ‘human-in-the-loop’ workflow to efficiently and progressively obtain more accurate models
179 and inferences of landmarks.

180 We manually placed five landmarks on 815 frames using the SLEAP GUI spanning 100
181 high-speed feeding videos including all three species and their hybrids. Frames were chosen for
182 labeling both by eye and automatically by the software to span highly divergent scenes spread

183 across the beginning, middle, and end of each strike. To train a model based on the labeled data,
184 after exploring various configurations we achieved the best performance using the multi-animal
185 bottom-up unet model with a receptive field of 156 pixels, max stride of 32 pixels, batch size of
186 3, input scaling of 0.75, and validation fraction of 0.1, resulting in a precision of 99% and mean
187 distance between labeled and inferred landmarks of 5 pixels (Fig. S2). We trained this model on
188 a laptop running a 16 Gb NVIDIA GeForce 3070 GPU, which completed training in less than
189 twelve hours. We then used this trained model to infer landmarks on each frame of a larger set of
190 strike videos using the flow cross-frame identity tracker, which shares information about
191 landmark places across frames for each individual strike video. We culled to a single instance
192 (i.e. one animal) per frame, given that fish were filmed individually, and connected single track
193 breaks. We predicted landmarks on 114,000 frames from 227 .mp4 videos (batch converted from
194 the original .cine files using Phantom Camera Control software) and spot-checked for accuracy
195 (Fig. 1). Coordinate data from each frame were exported in .hdf5 format, imported into R (Core
196 Team, 2021), and stored in any array using the rhdf5 package (Fischer et al., 2017).



197

198 **Fig. 1 High-speed video of gel-biting strikes with automated landmarks.** Strike sequences for
199 a scraping bite by a scale-eater (a-c), a miss by a hybrid (d-f), and an edge bite by a molluscivore
200 (g-i). Videos were filmed at 1,100 fps on a Phantom VEO 440S camera. Frames illustrate
201 approximately 20% of peak gape (first column), peak gape (second column), and jaw adduction
202 immediately after the bite (third column). Five yellow landmarks on each frame were placed
203 automatically using our SLEAP inference model and are illustrated as small yellow dots to
204 emphasize the accuracy of these inferred landmarks..

205

206 *Kinematic variables*

207 To quantify gel-biting strikes from coordinate data, we calculated five key kinematic variables
208 per strike: 1) peak gape, the distance from the anterior tip of the premaxilla to the anterior tip of
209 the dentary; 2) peak jaw protrusion, the distance from the center of the pupil to the anterior tip of

210 the premaxilla; 3) peak lower jaw angle, the minimum angle between the lower jaw, the
211 quadrate-articular point of jaw rotation, and the ventral surface of the fish beneath the
212 suspensorium measured from an anteroventral landmark on the preopercle. This measures the
213 maximum rotation of the oral jaws in a downward and outward direction toward the gel as
214 defined in St. John et al. (2020). Note that in Cyprinodontiformes oral jaw opening is decoupled
215 from jaw protrusion by the maxillomandibular ligament such that peak gape does not necessarily
216 occur simultaneously with peak jaw protrusion (Hernandez et al., 2009). 4) Time to peak gape
217 (TTPG) was the time in ms from 20% of peak gape to peak gape. 5) Ram speed (m/s) was
218 calculated as the distance from 20% of peak gape to peak gape (mm) divided by the time to peak
219 gape (ms). We calibrated each set of coordinates for each filming session using a ruler held at the
220 same distance from the camera as the gelatin cube. Milliseconds were calculated by counting
221 frames and multiplying by 0.909 to correct for the 1,100 frame rate per second. We then
222 subsetted to only those frames from initial start position to maximum distance from the start
223 position before the fish started to turn its head to the side post-bite so that kinematic variables
224 were only calculated from the start of the strike to the time of impact with the gelatin cube (see
225 Supplemental R code).

226

227 **Statistical analyses**

228 *Mixed-effect modeling*

229 Due to our repeated measures design of multiple strikes per fish, we used mixed-effects models
230 to compare kinematic variables and bite dimensions across species groups. We used the lme4
231 and lmerTest packages in R to fit linear mixed-effects models for each kinematic response
232 variable and bite metric with independent fixed effects for strike type and species (scale-eater,

233 molluscivore, generalist, or hybrid) plus the random intercept effect of individual ID. We
234 measured up to 16 strikes per individual fish. P-values were assessed for each factor level using
235 Satterthwaite's method (Kuznetsova et al., 2017). We used AIC to compare additional models
236 with random slopes and interactions among the fixed effects (Burnham et al., 2011).

237 Finally, due to the failure of this radiation to fit a tree-like model of evolution due to
238 extensive secondary gene flow (Richards et al., 2021), in addition to our inclusion of several
239 hybrid crosses, we did not correct for phylogenetic signal in our analyses. However, we note that
240 both generalist outgroup populations are more distantly related to each other than the scale-eater,
241 molluscivores, and hybrids on SSI (Martin and Feinstein, 2014), indicating that kinematic
242 variables and bite dimensions exhibit minimal phylogenetic signal (Losos, 2008).

