

Natural disasters generate heterogeneity in individual life histories

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Abstract

Extreme climatic events may influence individual-level variability in phenotypes, survival, and reproduction, and thereby drive the pace of evolution. Here, we quantify how experiencing major hurricanes influences individual life courses in the Cayo Santiago rhesus macaques. Our results show that major hurricanes increase heterogeneity in reproductive life courses despite an average reduction in mean fertility and survival, i.e. shortened life courses. In agreement with this, the population is expected to achieve stable population dynamics faster after a hurricane. Our work suggests that natural disasters force individuals into new niches to potentially reduce strong competition during poor environments where mean reproduction and survival are compromised. We also demonstrate that variance in lifetime reproductive success and longevity are differently affected by hurricanes, and such variability is mostly driven by survival.

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26 **Abstract.** Extreme climatic events may influence individual-level variability in phenotypes,
27 survival, and reproduction, and thereby drive the pace of evolution. Here, we quantify how
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29 macaques. Our results show that major hurricanes increase heterogeneity in reproductive life
30 courses despite an average reduction in mean fertility and survival, i.e. shortened life courses. In
31 agreement with this, the population is expected to achieve stable population dynamics faster after
32 a hurricane. Our work suggests that natural disasters force individuals into new niches to
33 potentially reduce strong competition during poor environments where mean reproduction and
34 survival are compromised. We also demonstrate that variance in lifetime reproductive success and
35 longevity are differently affected by hurricanes, and such variability is mostly driven by survival.

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49 **Introduction**

50 Hurricanes are rare, yet their extreme acute nature can have profound impacts on life
51 courses. They can reduce long-term population fitness through direct mortality (Batista & Platt,
52 2003; Wiley & Wunderle, 1993), suppressed fertility (Gannon & Willig, 1994; Luevano et al.,
53 2022; Morcillo et al., 2020), increased physiological stress (Behie & Pavelka, 2014), increased
54 social disorganization (Pavelka et al., 2003), and alterations in gene expression (Watowich et al.,
55 2022). Climate models predict increases in the frequency of intense hurricanes (Bloemendaal et
56 al., 2022; Holland & Webster, 2007), but how populations will respond to such sustained
57 environmental change remains unclear. This reflects limitations in our current ecological and
58 evolutionary understanding which mostly focuses on average population-level processes even
59 though individual heterogeneity is what evolution acts on (Vindenes et al., 2008; Vindenes &
60 Langangen, 2015). To predict the eco-evolutionary dynamics of populations under rapid
61 environmental change, we must first investigate whether and how extreme climatic events affect
62 the emergence and maintenance of individual heterogeneity (Caswell, 2001; Metcalf & Pavard,
63 2007). If adaptive, individual heterogeneity could support the persistence of populations by
64 counteracting the negative effects of extreme environmental conditions (Chevin & Hoffmann,
65 2017). However, if the observed individual heterogeneity is neutral or fully stochastic, its potential
66 for selection is hindered and the pace of evolutionary adaptation is significantly lowered,
67 compromising the viability of populations at risk (Chevin & Lande, 2010; Steiner et al., 2021;
68 Steiner & Tuljapurkar, 2012).

69 Here, we aim to understand how tropical cyclones influence individual heterogeneity in
70 life courses by contrasting individual performance during years of major hurricanes with
71 performance during ordinary years using a long-lived primate population. We first define

72 individual heterogeneity as differences in individual life courses and characterize it using metrics
73 of population entropy and stage persistence (Tuljapurkar et al., 2009). Starting at birth, life courses
74 can be described by the sequence of stages an individual experiences until death (Caswell 2001).
75 As life progresses through time, individuals may remain in the same stage (i.e., stasis) or transition
76 among developmental, morphological, reproductive, behavioral, and physiological stages.
77 Population entropy measures the rate at which these stage trajectories diversify with age
78 (Hernández-Pacheco & Steiner, 2017; Tuljapurkar, 1982; Tuljapurkar et al., 2009). Persistence in
79 any life stage instead measures the correlation time between an individual's current life stage and
80 their stage at a later age. Contrary to stable environments which are expected to reduce individual
81 heterogeneity because persistence of life stages is high (Tuljapurkar et al., 2009), extreme climatic
82 events on animal populations leading to unfavorable conditions are expected to increase individual
83 heterogeneity (Chambert et al., 2015; Jenouvrier et al., 2015). As major hurricanes result in
84 extremely poor environments that lead to strong intraspecific competition (Schaffner et al., 2012),
85 they may force individuals into new life history strategies or ecological niches increasing the
86 expected rate of diversification in stage trajectories (i.e., high entropy, low stage persistence).
87 Alternatively, major hurricanes may reveal potential trade-offs between survival and reproduction.
88 Here, individuals are driven to an optimal life history strategy by either allocating more energy to
89 maintenance processes to ensure future reproductive success after recovery (Morcillo et al. 2020)
90 or by increasing reproduction and growth in certain life stages where survival is uncertain
91 (Pascarella & Horvitz, 1998), thereby reducing entropy and increasing stage persistence.

92 Individual heterogeneity can also be described by the exact probability distributions of life
93 history traits and their higher moments, with life history traits being evaluated at the end of life
94 (Caswell, 2009, 2011; Tuljapurkar et al., 2020; van Daalen & Caswell, 2020). As individuals age,

95 their movement through the life cycle follows a random process determined by transition rates that
96 generates variation in life-history traits. We thus follow by defining the exact probability
97 distributions of lifetime reproductive success (LRS) and lifespan (Tuljapurkar et al., 2020), and
98 the variation in the lifetime number of visits to any transient state (i.e., individual stochasticity;
99 Caswell 2009). Given the negative effects of hurricanes on mean annual fertility and survival
100 across populations (Gannon & Willig, 1994; Morcillo et al., 2020; Wiley & Wunderle, 1993), we
101 predict that hurricanes shape the distributions of LRS and lifespan by reducing variance and
102 showing positive skewness (Tuljapurkar et al., 2020).

103 In this study, we evaluate the effects of major hurricanes on individual heterogeneity using
104 the Cayo Santiago rhesus macaques, a population whose mean fertility is reduced during hurricane
105 years likely to maintain high rates of survival (Morcillo et al., 2020). Here, we estimate multiple
106 metrics of individual heterogeneity using annual stage-structured matrix population models. First,
107 we estimate population entropy as a proxy of the expected annual rate of diversification in
108 reproductive life courses of individuals (Hernández-Pacheco & Steiner, 2017) during hurricane
109 years and contrast it to ordinary years. Next, we measure persistence time of life stages to
110 determine the correlation between an individual's current reproductive stage and its stage years
111 later (Tuljapurkar et al. 2009). Lastly, we quantify within-trajectory variation in LRS and lifespan
112 to obtain their exact probability distributions and higher moments (Caswell, 2009, 2011; Steiner
113 et al., 2010; Tuljapurkar et al., 2020). Our analysis revealed that major hurricanes influence eco-
114 evolutionary processes by increasing entropy and lowering correlation time despite evidence of
115 reductions in mean annual fertility and survival. We also find that the LRS and lifespan
116 distributions are positively skewed and are mostly driven by survival.

