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Runaway selection for cooperation and strict-and-severe punishment

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10 **Abstract**

11 Punishing defectors is an important means of stabilizing cooperation. When levels of coopera-
12 tion and punishment are continuous, individuals must employ suitable social standards for
13 defining defectors and for determining punishment levels. Here we investigate the evolution
14 of a social reaction norm, or psychological response function, for determining the punishment
15 level meted out by individuals in dependence on the cooperation level exhibited by their
16 neighbors in a lattice-structured population. We find that (1) cooperation and punishment can
17 undergo runaway selection, with evolution towards enhanced cooperation and an ever more
18 demanding punishment reaction norm mutually reinforcing each other; (2) this mechanism
19 works best when punishment is strict, so that ambiguities in defining defectors are small; (3)
20 when the strictness of punishment can adapt jointly with the threshold and severity of pun-
21 ishment, evolution favors the strict-and-severe punishment of individuals who offer slightly
22 less than average cooperation levels; (4) strict-and-severe punishment naturally evolves and
23 leads to much enhanced cooperation when cooperation without punishment would be weak
24 and neither cooperation nor punishment are too costly; and (5) such evolutionary dynamics
25 enable the bootstrapping of cooperation and punishment, through which defectors who never
26 punish gradually and steadily evolve into cooperators who punish those they define as defec-
27 tors.

28 **Keywords:** evolution, strict-and-severe punishment, cooperation, lattice-structured population,
29 reaction norm, social norm, psychological response, bootstrapping

30 **1. Introduction**

31 Understanding the evolution of cooperation is one of the greatest challenges in evolutionary
32 biology and the social sciences. Even though several general mechanisms are widely recog-
33 nized to facilitate the emergence and maintenance of cooperation (as reviewed, e.g., by
34 Nowak, 2006), many questions of a more detailed nature are still unresolved. Kin selection
35 (Hamilton, 1964) explains the evolution of altruism among relatives. Direct reciprocity in re-
36 peated interactions (Axelrod and Hamilton, 1981) and indirect reciprocity enabled by
37 reputation dynamics (e.g., Nowak and Sigmund, 1998; Leimar and Hammerstein, 2001;
38 Panchanathan and Boyd, 2003; Brandt and Sigmund, 2004; Ohtsuki and Iwasa, 2004; Na-
39 kamaru and Kawata, 2004; Takahashi and Mashima, 2006) promote the evolution of
40 cooperation among non-relatives. Group selection (e.g., Sober and Wilson, 1998) and selec-
41 tion shaped by local interactions (e.g., Matsuda, 1987; Nowak and May, 1992; Nakamaru et
42 al., 1997, 1998; Le Galliard et al., 2003, 2005; Ohtsuki et al., 2006) may advance cooperation
43 in ways that can often be interpreted as generalizations of kin selection (Lehmann et al.,
44 2007a).

45 Cooperation is promoted by the punishment of defectors (Axelrod, 1986; Boyd and Rich-
46 erson, 1992; Clutton-Brock and Parker, 1995; Henrich and Boyd, 2001; Rockenbach and
47 Milinski, 2006; Sigmund, 2007), and so-called altruistic punishment occurs when the direct
48 costs of punishing are outweighed by the indirect benefits of such behavior (Yamagishi, 1986;
49 Gintis, 2000; Sigmund et al., 2001; Fehr and Gächter, 2002; Boyd et al., 2003; Fehr and
50 Rockenbach, 2003; Bowles and Gintis, 2004; Fehr and Fischbacher, 2004a; Gardner and
51 West, 2004; Shinada et al., 2004; Fowler, 2005; Nakamaru and Iwasa, 2005, 2006; Brandt et
52 al., 2006; Henrich et al., 2006; Eldakar et al., 2007; Hauert et al., 2007; Lehmann et al.,
53 2007b; Eldakar and Wilson, 2008).