243

244 *Multivariate analyses of kinematic variation*

245 We calculated principal component analysis (princomp) from the correlation matrix of kinematic
246 variables. We also used linear discriminant analysis from the MASS package in R (Ripley et al.,
247 2013) to explore overall kinematic variation among species and strike type. We further
248 calculated classification accuracy using species or strike type as the grouping variable using all
249 five kinematic variables.

250

251 *Generalized additive modeling*

252 We used generalized additive semi-parametric models (GAMs) to test for nonlinear terms
253 relationships and interactions between kinematic variables and bite dimensions (Wood and
254 Augustin, 2002). Because we were interested in directly predicting bite performance from the
255 kinematic data for each strike, we treated strike as our unit of replication, not fish, and did not

256 control for individual in our statistical models. However, within-individual variation often
257 exceeded between-species variation and we were not directly interested in species kinematic
258 differences using this modeling framework, which we previously addressed explicitly using
259 mixed-effects models controlling for individual. Moreover, methods for fitting mixed-effects
260 GAM models in R are currently limited.

261 We fit GAM models using the `mgcv` package in R (Wood, 2012; Wood and Wood, 2015)
262 with the response variables of bite length, width, depth, or volume and independent covariates of
263 species and strike type, and independent continuous kinematic variables of peak gape, peak jaw
264 protrusion, peak lower jaw angle, time to peak gape, and ram speed. We used the REML method
265 for calculating smoothness of splines and Gaussian distributions for all models. We compared
266 the fit of each kinematic variable modeled as both a linear term or a smoothing spline using
267 model selection with the AIC criterion in R. We further allowed for shrinkage of each smoothing
268 spline within the full model to determine which kinematic variables were best modeled as spline
269 terms. We then systematically compared models with both two-way thin-plate splines or all one-
270 way splines to explore whether there were any nonlinear interactions between kinematic
271 variables. Model fits were visualized with the `mgViz` (Fasiolo et al., 2020) and `ggplot2` packages
272 (Wickham et al., 2016) in R. All coding was assisted by suggestions from ChatGPT 3.5 and 4.0
273 (OpenAI, Inc.).

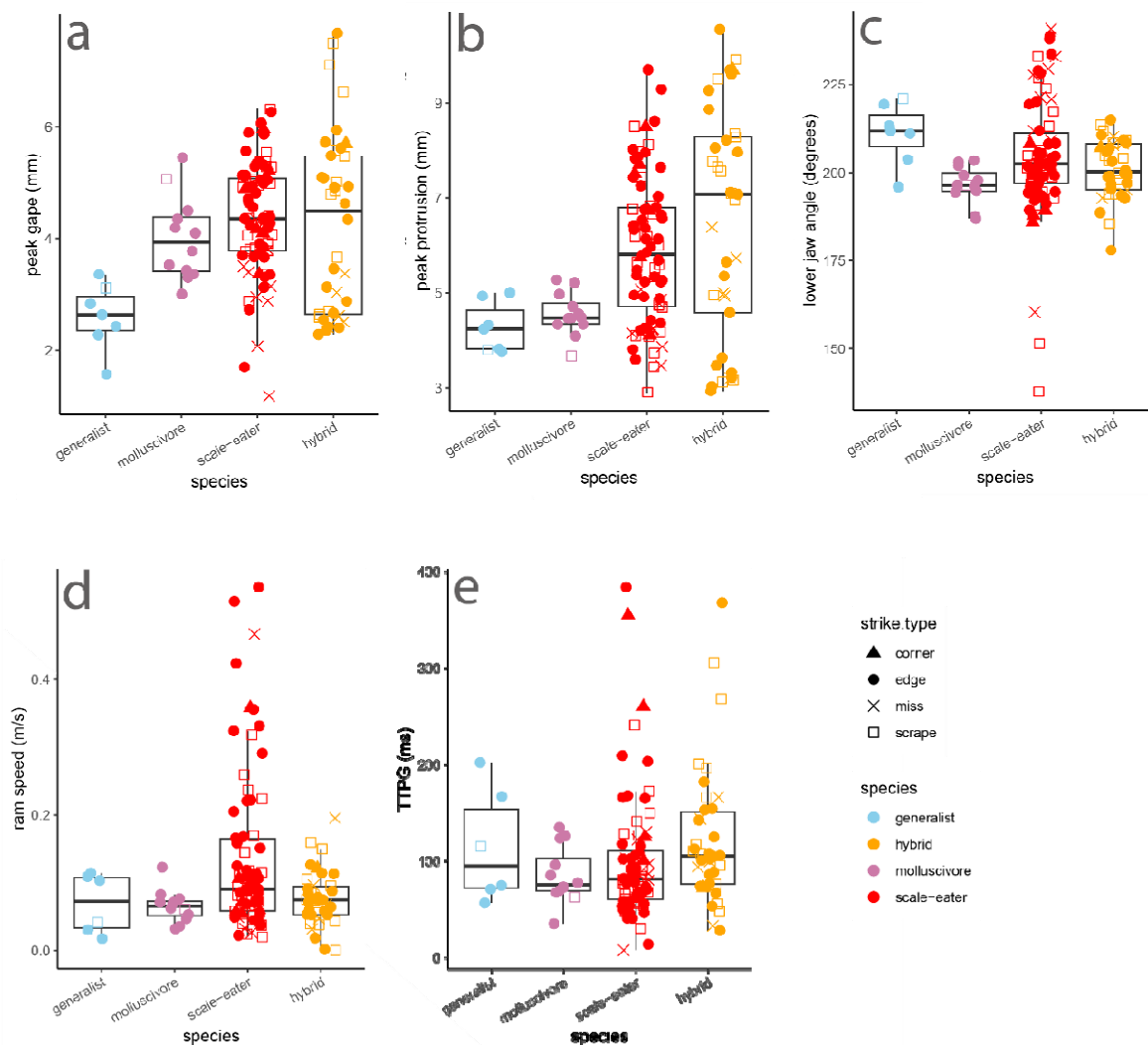
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275 **Results**