117 **Methods**

118 *Study population:*

119 Cayo Santiago is a 15.2ha sub-tropical island located 1km Southeast of Puerto Rico (lat.
120 18°09'N, long. 65°44'W) that serves as a biological station for behavioral primate studies. The
121 station was established in 1938 and is inhabited by free-ranging rhesus macaques (*Macaca*
122 *mulatta*). Monkeys spend 50% of their daily activities foraging on natural vegetation on the island
123 (Marriott et al., 1989) and are also provisioned ad libitum with water and approximately 0.23
124 kg/animal/day of commercial monkey chow. Since 1973 the entire population has been monitored
125 and a reliable longitudinal demographic database on all individuals has been maintained.
126 Individual data includes date of birth, sex, mother identification, social group membership, and
127 date of death or permanent removal from the island for all individuals (for details on Cayo Santiago
128 population data collection and management, see Hernández-Pacheco et al. (2016) and Ruiz-
129 Lambides et al. (2017)).

130 The Cayo Santiago rhesus macaque population has experienced the direct impact of three
131 major hurricanes (category ≥ 3) since the establishment of census records in 1956: Hugo (18
132 September 1989), Georges (21 September 1998) and Maria (20 September 2017; Kessler &
133 Rawlins, 2016; Morcillo et al., 2020). Hugo and Georges were category 3 hurricanes with
134 sustained wind speed of 201 km h⁻¹ and 185 km h⁻¹, respectively (NOAA, 2014; USGS, 1999).
135 Maria was a category 4 hurricane with sustained wind speed of 220 km h⁻¹ (NOAA, 2019).
136 Although food provisioning - and thus census taking - was resumed between 1 to 3 days after each
137 hurricane (Morcillo et al., 2020), there is evidence of significant hurricane effects on several
138 aspects of the population. Each hurricane caused 60-90% of canopy loss immediately after the
139 event (Morcillo et al., 2020), and changes in the social structure (Testard et al., 2021), adverse
140 demographic effects such as suppressed fertility (Luevano et al., 2022; Morcillo et al., 2020), as

141 well as alterations in immune cell gene regulation (Watowich et al., 2022) associated to hurricanes
142 have been reported.

143 *Demographic analysis:*

144 Our analysis was based on 46 years of individual data from 1973 to 2019. For each year,
145 we parametrized female-only, birth-pulse matrix population models employing post-breeding
146 censuses (Caswell 2001). Following Hernández-Pacheco and Steiner (2017), we defined the
147 annual structure in our analysis from 1 June at time t to 31 May at time $t + 1$ to avoid significant
148 overlap of birth seasons. Models were based on annual transition probabilities among
149 developmental and reproductive stages (Hernández-Pacheco & Steiner, 2017; Morcillo et al.,
150 2020). In a given year, we classified sexually immature females in one of three age-specific
151 developmental stages: infant (I; < 1 year of age), yearling (Y; 1-2 years of age) and juvenile (J; 2-
152 3 years of age). After reaching 3 years of age, we classified females in one of three reproductive
153 stages: nonbreeder (NB), failed breeder (FB) and successful breeder (B). Nonbreeders were adult
154 females who did not have an offspring a given year (i.e., birth season skipping). Failed breeders
155 were adult females whose offspring died before reaching 1 year of age. Successful breeders were
156 adult females whose offspring survived to 1 year of age (recruitment). Adult females transitioned
157 among these three reproductive stages until death or until being right censored due to permanent
158 removal from the population or if alive at the end of our study. Although transitions from J to B
159 or from J to FB are rare, they are expected to be non-zero as a small portion of 3-year-old females
160 reproduce (Hernández-Pacheco et al., 2013). We only considered reproductive performance of
161 females that survived to the census day, thus females dying during a given year were classified as
162 transitioning to the absorbing death state independently of their reproduction.

163 To address hurricane effects on individual heterogeneity, we parameterized four stage-
164 structured matrix models, one for each environment e , with stage-transition matrix, \mathbf{P}_e , and stage-
165 specific survival rates, $s_{je} = \sum_i^n P_{ije} = 1 - d_{je}$, where d_{je} is stage-specific mortality of stage j and
166 environment e , and n is the number of stages. The four environments e were defined by the annual
167 individual transitions belonging to (1) non-hurricane years (1973-1988, 1990-1997, 1999-2016,
168 2018-2019), (2) Hugo (1989-1990), (3) Georges (1998-1999), and (4) Maria (2017-2018). As only
169 stage B females contributed to reproduction, we set their fertility to 1 (100%) and NB and FB
170 fertility to 0. Survival of infants was set to 1, as only surviving infants were recruited into the
171 population (Morcillo et al., 2020). In our analysis, we assumed stage-specific mortality for adults
172 at any given time period to be non-zero. As no deaths were recorded among FB and B during
173 hurricane Hugo (Supporting Material, Table S1), we performed all analyses after adjusting the
174 Hugo matrix by adding a 1% of total mortality rate to FB and B. For completion, we present the
175 analysis using the empirical Hugo matrix in the Supporting Material (Table S2). For each matrix
176 model, we estimated the asymptotic growth rate (λ), the stable stage distribution (w), and the
177 reproductive values (v) by computing the dominant eigenvalue, and the corresponding right and
178 left eigen vectors, respectively (Caswell, 2001). 95% confidence intervals for λ were estimated
179 using bootstrap methods (Supporting Material). Transient dynamics can inform us on how strong
180 a perturbation disequilibrates the stage structure of the population and how fast these effects can
181 be reversed, i.e. how fast a population reaches a stable equilibrium. To contrast each hurricane
182 year with ordinary years, we compared the time the population takes to converge into stable
183 equilibrium following each environment by estimating the damping ratio $\rho_e = \frac{\lambda}{\lambda_1}$, where λ_1 is the
184 subdominant eigenvalue, for each matrix \mathbf{P}_e (Caswell 2001).

185 *Individual heterogeneity in stage trajectories:*

186 Individuals differ in their sequence of reproductive stages making up their life course. This
187 sequence defines an individual's reproductive trajectory (ω) and population entropy, H , describes
188 the rate at which these trajectories diversify with age (Tuljapurkar et al., 2009; Table 1). We
189 quantified this variation using matrix \mathbf{R}_e , a 3 x 3 submatrix of \mathbf{P}_e including adult stages only,
190 weighted by its corresponding quasi-stationary stage distribution, ω' (Hernández-Pacheco &
191 Steiner, 2017; Steiner et al., 2010). Here, we employed H as a proxy of the expected annual rate
192 of diversification in life courses of mature individuals (Hernández-Pacheco & Steiner, 2017). If
193 individuals follow the same sequence of reproductive stages across time, H is 0 (100% predictable
194 stage trajectory). On the contrary, if individuals are equally likely to transition from any given
195 state to any other state in the following year then entropy increases up to its maximum value $\ln(K)$,
196 where K is the number of reproductive stages (Tuljapurkar et al., 2009). For comparison across
197 environments, we scaled H to its maximum value and presented relative H which is bounded
198 between 0 and 1. We also simulated heterogeneity in fitness estimates (survival and reproduction)
199 using the mean population trajectory and the probability of observing a trajectory in transition
200 matrices. Without heterogeneity in life courses, every individual would follow the same
201 reproductive trajectory so that the populations reproductive variance becomes 0.

