

# 1 **Multiple coping strategies maintain stability of a small** 2 **mammal population in resource-restricted** 3 **environments**

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18 *Abstract* - In semi-arid environments, aperiodic rainfall pulses determine cycles of plant

19 production and resource availability for higher trophic levels, creating strong bottom-up

20 regulation. The influence of climatic factors on population vital rates often shapes the dynamics

21 of small mammal populations in such resource-restricted environments. Using a 21-year biannual

22 capture–recapture dataset (1993 to 2014), we examined the impacts of climatic factors on the

23 population dynamics of the brush mouse (*Peromyscus boylii*) in semi-arid oak woodland of

24 coastal-central California. We applied Pradel’s temporal symmetry model to estimate capture

25 probability ( $p$ ), apparent survival ( $\phi$ ), recruitment ( $f$ ), and realized population growth rate ( $\lambda$ ) of

26 the brush mouse, and examined the effects of temperature, rainfall, and El Niño on these

27 demographic parameters. The population was stable during the study period with a monthly

28 realized population growth rate of  $0.993 \pm \text{SE } 0.032$ , but growth varied over time from  $0.680 \pm$

29  $0.054$  to  $1.450 \pm 0.083$ . Monthly survival estimates averaged  $0.817 \pm 0.005$  and monthly

30 recruitment estimates averaged  $0.175 \pm 0.038$ . Survival probability and realized population

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31 growth were positively correlated with rainfall and negatively correlated with temperature. In  
32 contrast, recruitment was negatively correlated with rainfall and positively correlated with  
33 temperature. Brush mice maintained their population through multiple coping strategies,  
34 investing in high recruitment during warmer and drier periods and allocating more energy  
35 towards survival during cooler and wetter conditions. Although climatic change in coastal-central  
36 California will favor recruitment over survival, varying strategies may serve as a mechanism by  
37 which brush mice maintain resilience in the face of climate change. Our results indicate that  
38 rainfall and temperature are both important drivers of brush mouse population dynamics and will  
39 play a significant role in predicting the future viability of brush mice under a changing climate.

40

41 *Introduction* - Changes in abundance of small mammal populations can result from  
42 complex interactions among multiple factors, such as climate, plant production, food supply,  
43 vegetative cover, predation, and competition (Oli and Dobson 2003). The debate over the relative  
44 roles of endogenous (e.g., competition, predation) and exogenous (e.g., temperature, rainfall)  
45 factors in population dynamics has resulted in a general agreement that both influence population  
46 fluctuations (Turchin 2003). However, some studies in semi-arid systems show that exogenous  
47 factors outweigh endogenous factors in driving small mammal population dynamics and are key  
48 to understanding fluctuations (Gutierrez et al. 2010, Previtalli et al. 2009), particularly for  
49 evaluating the persistence of populations on the margins of a species' distribution (Gillespie et al.  
50 2008).

51 Semi-arid landscapes are highly variable in seasonal and annual rainfall patterns,  
52 typically with hot, dry summers and cool, wet winters. These systems are also resource-restricted,  
53 exhibiting pulse-like patterns of annual rainfall (averaging 25–50 cm) and large seasonal

54 fluctuations, with nearly all rainfall occurring in the fall and winter months (Peel et al. 2007).  
55 Fluctuations in rainfall may also be erratic, with some years of high and above-average  
56 precipitation (due to the El Niño effect in some areas), and other years of very little precipitation,  
57 leading to drought-like conditions (Previtali et al. 2009). In such systems, precipitation typically  
58 drives plant productivity, primarily through seed production and foliage growth (Brown and  
59 Ernest 2002, Heske et al. 1994, Meserve et al. 2003, Lima et al. 2002).

60 Many studies show that dramatic changes in precipitation have strong effects on  
61 population dynamics of small mammals, with rainfall pulses driving rodent dynamics indirectly  
62 through primary production, such as food availability and cover (Brown and Ernest 2002, Heske  
63 et al. 1994, Knapp et al. 2008, Meserve et al. 2003, Yates et al. 2002). Conversely, drought can  
64 have a detrimental effect on rodent population dynamics by reducing plant productivity (Brown  
65 and Ernest 2002, Meserve et al 2003), sometimes leading to population collapse (Facka et al.  
66 2010). However, we know relatively little about the effect of temperature on small mammal  
67 population dynamics in semi-arid climates, although some studies suggest that the effects of  
68 temperature are seasonal, showing a negative correlation with survival and recruitment in  
69 summer and a positive correlation in winter (Myers et al. 1985, Luis et al. 2010).

70 Due to their short life cycles, small mammals serve as ideal study systems for long-term,  
71 multi-generational studies. Short life spans and fast reproduction also translate to quick responses  
72 to changes in climatic conditions (Previtali et al. 2009). The brush mouse *Peromyscus boylii* has  
73 a wide distribution in the United States, occurring throughout much of the southwest and most of  
74 California, except for major grassland and desert areas (Baker 1968). This species is typically  
75 found in mature chaparral, oak woodland and hardwood conifer communities (Baker 1968). As  
76 its name suggests, the brush mouse prefers significant amounts of tree cover, dense and shrubby

77 vegetation, rock cover, and logs, which are important habitat structures that provide shelter from  
78 weather and predators, as well as nesting sites (Bradley and Schmidly 1999, Brehme et al. 2011,  
79 Gottesman et al. 2004). The brush mouse plays an important role in ecosystem function as a key  
80 prey species for the federally threatened Mexican spotted owl, *Strix occidentalis lucida* (Boyett  
81 2001). In the southwestern United States, the brush mouse is also a reservoir host for hantavirus  
82 (Abbott et al. 1999), making studies of population dynamics important for predicting rates of  
83 disease prevalence and spread. Anthropogenic impacts on brush mouse habitat include exurban  
84 development and increasing wildfire intensity (Brehme et al. 2011). From our knowledge, no  
85 study has yet provided estimates of apparent survival, recruitment, and realized population  
86 growth rate of the brush mouse, essential for understanding population dynamics of this species.

87         Our goal was to examine the effects of temperature, precipitation and El Niño on the  
88 population dynamics of the brush mouse in a coastal-central California mixed-oak-woodland. We  
89 applied temporal symmetry capture-mark-recapture (CMR) models (Pradel 1996; Nichols et al.  
90 2000; Williams et al. 2002) to a long-term (1993–2014) dataset to (1) estimate overall and  
91 seasonal patterns of capture probability, apparent survival, recruitment, and realized growth rate  
92 of the brush mouse in a coastal-central California mixed-oak woodland, (2) determine the  
93 relative contributions of survival and recruitment to population growth rate, and (3) explore the  
94 role of rainfall and temperature in explaining variations in population vital rates. Our results  
95 provide information on potential impacts of climate change on brush mouse population ecology,  
96 and provide important management information for the species.

