

TBD

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Abstract

Introduction

3 The adaptive landscape provides a powerful framework for understanding and predicting the evolution of biological diversity —from genes, to phenotypes, to species (Wright and Simpson). More than a metaphor, the adaptive landscape links quantitative genetic and phenotypic variation to evolution by natural selection (Lande 1979, Arnold papers). In particular, the shape of the adaptive landscape represents the presence and strength of different selection pressures imposed by the environment (Arnold’s work). Species interactions often play a key role in defining the selective environment, as evidenced by the influence of resource competition (Schluter 2000), mutualisms (Jordano 1987), and predation (Abrams 2000) on the evolution of biological diversity. Although there is clear evidence that pairwise interactions shape the adaptive landscape, we also know that most species interact with multiple species in an ecological community. Understanding how the adaptive landscape is shaped by community context represents a major challenge for both empiricists and theoreticians given the diversity of selection pressures that a population is likely to experience in a natural ecological community.

Species-interaction networks, such as a food web describing who eats whom, provide an explicit representation of the community context. These networks describe the interdependency of species within an ecological community, providing an effective framework for predicting how changes in the density of directly and indirectly connected species will impact other species, and in turn, community dynamics. Despite the importance of species-interaction networks for predicting ecological dynamics, these networks often lack quantitative information on how species interactions influence natural selection. Consequently, it is unclear how species-interaction networks shape the adaptive landscape, creating a major hurdle in our current ability to predict how changes in community context will affect phenotypic evolution.

Here, we integrate species-interaction networks and the adaptive landscape to predict how changes in network structure will affect phenotypic evolution. We did this by conducting a field experiment to characterize the effect of network structure on the adaptive landscape of a con-

stituent species. Specifically, we conducted a field experiment that manipulated the diversity of insect parasitoids that were able to impose selection on an abundant insect herbivore (*Iteomyia salicisverruca*) (Fig. 1). The larva of this herbivore species induce tooth-shaped galls when they feed on the developing leaves of willow trees (*Salix* sp., @Russo2006). These galls provide protection from generalist predators (e.g. ants, spiders), thus the network of interacting parasitoids provides a realistic representation of the biotic environment this insect herbivore is experiencing. Therefore, our manipulation of parasitoid diversity alters the diversity of interactions, or food-web complexity, that this insect herbivore experiences.

Changes in food-web complexity could influence a resource population's adaptive landscape in at least two ways. First, if a more diverse community of consumers is more effective at suppressing resource densities (@Ives2005), then this will result in lower mean fitness of the resource population. A reduction in mean fitness, all else equal, will intensify natural selection (@Hunter2018). On the other hand, if consumers impose different selection pressures on resource traits, then more diverse communities could dampen the strength of selection acting on a given trait. This is because a greater diversity in selection pressures is equivalent to greater uncertainty in the selective environment. Thus, a more diverse consumer community may relax the net selection pressures acting on resource traits. Here, we evaluate these hypothesized relationships through an experimental test of how changes in food-web complexity alters the adaptive landscape of a resource population.

Methods

Study Site

We conducted our study within a four-year old common garden of coastal willow (*Salix hookeriana*) located at Humboldt Bay National Wildlife Refuge (HBNWR) (40°40'53"N, 124°12'4"W) near Loleta, California, USA. This common garden consists of 26 different willow genotypes that were collected from a single population of willows growing around Humboldt Bay. Stem cuttings of each genotype (25 replicates per genotypes) were planted in a completely randomized design in two hectares of

a former cattle pasture at HBNWR. Willows in our garden begin flowering in February and reach
54 their peak growth in early August. During this study, willows had reached 5 - 9m in height. Further
details on the genotyping and planting of the common garden are available in @Barbour2015.

Food-web Manipulation

57 We setup our food-web manipulation across 128 plants soon after galls began developing on *S.
hookeriana* in early June of 2013. These 128 plants came from eight different plant genotypes,
spanning the range of trait variation observed in this willow population (@Barbour2015). On
60 treatment plants (8 replicates per genotype), we enclosed 14 galled leaves with 10x15cm organza
bags (ULINE, Pleasant Prairie, WI, USA) to exclude three parasitoid species that attack during
larva development (hereafter larval parasitoids). This treatment did not exclude the egg parasitoid
63 *Platygaster* sp. which attacks prior to gall initiation (note that in Cecidomyiid midges, larva
initiate gall development CITE). On control plants (8 replicates per genotype), we used flagging
tape to mark 14 galled leaves per plant (30 larva), allowing the full suite of parasitoids to attack
66 *Iteomyia*. Marking galls with flagging tape ensured that we compared control and treatment galls
with similar phenology when we collected galls later in the season. Our food-web manipulation
altered the average number of trophic interactions that *Iteomyia* was exposed to from BLANK on
69 control plants to BLANK on treatment plants. Thus, we refer to galls on control plants as being
exposed to a 'complex' food web, whereas galls on treatment plants were exposed to a 'simple'
food web. In late August, we collected marked and bagged galls from each plant, placed them
72 into 30 mL vials and kept them in the lab for 4 months at room temperature. We then opened
galls under a dissecting scope and determined whether larva survived to pupation (our measure
of fitness) or were parasitized. Since we were interested in selection imposed by interactions with
75 parasitoids, we restricted our data to larva that either survived to pupation, was parasitized by
an egg parasitoid (*Platygaster* sp.), or was parasitized by a larval parasitoid. For the food-web
treatment that excluded parasitoids, we further restricted our data by removing any instances of
78 parasitism by a larval parasitoid. This represented less than 3% of the observations in this food-