54 In this study, we investigate the evolution of a social reaction norm, or psychological re-
55 sponse function, for punishment. This norm determines the threshold of encountered
56 cooperation below which individuals punish, how strictly they apply such a threshold, and
57 how severely they punish when they do so. In addition, we allow individuals to choose their
58 level of cooperation from a continuum of strategies (Doebeli and Knowlton, 1998; Roberts
59 and Sherratt, 1998; Wahl and Nowak, 1999a, 1999b; Killingback et al., 1999; Killingback and
60 Doebeli, 2002; Le Galliard et al., 2003, 2005; Doebeli et al., 2004). In this way, we examine
61 the joint evolution of four continuous strategies determining, respectively, the cooperation
62 level and the threshold, strictness, and severity of punishment. Among other questions, this
63 allows us to appraise the potential for selfish punishment and strong reciprocity: selfish pun-

ishers do not cooperate but nevertheless punish non-cooperators, whereas strong reciprocators cooperate and punish non-cooperators. Our analysis of joint evolution also allows us to compare our results with a preceding theoretical study suggesting that in a metapopulation setting the joint evolution of cooperation and punishment leads to the collapse of cooperation unless cooperation and punishment are perfectly linked traits (Lehmann et al., 2007b).

Viscous populations, exhibiting local interactions on a lattice or a more general social network, have been shown to promote the evolution of continuous cooperation strategies (Killingback et al., 1999; Le Galliard et al., 2003, 2005), as well as the joint evolution of discrete strategies of cooperation and punishment (Brandt et al., 2003; Nakamaru and Iwasa, 2005, 2006). Our study extends this earlier work to the joint and gradual evolution of continuous strategies of cooperation and punishment. In this wider context, we examine adaptable social reaction norms for punishment, analyzing their evolutionary determinants and consequences.

2. Methods

We consider populations in which individuals occupy sites, not all of which in turn have to be occupied by individuals. To identify the effects of viscous population structure, we compare two situations. In well-mixed populations, individuals interact with n other individuals chosen at random from the entire population. In lattice-structured populations, sites are located on a lattice, with each individual occupying a site and interacting with individuals on n neighboring sites. We used a square lattice with periodic boundary conditions, 30×30 sites, and the von Neumann neighborhood of $n = 4$ nearest neighbors.

Each individual i possesses four adaptive traits (c_i , $c_{0,i}$, $p_{0,i}$, and s_i) that can all take continuous non-negative values. The cooperation level c_i determines how much individual i invests into cooperation with its neighbors: selfish individuals invest nothing or only a small amount, whereas cooperators invest a high amount. The punishment threshold $c_{0,i}$ determines the cooperation levels c that individual i deems sufficient or cooperative ($c > c_{0,i}$), as opposed to insufficient or selfish ($c < c_{0,i}$). Accordingly, selfish individuals with whom individual i interacts are confronted with levels of punishment by individual i that increase as their cooperation levels decrease. The punishment severity $p_{0,i}$ determines the punishment level individual i metes out to individuals with a cooperation level of zero. The punishment strictness s_i determines how sharply punishment by individual i changes around $c_{0,i}$.

Each individual i interacts with other individuals j on n neighboring sites in two steps: the interacting individuals cooperate according to their cooperation strategies and then punish

97 according to their punishment strategies. The cooperation strategy of individual i is given by
 98 its cooperation level c_i . For each investment c_i , individual i pays the cooperation cost

$$99 \quad C_c(c_i) = a_c c_i^{e_c} , \quad (1)$$

100 with non-negative parameters a_c and e_c . For $e_c < 1$ this cost function is decelerating, for
 101 $e_c = 1$ it is linear, and for $e_c > 1$ it is accelerating.

102 The punishment strategy of individual i is given by its punishment reaction norm,

$$103 \quad p_i(c) = p_{0,i} \exp(-(c/c_{0,i})^{s_i}) , \quad (2a)$$

104 and depends on its punishment threshold $c_{0,i}$, punishment severity $p_{0,i}$, and punishment
 105 strictness s_i . This reaction norm describes the punishment level $p_i(c)$ with which individual
 106 i responds to a cooperation level c . When punishment strictness s_i is high, cooperation lev-
 107 els $c > c_{0,i}$ receive very little punishment, while cooperation levels $c < c_{0,i}$ elicit almost the
 108 maximal punishment level $p_{0,i}$. When punishment strictness s_i is low, the punishment level
 109 still monotonically decreases as the cooperation level increases, but the transition to low pun-
 110 ishment is shallower around $c_{0,i}$. For testing the robustness of our results, we also considered
 111 two alternative parameterizations of punishment reaction norms,

$$112 \quad p_i(c) = p_{0,i}(1 - c/c_{0,i})^{1/s_i} \text{ if } c < c_{0,i} \text{ and } p_i(c) = 0 \text{ otherwise,} \quad (2b)$$

$$113 \quad p_i(c) = p_{0,i} / [1 - \exp(-s_i) + \exp(s_i(c/c_{0,i} - 1))] . \quad (2c)$$