97         We expected that brush mouse survival, recruitment, and realized population growth rate  
98 would exhibit seasonal fluctuations in response to the strongly seasonal patterns of rainfall on  
99 our study site. Specifically, we hypothesized that brush mouse population parameters would be

100 positively influenced by rainfall, because water is a limiting resource in semi-arid environments  
101 (Gutierrez et al. 2010, Previtali et al. 2009) and plant cover is an important habitat attribute for  
102 the brush mouse. We expected that temperature would negatively influence brush mouse vital  
103 rates due to higher energetic costs associated with foraging (Chen et al. 2015), although  
104 temperature could positively affect recruitment by creating more favorable conditions for  
105 reproduction (Andreo et al. 2009). We expected brush mouse vital rates to exhibit multi-annual  
106 fluctuations corresponding to El Niño, which affects rainfall pattern in California. Finally, we  
107 hypothesized that the population growth rate of the brush mouse would be influenced more by  
108 recruitment than by survival because of the fast life history of the species (Heppell et al. 2000;  
109 Oli and Dobson 2003; 2005; Oli 2004).

110         *Study Area* - We conducted the study at the Camp Roberts National Guard Post, a 17,000-  
111 ha military facility located in coastal-central California (Figure 1). Our study area was located in  
112 the backcountry of the Post, a roughly 4,000-ha matrix of undisturbed grassland, chaparral, and  
113 woodland. Climate of the study area is Mediterranean, with cool, wet winters and warm, dry  
114 summers. Annual rainfall is highly variable and is influenced by El Niño-La Niña oscillations.  
115 More than 95% of the rainfall typically falls between October and April. During the study, mean  
116 monthly rainfall during May to September was 0.33 cm (range: 0–4.75 cm, standard deviation:  
117 5.36 cm) compared to a monthly mean of 5.26 cm during October to April (range: 0–31.27 cm,  
118 standard deviation: 5.24 cm). The study area consisted of pure stands of blue oak (*Quercus*  
119 *douglasii*), or on the more mesic sites, a mix of blue oak and coast live oak (*Q. agrifolia*). The  
120 more mesic areas usually included a shrub layer of up to 35% cover (Tietje et al. 1997) and a  
121 ground layer of introduced Mediterranean annual grasses (*Avena* spp.) and forbs.

122            *Field Methods* - In summer 1993, we laid out nine square 5.8-ha plots in areas with at  
123 least 60% tree canopy cover. On each plot, we established with compass and tape a 17 x 17 grid  
124 with 15-m intersections. We marked each of the 289 intersections per plot with a stake and a  
125 survey flag with alphanumeric grid location. We trapped small mammals at each intersection in  
126 May and in October. From October 1993 to October 1996, we trapped for 5 nights each session  
127 for 7 trapping sessions (9 plots with 289 traps \* 5 nights \* 7 sessions = 91,035 trap nights). To  
128 increase the number of sampling grids on the study area, in winter of 1997 we established twelve  
129 1.1-ha plots with 8 x 8 trapping grid (Figure 1). Starting in May 1997, we trapped on these 12  
130 plots and on 8 x 8 corners (1.1 ha) or two diagonal corners of six of the 5.8-ha plots. From May  
131 1997 until May 2013, we sampled exclusively on these twenty-two 8 x 8 sampling grids for a  
132 total of 22 plots with 8 x 8 traps \* 3 nights \* 33 sessions = 139,392 trap nights. In October 2013,  
133 we trapped on 21 of the 22 plots for a total of 4,032 trap nights. Finally, in May 2014, we trapped  
134 on 9 of the 22 plots for a total of 1,728 trap nights. We will refer to the period between the  
135 October and May trapping sessions as a “season,” which is either a summer season (5-month  
136 period from May to September; the warm, dry season) or a winter season (the 7-month period  
137 from October to April; the cool, wet season).

138            During each May and October sampling session, we placed one Sherman live trap (3 x  
139 3.5 x 30-cm; H.B. Sherman Traps, Inc.) within 2 m of each grid intersection. To insulate trapped  
140 animals from overnight cold and from the heating of the interior of the trap by early morning  
141 sunshine, traps were placed in shade and covered with grass and other litter from the vicinity of  
142 the trap. We baited traps with a mixture of rolled oats, corn, and barley laced with molasses. On  
143 initial capture, we placed a laser-etched Monel 1005-1L1 animal tag and extra-long model XLK,  
144 7.7 x 9.5 x 30.5 cm trap (National Band and Tag Co., Newport, Kentucky) in the animal’s right

145 ear, and recorded trap location, tag number, species, sex, and age. Animals were released at site  
146 of capture. All handling of animals followed the guidelines of the University of California,  
147 Berkeley, Institutional Animal Care and Use Committee (UCB Permit # R-166). Trapping also  
148 met the guidelines of the American Society of Mammalogists (Sikes et al., 2011).

149       *Capture-mark-recapture (CMR) Analysis* - We used Pradel's (1996) temporal symmetry  
150 model to estimate apparent survival ( $\phi$ ), recapture ( $p$ ) probabilities, and realized population  
151 growth rate ( $\lambda$ ). First, we fitted a series of base models where we allowed  $\phi$ ,  $p$ , and  $\lambda$  to be  
152 affected by time (trapping session), year, season, and sex, and by the additive and interactive  
153 effects of these variables; we estimated model parameters using the most parsimonious model in  
154 the set (based on Akaike's information criterion corrected for small sample size AIC<sub>c</sub>; Burnham  
155 and Anderson 2002; Williams et al. 2002). After selecting the base model for each parameter, we  
156 tested for the effect of individual climatic covariates on that parameter using the base model.  
157 This second set of models allowed the model parameters to be affected by temperature, rainfall,  
158 and sea-surface temperature anomaly (a measure of El Niño Southern Oscillation). We used a  
159 similar modeling approach to test for the effects of climatic covariates on  $f$  using the  $\phi$ - $f$   
160 parameterization of Pradel's model (Williams et al. 2002). Since  $\lambda$  is based on the ratio of the  
161 population sizes between two sampling periods, the measured population size must represent the  
162 same sampled area. Otherwise, a bigger sampled area will yield a larger  $\lambda$  simply due to sampling  
163 a bigger area. Since sampling in our study area varied slightly over the study period and only  
164 remained constant from fall 1998 to spring 2013, we ran the  $\phi$ - $\lambda$  parametrization of Pradel's  
165 model to obtain  $\lambda$  estimates for fall 1998 to spring 2012. In addition, because the first estimate of  
166  $\lambda$  is typically inestimable in time-specific models, we omitted the first estimate of  $\lambda$ . We  
167 determined the relative contribution of  $\phi$  and  $f$  to  $\lambda$  by calculating the proportional contribution

168 parameter  $\gamma$  (Nichols and Hines 2002, Schorr 2012). If  $\gamma$  is greater than 0.5,  $\phi$  influences  $\lambda$  more  
169 than  $f$ . We did not consider spatial grid-to-grid variation and conducted a single analysis that  
170 combined data from all grids into one large population.

