

Mitonuclear divergence predicts gradual speciation in animal hybrid zones

Derek Eddo+, Bailey Rouse+, Silu Wang*
Department of Biological Sciences
State University of New York, Buffalo

+ Equal contributions

*Corresponding author: siluwang.biodiv@gmail.com

Abstract

The core of speciation is the genetic incompatibilities underlying the evolution of reproductive isolation. Hybrid zones provide unique opportunities to unravel the evolutionary rate of reproductive isolation in the origin of species. The selection against hybrids accrues with increased genetic incompatibilities and drives the evolution of reproductive isolation in the face of gene flow. There have been decades of debates over the relationship between the selection against hybrids and the genetic divergence between parental lineages. The debates occur primarily among three models: (1) the Additive Effect Model predicts a linear growth of selection against hybrids with the divergence of parental lineages; (2) the Snowball Effect Model predicts exponential growth of selection; whereas (3) the Slowdown Effect Model predicts a logarithmic growth of selection. Here, we tested the three models with animal hybrid zones worldwide. The Slowdown Effect Model (3) is best supported with the full dataset. We refined the three models to consider independent and interactive effects of mitochondrial (mtDNA) and nuclear genetic (nDNA) divergence on the selection against hybrids. The refined Snowball Model (2) was best supported by the data, revealing a significant effect of nuclear genetic distance and its interaction with mtDNA distance on the selection against hybrid DNA in the hybrid zones. Collectively, this data synthesis in the early stage of animal speciation reveals a gradual development of reproductive isolation with mitonuclear genetic divergence.

Keywords: Speciation rate, Hybrid zone, Snowball Effect, Genetic Incompatibility, Speciation, Hybridization, Cline.

45 **Introduction**

46 The evolution of reproductive isolation is underpinned by genetic incompatibilities, which
47 accrue as the diverging lineages accumulate lineage-specific substitutions (Dobzhansky 1934;
48 Orr 1995). The increased genetic incompatibilities contribute to the selection against hybrids
49 (Palopoli and Wu 1994; Orr 1995; Welch 2004), which attenuate gene flow between diverging
50 lineages and lead to reproductive isolation and speciation (Gavrilets 2004; Nosil 2008;
51 Westram et al. 2022). The rate at which lineage-specific substitutions drive the selection
52 against hybrids ultimately prescribes the speciation rate (Coyne and Orr 1989). The debates
53 over this relationship have inspired decades of investigation of genomic architecture and
54 mechanisms of speciation (Feder et al. 2014; Seehausen et al. 2014). If divergent substitutions
55 independently contribute to reproductive isolation, the genetic distance between the diverging
56 lineages should predict linear selection against gene flow. If two-loci epistatic incompatibilities,
57 or Dobzhansky & Muller Incompatibilities (DMIs)(Dobzhansky 1934; Muller 1942), are
58 prevalent, the selection against hybrids would increase exponentially as genetic distance
59 increases, reflecting a ‘Snowball Effect’ (Orr 1995). A greater acceleration in speciation rate
60 would be observed if the genetic distance results in abundant higher-order incompatibilities
61 (Orr 1995). In contrast, an early ‘slowdown’ might be observed if premating isolation is
62 established prior to substantial genetic incompatibilities (Gourbiere and Mallet 2010).

63 Several taxon-specific empirical studies have supported the Snowball Effect hypothesis,
64 where the increase of genetic distance between parental lineages exponentially suppresses
65 hybrid fitness (Presgraves 2002; Matute et al. 2010; Moyle and Nakazato 2010). However,
66 others reported a more gradual, linear accumulation of selection against hybrids in response to

67 genetic divergence (Lijtmaer et al. 2003; Gourbiere and Mallet 2010; Stelkens et al. 2010). A
68 logarithmic relationship was also observed, reflecting a Slowdown Effect (Gourbiere and Mallet
69 2010). Collectively, the taxon-specific patterns concur with the heterogeneous speciation rates
70 across the tree of life that await large-scale comparative studies to identify evolutionary
71 principles underlying the regularities about speciation rates.

72 The heterogeneous patterns of speciation rate are likely attributed to the variability of
73 locus-specific effect sizes and the probability of incompatibility in different parts of the genome.
74 For instance, the smaller effective population size in the mitochondrial genome than the
75 nuclear genome predicted greater drift, counteracting selection (Slatkin and Maruyama 1975).
76 In addition, mitochondrial deleterious mutations reportedly exhibit a greater average effect size
77 than nuclear mutations (Popadin et al. 2013). Moreover, the probability of incompatibility can
78 differ between intergenomic and intragenomic discordance.

79 More broadly, the heterogeneous patterns of selection against hybrids could also be
80 explained by the stages of speciation along the speciation continuum (Roux et al. 2016). In the
81 early stages of divergence, large-effect single-locus genetic barriers are more effective than
82 DMIs to overcome extensive gene flow (Felsenstein 1981; Wu 2001; Wang et al. 2020;
83 de Zwaan et al. 2022). If the additive effect sizes of single-locus barriers were constant through
84 time, the selection against hybrids should linearly increase as lineage-specific substitutions
85 accumulate over time (Fig. 1, Additive Effect). If the effect sizes were larger earlier (Fisher
86 1930) in speciation, when the lineage-specific substitutions were scarce and exhibiting
87 diminishing returns in response to further increments of genetic divergence, a Slowdown Effect
88 in the form of a logarithmic function is expected (Gourbiere and Mallet 2010) (Fig. 1, Slowdown

Effect). As the gene flow between diverging lineages attenuates, allowing DMIs to build up, a Snowball Effect can be observed (Orr 1995) (Fig. 1, Snowball Effect), where the selection against hybrids grows exponentially as lineage-specific substitutions accrue. The heterogeneous tension between genetic incompatibility and gene flow across the speciation continuum is naturally encapsulated in decades of hybrid zone studies around the world (Fig. 2).