202 We estimated the characteristic time τ and defined it as the correlation between an
203 individual's current reproductive stage at time t and its stage $t+1$ years later (Table 1). In this way,
204 τ is a timescale that measures the persistence of reproductive success or failure (Tuljapurkar et al.
205 2009). A low τ indicates that an individual's current reproductive stage little predicts its future
206 reproductive stage, and thus life trajectories are less deterministic.

207 *Probability distributions and higher moments of LRS and longevity*

208 We computed the exact probability distributions of LRS and lifespan (i.e., age-distribution
209 of death) for each matrix \mathbf{P}_e following methods based on discrete convolutions and discrete Fourier
210 transforms (Tuljapurkar et al., 2020; Supporting Material). The distributions were computed based
211 on stage-only models where individuals may visit a stage any number of times before dying. To
212 estimate higher moments in LRS and longevity, we computed the fundamental matrix \mathbf{N} from
213 matrix \mathbf{P}_e . Matrix \mathbf{N} allows us to estimate the expected mean number of visits to transient state i
214 an individual that starts in transient state j makes before death, regardless the order of occurrence
215 (sequence; Table 1, Caswell, 2009). In this way, we defined stage-specific mean LRS as the
216 expected mean number of visits (including stasis) an individual in stage j makes to the successful
217 breeder stage before death (i.e., last row of \mathbf{N}). Similarly, we defined longevity as the mean number
218 of visits an individual in stage j makes to all other stages before being absorbed in the death stage
219 (i.e., the sum of each column of \mathbf{N}). We also estimated the variance, skewness, and the coefficient
220 of variation (CV) for LRS (Table 1; Caswell, 2011, 2013; Varas Enríquez et al., 2022). A high
221 magnitude in skewness indicates that rare individuals experience an unusually short (negative
222 skew) or long (positive skew) life. A higher CV indicates that reproduction is highly spread over
223 the life stages (more variation), relative to the mean expectation. All analyses were performed
224 using the R software, version 4.1.3 (R Studio Team, 2022).

225 **Results**

226 Our analysis was based on 20,891 individual transitions from 4,075 females. During the
227 46-year period, λ was 1.119 (95% CI: 1.114, 1.123), for a mean annual population growth of
228 11.9%. During non-hurricane years, $\lambda = 1.121$ (1.116, 1.125), while during hurricane years λ was
229 reduced to 1.100 (1.084 1.114). Specifically, during hurricanes Hugo, Georges, and Maria λ was
230 1.128 (1.099, 1.152), 1.101 (1.058, 1.119), and 1.087 (1.060, 1.113), respectively. On average,

231 hurricane years revealed a higher proportion of NB and FB in the stable stage distribution relative
232 to non-hurricane years (Fig. 1). Contrary to this, hurricane years show a lower proportion of B and
233 I in the stable stage distribution (e.g., females transitioned less to the B stage and more to the NB
234 and FB stage; Fig. 1; Table 2). Moreover, FB showed the lowest survival during mean hurricane
235 years compared to non-hurricane years (0.865 and 0.912, respectively; Table S3). Years of major
236 hurricanes showed an increased damping ratio (Hugo = 1.532, Georges = 1.502, Maria = 1.504)
237 relative to years of no hurricanes ($\rho = 1.482$), suggesting that the population takes a shorter time
238 to converge into stable stage dynamics during poor environment years.

239 Population entropy increased during hurricane years ($H = 0.744$), relative to the non-
240 hurricane environment ($H = 0.718$) and such increase was related to hurricane intensity defined by
241 sustained wind speed. During Georges, the population entropy was $H = 0.663$, followed by Hugo
242 with $H = 0.757$, and Maria with $H = 0.784$. The different entropies characterizing our population
243 across environments can be further visualized with simulated cumulative reproduction (CR)
244 trajectories which show a higher diversification in reproductive trajectories as a function of age
245 (Fig. 2, top panel). Similarly, the corresponding simulated cumulative survival trajectories showed
246 a higher diversification with hurricane intensity (Fig. 2, bottom panel). In agreement with this,
247 stage persistence was reduced with increasing hurricane intensity (Georges = 0.014, Hugo = 0.009,
248 Maria = 0.006), while ordinary years showed the lowest value ($\tau = 0.0005$).

249 The LRS distribution across all environments was positively skewed with rare females
250 having an unusually high number of offspring (Fig. 3, top panel). This skewness was more
251 pronounced for Georges and Maria where the probability of having no offspring was greater (33%
252 and 35%, respectively) than that of the non-hurricane environment (23%). Visits to the breeder
253 stage was reduced during hurricanes Georges and Maria where females were expected to have a

254 mean LRS of 4 and 3 offspring, respectively (Table 3). During non-hurricane environments,
255 females were expected to have a mean LRS of 5 offspring. Variance in LRS from birth was highest
256 for non-hurricane years, followed by Georges and Maria (35.289, 25.003, 17.153, respectively,
257 Table 3). In contrast, during hurricane Hugo females had a mean LRS of 9 offspring with the
258 highest variance of 102.617. Immature stages showed the highest variability in LRS with respect
259 to their mean (CV; Table 3).

260 The distribution of lifespan was also positively skewed across all environments revealing
261 that individuals have a low probability of mortality past maturity (≥ 3 years of age) and that there
262 is no major differences in the likelihood to live long (Fig. 3, bottom panel). Hurricane Georges and
263 Maria show the highest probability of dying early in life relative to Hugo and ordinary years. We
264 observed a reduction in mean lifespan from birth during Georges (15.824 years) and Maria (15.555
265 years) relative to the non-hurricane environment (19.730 years). However, Hugo had a mean
266 lifespan of 38.227 years (Table 3). We found no evidence suggesting that Hugo was an unusually
267 good year as mean vital rates in 1989 were similar to other years (Fig. S1).

268 **Discussion**

269 Our study revealed that major hurricanes generate heterogeneity in individual life courses
270 despite an average reduction in mean fertility (i.e., decreased transitions to B) and survival (i.e.,
271 shortened life courses). Extreme climatic events may thus force individuals into new ecological
272 niches increasing the rate of diversification in reproductive stage trajectories and decreasing
273 reproductive stage persistence, while simultaneously shortening lifespans. By assessing life history
274 trait distributions, we also demonstrate that hurricanes have different effects on the variation in
275 LRS and lifespan and that such stochasticity is highly driven by survival in our long-lived primate
276 population.