web treatment and allowed us to focus our inferences of selection on those imposed by the egg parasitoid. Together, we had survival estimates for 1,306 larva from 607 galls, 111 plants, and 8
81 plant genotypes.

Measuring Gall Traits

We collected data on three different traits that we anticipated would experience selection based
84 on our previous work (@Barbour2016) and others work with Cecidomyiid midges (@Weis1983,
@Heath2018) (Heath et al. 2018). First, we measured gall diameter as the size of each gall chamber
to the nearest 0.01 mm at its maximum diameter (perpendicular to the direction of plant tissue
87 growth). Our previous work has shown that a larger gall diameter provides a refuge for larva from
parasitoid attack (@Barbour2016). <!-- Consider whether it is important to make the distinction
that diameter was measured at the multi-chambered gall level in previous work --> Second, we
90 measured the clutch size of adult female midges by counting the number of chambers in each gall
(@Weis1983). All larva collected from the same multi-chambered gall were scored with the same
clutch size. Third, we measured female preference for oviposition (egg-laying) sites as the density of
93 larva observed on a plant in an independent survey. Specifically, we randomly sampled five branches
per tree and summed the number of individual gall chambers observed. We then converted these
counts to a measure of gall density per 100 shoots by counting the number of shoots on the last
96 branch we sampled. All larva collected from the same plant were scored with the same female
preference. The measurement of larval densities on plants in the field is a commonly used index for
measuring oviposition preference (@Gripenberg2010); however, caution must be taken in inferring
99 'preference' as larval densities can be influenced by processes other than preference (@Singer1986).
Fortunately, a couple of features of our study system suggest that larval density on a plant may be a
good proxy for female preference. For example, since our data comes from a randomized placement
102 of willow genotypes in a common garden, there is no consistent bias in which willow genotypes that
females are exposed to while searching for oviposition sites. Also, egg predation is a minor source
of mortality for galling insects in general (@Hawkins1997), thus we do not expect any prior egg

predation to bias our estimates of observed larval densities.

Inferring the adaptive landscape from individual selection surfaces assumes that trait distributions are multivariate normal (Lande and Arnold 1983). To approximate this assumption, we log-transformed clutch size and added a small constant (1) to female preference before log transforming, since our surveys occasionally estimated zero larval densities. We then scaled all phenotypic traits to mean=0 and SD=1 in order to calculate standardized selection gradients that were comparable across traits and with other studies of natural selection.

Rather than imposing selection, parasitoids may themselves influence the expression of herbivore traits which would bias our measurements of selection. To estimate this bias, we subset our data to only include galls where there was variation in larval survival within the same gall (i.e. $1 > \text{mean survival} > 0$). Using the statistical methods discussed in the next section, we calculated the "apparent" relationship between chamber diameter and larva survival (α_{Diam}). We detected a positive bias ($\alpha_{\text{Diam}} = 0.35$), indicating that unadjusted analyses would overestimate the strength of selection on chamber diameter. To account for this bias, we subtracted 0.35 from our estimate of α_{Diam} with the full dataset prior to calculating directional selection on chamber diameter. Note that this analysis is based on the assumption that larva within each gall come from the same clutch and therefore should have similar chamber diameters regardless of whether they are parasitized. Our data supports this assumption, as we found that gall ID explains 54% of the variance in chamber diameter.

Quantifying the Adaptive Landscape

Our analyses consisted of three parts. First, we used generalized linear mixed models (GLMM) to quantify selection surfaces —linear and nonlinear relationships between absolute fitness (W) and phenotypic traits (z_i) of individuals —in each food-web treatment. Second, we translated our approximation of the selection surface into the scale of relative fitness (w) in order to calculate selection gradients. Third, we used our estimates of selection gradients to characterize the slope and curvature of the population's adaptive landscape.