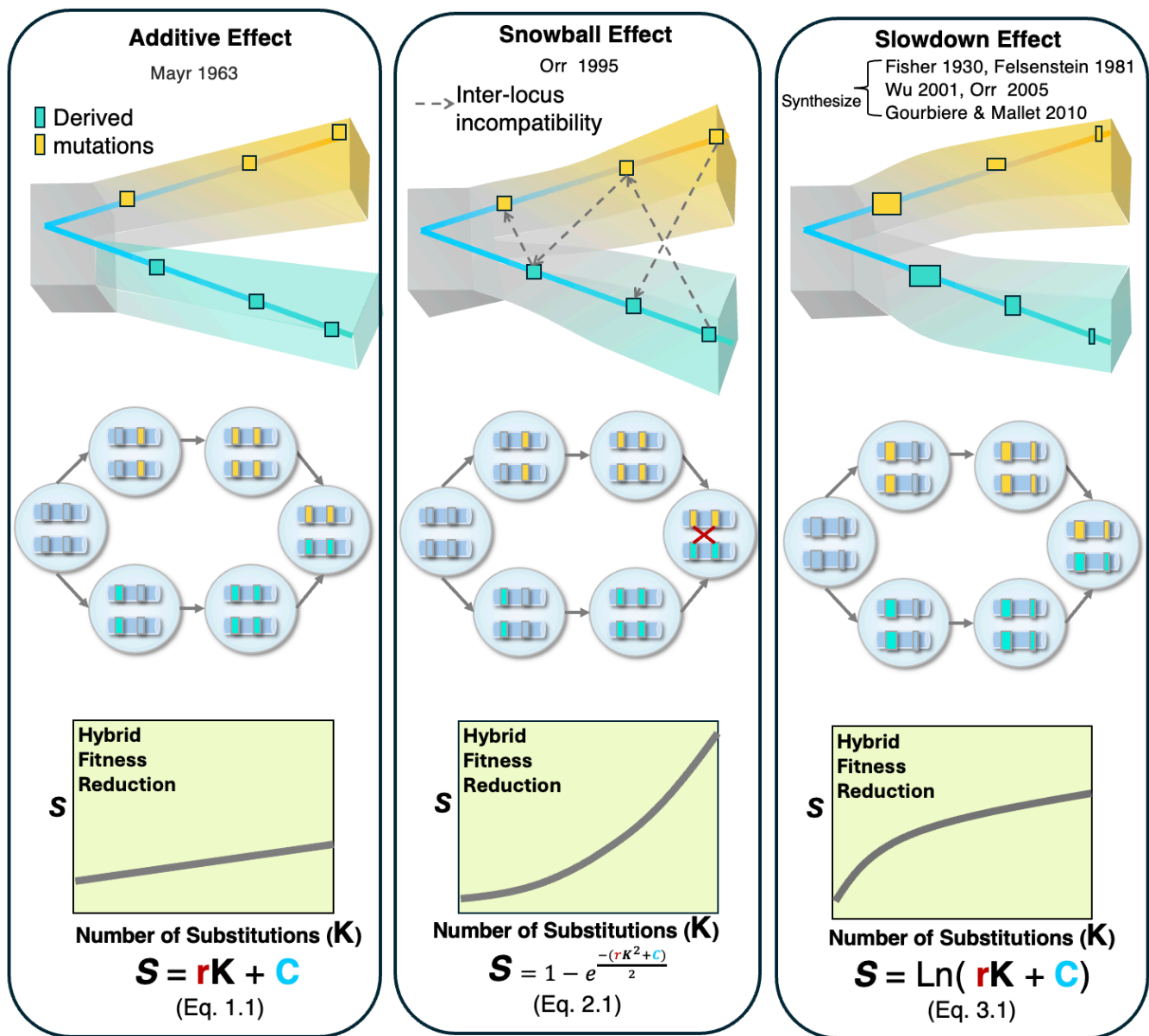


Fig. 1 Three speciation models: (1) Additive Effect, (2) Slowdown Effect, and (3) Snowball Effect.

Hybrid zones are geographic areas where diverging lineages hybridize (Barton 1979).

They provide unique opportunities to observe the progression of speciation as the tension

between genetic incompatibility and gene flow is resolved (Barton and Hewitt 1985). This

tension is effectively theorized as the tension zone model, where the width of the cline (w) is a

balance of selection (s) and dispersal (σ) (Fig. 2A) (Barton 1979). The selection (s) is a

composite expression of genetic incompatibility and divergent selection, or reproductive

isolation in the face of gene flow. The rate at which s increases with the extension of genetic

distance between diverging lineages ultimately translates to the speciation rate (Fig.2B).

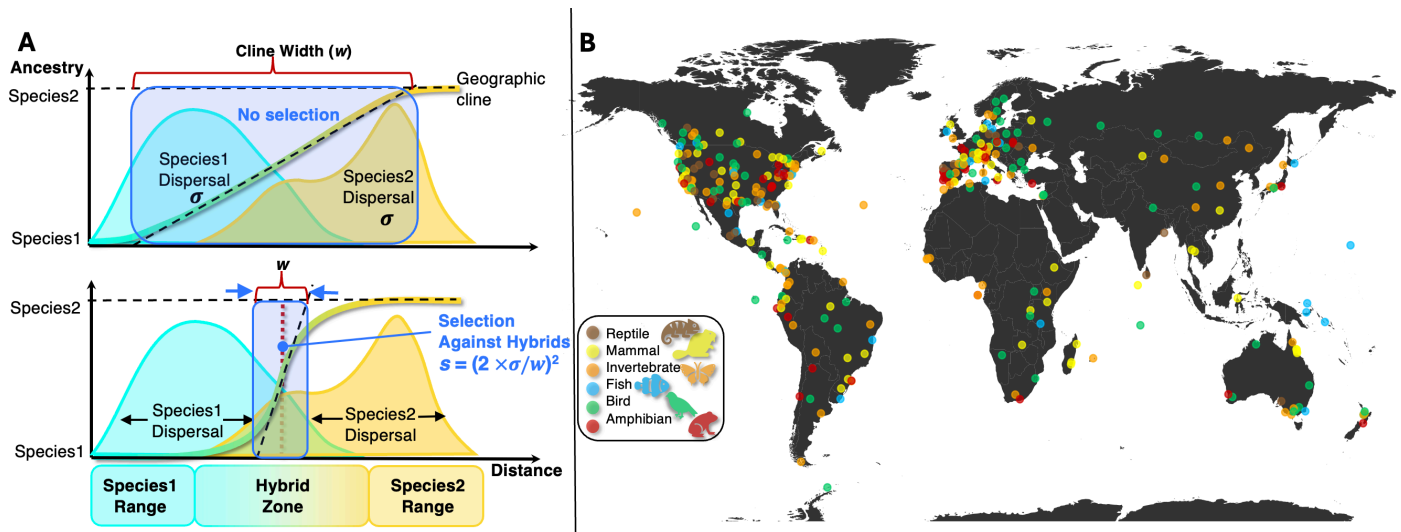


Fig.2 Hybrid zones are natural laboratories for understanding speciation. **A**, The cline width (w) of a hybrid zone is a balance of gene flow (σ) and selection (s) against hybrids. In the absence of s , σ will widen w over time as species 1 hybridizes with species 2 (top). Bottom: strong selection against hybrids counteracts with σ and shrinks w . **B**, World distribution of animal hybrid zones. Over 330 animal hybrid zones have been reported. Color categories correspond to six different organismal groups: mammals in yellow (65), birds in green (71), reptiles in brown (23), amphibians in red (45), invertebrates in orange (100), and fish in blue (26).