171 We performed all analyses with the program MARK (White and Burnham 1999) v. 6.2  
172 through RMark (Laake 2013) in program R v. 2.2.0 (R Core Team 2014). We determined the  
173 effect of climatic covariates by comparing  $AIC_c$  for models with and without a covariate, based  
174 on 95% confidence intervals for the slope parameter defining the relationship between a  
175 demographic parameter and the covariate(s).

176 *Climatic Covariates* - We extracted average daily temperature and precipitation for 1993–  
177 2015 from the Paso Robles City NACC weather station (NOAA 2016), located in Paso Robles,  
178 California, approximately 11 km southeast of the study area. We explored whether climatic  
179 conditions during the current season or the previous season (one-lag) affected California mouse  
180 vital rates (survival, recruitment, and growth). We used the following climatic variables: (i)  
181 average temperature (temp\_avg) (ii) coefficient of variation (CV) of temperature (temp\_cv), and  
182 CV of temperature with a one season lag (temp\_cv\_onelag), (iii) total seasonal rainfall  
183 (rain\_sum) and total seasonal rainfall with a one-season lag (rain\_sum\_onelag), and (iv) CV of  
184 rainfall (rain\_cv), and the CV of rainfall with a one-season lag (rain\_cv\_onelag). We reported the  
185 estimates of slope parameters ( $\beta$ ) based on the most parsimonious model that included a given  
186 covariate (temperature, rain) for each vital rate (survival, recruitment, growth), regardless of  
187 model structure for other demographic rates.

188 To examine effects of the El Niño Southern Oscillation (ENSO) on small mammal  
189 population dynamics, we used the Oceanic Niño Index (ONI), the standard used by NOAA to  
190 identify El Niño and La Niña events in the Pacific Ocean. An El Niño or La Niña is characterized

191 by five consecutive 3-month sea surface temperatures means above (for El Niño) or below (for  
 192 La Niña) a threshold of +0.5 °C (-0.5 °C), measured above the equatorial Pacific. We extracted  
 193 ONI values from the NOAA National Weather Service Climate Prediction Center  
 194 ([http://www.cpc.noaa.gov/products/analysis\\_monitoring/ensostuff/ensoyears.shtml](http://www.cpc.noaa.gov/products/analysis_monitoring/ensostuff/ensoyears.shtml)) and  
 195 information pertaining to the ENSO cycle from NOAA’s Climate website  
 196 (<https://www.climate.gov/enso>).

197

198 *Results* - During the study (1993–2014), we captured 3,258 (1,634 female and 1,624 male)  
 199 brush mice 6,351 times. Per 1,000 trap nights, the highest number of mice trapped per session  
 200 was 80.73 in October 1998, and the lowest number of mice trapped was 1.15 in October 1996  
 201 (Figure 2).

202 *Demographic parameters without covariate effects*—Pradel’s models indicated some  
 203 level of temporal variation in all demographic parameters: capture probability ( $p$ ), apparent  
 204 monthly survival ( $\phi$ ), apparent monthly recruitment ( $f$ ), and apparent monthly growth rate ( $\lambda$ ).  
 205 Average capture probability was higher for females than for males, and higher in summer than in  
 206 winter for both sexes (Table 1). Average  $\phi$  was  $0.789 \pm 0.005$ , and  $\phi$  was higher in winter ( $0.848$   
 207  $\pm 0.009$ ) than in summer ( $0.764 \pm 0.011$ ; Table 1). Estimates for  $\phi$  ranged from  $0.620 \pm 0.030$   
 208 (summer 1994) to  $0.946 \pm 0.034$  (winter 2005), and showed small fluctuations over the study  
 209 period, except when  $\phi$  decreased from 0.924 (winter 2006) to 0.528 (summer 2007), a 57%  
 210 decrease (Figure 3). Monthly recruitment ( $f$ ) varied substantially over time, ranging from  $0.029 \pm$   
 211  $0.047$  to  $0.538 \pm 0.063$  with an average of  $0.175 \pm 0.038$ , and was higher in summer ( $0.240 \pm$   
 212  $0.014$ ) than in winter ( $0.137 \pm 0.009$ ; Table 1). The largest estimated  $f$  ( $0.538 \pm 0.063$ ) occurred in  
 213 winter 1996, and other high  $f$  rates occurred in summer 2005 ( $0.511 \pm 0.097$ ) and winter 2011

214 (0.414 ± 0.107). Estimated recruitment was lowest in the winters of 2009 (0.029 ± 0.047), 2005  
 215 (0.050 ± 0.037), and 2010 (0.068 ± 0.086), and time-specific  $f$  parameters were un-estimable  
 216 (confidence intervals from 0 to 1) during three seasons (Figure 3).

217 Our study population was stable during the 21 years of study ( $\lambda = 0.993 \pm 0.032$ ,  $\phi$ - $f$   
 218 parametrization; Table 1), but estimated  $\lambda$  exhibited strong time-variation, particularly during  
 219 winter 1993 to summer 1997. The largest variation in estimated  $\lambda$  occurred when growth rate  
 220 increased from its lowest point during the 21-year period,  $0.672 \pm 0.052$  in summer 1996, to  $1.46$   
 221  $\pm 0.08$  in winter 1996 — a 118% increase (Figure 3). The most parsimonious model (based on  
 222 AIC<sub>c</sub>) included an additive effect of sex and season for  $p$ , and a time effect on  $f$ ,  $\phi$ , and  $\lambda$  (Table  
 223 2a). Estimates for the proportional contribution parameter ( $\gamma$ ) ranged from 0.515 to 0.969, with a  
 224 mean of 0.794 (SD = 0.097), and  $\gamma$  was consistently  $> 0.5$ .

225 *Individual climatic covariate effects on demographic parameters*—The most  
 226 parsimonious model for single covariate effects exhibited 92% of the AIC<sub>c</sub> weight and included  
 227 the effect of variation in rainfall on  $\phi$  and average temperature on  $f$  (Table 2b). Although the top  
 228 five models for  $\phi$  included only variation in rainfall, rainfall with a one season lag, and average  
 229 temperature, all other variables except El Niño had significant effects on  $\phi$ . Rainfall and variation  
 230 in rainfall with a one season lag had positive effects on  $\phi$ , while average temperature, variation in  
 231 temperature, variation in rainfall, and rainfall with a one season lag had negative effects on  $\phi$   
 232 (Table 3; Figure 4c, 4d).