114 In our model, punishment is costly. For each punishment level p_i , individual i pays the pun-
 115 ishment cost

$$116 \quad C_p(p_i) = a_p p_i^{e_p} , \quad (3)$$

117 with non-negative parameters a_p and e_p . For $e_p < 1$ this cost function is decelerating, for
 118 $e_p = 1$ it is linear, and for $e_p > 1$ it is accelerating.

119 The birth rate of individual i ,

$$120 \quad b_i = b_0 + \frac{1}{n} \sum_j c_j , \quad (4a)$$

121 is given by the intrinsic birth rate b_0 increased by the average cooperative investment individ-
 122 ual i receives from its neighboring sites (the sums in Eqs. (4) extend over all individuals j
 123 with whom individual i interacts, and thus naturally exclude empty sites in the neighborhood
 124 of individual i). The resultant offspring is placed at a randomly chosen site with which indi-
 125 vidual i is interacting, and is lost if that site is already occupied. Similarly, the death rate of
 126 individual i ,

127
$$d_i = d_0 + \frac{1}{n} \sum_j [p_j(c_i) + C_p(p_i(c_j)) + C_c(c_i)] . \quad (4b)$$

128 is given by the intrinsic death rate d_0 increased by the average punishment individual i re-
 129 ceives and by the average costs for punishment and cooperation individual i incurs.

130 Birth and death events occur asynchronously across the population and stochastically in
 131 time. After each such event, the waiting time until the next event is drawn from an exponen-
 132 tial distribution with mean $1/E$ with $E = B + D$, where B and D , respectively, are the
 133 current sums of all birth and death rates in the population. The event type is then chosen ac-
 134 cording to probabilities B/E and D/E , and the individual i undergoing the event is chosen
 135 according to probabilities b_i/B or d_i/D .

136 When an offspring is born, its traits may be mutated relative to those of its parent. For
 137 each trait, a mutation occurs with probability m . Mutated trait values are normally distributed
 138 around the corresponding parental trait values, with standard deviations σ_c for the traits c ,
 139 c_0 , and p_0 , and with standard deviation σ_s for the trait s . Mutated values of the traits c , c_0 ,
 140 p_0 , and s are constrained to minimal values 0 , 10^{-5} , 0 , and 0 , respectively. These boundaries
 141 are absorbing for c , c_0 , and p_0 , and reflective for s .

142 For testing the robustness of our results, we also considered errors in the implementation
 143 and perception of cooperation levels. With implementation errors, an implemented coopera-
 144 tion level differs from the actually intended cooperation level with a small error probability
 145 and with the difference being drawn from a normal distribution with a small standard devia-
 146 tion. With perception errors, a perceived cooperation level differs from the actually
 147 implemented cooperation level analogously.

148 **3. Results**

149 Fig. 1 shows how our model leads to runaway selection for costly cooperation and punish-
 150 ment in lattice-structured populations. Here punishment strictness s is not yet freely
 151 evolving, but instead is kept fixed at one and the same value for all individuals in the popula-
 152 tion. Evolution starts in the absence of any cooperation ($c = 0$) and of any punishment
 153 ($p_0 = 0$). All individuals are initially recognized as defectors ($c_0 = 10^{-5} > c$). In general, run-
 154 away selection among quantitative traits occurs when continual feedback between selection
 155 pressures and resultant evolutionary changes in the traits gradually leads to ever more extreme
 156 trait values. In our model, runaway selection occurs among the cooperation level c , the pun-
 157 ishment threshold c_0 , and the punishment severity p_0 , which are all increasing
 158 concomitantly. We see that the larger s is chosen, i.e., the stricter individuals apply their pun-
 159 ishment threshold c_0 , the faster these three traits evolve towards higher values. The