Here, we conduct a data synthesis to examine the buildup rate of selection against

hybrids (s) in animal hybrid zones (Fig. 2). We tested three alternative models that predict the

selection against hybrids with the genetic distance between parental lineages. Model 1

predicts a linear increase of s as genetic distance increases (Fig. 1, Additive Effect); Model 2

118 predicts an exponential growth of s (Fig. 1, Snowball Effect); while Model 3 predicts a
119 logarithmic relationship (Fig. 1, Slowdown Effect).

120 We further developed three refined models to dissect mitochondrial and nuclear genetic
121 effects on speciation. Since the mitochondrial and nuclear genomes have distinct substitution
122 rates (Ballard and Whitlock 2003) and coevolve to maintain critical eukaryotic energetics (Eddo
123 et al. 2025), ‘Endosymbiotic Snowball Effect’ was articulated (Wang et al. 2025). Therefore, we
124 further partitioned the genetic distance into mtDNA and nDNA distance to understand their
125 independent and interactive effects on speciation. Together, this study will inform the mode of
126 speciation in the origin of animal biodiversity.

127 **Method**
128 *Hybrid zones*

130 We studied 330 animal hybrid zones reported to date (McEntee et al. 2020; Nieto
131 Feliner et al. 2023) and categorized them into six taxonomic groups: amphibians, birds, fishes,
132 invertebrates, reptiles, and mammals (Fig. 2B). We filtered the data with the following criteria:
133 (1) there has been sufficient spatial sampling to fit the genetic and/or phenotypic clines across
134 the hybrid zone; (2) there have been direct estimations of cline widths with genetic markers; (3)
135 there is sufficient information on dispersal distance in the unit of Kilometer/generation or
136 Kilometer/generation^{0.5}; (4) for the dispersal distance reported in Kilometer/generation, there is
137 information on generation time to calculate the Kilometer/generation^{0.5}. These criteria resulted
138 in 148 animal hybrid zones for model testing (see below).

139 *Mitonuclear genetic distance*

140 To infer the genetic distance between hybridizing species, we calculated the genetic
141 distance of the most sequenced mitochondrial and nuclear genes in animals. We searched for

COI, ND2, ND3, ND4, CYTB, and MSH1 for mitochondrial genes and RAG1, RAG2, MC1R, BDNF, ODC1, ASIP, and ACTB for nuclear genes. We fetched DNA sequences of the target genes for each parental species from GenBank and only kept the genes if both species had at least three sequences per species. To ensure even sampling depths among the species pairs in our meta-analysis, for species with more than 10 sequence records of a gene, we randomly sampled 10 sequences among all the records. We then aligned all the sequences of each gene in each species pair with the msa function in the msa package (Bodenhofer et al. 2015) in R (R Core Team 2023). Specifically, we used MUSCLE (Edgar 2004) with gapOpening = 20 and gapExtension = 10. With the alignment, we calculated the mean pairwise distance between parental species with dist.dna function with the TN93 substitution rate model in the ape package (Paradis and Schliep 2019). Then, we averaged the between-species distance of all the mitochondrial genes and nuclear genes separately to result in the mean pairwise mtDNA distance ($\overline{K_m}$) and the mean pairwise nDNA distance ($\overline{K_n}$) between each pair of parental species. Finally, we calculated the mean pairwise genetic distance (\overline{K}) between each pair of parental species by averaging the nDNA and mtDNA distances. To account for the uncertainty in \overline{K} estimations, we calculated the standard deviation of $\overline{K_m}$, $\overline{K_n}$, and \overline{K} . Among the 148 hybrid zones, 77 hybrid zones had sufficient genetic information to calculate the genetic distance between parental species. To evaluate the reliability of our genic estimation of parental species divergence, we computed the ratio of between-species distance over the sum of between- and within-species distance and conducted correlation tests of F_{ST} estimates for several pairs of species reported in the literature (see supplementary data).

164 *Selection against hybrids*

165 We extracted the dispersal, center, and width of hybrid zones from McEntee et al.
166 (2020), Nieto Feliner et al. (2023), and additional literature (supplementary data). We unified
167 the unit of dispersal rate as kilometers (KM) $/\sqrt{generation}$ with the average generation time
168 information of each species pair. To account for sex-specific dispersal, we conducted another
169 round of literature search for female vs male dispersal rate in each species pair
170 (supplementary information). We calculated the average cline width (w) and center for mtDNA
171 and nDNA data, as well as phenotypic data separately. We calculated selection (s) against
172 heterozygotes as a function of cline width (w) and dispersal (σ): $s = (2 \times \frac{\sigma}{w})^2$ (Fig. 1) (Barton
173 and Gale 1993). To infer the cumulative selection against hybrids, we substituted w with
174 averaged cline width (\bar{w}) among mtDNA and nDNA clines (Fig. 3B). To account for uncertainty
175 of selection estimates due to heterogeneous dispersal rates and cline width estimates, we
176 monitored sex-specific dispersal and the variance of cline width estimates for each hybrid
177 zone. For hybrid zones with known sex-specific dispersal rates, we calculated selection on
178 mtDNA and nDNA, respectively, as $s_{mt} = (2 \times \frac{\sigma(female)}{w})^2$ and $s_{nt} = (2 \times \frac{\sigma(average)}{w})^2$. We derived
179 the standard deviation of s , $SD(s)$, as $(8\sigma^2 / w^3) SD(w)$.