277 Prior evidence show that hurricanes affect negatively the dynamics of animal populations
278 mainly through changes in food and habitat structure (Klinger, 2006; Pavelka & Behie, 2005;
279 Woolbright, 1991), but few studies have shown the long-term effects on fertility and mortality
280 (Luevano et al., 2022; Morcillo et al., 2020; Pavelka et al., 2007). The effects of hurricanes on the
281 population dynamics of Cayo Santiago rhesus macaques are mostly driven by reductions in mean
282 annual fertility, suggesting that survival is maintained at the expense of reproduction (Morcillo et
283 al. 2020). Yet, these population-level processes reveal no information about the role of these
284 climatic events and consequent environmental changes on the emergence and maintenance of
285 individual heterogeneity and trait variances. Our analysis shows that population entropy increased
286 with hurricane intensity, demonstrating that natural disasters generate heterogeneity in individual
287 life courses and that such heterogeneity is independent of potential trade-offs between
288 reproduction and longevity. In contrast to this, stage persistence was reduced with increasing
289 hurricane intensity, thus the more extreme the environmental impact the more unpredictable life
290 stages are given a female's current stage (i.e., low correlation time). An increased population
291 entropy and a decreased stage persistence during hurricanes could be the result of a strategy to
292 avoid strong intraspecific competition during bad years through the exploration of ecological
293 niches or demographic roles (Bolnick, 2001; Coulson et al., 2001). On the other hand, high-quality
294 females (individuals that survive and breed successfully) could sustain breeding despite bad
295 environmental conditions, whereas poor quality females might need to transition into other stages
296 potentially contributing to the observed variability (Jenouvrier et al., 2022). In alignment with
297 entropy, years of major hurricanes had the highest rate of convergence to a stable stage structure.
298 This suggests that populations in which individuals explore stages that can maximize fitness at

299 higher rates may reach equilibrium sooner than populations where individuals are performing less
300 randomly.

301 Our findings also reveal that other unknown factors are contributing to individual
302 heterogeneity as non-hurricane years had a relatively high entropy. For example, unmeasured
303 physiology (Plard et al., 2015), as well as maternal and genetic (fixed) heterogeneity (Peripato et
304 al., 2002) may contribute to differences in life history outcomes. Although our population exhibits
305 negative density-dependence in fertility (Hernández-Pacheco et al., 2013), entropy was not found
306 to be driven by density (Hernández-Pacheco & Steiner, 2017). We also found stage persistence to
307 be the lowest during ordinary years. Such low correlation between current stage and a future stage
308 could be due to individuals randomly exploring new stages to maximize the available variability
309 for eco-evolutionary processes. Future research has yet to explore what other factors drive
310 population entropy and stage persistence in ordinary environments.

311 In agreement with previous studies, our analysis shows that life history trait distributions
312 are highly skewed (Colchero et al., 2016; Tatarenkov et al., 2008; Tuljapurkar et al., 2020; Fig. 3).
313 In our population, only rare females produce a large number of daughters mainly because most
314 females do not live their entire reproductive life (3-24 years of age) as they die at younger ages
315 (Hernández-Pacheco et al., 2013). In particular, the LRS distributions show that hurricanes can
316 increase the probability of having no offspring as Georges and Maria had the largest probability
317 of 0 LRS, a larger skewness, and a reduced mean and variance, compared to ordinary years. The
318 lifespan distributions revealed that the probability of mortality is highest before sexual maturity
319 with one single large mortality schedule, in contrast to other mammal populations exhibiting one
320 mortality schedule during juvenility and a second during old ages (belugas: Schindler et al., 2012;
321 humans: Edwards & Tuljapurkar, 2005). This mortality schedule was more accentuated during

322 Georges and Maria (Fig 3., bottom panel). However, this effect was eliminated when adult survival
323 was not negatively affected as during Hugo. Similar to hurricanes Georges and Maria in which
324 transitions into the breeder stage were reduced (Table 2), the population suffered a reduction in
325 mean annual fertility during hurricane Hugo, however during Hugo mortality was also reduced
326 among adults. As a result, this hurricane event showed a substantial increase in mean and variance
327 of life history traits, supporting prior evidence of the unbalanced contribution of survival and
328 fertility to variability in this long-lived primate population (Morcillo et al., 2020). The fact that
329 Hugo showed high survival was not surprising as no evidence that hurricanes affect survival was
330 found previously (Morcillo et al., 2020). Rather than a rare year, during Hugo females showed
331 similar survival rates as many other years in the history of the population (Fig. S1).

332 Our findings support the hypothesis that intraspecific competition drives niche exploration,
333 and this effect is more pronounced during extreme climatic events as seen by an increased
334 heterogeneity in life histories following major hurricanes. Future studies investigating the viability
335 of vulnerable populations should address the effects of extreme climatic events on individual
336 heterogeneity. This will help to better understand in what direction significant environmental
337 changes drive individual life courses and trait distributions, especially for traits (heritable or
338 plastic) that have high adaptive potential (Conner & White, 1999). In particular, quantifying these
339 changes may reveal if populations are at risk of extinction, especially if extreme events are drivers
340 of evolutionary change (Grant et al., 2017). Future directions in modelling the effects of extreme
341 climatic events on demography also include the description of probability distributions of life
342 history traits (Schindler et al., 2012). Such approach can be used to further understand how the
343 long-term fitness of a population (i.e., growth of a stable population) at risk is shaped by the LRS
344 distribution and the age-distribution of death.

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356 **Reference list**

- 357 Behie, A. M., & Pavelka, M. S. M. (2014). Interacting roles of diet, cortisol levels, and parasites
358 in determining population density of Belizean howler monkeys in a hurricane damaged
359 forest fragment. *Primates in Fragments: Complexity and Resilience*, April, 447–456.
360 https://doi.org/10.1007/978-1-4614-8839-2_30
- 361 Bloemendaal, N., Moel, H. De, Martinez, A. B., Muis, S., Haigh, I. D., Wiel, K. Van Der,
362 Haarsma, R. J., Ward, P. J., Roberts, M. J., Dullaart, J. C. M., & Aerts, J. C. J. H. (2022). A
363 globally consistent local-scale assessment of future tropical cyclone risk. *Science Advances*,
364 8438, 1–14.
- 365 Bolnick, D. I. (2001). *Intraspecific competition favours niche width expansion in Drosophila*
366 *melanogaster*. <http://www.nature.com>
- 367 Caswell, H. (2001). Matrix Population Models: Construction, Analysis, and Interpretation. In

368 *Sinauer Associates, Sunderland, MA* (Issue 2nd edn.).