Selection surface: Since larva survival was our measure of absolute fitness we used a GLMM

that assumed a binomial error distribution (and logit-link function). To approximate the selection surface, we modelled larva survival as a function of food-web treatment as well as linear (α_{z_i}), quadratic ($\alpha_{z_i:z_i}$), and linear interactions ($\alpha_{z_i:z_j}$) between each trait. Other approaches have been advocated for approximating the selection surface (Schluter 1988); however, our approach enables us to calculate directional (β_{z_i}), quadratic (γ_{z_i,z_i}), and correlational selection (γ_{z_i,z_j}) gradients, and thus is more appropriate for our approximating the adaptive landscape (cite Arnold). We also allowed these α coefficients to vary between food-web treatments. To account for the nonindependence of individual estimates of clutch size (gall level) and female preference (plant level) as well as any independent effects of willow genotype on larva survival, we specified gall ID nested within plant ID nested within plant genotype as random effects. Including these random effects helps alleviated biased estimates of selection due to environmental covariances between traits and fitness (Rauscher 1992 Evolution). With our full statistical model, we used parametric bootstrapping (1,000 replicates) to test whether our food-web treatment altered each α coefficient. If the 95% confidence interval of this food-web: α effect overlapped with zero, then we removed this statistical interaction and refit the model. Note that our final model of the selection surface still included estimates (and 95% CI) of our food-web treatment and all possible α coefficients, even if they overlapped with zero, but only a subset of possible food-web: α coefficients. Note also that we estimated linear coefficients (α_{z_i}) by excluding quadratic and linear interaction terms in the statistical model (Lande and Arnold 1983).

Selection gradients: Since we were interested in characterizing the adaptive landscape – the relationship between mean trait values and population mean fitness – we assumed the mean value of our random effects (i.e. setting them to zero) when calculating selection gradients. We then used the method of Janzen and Stern (1998) to translate our estimates (and 95% CI) of each α coefficient into the scale of relative fitness in order to calculate directional (β_{z_i}), quadratic (γ_{z_i,z_i}), and correlational (γ_{z_i,z_j}) selection gradients.

Adaptive landscape: We took advantage of existing theory that translates estimates of se-

lection gradients to the population's adaptive landscape (Phillips and Arnold 1998, Arnold 2003).

159 Specifically, the slope of the adaptive landscape corresponds to the column vector of directional selection gradients:

$$\text{Slope} = \beta = \begin{pmatrix} \beta_{\text{Diam}} \\ \beta_{\text{Clutch}} \\ \beta_{\text{Pref}} \end{pmatrix}$$

whereas the curvature (C) of the adaptive landscape is a matrix that is determined by directional, quadratic, and correlational selection gradients:

$$\text{Curvature} = C = \gamma - \beta\beta^T$$

where γ represents the matrix of quadratic (diagonal) and correlational (off-diagonal) selection gradients:

$$\gamma = \begin{pmatrix} \gamma_{\text{Diam:Diam}} & & \\ \gamma_{\text{Clutch:Diam}} & \gamma_{\text{Clutch:Clutch}} & \\ \gamma_{\text{Pref:Diam}} & \gamma_{\text{Pref:Clutch}} & \gamma_{\text{Pref:Pref}} \end{pmatrix}$$

165 Note that we omitted the upper triangle of the matrix for clarity since it is simply the reflection of the lower triangle. Assuming that there is additive genetic variance and covariance between these traits, then the slope and curvature of the adaptive landscape give insight to changes in the population's mean trait value between generations ($\Delta z = G\beta$) and G-matrix structure within a generation ($\Delta G = G(\gamma - \beta\beta^T)G$).

Although quantitative predictions about trait and G-matrix evolution require knowledge of a population's G-matrix, the adaptive landscape can still yield valuable qualitative insight. In particular, we can infer selective constraints acting on a population by inspecting the structure of the matrix describing the curvature of the adaptive landscape. For example, the sign of the values along the diagonal of the curvature matrix dictate whether selection will increase (+), decrease (-),

or cause no change (0) in the additive genetic variance of a trait. Similarly, any nonzero covariance terms (off-diagonal) are indicative of trait integration. We propose that you can infer selective constraints acting on a population by counting the number of negative signs along the diagonal (decrease in additive genetic variance) and the number of nonzero terms along the off-diagonal of the matrix (trait integration).

*Measuring selection on *Platygaster* sp.*

Our experimental manipulation of food-web complexity enables us to also measure selection on the egg parasitoid (*Platygaster* sp.), albeit with a different approach. Specifically, we can compare the distribution of gall traits that the egg parasitoid is observed in simple vs. complex food webs. In essence, the gall phenotype becomes the extended phenotype of the egg parasitoid, since its attributes can influence the egg parasitoid's survival in the face of intraguild predation (larval parasitoid guild). This cross-sectional analysis allows us to estimate directional and quadratic selection differentials, but not selection gradients.