180

181 *Speciation models*

182 To investigate the relationship between genetic distance and strength of selection
183 against hybrids (s) in the hybrid zones, we test three conventional hybrid incompatibility
184 models: (1) Linear Effect (Eq. 1.1), (2) Snowball Effect (Eq. 2.1; Fig. S1), and (3) Slowdown
185 Effect (Eq. 3.1). In each model, selection (s) is a function of genetic distance (K) with a rate

parameter r and a constant parameter C as divergent plasticity or genetic assimilation (Pigliucci and Murren 2003; Ehrenreich and Pfennig 2016; Mueller et al. 2025). The rate parameter corresponds to the average effect size of each pair of incompatibilities.

$$s = rK + C \quad (\text{Eq. 1.1})$$

$$s = 1 - e^{-(rK^2 + C)} \quad (\text{Eq. 2.1})$$

$$s = \ln(rK + C) \quad (\text{Eq. 3.1})$$

To account for the effects of mtDNA and nuclear genetic distance between the parental species (Fig. 3), we further refined the conventional expressions (Eq. 1.1, 2.1, and 3.1) into mitonuclear models (Eq. 1.2, 2.2, and 3.2) to partition K into K_m (mtDNA distance) and K_n (nuclear DNA distance) that correspond to rate parameters r_m and r_n , respectively, besides a constant of C .

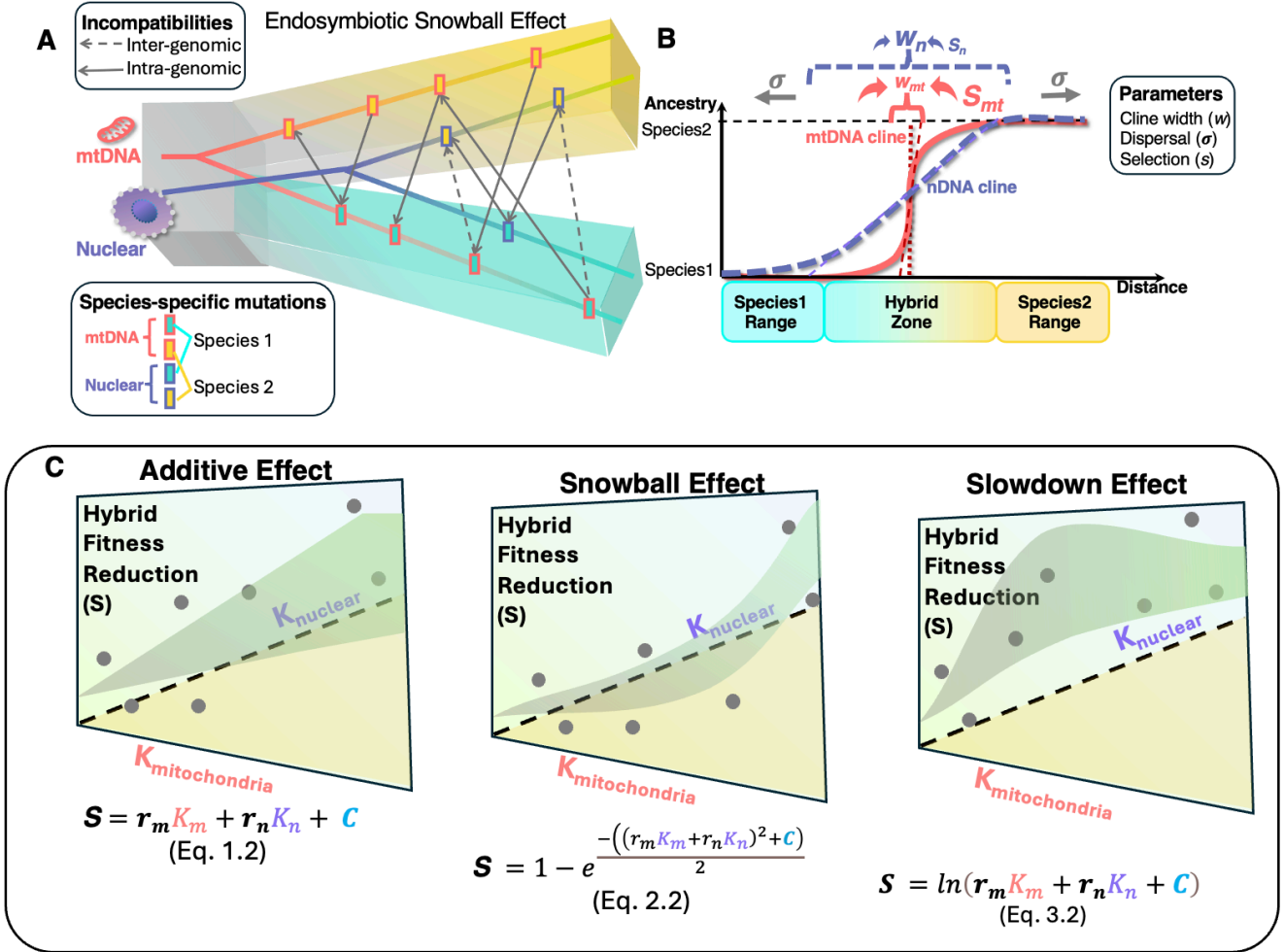
$$s = r_m K_m + r_n K_n + C \quad (\text{Eq. 1.2})$$

$$s = 1 - e^{-((r_m K_m + r_n K_n)^2 + C)} \quad (\text{Eq. 2.2})$$

$$s = \ln(r_m K_m + r_n K_n + C) \quad (\text{Eq. 3.2})$$

We conducted model fitting of the above three models with the nls2 function of nls2 package (Grothendieck, 2007) in R (R Core Team 2023). With the default algorithm and starting value of $r = 0.01$, $C = 0.1$, model fitting was conducted. The AIC of each fitted model is obtained with the AIC function. In each conventional model fitting (Eq. 1.1, 2.1, 3.1), we estimated the rate parameter r and constant C . To evaluate taxon-specific patterns, we conducted conventional model fitting (1.1, 2.1, 3.1) within each taxonomic group (Amphibians, Birds, Invertebrates, Mammals, and Reptiles), except for Fishes, due to data insufficiency. Furthermore, we estimated r_m , r_n , and C in the refined model fitting (1.2, 2.2, 3.2). Only 18

211 hybrid zones had sufficient mtDNA and nDNA sequence data to participate in refined model
 212 fitting, thus further taxon-specific modeling fitting was omitted.