369 Caswell, H. (2009). *Stage, Age and Individual Stochasticity in Demography* (Vol. 118, Issue 12).
370 <https://www.jstor.org/stable/27759794?seq=1&cid=pdf->

371 Caswell, H. (2011). Beyond R0: demographic models for variability of lifetime reproductive
372 output. *PloS One*, 6(6). <https://doi.org/10.1371/JOURNAL.PONE.0020809>

373 Caswell, H. (2013). Sensitivity analysis of discrete Markov chains via matrix calculus. *Linear*
374 *Algebra and Its Applications*, 438(4), 1727–1745.
375 <https://doi.org/10.1016/J.LAA.2011.07.046>

376 Chambert, T., Rotella, J. J., & Garrott, R. A. (2015). Female Weddell seals show flexible
377 strategies of colony attendance related to varying environmental conditions. *Ecology*, 96(2),
378 479–488.

379 Chevin, L. M., & Hoffmann, A. A. (2017). Evolution of phenotypic plasticity in extreme
380 environments. *Philosophical Transactions of the Royal Society B: Biological Sciences*,
381 372(1723). <https://doi.org/10.1098/RSTB.2016.0138>

382 Chevin, L. M., & Lande, R. (2010). When do adaptive plasticity and genetic evolution prevent
383 extinction of a density-regulated population? *Evolution*, 64(4), 1143–1150.
384 <https://doi.org/10.1111/J.1558-5646.2009.00875.X>

385 Colchero, F., Rau, R., Jones, O. R., Barthold, J. A., Conde, D. A., Lenart, A., Nemeth, L.,
386 Scheuerlein, A., Schoeley, J., Torres, C., Zarulli, V., Altmann, J., Brockman, D. K.,
387 Bronikowski, A. M., Fedigan, L. M., Pusey, A. E., Stoinski, T. S., Strier, K. B., Baudisch,
388 A., Vaupel, J. W. (2016). The emergence of longevous populations. *Proceedings of the*
389 *National Academy of Sciences of the United States of America*, 113(48), E7681–E7690.
390 <https://doi.org/10.1073/PNAS.1612191113/->

391 /DCSUPPLEMENTAL/PNAS.201612191SI.PDF

392 Conner, M. M., & White, G. C. (1999). Effects of individual heterogeneity in estimating the
393 persistence of small populations. *Natural Resource Modeling*, 12(1), 109–127.
394 <https://doi.org/10.1111/J.1939-7445.1999.TB00005.X>

395 Edwards, R. D., & Tuljapurkar, S. (2005). Inequality in life spans and a new perspective on
396 mortality convergence across industrialized countries. *Population and Development Review*,
397 31(4), 645–674. <https://doi.org/10.1111/j.1728-4457.2005.00092.x>

398 Gannon, M. R., & Willig, M. R. (1994). The Effects of Hurricane Hugo on Bats of the Luquillo
399 Experimental Forest of Puerto Rico. *Association for Tropical Biology and Conservation*,
400 26(3), 320–331.

401 Grant, P. R., Rosemary Grant, B., Huey, R. B., Johnson, M. T. J., Knoll, A. H., & Schmitt, J.
402 (2017). Evolution caused by extreme events. *Philosophical Transactions of the Royal
403 Society B: Biological Sciences*, 372(1723). <https://doi.org/10.1098/RSTB.2016.0146>

404 Hernández-Pacheco, R., Delgado, D. L., Rawlins, R. G., Kessler, M. J., Ruiz-Lambides, A. V.,
405 Maldonado, E., & Sabat, A. M. (2016). Managing the Cayo Santiago Rhesus Macaque
406 Population: The Role of Density HHS Public Access. *Am J Primatol*, 78(1), 167–181.
407 <https://doi.org/10.1002/ajp.22375>

408 Hernández-Pacheco, R., Rawlins, R. G., Kessler, M. J., Williams, L. E., Ruiz-Maldonado, T. M.,
409 González-Martínez, J., Ruiz-Lambides, A. V., & Sabat, A. M. (2013). Demographic
410 variability and density-dependent dynamics of a free-ranging rhesus macaque population.
411 *American Journal of Primatology*, 75(12), n/a-n/a. <https://doi.org/10.1002/ajp.22177>

412 Hernández-Pacheco, R., & Steiner, U. K. (2017). Drivers of diversification in individual life
413 courses. *American Naturalist*, 190(6), E132–E144. <https://doi.org/10.1086/694317>

414 Holland, G. J., & Webster, P. J. (2007). Heightened tropical cyclone activity in the North
415 Atlantic: Natural variability or climate trend? *Philosophical Transactions of the Royal*
416 *Society A: Mathematical, Physical and Engineering Sciences*, 365(1860), 2695–2716.
417 <https://doi.org/10.1098/rsta.2007.2083>

418 Jenouvrier, S., Aubry, L., Daalen, S. van, Barbraud, C., Weimerskirch, H., & Caswell, H. (2022).
419 When the going gets tough, the tough get going: Effect of extreme climate on an Antarctic
420 seabird's life history. *Ecology Letters*, 00, 1–12. <https://doi.org/10.1111/ELE.14076>

421 Jenouvrier, S., Péron, C., & Weimerskirch, H. (2015). Extreme climate events and individual
422 heterogeneity shape life-history traits and population dynamics. *Ecological Monographs*,
423 85(4), 605–624. <https://doi.org/10.1890/14-1834.1>

424 Kessler, M. J., & Rawlins, R. G. (2016). A 75-year pictorial history of the Cayo Santiago rhesus
425 monkey colony. *American Journal of Primatology*, 78(1), 6–43.
426 <https://doi.org/10.1002/ajp.22381>

427 Klinger, R. (2006). The interaction of disturbances and small mammal community dynamics in a
428 lowland forest in Belize. *Journal of Animal Ecology*, 75, 1227–1238.
429 <https://doi.org/10.1111/j.1365-2656.2006.01158.x>

430 Luevano, L., Sutherland, C., Gonzalez, S. J., & Hernández-Pacheco, R. (2022). Rhesus
431 macaques compensate for reproductive delay following ecological adversity early in life.
432 *Ecology and Evolution*, 12(1). <https://doi.org/10.1002/ECE3.8456>

433 Marriott, B. M., Roemer, J., & Sultana, C. (1989). An overview of the food intake patterns of the
434 Cayo Santiago rhesus monkeys (*Macaca mulatta*): report of a pilot study. *Puerto Rico*
435 *Health Sciences Journal*, 8(1), 87–94. <https://europepmc.org/article/med/2780973>

436 Metcalf, C. J. E., & Pavard, S. (2007). Why evolutionary biologists should be demographers.

437 *Trends in Ecology and Evolution*, 22(4), 205–212.
438 <https://doi.org/10.1016/J.TREE.2006.12.001>

439 Morcillo, D. O., Steiner, U. K., Grayson, K. L., Ruiz-Lambides, A. V., & Hernández-Pacheco, R.
440 (2020). Hurricane-induced demographic changes in a non-human primate population. *Royal*
441 *Society Open Science*, 7(8), 200173. <https://doi.org/10.1098/rsos.200173>

442 Pascarella, J. B., & Horvitz, C. C. (1998). Hurricane disturbance and the population dynamics of
443 a tropical understory shrub: Megamatrix elasticity analysis. *Ecology*, 79(2), 547–563.
444 [https://doi.org/10.1890/0012-9658\(1998\)079\[0547:HDAATPD\]2.0.CO;2](https://doi.org/10.1890/0012-9658(1998)079[0547:HDAATPD]2.0.CO;2)