To quantify directional selection differentials, we used separate linear mixed models with the standardized gall trait as the response variable, food-web treatment as the fixed effect, and the same suite of random effects as for our primary analyses of *Iteomyia* survival (i.e. gall ID, nested within plant ID, nested within plant genotype).

To quantify quadratic selection differentials, we can calculate the difference in variance between simple and complex food webs. We then used Levene's test for homogeneity of variance to determine the statistical significance of these quadratic selection differentials. Note that this result should be interpreted cautiously, since we cannot account for the nonindependence of gall phenotypes due to plant genotype, plant ID, or gall ID.

All analyses and visualizations were conducted in R (@R2018). Note that for visualizing the adaptive landscape we restrict trait axes to ± 1 SD of the mean trait value. This emphasizes the fact that we can only reliably estimate the shape of the adaptive landscape near the mean trait values of the population (Arnold 2001). We also plot mean larva survival on a natural log scale to

201 accurately reflect the shape of the adaptive landscape.

Results

Slope of the adaptive landscape

204 We found that food-web complexity altered the slope of the adaptive landscape (β) in two key ways. First, there were fewer traits under directional selection in the complex (2 of 3) vs. simple food web (3 of 3). For example, in both simple and complex food webs there was directional selection for
207 larger chamber diameters and for adult females to avoid ovipositing on trees with high densities of conspecifics (fig. 2; fig. 3). In the simple food web, we also observed directional selection for smaller clutch sizes, but there was no evidence of selection acting on this trait in the complex food web
210 (fig. 2; fig. 3). The absence of selection on clutch size in the complex food web appeared to be a result of conflicting selection pressures imposed by each guild of parasitoids. Specifically, when we subset our data to focus on differences between parasitoid guilds, we found that larval parasitoids
213 actually impose directional selection for larger clutch sizes (fig. 2).

The second key effect of food-web complexity was to steepen the slope of the adaptive landscape (i.e. increase strength of selection). For example, the magnitude of directional selection in complex
216 vs. simple food webs was ~60% higher for both chamber diameter and female preference (fig. 2; fig. 3). This result was not due to any difference in the relationship between larva survival and gall phenotype (Simple:coef_{Diam} = 0.02 [-0.56, 0.60]; Simple:coef_{Pref} = -0.16 [-0.83, 0.61]), but was
219 due to the fact that the average probability of larval survival was lower in complex (0.44 [0.30, 0.58]) vs. simple food webs (0.69 [0.54, 0.80]).

Curvature of the adaptive landscape

222 Unlike the slope, the curvature of the adaptive landscape ($\gamma - \beta\beta^T$) is influenced by directional (β_{z_i}), quadratic (γ_{z_i, z_i}), and correlational (γ_{z_i, z_j}) selection gradients. Our food-web treatment did not appear to alter correlational selection for any combination of traits and there was no strong

225 evidence that any of these selection gradients differed from zero (fig. 2). Similarly, our food-web
treatment did not influence quadratic selection on chamber diameter or clutch size and there was no
strong evidence that these selection gradients differed from zero (fig. 2). In contrast, our food-web
228 treatment did alter quadratic selection acting on female preference. Specifically, we found evidence
for disruptive selection in the complex food web, but no evidence of disruptive or stabilizing selection
in the simple food web (fig. 2; fig. 3). Disruptive selection in the complex food web appeared to be
231 due to the presence of larval parasitoids enhancing the relative fitness of larva occurring on either
low or high density trees (fig. 2).

We can quantify the curvature (C) of the adaptive landscape by subtracting the matrix of direc-
234 tional selection gradients ($\beta\beta^T$) from the matrix of quadratic and correlational selection gradients
(γ):

$$C = \begin{pmatrix} C_{Diam,Diam} & & \\ C_{Clutch,Diam} & C_{Clutch,Clutch} & \\ C_{Pref,Diam} & C_{Pref,Clutch} & C_{Pref,Pref} \end{pmatrix}$$

We omit the upper triangle of the matrix for clarity, because it is a mirror image of the lower
237 triangle. Each component of this matrix describes how the adaptive landscape changes additive
genetic (co)variances in phenotypic traits within a generation. To empirically estimate the curva-
ture, we only retained nonzero selection gradients (i.e. 95% CI does not overlap with zero). We
240 found that the curvature of the adaptive landscape in the complex and simple food web exhibited
the following structures:

$$C_{Complex} = \begin{pmatrix} -0.13 & & \\ 0 & 0 & \\ 0.09 & 0 & 0.33 \end{pmatrix}$$

$$C_{Simple} = \begin{pmatrix} -0.05 & & \\ 0.02 & -0.01 & \\ 0.03 & -0.01 & -0.02 \end{pmatrix}$$