213 **Fig. 3 Mitonuclear genetic divergence underlying the selection against hybrids.** **A.** Endosymbiotic Snowball
 214 Effect (Wang et al. 2025): Lineage-specific substitutions in the mitochondrial genome (red) and nuclear genome
 215 (purple) can lead to intragenomic and intergenomic Dobzhansky-Muller Incompatibilities, which emergently
 216 contribute to selection against hybrids. **B.** This selection counteracts gene flow (σ) across hybrid zones and
 217 reduces the cline width (w). Greater substitution rates in the mitochondrial genome than the nuclear genome can
 218 result in a greater rate of incompatibility and thus stronger selection on mtDNA clines than the nDNA clines. **C.**
 219 Refined speciation models with independent and interactive mitonuclear effects: (1) Additive (Eq. 1.2), Snowball
 220 (Eq. 2.2), Slowdown (Eq. 3.2).
 221
 222

223 To test the significance of observed regression coefficients, we conducted 1,000
 224 bootstrap iterations to generate a bootstrap distribution of regression coefficients \hat{r} and \hat{C} . In
 225 each bootstrap iteration, we randomly sampled $N = 148$ hybrid zones with replacement from

the full dataset. For each bootstrap sample of $K^* = [K_1^*, K_2^*, \dots, K_i^*, \dots, K_{148}^*]$ and $s^* = [s_1^*, s_2^*, \dots, s_i^*, \dots, s_{148}^*]$, we simultaneously fitted the sample to (1) Additive, (2) Snowball, and (3) Slowdown Effect Models (Fig.1; Eq. 1.1, 2.1, & 3.1) to calculate \hat{r} and \hat{C} . To account for the uncertainty in K and s , we randomly sampled K_i^* and s_i^* respectively from normal distributions: $N(K_i, sd(K_i))$ and $N(s_i, sd(s_i))$. To generate a null distribution of \hat{r} and \hat{C} , we randomly shuffled and mismatched K^* and s^* in each bootstrap sample and simultaneously fitted the three models (Eq. 1.1, 2.1, & 3.1) again. We used paired-T tests to compare the observed distribution of r and C , r_Obs and C_Obs , to the corresponding null distributions, r_H0 and C_H0 .

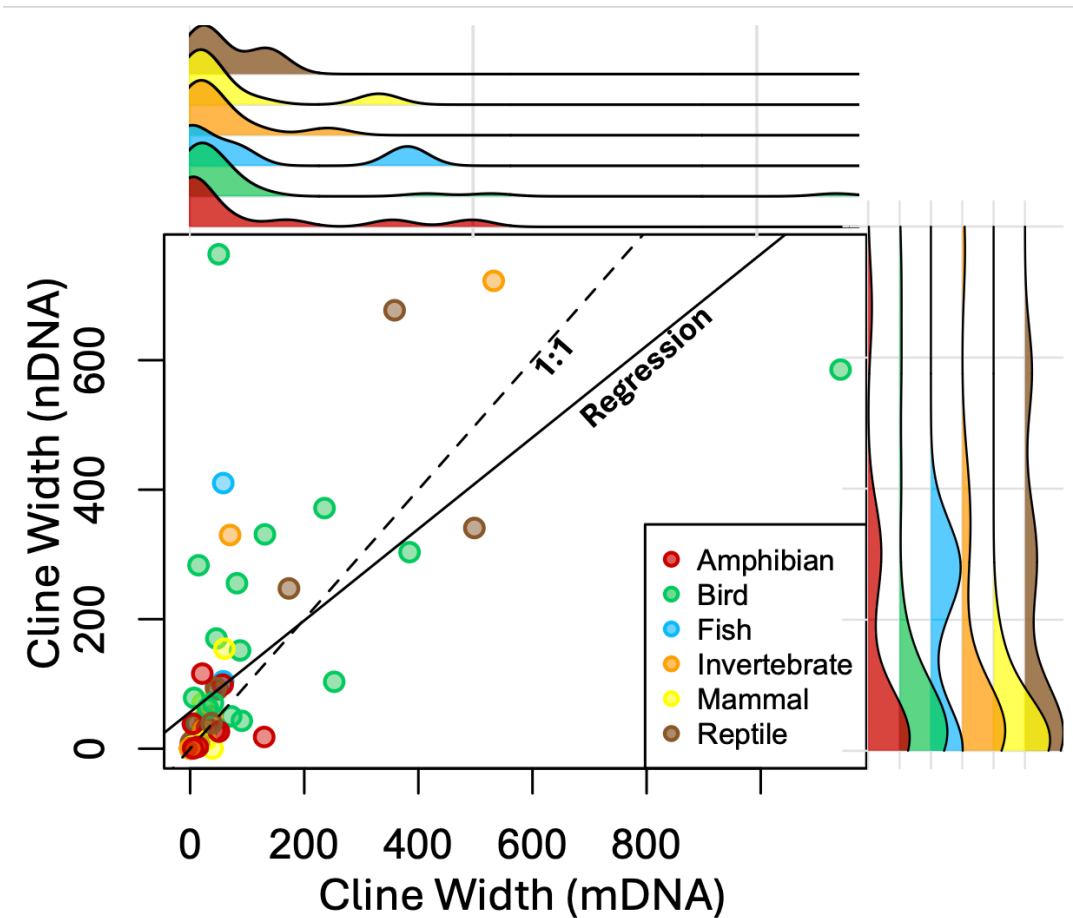
A similar bootstrapping test was applied to the refined models (Fig.3C; Eq. 1.2, 2.2, & 3.2), where (1) Additive, (2) Snowball, (3) Slowdown Models were compared in terms of AICs, and the observed bootstrap distribution of \hat{r}_m , \hat{r}_n , and \hat{C} , r_m_Obs , r_n_Obs and C_Obs , was respectively compared to r_m_H0 , r_n_H0 , and C_H0 .

Results

Discordant mitonuclear selection

There is significantly greater genetic distance in mtDNA than nDNA between parental species (mean difference = 0.036, 95% CI: 0.023-0.049, $p < 10^{-11}$). Concordantly, there is greater selection on mitochondrial clines than nuclear clines in hybrid zones (mean difference = 0.110, 95% CI: 0.024-0.196, $p < 0.05$), reflected by the narrower mtDNA cline width (Fig. 4). The discrepancy of selection on mitochondrial and nuclear genetic clines is significantly lower when there is greater genetic distance between parental species ($R^2 = 0.08$, $p = 0.04$). Our

248 mitonuclear divergence estimates were significantly correlated with F_{ST} estimates from
 249 heterogeneous sequencing datasets reported in the literature ($r = 0.50$, $p < 0.05$).