233 Although the top five most parsimonious models for  $f$  included average temperature,  
 234 rainfall, and variation in rainfall with a one season lag, all other variables except El Niño had a  
 235 significant effect on  $f$ . Average temperature, variation in temperature, variation in rainfall, and  
 236 rainfall with a one season lag had a positive effect on recruitment, while rainfall, rainfall with a

237 two season lag, and variation in rainfall with a one season lag had negative effects on  $f$  (Table 3;  
238 Figure 4a, 4b). The effect of climatic covariates on  $\lambda$  was similar to the effect of climatic  
239 covariates on  $\phi$ . However, the overall strength of these effects was lower for  $\lambda$  than for  $\phi$ .  
240 Rainfall and variation in rainfall with a one season lag had a positive effect on  $\lambda$ , while average  
241 temperature and variation in temperature had a negative effect on  $\lambda$  (Table 3).

242

243 *Discussion - Effect of rainfall on  $\phi$ .*—Although the brush mouse population was stable  
244 throughout the study period, there was substantial temporal variation in survival and recruitment,  
245 and we posit that these fluctuations could be driven by climatic factors. Single-covariate models  
246 showed a positive effect of rainfall on  $\phi$ , and fluctuations in  $\phi$  were highly correlated with  
247 rainfall ( $r = 0.44, p = 0.005$ ). The effects of precipitation on small mammal population vital rates  
248 in semi-arid systems are generally well studied (Heske et al. 1994, Brown and Ernest 2002, Lima  
249 et al. 2002, Meserve et al. 2003, Letnic and Dickman 2005), and a widely accepted hypothesis  
250 posits that higher precipitation in semi-arid systems increases primary production, leading to an  
251 increase in survival of small mammal populations (Heske et al. 1994, Lima et al. 2002, Shenbrot  
252 and Krasnov 2001, Letnic and Dickman 2005, Yates et al 2002). Studies have associated  
253 increased rodent densities with higher precipitation in various habitats (Brown and Ernest 2002,  
254 Kuenzi et al. 2007, Meserve et al. 1995). Because the brush mouse is a shrub-habitat specialist,  
255 its survival will be especially affected by precipitation and its effect on primary production,  
256 which directly impacts food availability and refuge from predators (Bradley and Schmidly 1999,  
257 Baker 1968, Kalcounis-Ruppell and Millar 2002). As an omnivore, the brush mouse consumes  
258 fruits and seeds of a wide variety of plant species as well as insects, which are positively affected  
259 by rainfall in resource-restricted semi-arid environments (Fuentes and Campusano 1985, Yang et

260 al. 2011). The brush mouse also depends on closed and dense cover for foraging and refuge from  
261 weather and predators (Bradley and Schmidly 1999, Baker 1968, Kalcounis-Ruppell and Millar  
262 2002). Emphasizing the crucial importance of vegetative cover for this species, a wildfire that  
263 appreciably consumed brush cover led to a 90% decrease in brush mouse populations (Brehme et  
264 al. (2011). Rainfall typically increases the amount of understory shrub and chaparral cover,  
265 suggesting that the ability of the brush mouse to find food, survive in inclement weather, and  
266 evade predators will increase with higher precipitation and lead to higher survival rates,  
267 suggesting that abundance and access to food during high precipitation events could directly lead  
268 to an increase in survival. Survival is thus an important population driver for the brush mouse  
269 during periods of high rainfall.

270 *Effect of rainfall on f.*— In contrast to its effect on survival, covariate models showed a  
271 negative influence of rainfall on recruitment. In our study, rainfall may have had a negative effect  
272 on recruitment due to increased energetic costs incurred by reproductive females, or the  
273 destruction of food stores and nesting sites used by reproductive females during times of heavy  
274 precipitation. Although this is contrary to the general expectation that typically shows higher  
275 recruitment in response to increasing precipitation (e.g., see Previtalli et al. 2009, Shenbrot et al.  
276 2010, Thibault et al. 2010b), the effect of precipitation is not necessarily linear or simple.  
277 Extreme rainfall events can lead to catastrophic declines of small mammals, attributed to the  
278 destruction of food stores and nesting sites (Valone and Brown 1995, Thibault and Brown 2008),  
279 especially when located underground or just above the ground (shrubs or logs), habitats typically  
280 utilized by brush mice for nesting. High rainfall, especially during cold periods, can also cause  
281 abrupt declines in mouse populations (Calisher et al. 2005, Mills 2005, Llewellyn and Vessey  
282 1998, Garsd and Howard 1981, de Villafane and Bonaventura 1987, Mills et al. 1992). Such

283 unfavorable climatic conditions affect reproduction through mortality from direct exposure or  
284 lack of access to food or shelter. Furthermore, exposure to heavy rainfall can result in  
285 populations with smaller and fewer litters, possibly due to the higher energetic costs of  
286 microclimate regulation for the female (Myers et al. 1985). It is important to note that in our  
287 semi-arid study area, 96% of all precipitation occurred in the winter, and rainfall and temperature  
288 showed high negative correlation ( $r = -0.81$ ,  $p < 0.001$ ). This indicates that precipitation  
289 primarily occurred when it was colder, which could have had direct and indirect adverse effects  
290 on recruitment as recruitment was positively correlated with temperature. Although the effect of  
291 precipitation in winter could potentially be diluted by the lack of precipitation in summer,  
292 conversely, a small amount of rain in summer can potentially have a big effect because of such  
293 low amounts of precipitation in summer.

294 *Effect of temperature on  $\phi$  and  $f$ .*—Temperature was negatively correlated with  $\phi$ . We  
295 suspect this is because foraging and juvenile dispersal would become energetically costly and  
296 difficult to perform with higher temperatures, potentially leading to decreased  $\phi$  (Bradley and  
297 Schmidly 1999). Interestingly, average temperature had a significant positive effect on  $f$ . This  
298 could be attributed to two plausible reasons: (1)  $f$  could likely increase with warmer temperatures  
299 because brush mice can breed more frequently, especially during the winter (California  
300 Department of Fish and Wildlife 2014, Bradley and Schmidly 1999); (2) higher temperatures in  
301 spring and early summer increase primary productivity and allow female mice to expend less  
302 energetic costs for foraging and ensuring a favorable microclimate for litters (through increased  
303 vegetative cover for nests, which are typically found in trees and shrubs). Season and  
304 temperature are known to have interactive and sometimes contradictory effects on small mammal

305 populations (Luis et al. 2010), however our results did not support an interactive effect of season  
306 and temperature on brush mouse recruitment.