Comparing these matrices, we find evidence of selective constraints acting on 2 of 6 trait
 243 (co)variances in the complex food web, whereas 6 of 6 trait (co)variances experience selective
 constraints in the simple food web. Specifically, we found that simple food webs acted to decrease
 genetic variance for all three phenotypic traits (negative diagonal terms), whereas only one trait
 246 (chamber diameter) experienced a decrease in additive genetic variance in the complex food web.
 For genetic covariances, we found that the simple food web favored integration among all three
 phenotypic traits (nonzero off-diagonal terms), and thus constraints along all three axes of covari-
 249 ance (fig. 4). In contrast, the complex food web only constrained genetic covariance by favoring
 integration between chamber diameter and female preference (fig. 4).

Selection on the extended phenotype of egg parasitoid

252 We did not observed any evidence of directional or quadratic selection on chamber diameter
 ($s_{Diam} = 0.06[-0.14, 0.27]$; $s_{Diam:Diam} = -0.11, P = 0.28$) or clutch size ($s_{Clutch} = 0.21[-0.10, 0.51]$; $s_{Clutch:Clutch}$
 $-0.24, P = 0.17$) that the egg parasitoid was associated with. However, we did observed directional
 255 selection for egg parasitoids to attack larva on high density trees ($s_{Pref} = 0.20[-0.10, 0.51]$) and an associated decrease
 $-0.91, P < 0.001$).

Discussion

258 Two key patterns emerged from our study. First, selection was stronger and thus the potential
 speed of evolution in complex compared to simple food webs. Second, the adaptive landscape was
 less constrained in the complex vs. the simple food web.

Conclusion

Acknowledgments

Literature Cited

- ²⁶⁴ Heath, J. J., P. Abbot, and J. O. Stireman, 3rd. 2018. Adaptive divergence in a defense symbiosis driven from the top down. *Am. Nat.* 192:E21–E36.