160 population's average cooperation level c always evolves to be slightly larger than the average
161 punishment threshold c_0 , so that most individuals are recognized as cooperators by most other
162 individuals. Cooperation levels are driven up by evolutionary increases in punishment thresh-
163 olds and vice versa. In other words, as the population evolves to become increasingly
164 cooperative, the social demands on individuals to be recognized as cooperators rise concomi-
165 tantly. Also the punishment severity p_0 increases with the punishment strictness s . The
166 speed of runaway selection thus increases with punishment strictness. Hence, stricter punish-
167 ment indirectly favors both more severe punishment and higher cooperation levels.

168 Fig. 2 shows what happens when punishment strictness s is allowed to evolve together
169 with the three other adaptive traits c , c_0 , and p_0 . Again, evolution starts in the absence of
170 any cooperation and of any punishment. In addition, individuals are assumed to be initially
171 indiscriminating ($s = 0$). When the evolution of s is sufficiently fast (i.e., when σ_s is suffi-
172 ciently large compared to σ_c), punishment strictness rises together with all other adaptive
173 traits, resulting in a cooperative regime with strict-and-severe punishment. As in Fig. 1, the
174 social requirements for avoiding punishment escalate with increasing cooperation. By con-
175 trast, when evolution of s starts out from 0 but is too slow, punishment strictness remains
176 low. Individuals thus continue to be indiscriminating, and runaway selection for cooperation
177 and punishment cannot occur (results not shown). However, even when evolution of s is
178 slow, a sufficiently high initial value of s reinstates the phenomenon of runaway selection, in
179 line with the results already documented in Fig. 1.

180 Fig. 3 shows a systematic evaluation of the consequences of cooperation costs and pun-
181 ishment costs for the joint evolution of cooperation and punishment. Without punishment
182 (i.e., for p_0 fixed at 0), cooperation evolves only when cooperation costs are sufficiently de-
183 celerating (Fig. 3a). Even then, resultant cooperation levels remain relatively low. Evolving
184 punishment, by contrast, can lead to much higher levels of cooperation. This occurs when
185 punishment costs are decelerating or linear and cooperation costs are roughly linear (Fig. 3b).
186 A look at the three traits determining the punishment strategy (Figs. 3c to 3e) confirms that
187 these high levels of cooperation are enabled by the evolution of strict-and-severe punishment:
188 the average punishment threshold (Fig. 3c) is again just slightly lower than the average coop-
189 eration level (Fig. 3b), the average punishment severity is high (Fig. 3d), and the average
190 punishment strictness is also high (Fig. 3e).

191 We can categorize and understand these outcomes in terms of four cost scenarios. First,
192 when cooperation is too cheap (i.e., cooperation costs are decelerating and e_c is lower than
193 about 0.5), the population's lattice structure alone is sufficient for promoting cooperation, so
194 that costly punishment is not favored. Second, when cooperation is too expensive (i.e., coop-

195 eration costs are accelerating and e_c is higher than about 1.25), cooperation evolution is hin-
196 dered by these costs, independently of the costs of punishment. Third, when punishment is too
197 expensive (i.e., punishment costs are accelerating and e_p is higher than about 1.25), punish-
198 ment evolution is hindered by these costs and no enhanced cooperation can thus occur.
199 Fourth, when punishment is not too expensive (i.e., punishment costs are linear or decelerat-
200 ing so that e_p is lower than about 1.25) and cooperation is neither too cheap nor too
201 expensive (i.e., cooperation costs are roughly linear so that e_c lies between about 0.5 and
202 1.25), runaway selection for cooperation and punishment occurs and results in greatly en-
203 hanced cooperation.

204 To test the robustness of our results, we changed the intrinsic birth and death rates, b_0 and
205 d_0 , without observing any qualitative differences. The patterns reported above also remain
206 intact when we use the alternative parameterizations of punishment reaction norms in Eqs.
207 (2b) and (2c), instead of the one in Eq. (2a). Also the introduction of implementation and per-
208 ception errors did not lead to any qualitative changes in the observed evolutionary dynamics.
209 When increasing the mutation probability and the mutational standard deviations, we could
210 confirm earlier results by Le Galliard et al. (2003) that showed how such changes in the muta-
211 tion process facilitate the evolution of continuous cooperation strategies.