307 *Trade-offs between  $\phi$  and  $f$  in the maintenance of population stability.*— Brush mouse  
308 population growth rate ( $\lambda$ ) was stable throughout the study ( $\lambda = 0.993 \pm 0.032$ ). While  $\phi$  was  
309 positively affected by rainfall and negatively affected by temperature,  $f$  was positively affected  
310 by temperature and negatively affected by rainfall (Figure 4). These contrasting effects of  
311 climatic factors on vital rates suggests that the brush mouse employs alternating strategies in  
312 maintaining population stability dependent on climatic conditions. Additionally,  $\phi$  and  $f$   
313 estimates were highly negatively correlated ( $r = -0.61, p = 0.002$ ), implying a trade-off between  
314  $\phi$  and  $f$ . During hotter, drier seasons, the brush mouse invests in higher  $f$  as a means of  
315 maintaining the population, whereas during cooler, wetter seasons, the brush mouse allocates  
316 more energy towards  $\phi$  to maintain population  $\lambda$ . Considered together, the brush mouse appears  
317 to adopt a combination of coping mechanisms to ensure population stability.

318 The contribution of  $\phi$  to  $\lambda$  was consistently higher than  $f$ , as inferred from the  
319 proportional contribution parameter (Nichols and Hines 2002, Schorr 2012). The estimate of  $\gamma$   
320 averaged 0.82, i.e., on average, 82% of individuals in the current season are individuals that  
321 survived from the previous season. Although small mammals are typically r-selected, where  
322 population increase is fueled more by  $f$  than  $\phi$  (Heppell et al. 2000b, Oli and Dobson 2003), the  
323 brush mouse seems to be less r-selected than other small mammal species (Schorr 2012), due to  
324 the relatively higher contribution of  $\phi$  to  $\lambda$  for the species. This is interesting considering that  
325 brush mouse life expectancy is typically less than one year (Abbott et al. 1999). The relatively  
326 higher contribution of  $\phi$  towards  $\lambda$  was also supported by the overall climatic effects on vital  
327 rates. The effects of rainfall and temperature were similar for  $\phi$  and  $\lambda$  (a positive effect of rainfall,

328 and a negative effect of temperature on both parameters). However, these effects were reversed  
329 for  $f$  (a negative effect of rainfall, a positive effect of temperature). The relatively higher  
330 contribution of  $\phi$  towards  $\lambda$  than what is typically observed for small mammals is also supported  
331 by the fact that the brush mouse prioritizes  $\phi$  as a coping strategy in addition to  $f$ , and does not  
332 rely primarily on  $f$  to ensure population stability.

333       *Climate change.*—Climate change forecast models suggest that climatic conditions will  
334 become increasingly hotter and drier, with long dry periods punctuated by dramatic rainfall  
335 events (Masson-Delmotte et al. 2018). These changes in climatic conditions will have differing  
336 effects on brush mouse vital rates. Increasingly hotter and drier conditions may not bode well for  
337 brush mice survival, especially because dry conditions typically worsen the effect of increased  
338 temperatures. Our results suggest that brush mouse survival might decrease with climatic change.  
339 Recruitment, on the other hand, might increase with hotter and drier conditions and become the  
340 more dominant coping strategy in maintaining population stability. However, dramatic rainfall  
341 events might have a negative effect on brush mouse recruitment, suggesting that an increase in  
342 recruitment due to hotter and drier conditions might be tempered by more frequent, heavy  
343 rainfall. Brush mice are also particularly dependent on ground and vegetative cover, mesic  
344 environments and shrubs and trees for foraging and nesting (California Department of Fish and  
345 Wildlife 2014). Since primary productivity is negatively affected by an increase in dry  
346 conditions, climate change associated with more frequent and more intense droughts could be  
347 detrimental to structural habitats that are important for the species. Climate models also predict  
348 an increase in the frequency and magnitude of wildfires, especially in southern California and the  
349 western Sierras (Westerling and Bryant 2008), which are extremely detrimental to brush mouse  
350 populations. In addition to these, the invasive pathogen *Phytophthora ramorum*, which causes

351 Sudden Oak Death in tanoak and other oak species (McPherson et al. 2010), is one of the  
352 pathogens that may become more widespread and harder to control with changing climatic  
353 conditions (Brown and Allen-Diaz 2006). The spread of this pathogen could be especially  
354 problematic for brush mice, which rely on tanoak acorns (Reid et al. 2013) and coast live oak  
355 acorns (Kalcounis and Ruppell 2002), as a primary food source. However, the varying coping  
356 strategies employed by the brush mouse in response to varying environmental conditions may  
357 serve as a mechanism by which the species maintains resilience in the face of climate change.  
358

359         *Conclusions* - Our study was the first to apply Pradel's temporal symmetry models to  
360 brush mouse populations and demonstrated the implementation of a modern demographic  
361 modeling framework to quantify the effect of  $f$  and  $\phi$  on population  $\lambda$  for a brush mouse  
362 population. Although brush mouse population dynamics were influenced by localized climatic  
363 effects, the overall population size remained stable and appeared resilient to annual and multi-  
364 annual fluctuations. The brush mouse utilized a trade-off between survival and recruitment to  
365 maintain population stability, using high  $f$  during warmer, drier seasons and high  $\phi$  during cooler,  
366 wetter seasons to sustain population size. Our results suggest that brush mouse survival might  
367 decrease with climate change, while recruitment might increase with climate change and become  
368 the more dominant coping strategy in maintaining population stability. However, since dramatic  
369 rainfall events negatively affect brush mouse recruitment, an increase in recruitment due to hotter  
370 and drier conditions might be tempered by more frequent, heavy rainfall. The varying coping  
371 strategies employed by the brush mouse in response to varying environmental conditions may  
372 serve as a mechanism by which the species maintains resilience in the face of climate change.  
373 Brush mice play an important role in ecosystem processes, as they are extremely important prey

374 for a large variety of terrestrial and avian predators, including the federally threatened Mexican  
375 spotted owl (Boyett 2001). Brush mouse are also reservoir hosts for the hantavirus, raising  
376 important health concerns associated with rates of spread. Future investigations are needed to  
377 build on our findings and examine other aspects such as predation, intra-guild competition and  
378 the potential influence of spatial attributes on population dynamics, factors that, in addition to  
379 our results, will likely have implications for the management and conservation of the species.  
380 Furthermore, the trade-off between survival and recruitment could pose important consequences  
381 for broader small mammal population dynamics and understanding mechanisms of species  
382 resilience with climate change.