445 Pavelka, M. S. M., & Behie, A. M. (2005). The effect of Hurricane Iris on the food supply of
446 black howlers (*Alouatta pigra*) in southern Belize. *Biotropica*, 37(1), 102–108.
447 <https://doi.org/10.1111/j.1744-7429.2005.03102.x>

448 Pavelka, M. S. M., Brusselers, O. T., Nowak, D., & Behie, A. M. (2003). Population Reduction
449 and Social Disorganization in *Alouatta pigra* Following a Hurricane. *International Journal*
450 *of Primatology*, 24(5), 1037–1055. <https://doi.org/10.1023/A:1026276228635>

451 Pavelka, M. S. M., McGoogan, K. C., & Steffens, T. S. (2007). Population Size and
452 Characteristics of *Alouatta pigra* Before and After a Major Hurricane. *International Journal*
453 *of Primatology*, 28(4), 919–929. <https://doi.org/10.1007/s10764-007-9136-6>

454 Peripato, A. C., De Brito, R. A., Vaughn, T. T., Pletscher, L. S., Matioli, S. R., & Cheverud, J.
455 M. (2002). Quantitative trait loci for maternal performance for offspring survival in mice.
456 *Genetics*, 162(3), 1341–1353. <https://doi.org/10.1093/GENETICS/162.3.1341>

457 Plard, F., Gaillard, J. M., Coulson, T., Delorme, D., Warnant, C., Michallet, J., Tuljapurkar, S.,
458 Krishnakumar, S., & Bonenfant, C. (2015). Quantifying the influence of measured and
459 unmeasured individual differences on demography. *Journal of Animal Ecology*, 84(5),

460 1434–1445. <https://doi.org/10.1111/1365-2656.12393>

461 Ruiz-Lambides, A. V., Weiß, B. M., Kulik, L., Stephens, C., Mundry, R., & Widdig, A. (2017).
462 Long-term analysis on the variance of extra-group paternities in rhesus macaques.
463 *Behavioral Ecology and Sociobiology*, 71(4). <https://doi.org/10.1007/s00265-017-2291-7>

464 Scatena, A. F. N., Larsen, M. C., & Url, S. (2011). Physical Aspects of Hurricane Hugo in Puerto
465 Rico Responses to Hurricanes in the Caribbean (Dec ., 1991), pp . 317-323 Published by :
466 The Association for Tropical Biology and Conservation Physical Aspects of Hurricane
467 Hugo in Puerto Rico1. *Biotropica*, 23(4), 317–323.

468 Schaffner, C. M., Rebecchini, L., Ramos-Fernandez, G., Vick, L. G., Aureli, F., Schaffner, C.
469 M., Rebecchini, : L, Aureli, F., Ramos-Fernandez, G., & Vick, L. G. (2012). Spider
470 Monkeys (*Ateles geoffroyi yucatenensis*) Cope with the Negative Consequences of
471 Hurricanes Through Changes in Diet, Activity Budget, and Fission-Fusion Dynamics.
472 *International Journal of Primatology*, 33:922-936. [https://doi.org/10.1007/s10764-012-](https://doi.org/10.1007/s10764-012-9621-4)
473 [9621-4](https://doi.org/10.1007/s10764-012-9621-4)

474 Schindler, S., Tuljapurkar, S., Gaillard, J. M., & Coulson, T. (2012). Linking the population
475 growth rate and the age-at-death distribution. *Theoretical Population Biology*, 82(4), 244–
476 252. <https://doi.org/10.1016/j.tpb.2012.09.003>

477 Steiner, U.K., & Tuljapurkar, S. D. (2012). *Neutral theory for life histories and individual*
478 *variability in fitness components*. 109(12). <https://doi.org/10.1073/pnas.1018096109>

479 Steiner, U.K., Tuljapurkar, S. D., & Roach, D. A. (2021). Quantifying the effect of genetic,
480 environmental and individual demographic stochastic variability for population dynamics in
481 *Plantago lanceolata*. *Scientific Reports*, 11, 23174. [https://doi.org/10.1038/s41598-021-](https://doi.org/10.1038/s41598-021-02468-9)
482 [02468-9](https://doi.org/10.1038/s41598-021-02468-9)

483 Steiner, Ulrich K, Tuljapurkar, S. D., & Orzack, S. H. (2010). Dynamic heterogeneity and life
484 history variability in the kittiwake. *Journal of Animal Ecology*, 79(2), 436–444.
485 <https://doi.org/10.1111/j.1365-2656.2009.01653.x>

486 Tatarenkov, A., Healey, C. I. M., Grether, G. F., & Avise, J. C. (2008). Pronounced reproductive
487 skew in a natural population of green swordtails, *Xiphophorus helleri*. *Molecular Ecology*,
488 17(20), 4522–4534. <https://doi.org/10.1111/j.1365-294X.2008.03936.x>

489 Testard, C., Larson, S. M., Watowich, M. M., Kaplinsky, C. H., Bernau, A., Faulder, M.,
490 Marshall, H. H., Lehmann, J., Ruiz-Lambides, A., Higham, J. P., Montague, M. J., Snyder-
491 Mackler, N., Platt, M. L., & Brent, L. J. N. (2021). Rhesus macaques build new social
492 connections after a natural disaster. *Current Biology*, 31(11), 2299-2309.e7.
493 <https://doi.org/10.1016/J.CUB.2021.03.029>

494 Tuljapurkar, S. D. (1982). Why Use Population Entropy? It Determines the Rate of
495 Convergence. In *J. Math. Biology* (Vol. 13).

496 Tuljapurkar, S. D., Steiner, U. K., & Orzack, S. H. (2009). Dynamic heterogeneity in life
497 histories. *Ecology Letters*, 12(1), 93–106. <https://doi.org/10.1111/j.1461-0248.2008.01262.x>

498 Tuljapurkar, S. D., Zuo, W., Coulson, T., Horvitz, C., & Gaillard, J.-M. (2020). Skewed
499 distributions of lifetime reproductive success: beyond mean and variance. *Ecology Letters*.
500 <https://doi.org/10.1111/ele.13467>

501 van Daalen, S. F., & Caswell, H. (2020). Variance as a life history outcome: Sensitivity analysis
502 of the contributions of stochasticity and heterogeneity. *Ecological Modelling*, 417, 108856.
503 <https://doi.org/10.1016/J.ECOLMODEL.2019.108856>

504 Varas Enríquez, P. J., Van Daalen, S., & Caswell, H. (2022). Individual stochasticity in the life
505 history strategies of animals and plants. *PloS One*, 17(9), e0273407.