250
 251 **Fig. 4 Mitonuclear cline widths in animal hybrid zones.** The cline widths of mtDNA were
 252 significantly narrower than nDNA (one-tailed paired-T test, $p < 0.05$). The two variables are
 253 significantly correlated ($r = 0.67$, $p < 10^{-9}$).
 254

255 *Speciation models*

256 The Slowdown Effect model (3.1) (Fig. 1, 5A) has the lowest AIC (Table 1). The
 257 Slowdown model continues to fit superiorly for all the taxon-specific data except for birds and
 258 mammals, where the Snowball Effect and Linear model were respectively selected (Table 1).
 259 The regression coefficient of r of the Slowdown Effect model fitted to the full data was 3.078
 260 (Table 1). Similarly, the bootstrap mean of r_{Obs} in the Slowdown Effect model (Eq. 3.1) was

3.534 (Fig. 5B), which was significantly different from r_{H0} (mean difference: 3.41, 95%CI of the difference: 3.09 to 3.74; $p < 10^{-15}$), indicating a significantly positive effect of genetic divergence on selection against hybrids. However, the estimates of r were heterogeneous among taxonomic groups (Table S1), with birds ($r = 17.25$) and mammals ($r = 4.82$) being the highest, invertebrates ($r = 2.59$) and reptiles ($r = 1.34$) intermediate, and lower for amphibians ($r = 0.71$) (Table 1).

Table 1 The effect of genetic divergence on the selection against hybrids among animal taxa. Standard errors (SE) of the regression coefficients for the rate parameter(r) and constant (C). The models (Fig. 1, Eq 1.1-3.1) with the lowest AICs are in bold. The Slowdown model (equation 3.1) has the lowest AIC fitted to the full dataset.

Taxa	Models	r	SE (r)	C	SE (C)	AIC
All	Additive	2.54	1.16	0.03	0.04	15.03
	Snowball	5.25	1.55	0.07	0.04	16.93
	Slowdown	3.08	1.40	1.03	0.05	14.78
Amphibian	Additive	0.60	2.05	0.09	0.11	-5.04
	Snowball	1.04	11.44	0.12	0.09	-4.93
	Slowdown	0.71	2.30	1.09	0.12	-5.05
Bird	Additive	5.87	3.62	0.04	0.11	26.04
	Snowball	17.25	5.73	0.03	0.11	25.28
	Slowdown	8.34	5.40	1.02	0.13	25.80
Invertebrate	Additive	2.20	2.50	0.02	0.11	4.62
	Snowball	4.81	3.40	0.06	0.10	4.94
	Slowdown	2.59	2.91	1.02	0.12	4.58
Mammal	Additive	4.82	1.58	-0.03	0.04	-24.05
	Snowball	8.81	1.72	0.02	0.03	-23.89
	Slowdown	5.46	1.97	0.96	0.04	-23.89
Reptile	Additive	1.17	2.11	0.02	0.11	0.55
	Snowball	2.44	4.42	0.05	0.09	0.83
	Slowdown	1.34	2.31	1.01	0.11	0.53

When considering the effects of mitochondrial and nuclear genetic distance separately (Fig. 3), the Snowball Effect model (Eq. 2.2) has the lowest AIC (Table 2; Fig. 5 C-D). We observed a significantly positive effect of nuclear DNA distance, a small negative effect of

275 mDNA distance independently, but a positive interaction of mitonuclear distance (Fig. 5C) on
 276 the selection against hybrid DNA. Specifically, the r_n was estimated to be 31.18 (Table 2). In
 277 the Snowball Effect Model, bootstrap mean of r_n , 26.82, was significantly different from r_{n_H0}
 278 (mean difference: 13.27, 95%CI of the difference: 11.09 to 15.44; $p < 10^{-15}$) (Fig. 5D). In
 279 contrast, the regression coefficient of r_m was -1.85, similar to the bootstrap mean (-2.42, Fig.
 280 5D), which was significantly more negative than r_{m_H0} (mean difference: -1.43, 95%CI of the
 281 difference: -1.76 to -1.09; $p < 10^{-15}$).

282

283 **Table 2** Estimates and standard errors (SE) of the r_m , r_n , and C of the refined models. The
 284 Snowball Effect model has the lowest AIC.

Models	r_m	SE (r_m)	r_n	SE (r_n)	C	SE (C)	AIC
Additive	-0.65	1.16	7.57	7.49	0.07	0.06	-19.64
Snowball	-1.85	6.02	31.18	18.29	0.06	0.03	-20.76
Slowdown	-0.71	1.24	7.95	8.54	1.07	0.06	-19.58

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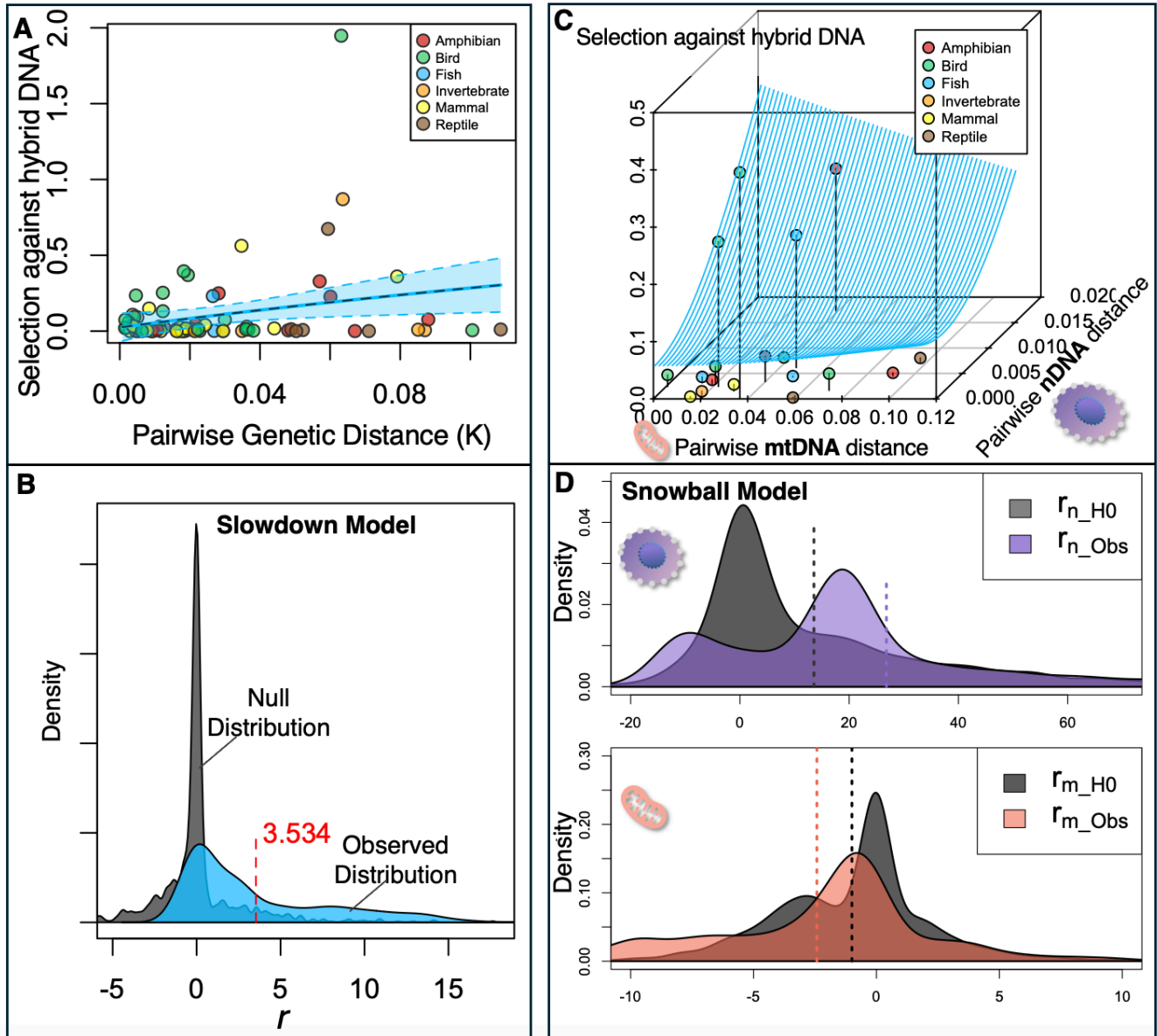