383

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385 collected the data; AYP, AS, MKO, JEH, and VR analyzed the data; AYP and WDT led the  
386 writing of the manuscript. All authors contributed critically to the drafts and gave final approval  
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574 **Table 1.** Overall, sex-specific and season-specific estimates of monthly apparent survival ( $\phi$ ),  
 575 capture probability ( $p$ ), monthly recruitment ( $f$ ), and realized monthly growth rate ( $\lambda$ ) without  
 576 covariate effects estimated using Pradel's model fitted to brush mice capture-mark-recapture  
 577 data. Overall, sex-specific and season-specific estimates of  $\phi$  were based on the third and fifth-  
 578 ranked model from Table 2a. Sex-specific and season-specific estimates of  $f$  were based on the  
 579 first, fourth, and twenty-fifth ranked model from Table 2a. Sex-specific and season-specific  
 580 estimates of  $\lambda$  were based on the first and third-ranked model from Supplementary Table S1.

	Survival ( $\phi$ )	Recruitment ( $f$ )	Growth Rate ( $\lambda$ )	Capture probability ( $p$ )
<b>Female</b>	0.815 $\pm$ 0.005	0.177 $\pm$ 0.038	0.993 $\pm$ 0.033	0.700 $\pm$ 0.055
<b>Male</b>	0.815 $\pm$ 0.005	0.173 $\pm$ 0.038	0.993 $\pm$ 0.033	0.468 $\pm$ 0.046
<b>Winter</b>	0.848 $\pm$ 0.009	0.137 $\pm$ 0.009	1.009 $\pm$ 0.010	0.526 $\pm$ 0.048
<b>Summer</b>	0.764 $\pm$ 0.011	0.240 $\pm$ 0.014	0.973 $\pm$ 0.014	0.641 $\pm$ 0.053
<b>Overall</b>	0.817 $\pm$ 0.005	0.175 $\pm$ 0.038	0.993 $\pm$ 0.032	0.584 $\pm$ 0.051

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594 **Table 2a-b.** Model selection results for the Pradel's model fitted to capture-mark-recapture data  
 595 for brush mice (*Peromyscus boylii*), testing for the effect of time (trapping session), year, season  
 596 (winter or summer), and sex (male or female). Parameters are:  $\phi$  = apparent survival probability;  
 597  $p$  = capture probability; and  $\lambda$  = realized population growth rate. The number of parameters ( $K$ ),  
 598 difference in Akaike's information criterion corrected for small sample size between a given  
 599 model and the top-ranked model ( $\Delta\text{AICc}$ ), and the relative model weight are also given. The five  
 600 best-supported models are presented. A plus sign (+) indicates additive and an asterisk (\*)  
 601 indicates both additive and interactive effects of the covariates involved. (A) Models for monthly  
 602 apparent survival ( $\phi$ ) and recruitment ( $f$ ) rates without covariate effects. (B) Models testing for  
 603 the singular effect of climatic covariates on  $\phi$  and  $f$ .

604 **Table 2a.**

Model	$K$	$\Delta\text{AICc}$	Weight
$\phi(\sim\text{time}) p(\sim\text{season} + \text{sex}) f(\sim\text{time})$	85	0	0.441
$\phi(\sim\text{time}) p(\sim\text{season} * \text{sex}) f(\sim\text{time})$	86	1.808	0.179
$\phi(\sim\text{time} + \text{sex}) p(\sim\text{season} + \text{sex}) f(\sim\text{time})$	86	2.182	0.148
$\phi(\sim\text{time} + \text{sex}) p(\sim\text{time}) f(\sim\text{time})$	86	3.772	0.067
$\phi(\sim\text{season}) p(\sim\text{season} * \text{sex}) f(\sim\text{time})$	87	3.8528	0.064

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606 **Table 2b.**

Model	$K$	$\Delta\text{AICc}$	Weight
$\phi(\sim\text{rain\_cv}) p(\sim\text{season} + \text{sex}) f(\sim\text{temp\_avg})$	46	0	0.921
$\phi(\sim\text{rain\_cv}) p(\sim\text{season} + \text{sex}) f(\sim\text{rain\_sum})$	46	5.977	0.046389
$\phi(\sim\text{rain\_sum\_onelag}) p(\sim\text{season} + \text{sex})$ $f(\sim\text{rain\_cv\_onelag})$	46	7.172	0.025523
$\phi(\sim\text{rain\_sum\_onelag}) p(\sim\text{season} + \text{sex})$ $f(\sim\text{temp\_avg})$	46	12.822	0.001514
$\phi(\sim\text{temp\_avg}) p(\sim\text{season} + \text{sex}) f(\sim\text{temp\_avg})$	46	13.057	0.001346

607 **Table 3.** The effect of climatic covariates on apparent survival ( $\phi$ ), recruitment ( $f$ ), and realized  
 608 population growth rate ( $\lambda$ ) of brush mice (*Peromyscus boylii*) in a coastal-central California  
 609 mixed-oak woodland. We report the estimate of slope parameters ( $\beta \pm SE$ ) based on the most  
 610 parsimonious model that included a given covariate for each demographic rate, regardless of  
 611 model structure for other demographic rates. Estimates in bold indicate that 95% CI for  $\beta$  do not  
 612 include zero.

<b>Demographic parameter</b>	<b>Climate covariate</b>	<b><math>\beta \pm SE</math></b>
Survival ( $\phi$ )	Temp_avg	<b>-0.12 <math>\pm</math> 0.05</b>
	Temp_cv	<b>-0.07 <math>\pm</math> 0.03</b>
	Rain_sum	<b>0.15 <math>\pm</math> 0.04</b>
	Rain_sum_onelag	<b>-0.13 <math>\pm</math> 0.03</b>
	Rain_cv	<b>-0.14 <math>\pm</math> 0.03</b>
	Rain_cv_onelag	<b>0.08 <math>\pm</math> 0.04</b>
	El Niño	-0.03 $\pm$ 0.02
<b>Demographic parameter</b>	<b>Climate covariate</b>	<b><math>\beta \pm SE</math></b>
Recruitment ( $f$ )	Temp_avg	<b>0.27 <math>\pm</math> 0.05</b>
	Temp_cv	<b>0.076 <math>\pm</math> 0.037</b>
	Rain_sum	<b>-0.17 <math>\pm</math> 0.045</b>
	Rain_sum_onelag	<b>0.193 <math>\pm</math> 0.031</b>
	Rain_cv	<b>0.147 <math>\pm</math> 0.034</b>
	Rain_cv_onelag	<b>-0.091 <math>\pm</math> 0.034</b>
	El Niño	-0.004 $\pm$ 0.02
<b>Demographic parameter</b>	<b>Climate covariate</b>	<b><math>\beta \pm SE</math></b>
Growth rate ( $\lambda$ )	Temp_avg	0.008 $\pm$ 0.009
	Temp_cv	-0.003 $\pm$ 0.003
	Rain_sum	<b>0.033 <math>\pm</math> 0.005</b>
	Rain_sum_onelag	<b>0.070 <math>\pm</math> 0.006</b>
	Rain_cv	<b>-0.014 <math>\pm</math> 0.004</b>
	Rain_cv_onelag	<b>-0.019 <math>\pm</math> 0.004</b>
	El Niño	0.009 $\pm$ 0.01