212 For well-mixed populations, the joint evolution of costly cooperation and punishment
213 never occurs, as can be shown analytically (see appendix) and corroborated by individual-
214 based simulations. This result can be understood intuitively: since punishing is costly to the
215 punisher, and since in well-mixed populations this cost is the only selection pressure acting on
216 punishment severity (see appendix), punishment – and, in its wake, cooperation – are invaria-
217 bly eliminated from well-mixed populations.

218 **4. Discussion**

219 Here we have shown that the joint and gradual evolution of cooperation and punishment can
220 greatly promote cooperation levels in lattice-structured populations, even when cooperation
221 and punishment are entirely absent initially. This promotion is driven by runaway selection,
222 through which cooperation level, punishment threshold, and punishment severity rise con-
223 comitantly. The pace of the runaway process increases with punishment strictness. When
224 punishment strictness is allowed to evolve, evolution often leads to strict-and-severe punish-
225 ment accompanied by high cooperation levels. This process is again driven by runaway
226 selection, now for all four traits. The enhancement of cooperation levels through the evolution
227 of strict-and-severe punishment is largest when neither cooperation nor punishment are too
228 costly and cooperation levels in the absence of punishment would be low. Our results explain

229 the bootstrapping of cooperation and punishment, in the sense that defectors who rarely or
230 only indiscriminately punish gradually and steadily evolve into cooperators who strictly and
231 severely punish those they define as defectors.

232 The evolutionary mechanisms underlying these findings can be understood in intuitive
233 terms. In general, any process of runaway selection requires positive feedback between selec-
234 tion pressures and resultant evolutionary changes in one trait and selection pressures and
235 resultant evolutionary changes in another trait. In our model, such mutual reinforcement can
236 occur among all four evolving traits, as we have schematically summarized in Fig. 4. We start
237 our explanation by recalling that lattice-structured populations enable the evolution of low
238 levels of cooperation even in the absence of punishment (arrow a in Fig. 4). When punishment
239 strictness is small but does not vanish completely, these cooperation levels favor increased
240 punishment severity (arrow b). Under these conditions, punishment locally reduces the fre-
241 quency of individuals with relatively low cooperation level, by differentially burdening them
242 with a fitness disadvantage. Consequently, any region on the lattice in which punishment se-
243 verity slightly differs from zero can expand into adjacent regions with vanishing punishment
244 severity. Increased punishment severity then favors increased cooperation levels (arrow c),
245 since these are advantageous when punishment reduces the exposure of more cooperative in-
246 dividuals to exploitation by less cooperative individuals. In turn, increased cooperation levels
247 again favor increased punishment severity (arrow b), since this maintains the relative impact
248 of punishment on fitness after cooperation levels have risen. Increased cooperation levels also
249 favor increased punishment thresholds (arrow d), since this maintains the discriminating of
250 individuals with relatively low cooperation levels after cooperation levels have risen. In turn,
251 increased punishment thresholds favor increased cooperation levels (arrow e), since individu-
252 als must then cooperate more to escape punishment. Under these conditions, selection favors
253 an increase in punishment strictness (arrow f), since this enables a better targeting of punish-
254 ment to individuals with relatively low cooperation levels. In turn, stricter punishment
255 strengthens the already described selection pressures on cooperation level, punishment
256 threshold, and punishment severity (arrow g), since stricter punishment selects for enhanced
257 cooperation and tougher punishment.