Figure legends

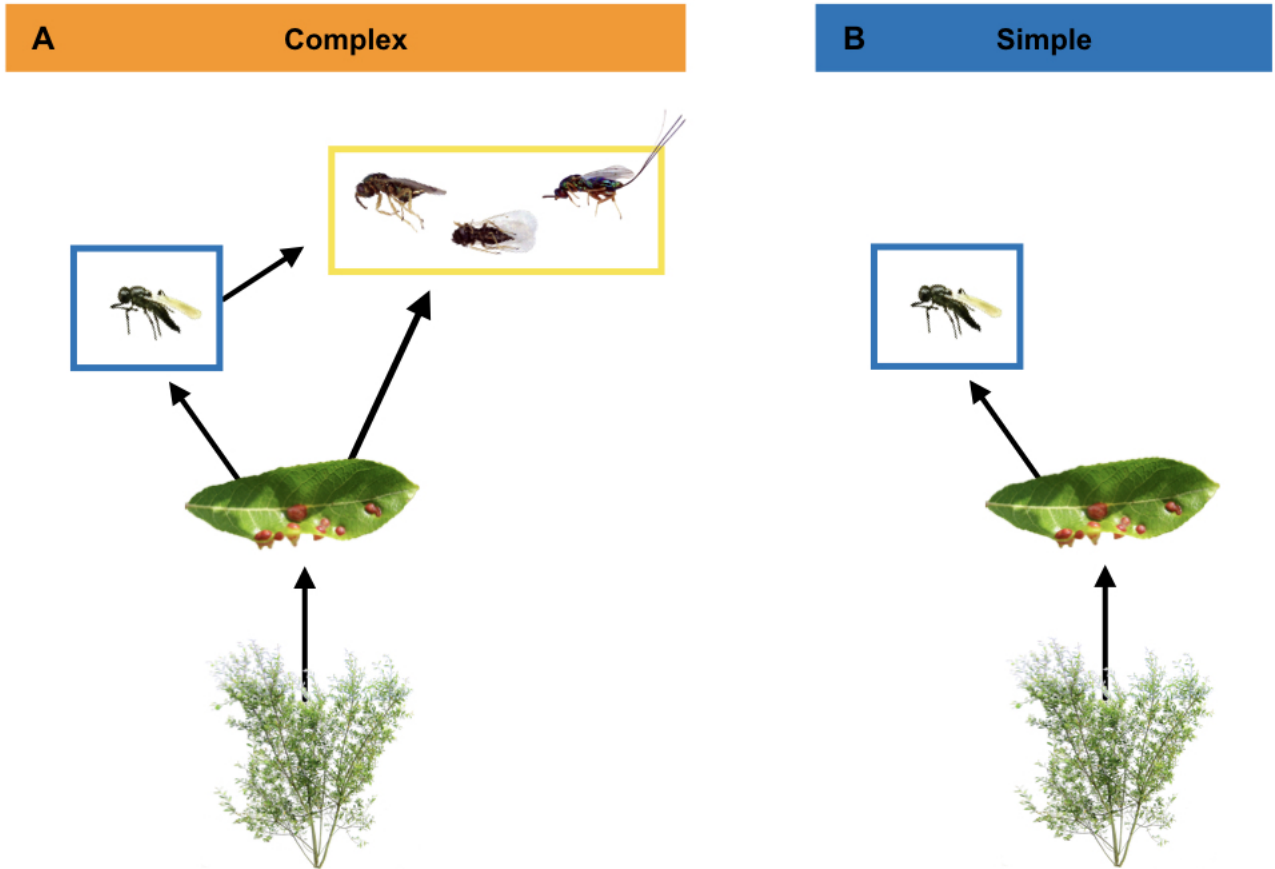


Figure 1: Illustrations of complex (A) and simple (B) food webs associated with the insect herbivore, *Iteomyia salicisverruca*. Black arrows denote the flow of energy in this network of trophic interactions.