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615 **Figure 1.** Map of the study area. Trapping was carried out on nine 5.8-ha 17x17 trapping grids  
616 (open, larger squares) from 1993 to 1996, and on twenty two 1.1-ha 8 ×8 trapping grids (solid,  
617 smaller squares) from 1997 to 2014. Inset map of California (adapted from California wildlife  
618 habitat relationships system range maps, California Department of Fish and Wildlife, California  
619 Interagency Wildlife Task Group, 2018). Location of the study area (solid circle) and the  
620 distribution of the brush mouse (*Peromyscus boylii*; grey shading) in California, USA

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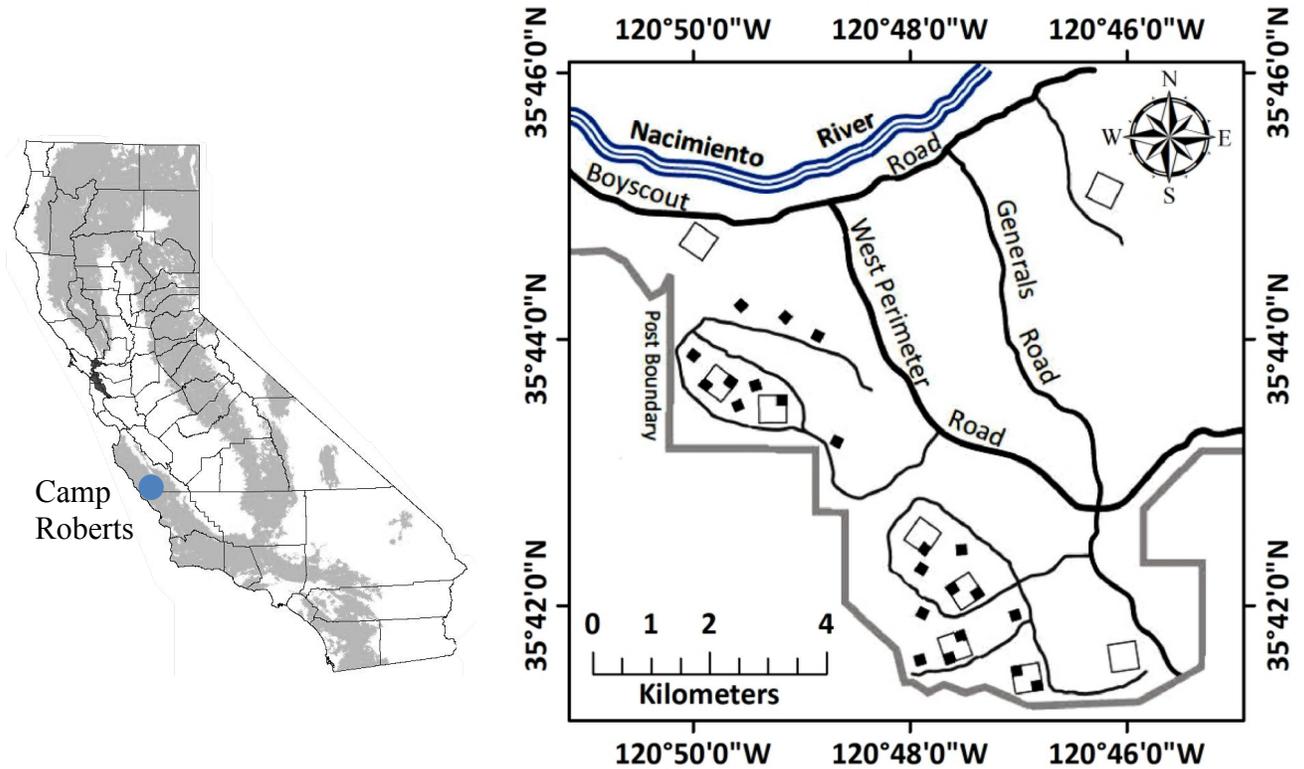
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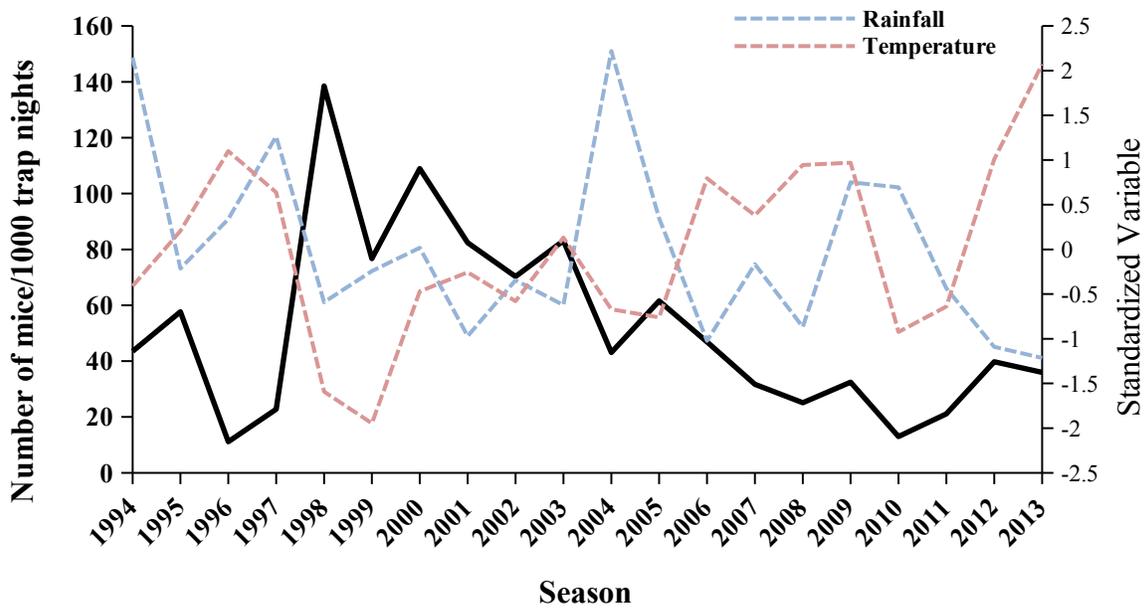
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640 **Figure 2.** The minimum number of brush mice/1000 trap nights captured during spring and fall  
 641 sessions on nine 5.8-ha study plots (17 x 17 sampling grid) during October 1993 to October  
 642 1996, on 22 1.1-ha study plots (8 x 8 sampling grid) during May 1997 to May 2014 for a total of  
 643 232,155 total trap nights at Camp Roberts, CA. The blue and orange dotted lines represent a  
 644 standardized rainfall and temperature index.