276 *Scale-eaters displayed increased gape size and bite length*

277 In mixed-effects models controlling for species, strike type, and repeated sampling of each
278 individual, we found no effect of species on bite depth or width (effect of scale-eater factor level:

279 $P = 0.062$); however, there was a significant positive effect of scale-eater species identity on bite
280 length ($P = 0.0060$). Among the five kinematic variables, only increased peak gape was
281 significantly associated with scale-eater strikes ($P = 0.040$). No kinematic variables were
282 significantly associated with any other species or hybrids across strike types (Fig. 2), even when
283 comparing missed strikes to successful bites.
284



285
286 **Fig. 2 Scale-eaters only differ in increased peak gape during gel-biting strikes.** Boxplots
287 overlaid with raw data show five kinematic variables measured during gel-biting strikes
288 measured from automated landmarking of 227 videos. Species or hybrid cross is indicated by

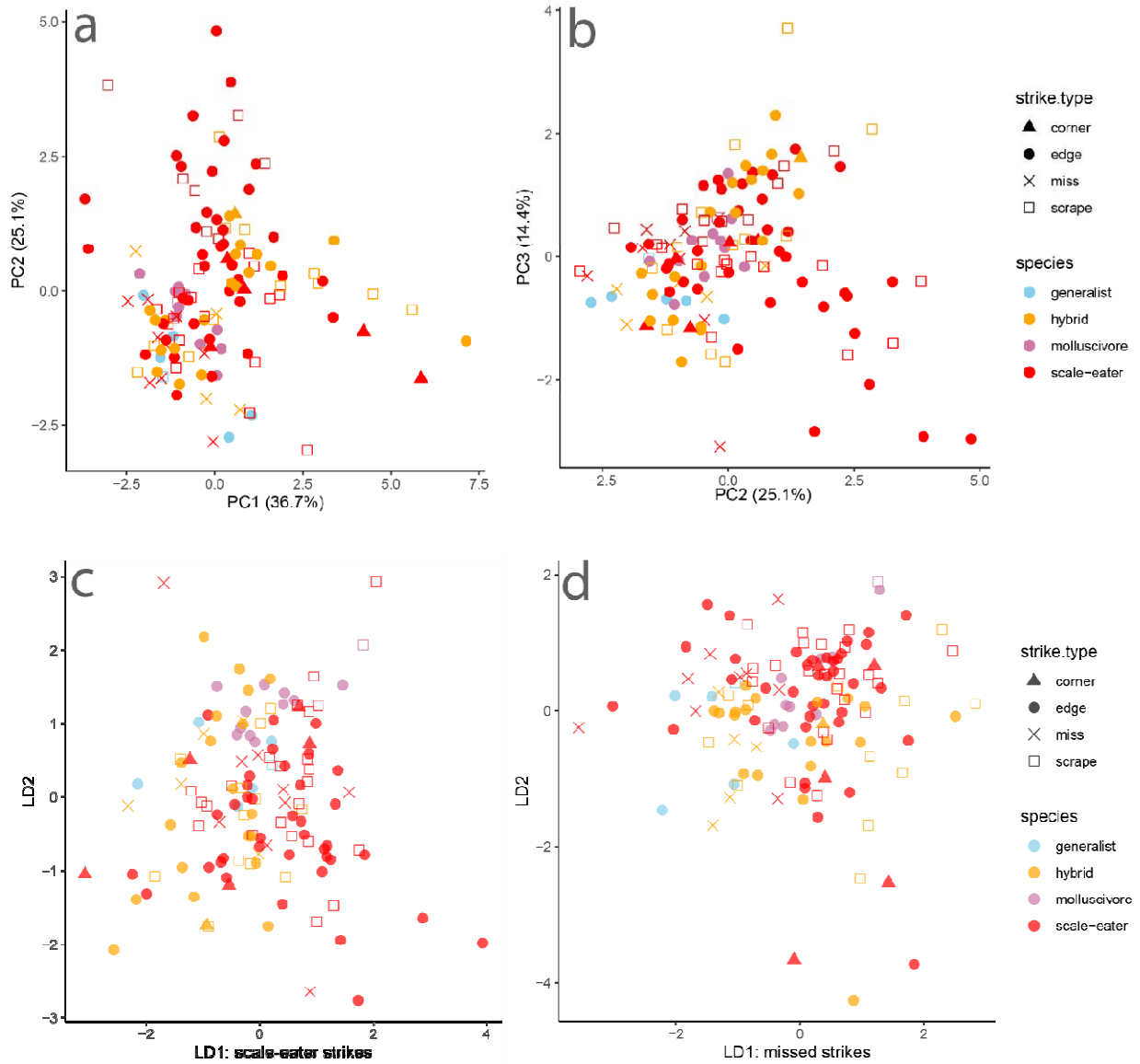
289 color and strike type is indicated by shape. TTPG: time to peak gape. Lower jaw angle is the
290 minimum angle between the lower jaw and suspensorium from 20% to peak gape. Ram speed
291 was calculated from the distance traveled between 20% and peak gape.