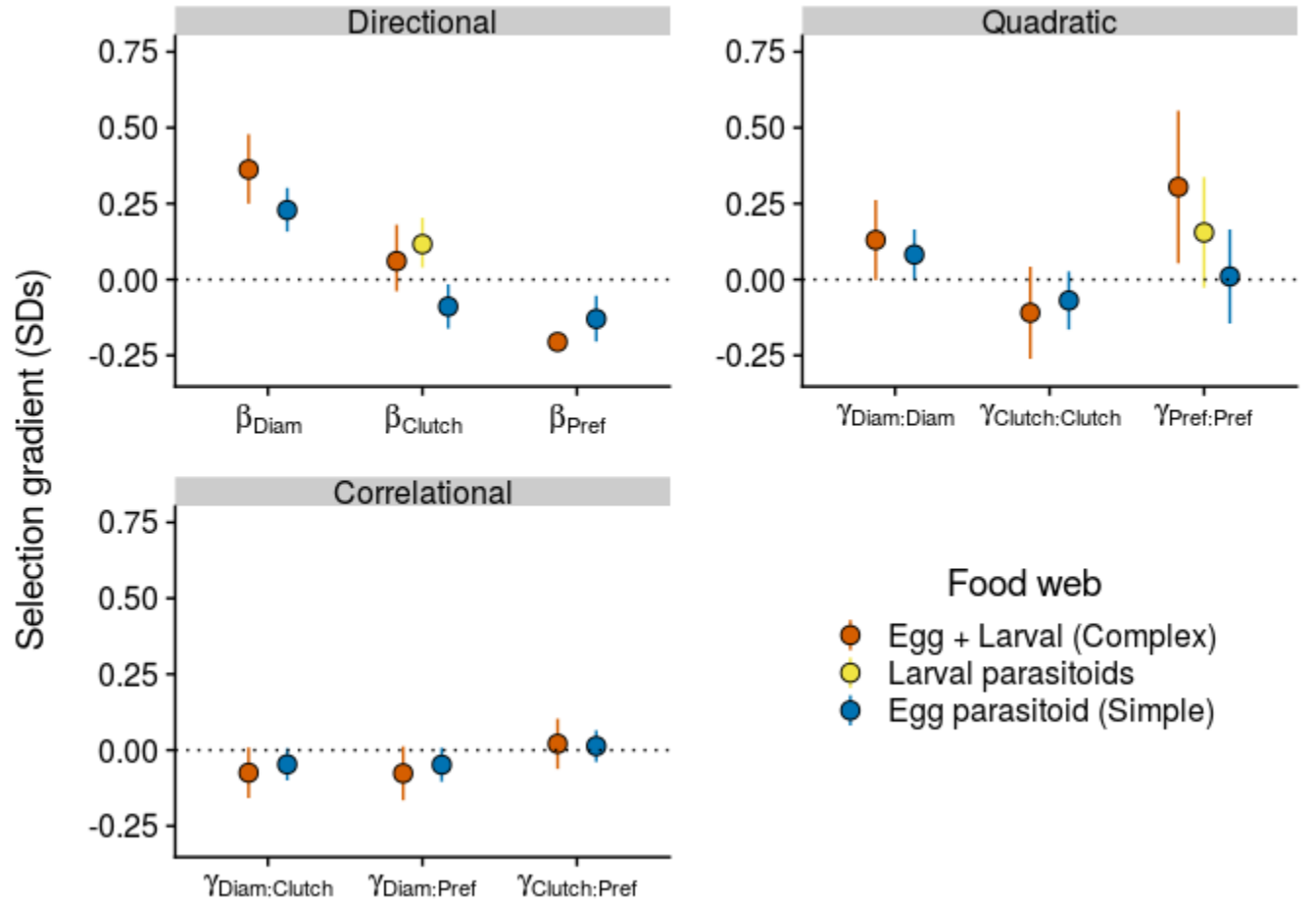


Figure 2: Selection gradients acting on *Iteomyia* phenotypes. Points and lines correspond to estimated means and 95% confidence intervals, respectively. Orange and blue colors correspond to estimated gradients in complex and simple food webs, respectively. Yellow colors correspond to selection gradients imposed by larval parasitoids. Note that we only plot selection gradients for larval parasitoids when they differed from egg parasitoids (simple food web).

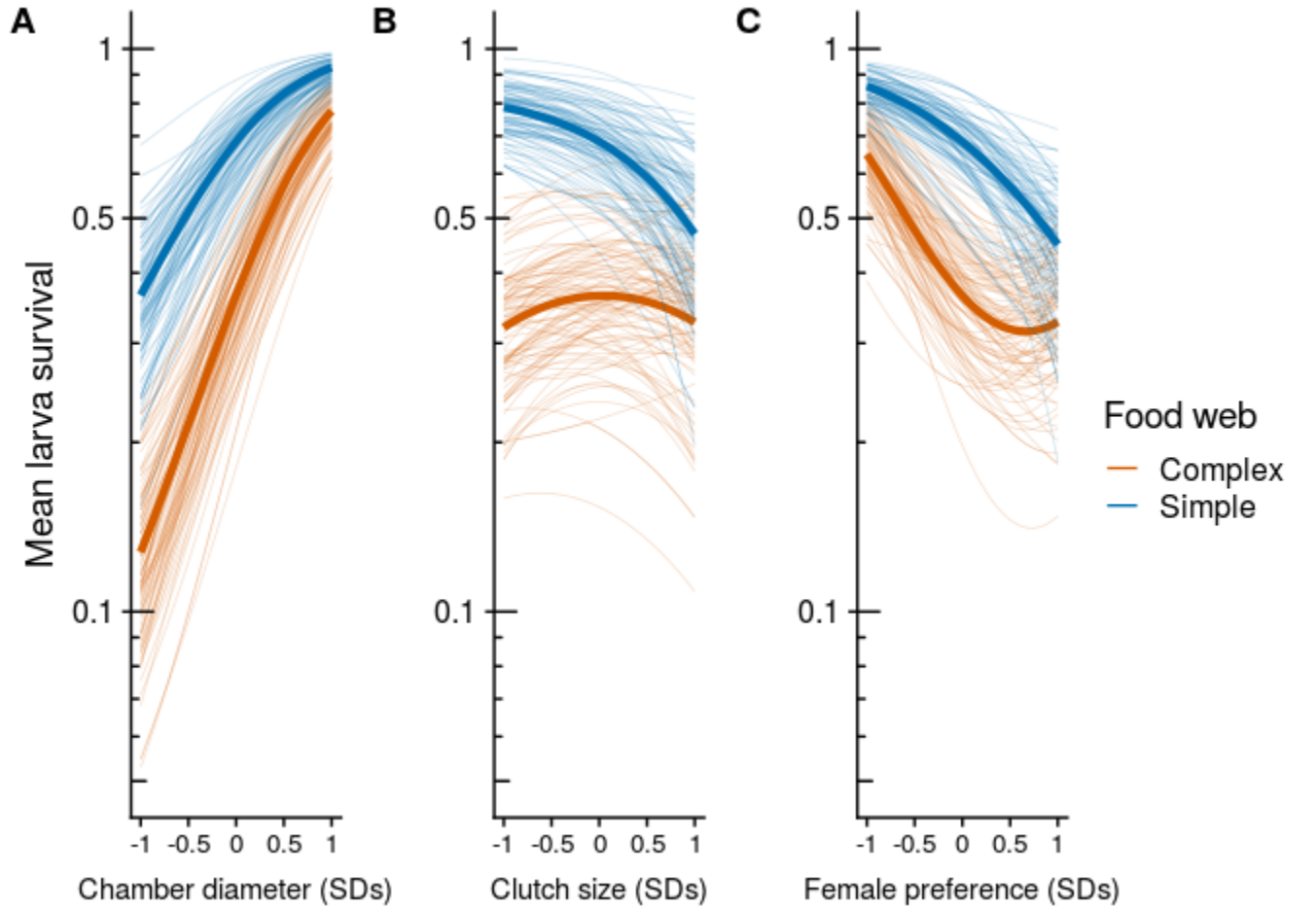


Figure 3: Adaptive landscape of *Iteomyia* phenotypes in complex vs. simple food webs. Each panel corresponds to a different phenotypic trait: chamber diameter (A); clutch size (B); and female preference (C). Solid lines represent the estimated gradients in complex (orange) and simple (blue) food webs. Transparent lines represent bootstrapped replicates to show the uncertainty in estimated gradients. For clarity, we only display 100 bootstraps even though inferences are based on 1,000 bootstrapped samples. Note the mean larva survival is plotted on a natural log scale to accurately reflect the shape of the adaptive landscape.