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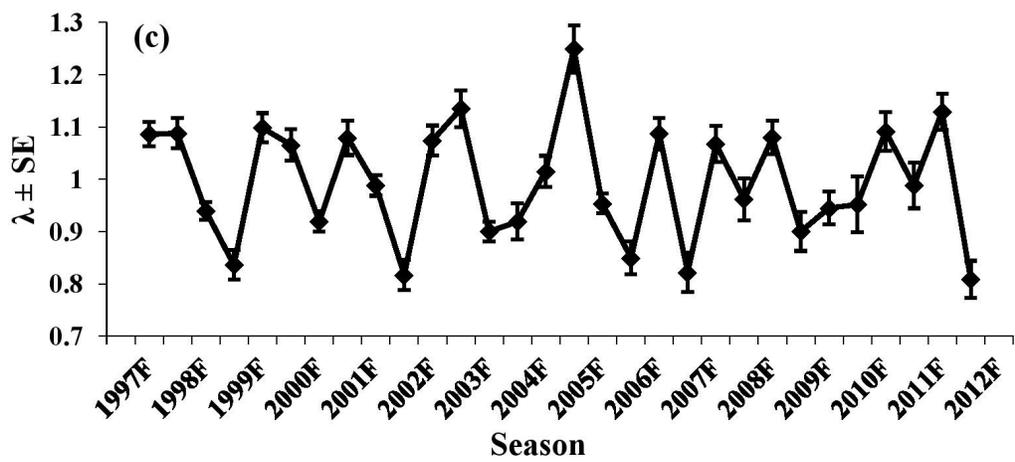
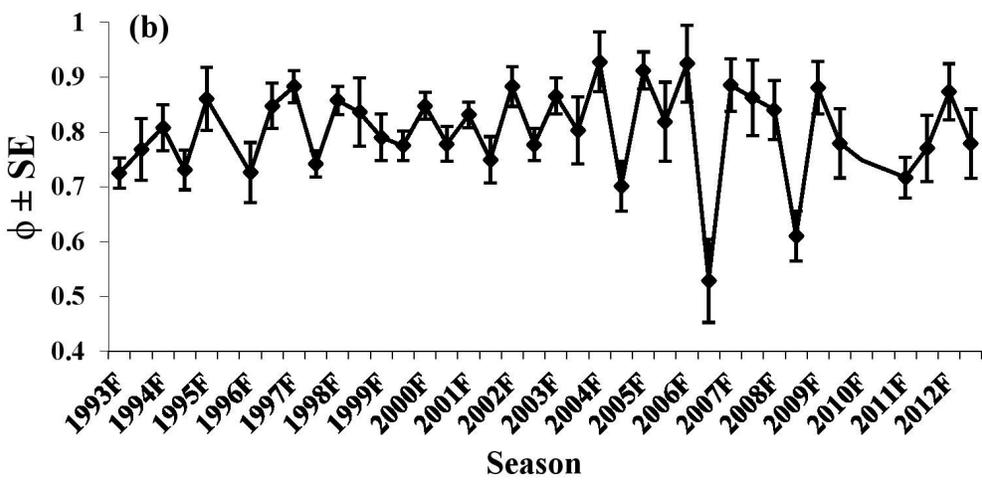
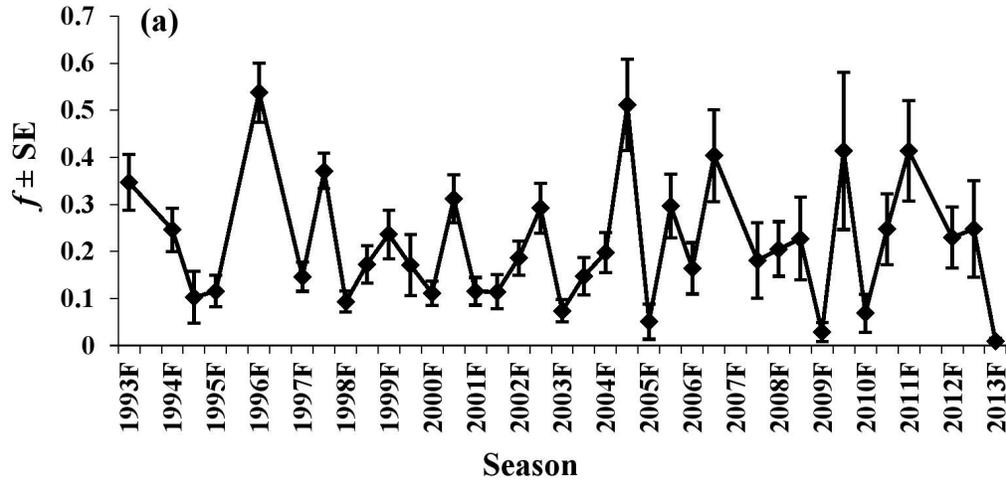
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656 **Figure 3a-c.** Estimates of (a) monthly apparent survival ( $\phi$ ) (b) recruitment rate ( $f$ ), and (c)  
657 realized population growth rate ( $\lambda$ ) of brush mice (*Peromyscus boylii*) at Camp Roberts,  
658 California, from winter 1993 to winter 2013 for  $\phi$  and  $f$ , and from summer 1997 to winter 2012  
659 for  $\lambda$ , based on the most parsimonious model (Table 2a, 2b). Sampled area varied over the course  
660 of the study, but was constant from winter 1997 to summer 2012. Thus, estimates for  $\lambda$  were  
661 produced for a shorter time interval because  $\lambda$  must be based on constant-size areas. The first  
662 estimate for  $\lambda$  is excluded because it is typically inestimable in time-specific models. Spring  
663 season estimates follow winter seasons and are not labeled on the axis.



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666 **Figure 4.** Effects of climatic variables (z-transformed) on survival and recruitment of brush mice  
 667 (*Peromyscus boylii*) at Camp Roberts, California based on the most parsimonious model that  
 668 included the given covariate for each demographic rate; (a) relationship between rainfall and  
 669 recruitment ( $f$ ); (b) relationship between temperature and recruitment ( $f$ ); (c) relationship  
 670 between rainfall and survival ( $\phi$ ); (d) relationship between temperature and survival ( $\phi$ ).

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