292

293 *Substantial similarity in strike kinematics among species and strike types*

294 Similarly, we found substantial overlap in kinematic variation across species and strike types.
295 Principal component analysis showed the strongest loadings of peak gape and peak jaw
296 protrusion on PC1, explaining 36.7% of kinematic variance among strikes (Fig. 3).

297 Linear discriminant analysis by strike type successfully classified strikes at a rate of only
298 50.8%. The kinematic variable best separating misses from other strike types on discriminant
299 axis one was peak gape. Linear discriminant analysis by species successfully classified species or
300 hybrids based on their strike kinematics at a rate of only 32.3%. The kinematic variables best
301 separating scale-eaters from other groups on discriminant axis one was again peak gape and peak
302 jaw protrusion while TTPG had the weakest effect on classification of species by kinematic
303 variables. Although plots of the first two principal components and discriminant axes indicate
304 greater variation within scale-eater and hybrid strike kinematics, there was also clearly
305 substantial overlap among species and hybrids (Fig. 3).



306

307 **Fig. 3 Principal component and linear discriminate analyses illustrate substantial overlap**

308 **by strike type and species.** Species or hybrid is indicated by color and strike type is indicated by

309 shape. Multivariate analyses were based on five kinematic variables: peak gape, peak jaw

310 protrusion, minimum lower jaw angle with the suspensorium, time to peak gape (TTPG), and

311 ram speed.

312

313

314

315 ***Multi-peak performance landscape for gel-biting***

316 We used generalized additive modeling to explore the relationship between kinematic variables
317 and bite size dimensions. The best fit model (Table 1) included a two-dimensional thin-plate
318 spline for peak gape and peak jaw protrusion along with fixed linear predictors for species, strike
319 type, peak lower jaw angle, time to peak gape (TTPG), and ram speed. The nonlinear interaction
320 between peak gape and peak jaw protrusion was significantly associated with both the bite length
321 (edf = 10.82, $P = 9e^{-7}$) and the overall gel volume removed (edf = 8, $P = 0.0008$) and displayed a
322 bimodal surface with two isolated performance peaks (Fig. 4). The best fit model for bite length
323 included additional significant linear effects of ram speed ($P = 0.008$) and peak lower jaw angle
324 ($P = 0.007$), but not TTPG ($P = 0.440$) in addition to significant factor levels of missed strikes (P
325 = $1.12e^{-10}$) and molluscivore species ($P = 0.012$). Models without TTPG fit the data equally well
326 ($\Delta AIC < 2$).

327 Even after excluding missed strikes that made contact but left no mark on the gel and
328 scraping bites in which the jaws did not fully occlude, the interaction between peak gape and jaw
329 protrusion was still significantly associated with edge and corner bite length (edf = 10.34, $P = 9e^{-$
330 6) and volume (edf = 11.17, $P = 4.89e^{-5}$). Bite width was significantly linearly associated with the
331 interaction between peak gape and peak jaw protrusion resulting in a flat performance landscape
332 (edf = 1.759, $P = 0.0002$), which is largely controlled by the morphological dimensions of the
333 oral jaws of each fish, rather than kinematic variables. Bite depth was not significantly associated
334 with peak gape and peak jaw protrusion nor any linear variable in this model except for the
335 factor of scraping strikes as expected based on the shallow dimensions of these strikes ($P =$
336 0.046).