506 <https://doi.org/10.1371/journal.pone.0273407>

507 Vindenes, Y., Engen, S., & Sæther, B. (2008). Individual Heterogeneity in Vital Parameters and
508 Demographic Stochasticity. *The American Naturalist*, 171(4), 455–467.
509 <https://doi.org/10.1086/528965>

510 Vindenes, Y., & Langangen, Ø. (2015). Individual heterogeneity in life histories and eco-
511 evolutionary dynamics. *Ecology Letters*, 18(5), 417–432.
512 <https://doi.org/10.1111/ELE.12421>

513 Watowich, M. M., Chiou, K. L., Montague, M. J., Simons, N. D., Horvath, J. E., Ruiz-Lambides,
514 A. V., Martinez, M. I., Higham, J. P., Brent, L. J. N., Platt, M. L., & Snyder-Mackler, N.
515 (2022). Natural disaster and immunological aging in a nonhuman primate. *Proceedings of*
516 *the National Academy of Sciences of the United States of America*, 119(8).
517 https://doi.org/10.1073/PNAS.2121663119/SUPPL_FILE/PNAS.2121663119.SD11.XLSX

518 Wiley, J. W., & Wunderle, J. M. (1993). The effects of hurricanes on birds, with special
519 reference to Caribbean islands. *Bird Conservation International*, 3(4), 319–349.
520 <https://doi.org/10.1017/S0959270900002598>

521 Woolbright, L. L. (1991). The Impact of Hurricane Hugo on Forest Frogs in Puerto Rico.
522 *Biotropica*, 23(4), 462–467.

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529 **Tables**

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531 **Table 1.** Parameters and definitions. Taken from Caswell 2009 and Tuljapurkar et al. 2009.

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Parameter	Definition
$\tau = \frac{-1}{\ln(\lambda_1)}$	Correlation time Subdominant eigenvalue ($\lambda_1 \neq 1$; solves $\det(\lambda I - \mathbf{R}) = 0$)
$H = -\sum_{i=1}^K \sum_{j=1}^K \pi_i \mathbf{R}_{ij} \log \mathbf{R}_{ij}$ (with $0 \log(0) = 0$)	Population entropy
$V(v_{ij}) = (2\mathbf{N}_{diag} - I)\mathbf{N} - \mathbf{N}^2$	Variance in stage occupancy time LRS
$SD(v_{ij}) = \sqrt{V(v_{ij})}$	Standard deviation in LRS
$\eta_2 = (2\mathbf{N}_{diag} - I)\mathbf{N}$	Second moment in LRS
$\eta_3 = (6\mathbf{N}_{diag}^2 - 6\mathbf{N}_{diag} + I)\mathbf{N}$	Third moment in LRS
$\text{Skew} = (\eta_3 - 3\eta_2\mathbf{N} + 2\mathbf{N}^3)/SD(v_{ij})^3$	Skewness in LRS
$CV = SD(v_{ij})/\text{mean}$	Coefficient of variation

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544 **Table 2.** Mean transition matrices \mathbf{P}_e for environment e where the columns represent stage j at
545 time t and rows represent stage i at time $t + 1$. I = immatures; Y = yearlings; J = juveniles; NB =
546 nonbreeders; FB = failed breeders; B = successful breeders; q_x represents stage-specific mortality
547 calculated after rounding. Bolded transition probabilities represent the 3x3 submatrix \mathbf{P}_e used to
548 estimate matrix \mathbf{R}_e .

	I	Y	J	NB	FB	B
<i>Non-hurricane years</i>						
I	0	0	0	0	0	1
Y	1	0	0	0	0	0
J	0	0.948	0	0	0	0
NB	0	0	0.968	0.606	0.533	0.605
FB	0	0	0.003	0.042	0.070	0.039
B	0	0	0.006	0.294	0.310	0.311
q_x	0	0.052	0.023	0.058	0.087	0.045
<i>Hurricane Hugo</i>						
I	0	0	0	0	0	1
Y	1	0	0	0	0	0
J	0	0.975	0	0	0	0
NB	0	0	0.980	0.628	0.663	0.659
FB	0	0	0	0.071	0.163	0.069
B	0	0	0	0.265	0.163	0.262
q_x	0	0.025	0.020	0.036	<i>0.011</i>	<i>0.010</i>
<i>Hurricane Georges</i>						
I	0	0	0	0	0	1
Y	1	0	0	0	0	0
J	0	0.825	0	0	0	0
NB	0	0	0.971	0.583	0.643	0.694
FB	0	0	0	0.021	0.071	0.035
B	0	0	0	0.333	0.214	0.212
q_x	0	0.175	0.029	0.063	0.072	0.059
<i>Hurricane Maria</i>						
I	0	0	0	0	0	1
Y	1	0	0	0	0	0
J	0	0.963	0	0	0	0
NB	0	0	0.929	0.588	0.574	0.586
FB	0	0	0	0.110	0.019	0.080
B	0	0	0	0.227	0.241	0.310
q_x	0	0.037	0.071	0.075	0.166	0.024

549 **Note:** numbers in italics represent simulated mortality of 1% for NB and B stages.

550 **Table 3.** Mean, variance, skewness, and coefficient of variation of lifetime reproductive output. I
 551 = infants; Y = yearlings; J = juveniles; NB = nonbreeders; FB = failed breeders; B = successful
 552 breeders.

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	I	Y	J	NB	FB	B
<i>Non-hurricane years</i>						
Mean	5.035	5.036	5.310	5.430	5.282	6.522
Variance	35.288	35.288	35.754	35.911	35.714	36.011
Skew	2.062	2.062	2.027	2.015	2.030	2.007
CV	1.180	1.180	1.126	1.104	1.131	0.920
<i>Hurricane Hugo</i>						
Mean	9.040	9.040	9.271	9.457	9.586	10.696
Variance	102.617	102.617	103.099	103.407	103.581	103.703
Skew	2.032	2.032	2.019	2.011	2.006	2.002
CV	1.121	1.121	1.095	1.075	1.062	0.952
<i>Hurricane Georges</i>						
Mean	3.827	3.827	4.639	4.775	4.617	5.680
Variance	25.003	25.003	26.542	26.671	26.518	26.585
Skew	3.492	3.492	3.197	3.185	3.286	3.140
CV	1.307	1.307	1.111	1.082	1.115	0.908
<i>Hurricane Maria</i>						
Mean	3.134	3.134	3.255	3.502	3.227	4.804
Variance	17.153	17.153	17.420	17.878	17.361	18.270
Skew	2.177	2.177	2.138	2.072	2.147	2.014
CV	1.321	1.321	1.282	1.207	1.291	0.890