Fig. 5 Coefficients of estimates in the best-fitted models for simplified versus refined models of speciation. **A**, Conventional models (equations 1.1, 2.1, and 3.1) considered the total genetic distance as the predictor for selection against hybrids. The Slowdown Effect model (dotted line) has the lowest AIC. **B**, Bootstrap distributions of the rate parameter r (equation 1.3) estimations based on observed relationship between K and s across animal hybrid zones (blue) versus r estimates based on randomly mismatched K and s (grey). Estimates of r are from the best-fitted conventional model. **C**, Refined models considered the effect of mtDNA and nDNA divergence and their interactions separately. The Snowball Effect model has the lowest AIC. **D**, The bootstrap distribution of rate parameter estimations of r_m and r_n , which corresponds to the effects of mtDNA and nDNA distance separately. Estimates of r_m and r_n are from the best-fitted refined model.

296 Discussion

297 The accumulation of selection against hybrids in the hybrid zones between diverging
298 lineages is key to unraveling speciation rates *in situ*. Here, we leveraged decades of hybrid
299 zone studies to understand the evolution of reproductive isolation in the face of gene flow.
300 Overall, the selection against hybrids increased with genetic divergence between the parental
301 species, despite the heterogeneous rates of this relationship (Table 1, 2; Fig. 5).

302 The Slowdown model (equation 3.1) best fitted the full data among the three
303 conventional models (equation 1.1, 2.1, 3.1) (Fig. Table 1, 4 A-B). This result is consistent with
304 the Slowdown Effects of parental genetic distance previously observed in sympatric starfish
305 and fruit fly (Gourbiere and Mallet 2010). However, the delta AIC was small (Table 1), which
306 suggests that the superiority of the Slowdown model over the other components was weak.
307 The slowdown model predicts an opposite pattern from the Snowball Effect (2.1) expectation at
308 the early stage of speciation, with an early decelerating buildup of selection against hybrids as
309 the parental lineages diverge. One explanation for this early deceleration is the establishment
310 of premating isolation prior to extensive genetic divergence (Gourbiere and Mallet 2010). Early
311 premating isolation would contribute to the reduction in gene flow, leading to narrower cline
312 width over dispersal across the hybrid zones. The early premating isolation without widespread
313 genetic divergence between the parental lineages can be underpinned by a single or few
314 large-effect substitutions (Nosil and Schluter 2011; Servedio et al. 2011), genetic assimilation
315 (Waddington 1961), and/or cultural evolution (Lachlan and Servedio 2004). These
316 mechanisms might experience organismal constraints and thus are heterogeneous among
317 taxonomic groups. For example, vocal learning in songbirds might uniquely increase the

318 likelihood of cultural evolution as premating isolation (Slabbekoorn and Smith 2002; Lachlan
319 and Servedio 2004), though it can also disrupt reproductive isolation. Indeed, we observed the
320 heterogeneity of rate parameter estimations among taxonomic groups (Table 1).

321 In fact, the rate parameter was higher in endotherms (Table 1), birds, and mammals,
322 with avian hybrid zones fitting the Snowball model better. The relatively lower rates in
323 ectotherms (Table 1), amphibians, reptiles, and invertebrates could indicate a relatively smaller
324 genetic effect on reproductive isolation at the early stage of speciation in these groups. High
325 basal metabolic rates in endotherms could require stronger selection on genetic
326 incompatibility, especially between the mitochondrial and nuclear genomes (Hill 2019). As
327 endotherms exhibit faster climatic niche shifts than ectotherms (Rolland et al. 2018), the
328 underlying genetic divergence between endothermic sister populations could
329 disproportionately divergent adaptation. Both the proximate and ultimate differences in
330 endotherms vs ectotherms could lead to greater effect sizes of genetic divergence on the
331 selection against hybrids.

332 When we consider the mtDNA and nDNA divergence separately (Eq. 1.2, 2.2, 3.2),
333 Snowball Effect model (Eq. 2.2) fits the data best (Fig. 1). In this model, nuclear genetic
334 divergence and its interaction with mitochondrial divergence showed large positive effects,
335 while the independent mitochondrial divergence exhibited a mild negative effect (Fig. 5C,
336 Table 2). This result is concordant with the conventional model (Fig. 5A) in terms of the overall
337 positive effect of genetic divergence on selection against hybrids, but discordant in the modes.
338 Such discordance illuminates the heterogeneity of effects in different parts of the genomes.
339 Combining a rapid, exponentially growing nDNA effect with a mildly negative mtDNA effect can

340 give rise to a Slowdown cumulative genetic effect. The three speciation models (Fig. 1) are not
341 necessarily mutually exclusive and could be parallel, sequential, or hierarchical in the genome
342 evolution underlying speciation. Future combinatorial examination of the three reductionist
343 models in different parts of the nuclear genome could further unveil the evolutionary genomic
344 mechanism of speciation.