337

338 **Table 1.** Model selection of generalized additive models ($n = 130$ strikes with complete
 339 kinematic and gel bite data) predicting bite length or total bite volume removed from Repashy
 340 gelatin cubes by a single strike. sp: species; g: peak gape, jp: peak jaw protrusion; ja: peak lower
 341 jaw angle with the suspensorium; TTPG: time to peak gape; rs: ram speed. Strike type included
 342 full bites from the gelatin edge, corner, scraping bites, and misses in which the jaws made

model	deviance explained	df	AIC	Δ AIC
length ~ species + strike type + s(g, jp) + ja + rs + TTPG	79.2%	23.1	322.7	--
length ~ sp. + strike type + s(g, jp) + ja + rs	79.2%	22.3	321.2	--
length ~ sp. + strike type + s(g) + s(jp) + ja + TTPG + rs	77.4%	19.7	327.0	6
length ~ sp. + strike type + s(g) + s(jp) + s(ja) + s(TTPG) + s(rs)	74.7%	14.9	331.6	10
length ~ sp. + strike type + g + jp + ja + TTPG + rs	70.7%	13	347.1	16
length ~ sp. + s(g, jp) + ja + TTPG + rs	50.3%	17.4	424.5	103
length ~ strike type + s(g, jp) + ja + TTPG + rs	72.3%	18.6	351.0	30
volume ~ species + strike type + s(g, jp) + ja + TTPG + rs	41.5	22.2	669.8	--

343 contact with the gelatin but left no impression. Significant terms within each GAM model ($P <$
 344 0.05) are highlighted in bold.

volume ~ species + strike type + s(g, jp) + ja + rs	42.7	22.2	667.6	--
volume ~ sp. + strike type + s(g) + s(jp) + ja + TTPG + rs	28.9	12.1	674.8	7
volume ~ sp. + strike type + s(g) + s(jp) + s(ja) + s(TTPG) + s(rs)	28.6	9.9	671.1	14
volume ~ sp. + strike type + g + jp + ja + TTPG + rs	28.9	13	676.7	19
volume ~ sp. + s(g, jp) + ja + TTPG + rs	13.7	9.7	695.5	28
volume ~ strike type + s(g, jp) + ja + TTPG + rs	24.5	9.7	677.9	10

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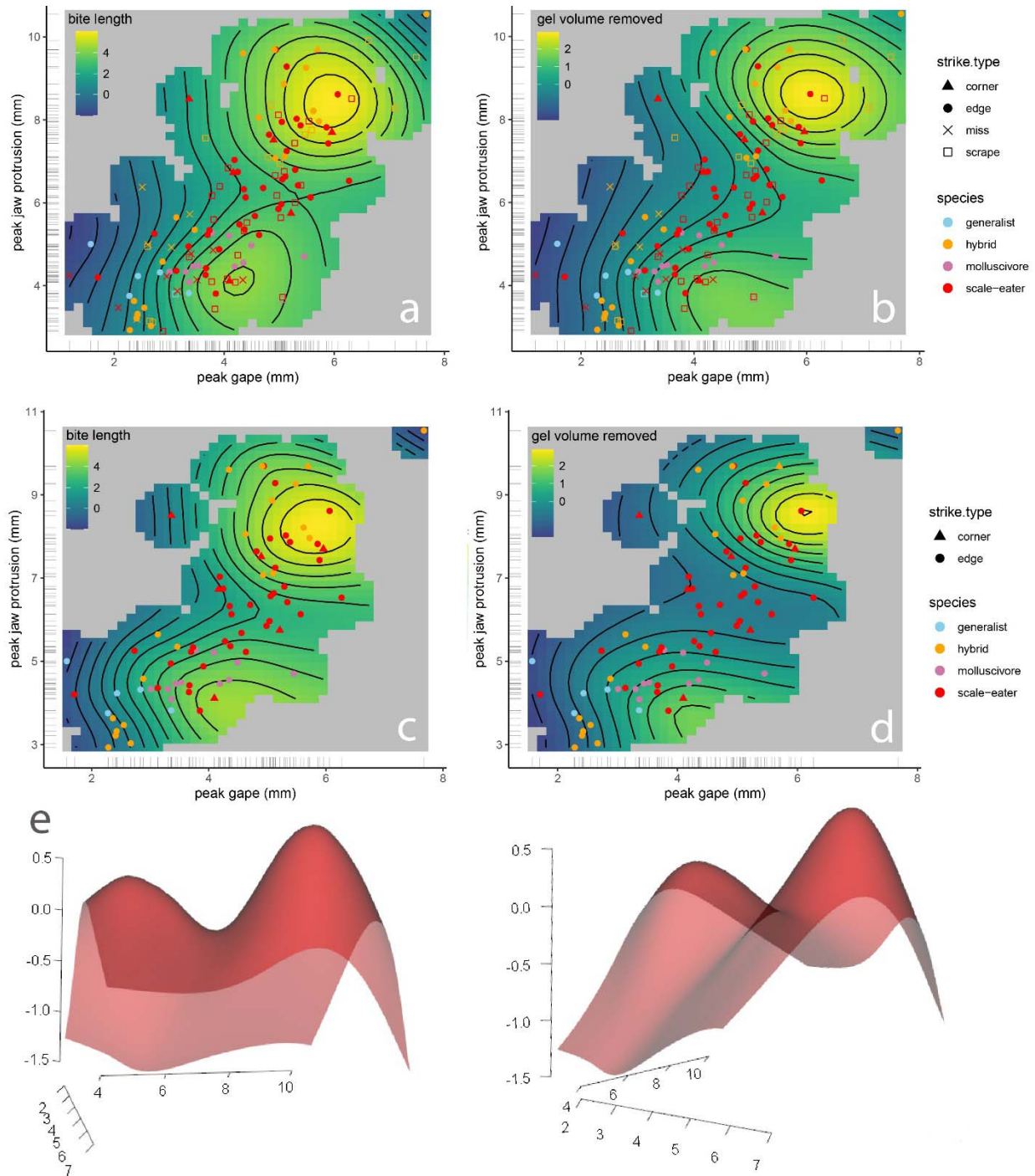
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353 **Fig. 4 Generalized additive modeling supports a nonlinear interaction between peak gape**