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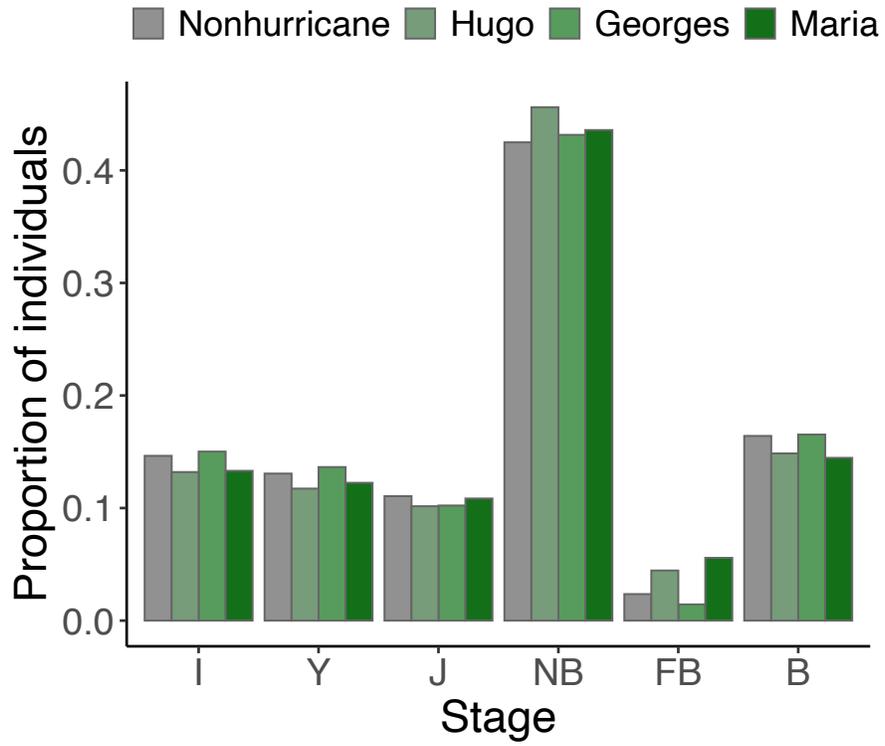
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562 **Figures**
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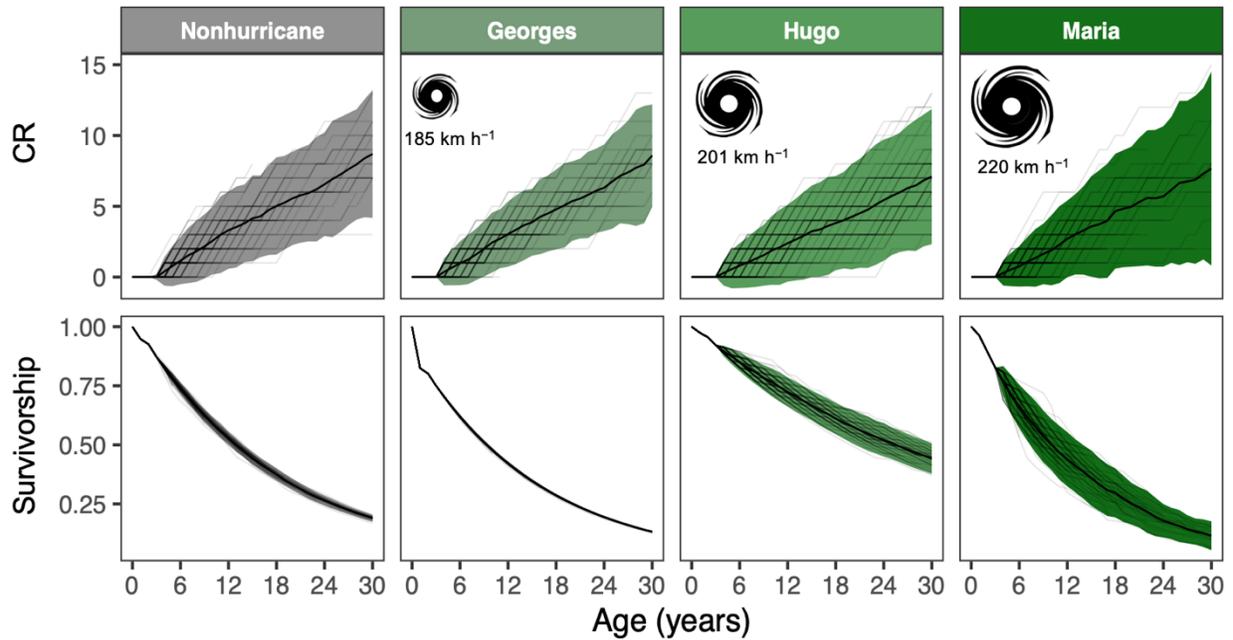


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566 **Figure 1.** Stable stage distribution during non-hurricane and hurricane years.

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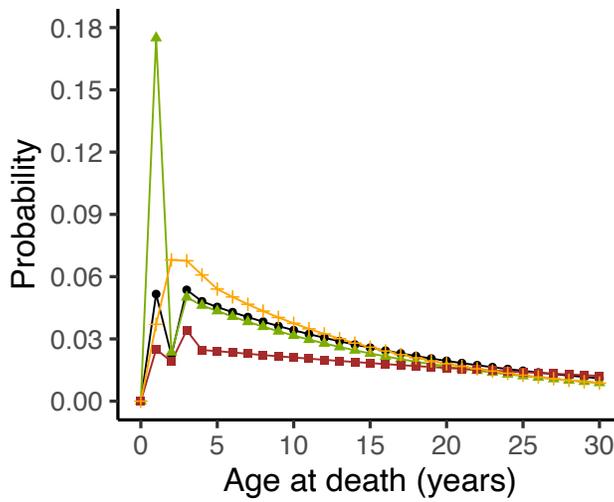
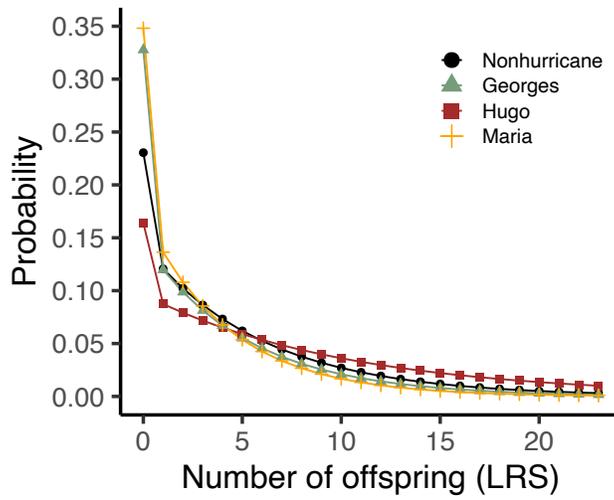


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570 **Figure 2.** Simulated cumulative reproduction (CR, top panel) and survivorship (bottom panel)
 571 for non-hurricane and hurricane environments. Hurricanes are presented in increasing order of
 572 intensity (sustained wind speed; Georges: $185 \text{ km} \cdot \text{h}^{-1}$; Hugo, $201 \text{ km} \cdot \text{h}^{-1}$; Maria $220 \text{ km} \cdot \text{h}^{-1}$).
 573 Lines represent mean trajectories (thick line) and individual trajectories (thin lines). Ribbons
 574 indicate 95% confidence intervals based on 1000 simulated trajectories.

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Figure 3. Probability distributions of lifetime reproductive success (LRS) and lifespan across environments.