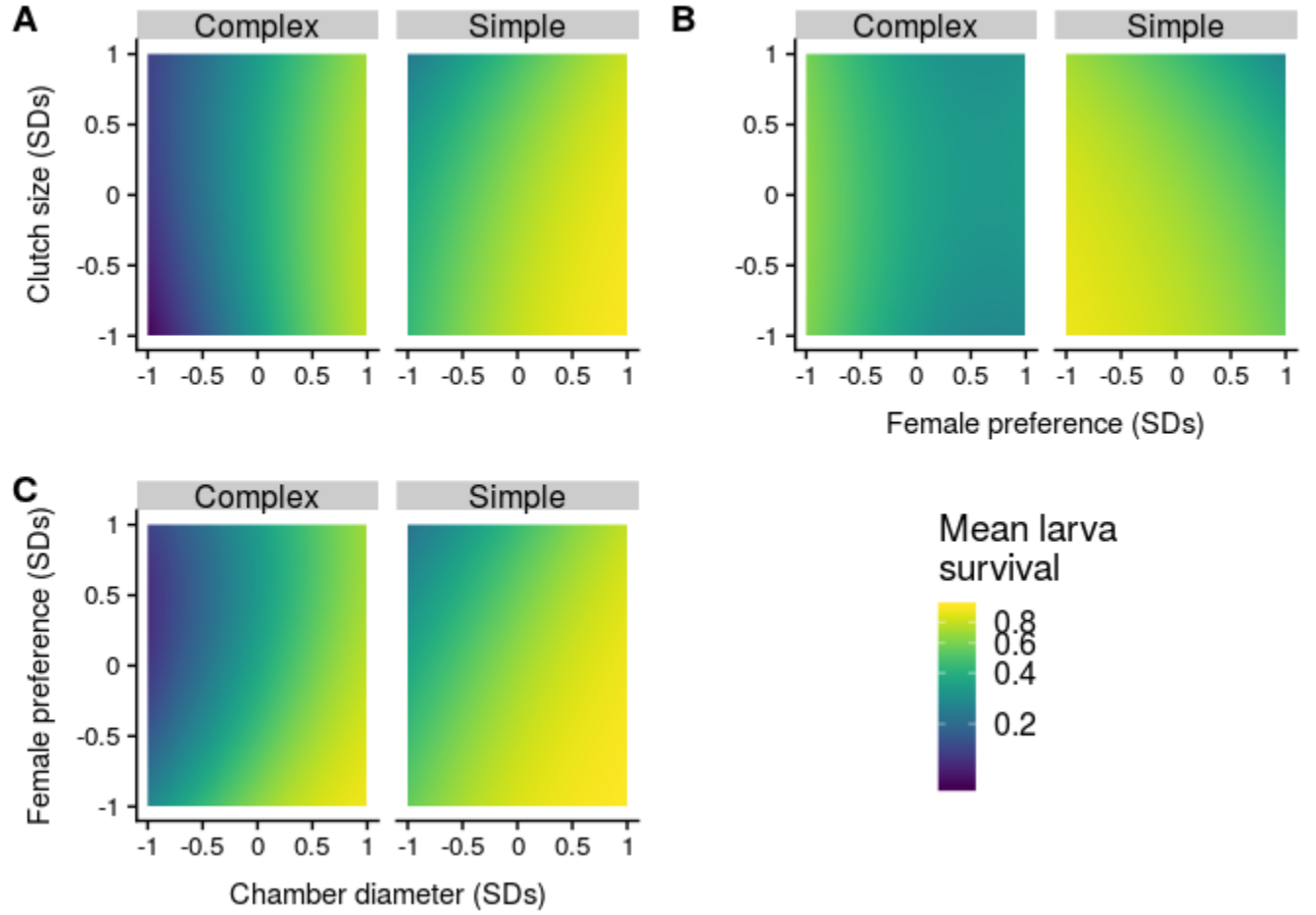


Figure 4: Two dimensional view of adaptive landscapes in complex vs. simple food webs. Each panel corresponds to a different combination of traits: clutch size and chamber diameter (A); clutch size and female preference (B); female preference and chamber diameter (C). Note the mean larva survival is plotted on a natural log scale to accurately reflect the shape of the adaptive landscape.