354 **and peak protrusion for bite performance.** Thin-plate splines from the best-fitting GAM

355 model (Table 1) for the response variable of bite length (first column: **a,c**) or total bite size (gel

356 volume removed: second column: **b,d**). First row includes all strike types and second row

357 excludes scraping and missed strikes. **e.** Performance landscape for bite length in 3D perspective
358 view. Species or hybrid cross is indicated by color and strike type is indicated by shape for 130
359 filmed strikes with data for gel bite dimensions. Gel volume removed was calculated from
360 length x width x depth of gelatin bite measured with a digital caliper under a stereomicroscope.

361

362 **Discussion**

363 We estimated a surprisingly complex performance landscape for the unusual trophic niche of
364 lepidophagy from high-speed videos of gelatin-removing bites. In contrast to studies of suction-
365 feeding performance on evasive, attached, and strain-sensitive (e.g. zooplankton) prey (Holzman
366 et al., 2008; Holzman et al., 2022; Keren et al., 2018; Olsson et al., 2020), we found no effect of
367 kinematic timing variables such as TTPG or even ram speed on the performance of scale-biting
368 strikes, measured by the length, width, depth, and total volume of gelatin removed. Instead,
369 successful scale-biting appears to require strike coordination between jaw opening (peak gape)
370 and jaw protrusion and, surprisingly, this interaction resulted in two distinct performance optima:
371 1) individuals with small peak gapes removed the greatest amount of material per bite at small
372 jaw protrusion distances; 2) individuals with large peak gapes removed the greatest volumes at
373 large jaw protrusion distances; and 3) these two performance optima were surrounded by
374 reduced bite performance in all directions including more extreme values and intermediate
375 values of peak gape and protrusion. This resulted in two distinct performance peaks on the two-
376 dimensional thin-plate spline for jaw protrusion and peak gape in the model best supported by
377 the data (Fig. 4, Table 1). Thus, the strikes with the largest peak gapes and jaw protrusion
378 distances observed suffered a performance decline, in line with a few datapoints from our initial
379 kinematic study of biting in this system (St. John et al., 2020); similarly, the strikes with the

380 smallest peak gapes and jaw protrusion distances also suffered a performance decline. Therefore,
381 we unexpectedly found evidence of a multi-peak performance landscape for the relatively
382 straightforward functional task of biting, a well-studied functional task which is often viewed as
383 a simple mechanical system (Herrel et al., 2008; Wainwright and Richard, 1995; Westneat,
384 2005).

385

386 *Two distinct performance optima for biting rather than a linear ridge*

387 Surprisingly, there was no simple linear ridge for the interaction between peak gape and peak
388 jaw protrusion in relation to bite volume or bite length; both extreme and intermediate values of
389 these kinematic variables resulted in poor performance, i.e. reduced gelatin bite sizes (Fig. 4).
390 Only strikes by scale-eating specialists and some hybrid strikes resided on the second
391 performance peak with larger gapes and jaw protrusion while all generalists and molluscivores
392 occurred on or near the first performance peak. This suggests that a performance valley isolates
393 the recently evolved scale-eating specialist *C. desquamator* from its generalist ancestor.
394 Explanations for this performance landscape must also account for the poor performance of
395 intermediate strike values observed, rather than just a simple performance ridge indicating that
396 coordination between peak gape and peak jaw protrusion is important.

397 One possible explanation is a biomechanical tradeoff in precision and targeted bite area
398 with the most adverse effects on gel-biting performance at intermediate values. Interactions
399 between oral jaw scraping and biting with the gelatin surface may only be effective within two
400 different kinematic regimes. Smaller peak gapes with less jaw protrusion may allow for precise
401 targeting and higher mechanical advantage for removing more gelatin. Larger peak gapes with
402 greater jaw protrusion may reduce precision and mechanical advantage of the bite, but cover a

403 large area, resulting in more gelatin removed per bite. Intermediate values may suffer the costs of
404 less precise biting and less area covered per bite. It is tempting to speculate that strike speed or
405 lower jaw angle play a role in this precision/target area tradeoff. However, while ram speed and
406 lower jaw angle with the suspensorium both had strong linear effects on bite performance, there
407 was no evidence of any nonlinear interactions with peak gape or peak jaw protrusion distance
408 (Table 1). Similarly, timing (TTPG) seems to play no role in bite performance, which would
409 seem surprising if precision is important for gelatin removal since faster time to peak gape
410 should reduce bite precision. However, biting strikes generally achieved peak gape substantially
411 before contact with the gel, suggesting that the time to reach peak gape does not seem to be
412 important as long as the jaw is open at some point before contact with the target. This is the
413 consistent with the ‘plateau effect’ observed in the scale-eating piranha, the only other direct
414 study of scale-biting kinematics in other systems (Janovetz, 2005).