345 The interaction of mtDNA and nDNA distance showed a positive effect on selection
346 against hybrids (Fig. 5C). This pattern agrees with the growing evidence that mitonuclear
347 incompatibility is important for early-stage speciation (Hill 2017; Wang et al. 2021; Burton
348 2022; Moran et al. 2024). However, the mildly negative mitochondrial independent effect is
349 surprising, as the greater mutation rates of mtDNA than nDNA in animal cells (Ballard and
350 Whitlock 2003; Lynch et al. 2006) were expected to contribute to greater interspecific
351 incompatibility resulting from mitochondrial substitutions (Orr 1995). In addition, the roughly 5-
352 fold greater effect sizes of mitochondrial deleterious mutations than nuclear mutations
353 (Popadin et al. 2013) could further contribute to hybrid unfitness. However, greater purifying
354 selection (Popadin et al. 2013) in the mitochondrial genome and compensatory evolution of the
355 nuclear genome (Princepe and De Aguiar 2024) might drastically dampen the effect size of
356 independent mitochondrial substitutions on hybrid fitness, which is consistent with our
357 observation (Fig. 5C). This notion is also supported by the narrower mtDNA cline than the
358 nDNA cline (Fig. S1) despite the smaller effective population size of mtDNA than nDNA
359 expected greater drift and a wider cline under neutrality (Slatkin and Maruyama 1975).

360 *Caveats*

361 The available 330 animal hybrid zone studies have been concentrated in the global
362 north (Fig. 2B), with relatively fewer studies in the more biodiverse global south. Our
363 observation could be predominantly representing animal speciation patterns in temperate
364 ecoregions, which may be different from those in tropical ecoregions. There are far fewer
365 aquatic hybrid zones than terrestrial hybrid zones (Fig. 2B), which may constrain our
366 conclusion within the terrestrial domain. Life history information, such as independently
367 measured generation time and dispersal rate, was unavailable for a large proportion of the
368 hybrid zones, with only 148 of the 330 cases containing sufficient information to calculate
369 selection against hybrids. For the 148 hybrid zones for which we could calculate selection,
370 there were insufficient sequence data in GenBank for commonly sequenced genes in the
371 nuclear genomes of the species pairs. This sequencing gap limits the inference from the
372 refined models to a small sample size. Despite the limited dataset, the significant mitonuclear
373 effect on the selection against hybrid DNAs supported the Endosymbiotic Snowball Effect
374 (Wang et al. 2025). Future studies that broaden the sequencing scope of hybrid zone studies
375 could provide valuable modeling resolution to illuminate speciation heterogeneity across the
376 genomes. Since the understanding of organismal gene flow is crucial for biodiversity science
377 and conservation in the increasingly fragmented biosphere (Wang and Yoder 2025), a
378 standardized global hybrid zone database could be productive for speciation and conservation
379 science.

380 *Conclusion*

381 We investigated the evolutionary mode of reproductive isolation in the early stage of
382 speciation, empowered by decades of animal hybrid zone studies. We found a gradual buildup

383 of selection against hybrids as the parental lineages diverged. Further dissection of
384 mitochondrial and nuclear genetic divergence revealed exponential growth of selection against
385 hybrid DNA, consistent with the Endosymbiotic Snowball Effect (Wang et al. 2025). Future
386 dissection of the genomic heterogeneity of hybrid fitness effects would be indispensable for
387 refining our understanding of speciation rates across the tree of life.

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Data Accessibility

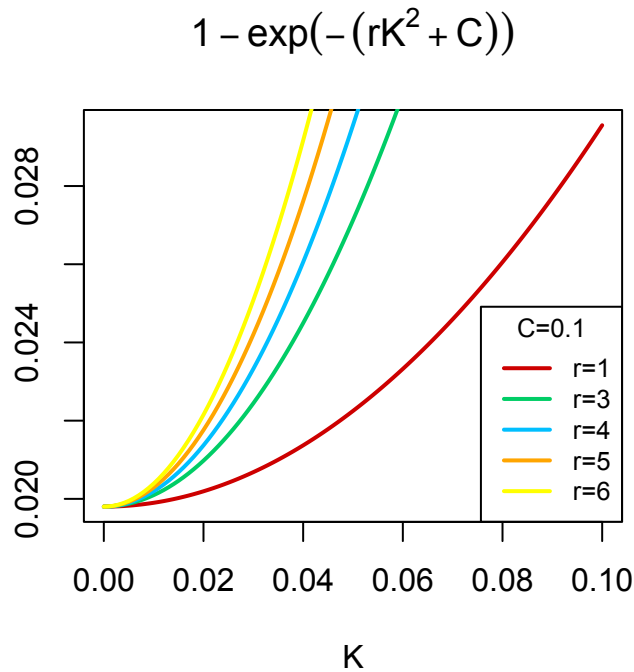
507 All the data involved in the analysis will be submitted to Dryad upon manuscript acceptance [e.](https://github.com/setophaga/hybridzone.snowball/tree/7dcd583defec2227a624b7c9a30b2d1806e16331/Final.data.code)
508 [https://github.com/setophaga/hybridzone.snowball/tree/7dcd583defec2227a624b7c9a30b2d18](https://github.com/setophaga/hybridzone.snowball/tree/7dcd583defec2227a624b7c9a30b2d1806e16331/Final.data.code)
509 06e16331/Final.data.code

510
511

512 **Author Contributions**

513 SW conceived the idea, and BR and DE conducted data collection under the supervision of
514 SW. SW analyzed the data with assistance from DE and BR. All authors contributed for
515 literature data extraction and cleaning. Specifically, BR curated the GPS coordinates of the
516 hybrid zones and DE conducted the GenBank sequence extraction and alignments. SW
517 conducted speciation modeling while teaching DE and BR.

518 **Supplement**



519 **Fig. S1. Modified Snowball Effect.** Orr (1995) derived the cumulative effect of K substitutions
520 with average effect size r on reduction of hybrid fitness, or the “strength of reproductive
521 isolation” as $1 - e^{-rK^2/2}$ (see Eq. 8 in Orr 1995). To incorporate epigenetic effects and be
522 directly comparable to Additive and Slowdown effects (Eq. 1.1 and 1.3), we modified the
523 original form and examined its form with different r values at moderate C .
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