415 Alternatively, it is striking that none of the generalist or molluscivore species exhibited
416 feeding strikes with jaw protrusion distances within range of the second performance peak (Fig.
417 4). Similarly, only hybrids with scale-eater ancestry were capable of producing feeding strikes
418 with jaw protrusion distances in this range (> 6.5 mm; see supplemental raw data). Thus, there
419 appears to be a genetic basis underlying the two performance peaks: only scale-eaters and
420 hybrids with scale-eater ancestry can protrude their jaws sufficiently to reach the second
421 performance optimum. This may be due to additional anatomical properties of their oral jaws
422 that allow for greater extension during strikes, such as different ratios of muscle fiber types
423 within the adductor mandibulae (Ono and Kaufman, 1983; Summers and Long, 2005), along
424 with unmeasured aspects of their behavior or strike kinematics. Indeed, scale-eaters exhibit
425 significant differences in their boldness and exploratory behaviors (St. John et al., 2020).

426 Genome-wide association scans for oral jaw length also identified collagen genes with fixed
427 regulatory differences between scale-eaters and molluscivores, including collagen type XV alpha
428 1 (col15a1), suggesting that the elasticity of jaw opening may be under selection in this species
429 (Martin et al., 2019; McGirr and Martin, 2017). Greater peak gapes are possible due to the two-
430 fold larger oral jaws of the scale-eater. However, scale-eaters still do not open their jaws as wide
431 as possible during strikes (St. John et al., 2020) or achieve 180° angles with their open jaws as in
432 other scale-eating specialists such as the scale-eating piranha (Janovetz, 2005), indicating
433 adaptive behavioral compensation for their extreme oral anatomy during strikes.

434 Finally, we cannot rule out more esoteric explanations for the unexpected fitness valley
435 between bite performance optima. Sensory perception during strikes may be limited at
436 intermediate strike distances due to the blind spot caused by the positioning of the vertebrate
437 optic nerve in front of the retina, although biomechanical implications of this in fishes are
438 unknown (Gregory and Cavanagh, 2011). Alternatively, intermediate jaw protrusion may be an
439 indirect effect of premature suspension of strike behavior or lack of motivation during the strike.
440 However, excluding missed strikes did not alter the observed two-peak performance landscape
441 (Fig. 4).

442

443 *Similarity between the performance and fitness landscapes*

444 Both field measurements of fitness and laboratory measurements of scale-biting performance
445 support a two-peak landscape. Repeated field experiments in this system measured the fitness
446 landscape from the growth and survival of advanced generation hybrids placed within 3-4 m
447 enclosures in their natural hypersaline lake environments for 3 – 11 months and estimated two
448 fitness peaks separated by a fitness valley (Martin and Wainwright, 2013b). Surprisingly, these

449 landscapes remained relatively stable and exhibited a similar two-peak topography across years,
450 lakes, and frequency-manipulations of hybrids (Martin, 2016a; Martin and Gould, 2020).
451 However, in all cases where it was detected, the second peak corresponded to the phenotype of
452 molluscivores; whereas hybrids resembling the scale-eater survived and grew at the lowest rates
453 across all fitness experiments. Thus, a single fitness peak corresponds to generalist morphology
454 and kinematics for both fitness and performance, whereas the second scale-biting performance
455 peak has no analog in field enclosures, perhaps due to a lack of hybrids with the necessary
456 combination of scale-biting kinematics and morphology. However, these laboratory estimates do
457 suggest that F2 hybrids with scale-eater ancestry display sufficient jaw protrusion distances
458 during gel-biting strikes to occupy the second performance peak.

459 Ultimately, the goal of connecting morphological fitness landscapes to performance
460 landscapes is difficult due to the many-to-one mapping of hybrid morphologies onto biting
461 kinematics. However, better understanding of the genetic basis of kinematic variables, such as
462 jaw protrusion distance, may enable connecting the genetic regulatory networks underlying
463 morphological, kinematic, and behavioral traits through genotypic fitness networks informed by
464 laboratory performance and field fitness experiments. Our initial study of genotypic fitness
465 networks found rare but accessible pathways in genotype space (i.e. equal or increasing in fitness
466 at each step) connecting scale-eaters to other species, but so far we have found no associations
467 between fitness and any behavioral genes (Patton et al., 2022). Also see the role of *sox9b* in
468 cichlid foraging performance, in which correcting for genotype improves the form-function
469 relationship (Matthews et al., 2023).

470

471 *Conclusion*

472 Here we explore the biomechanics of a highly specialized trophic niche and demonstrate the
473 power of machine-learning approaches to analyze kinematic data. We estimated a surprisingly
474 complex two-peak performance landscape for biting that indicates that the highly protrusible
475 jaws of scale-eating specialists may provide a performance benefit for scale-eating. This study
476 provides a new framework for understanding bite mechanics in fishes – particularly scraping
477 dynamics – and a foundation for dissecting the genetic basis of these predatory behaviors and
478 their relationship to fitness landscapes driving rapid adaptive radiation in the wild.

479

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489

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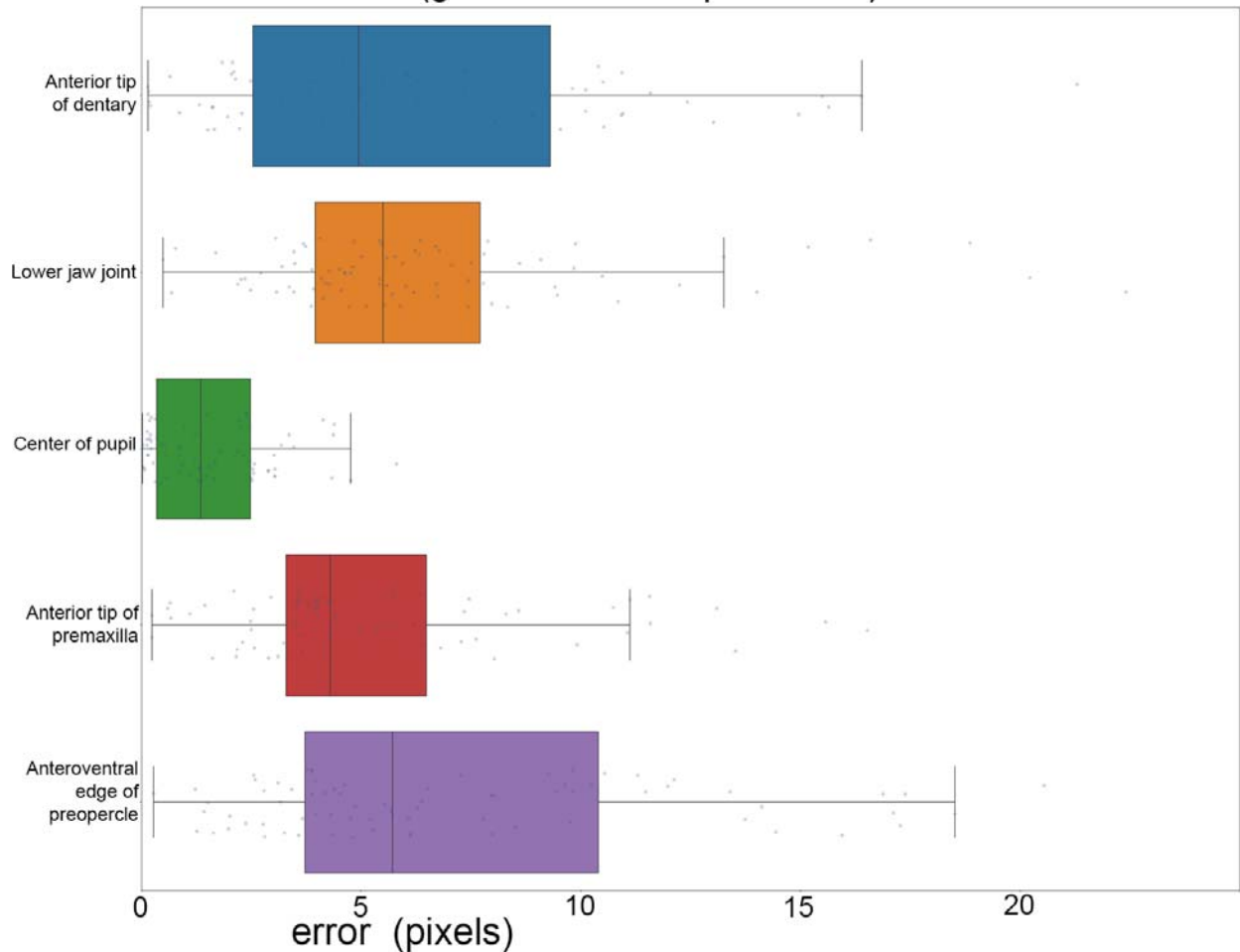
705

706 **Fig. S1** Top-down and side views of typical single bites from Repashy gelatin cubes. Both bites

707 pictured were classified as edge bites, the most common type of bite recorded.

708

Node distances (ground truth vs. prediction)



709

710 **Fig. S2 Error distribution for each landmark in our best performing model.** Trained model
711 was based on 815 labeled frames across 100 high-speed videos. We used the multi-animal
712 bottom-up unet model with a receptive field of 156 pixels, max stride of 32 pixels, batch size of
713 3, input scaling of 0.75, and validation fraction of 0.1. The mean distance between labeled data
714 and inferred landmark positions was 5.80 pixels. Precision obtained was 0.990 based on four
715 false positives, five false negatives, and 397 true positives in the model summary output from
716 SLEAP.