258 These explanations help us to appreciate why runaway selection for cooperation and
259 strict-and-severe punishment does not occur for all parameter values and initial conditions
260 considered in our analysis. First, when the costs of cooperation or punishment are too high
261 (upper and right regions in Figs. 3b to 3e), the selection pressures described above (arrows b
262 to e in Fig. 4) are counteracted by those directly resulting from the costs, thus stalling the run-
263 away process at low levels of cooperation and punishment. Second, when cooperation levels

264 are high already in the absence of punishment (left regions in Figs. 3b to 3e), the relative ad-
265 vantages of punishment, and therefore the corresponding selection pressures on punishment
266 (arrows b and d in Fig. 4), are low, thus stalling punishment evolution at low levels. Third, the
267 initial selection pressure on punishment severity (arrow b in Fig. 4) occurs unless punishment
268 is totally absent from the initial population. For the punishment reaction norms in Eq. (2b) the
269 initial punishment threshold must thus exceed the initial cooperation level, since otherwise no
270 punishment occurs at all. Fourth, for selection to favor stricter and severer punishment (ar-
271 rows b, d, and f in Fig. 4), more cooperation has to result in less punishment, which implies
272 that the punishment reaction norm must be a decreasing function. A vanishing punishment
273 severity translates into a flat punishment reaction norm (Eqs. 2), which prevents the runaway
274 process from taking off. Conversely, this explains why increased punishment strictness accel-
275 erates the runaway process of the three other traits (Fig. 1) and why rapidly evolving
276 punishment strictness facilitates the runaway process of all four traits (Fig. 2).

277 Our representation of cooperation and punishment strategies as continuous quantitative
278 traits and the consideration of their gradual evolutionary dynamics play an important role for
279 the findings reported here. In particular, the evolutionary mechanisms underlying the runaway
280 process cause the steady and gradual adjustment of trait values driven by the subtle mutual
281 reinforcement of selection pressures. In contrast, large sudden increases in punishment
282 threshold or severity might not be selectively advantageous, since the resultant costs may
283 outweigh the resultant benefits. Likewise, large sudden increases in cooperation levels are
284 unlikely to be favored, since these would not be backed up by a corresponding orchestration
285 of the punishment strategy. This highlights why cooperation games with continuous strategies
286 and gradual trait evolution can reveal qualitative phenomena, such as the runaway selection
287 for cooperation and strict-and-severe punishment reported here, that might be fundamentally
288 obscured in corresponding games with discrete strategies.

289 Our results provide an evolutionary explanation for the widely observed appreciation of
290 “strict but fair” punishment. This common cultural predisposition is an integral part of many
291 moral systems and legal codes, and is often touted as a highly effective approach to education,
292 reeducation, military discipline, and the preservation of public order. Strict-and-severe pun-
293 ishment is closely related to the “zero tolerance” approach to law enforcement, by which
294 already small infractions of accepted rules are subjected to significant punishment. In our
295 model, these ethical considerations have their counterpart in the emergence of high punish-
296 ment strictness, elevated punishment severity, and of punishment thresholds finely tuned to
297 majority behavior. In fact, our results presented in Figs. 1 to 3 make it clear that effective pun-
298 ishment must operate on shifting baselines, with the criterion for punishment being

299 continually refined as majority behavior evolves. Like in many other models of cooperation
300 and punishment, these outcomes arise, gradually and naturally, from evolutionary dynamics
301 solely driven by the selfish interests of individuals.

302 Based on these insights, we can revisit two conditions that could be perceived as limiting
303 the bootstrapping of cooperation and punishment in our model. We had already explained
304 above why runaway selection is hindered by vanishing initial punishment strictness, and,
305 while punishment strictness is still low, by its low evolutionary rate. Notice that these obser-
306 vations only apply when punishment strictness is zero or very low initially. We can now
307 question whether that would indeed be a realistic assumption. At least in humans, it seems fair
308 to assume, instead, that innate or cultural circumstances are causing punishment strictness to
309 start out from some intermediate level, even when punishment severity and punishment
310 threshold start out from zero. Our results and explanations above make it clear that, under
311 such conditions, runaway selection for cooperation and strict-and-severe punishment is
312 greatly facilitated.

313 Here we have studied situations in which the punishment that individuals mete out simply
314 depends on the cooperation levels of the individuals they interact with. Yet, punishment re-
315 sponses may be affected by many other factors. For example, breaking a social norm that is
316 widely shared among members of a group may invite punishment (Gintis, 2000; Fehr and Fis-
317 chbacher, 2004b), an effect that may be superimposed on the punishment responses
318 considered here. Also emotions can influence punishment behavior, and may compel indi-
319 viduals to punish cheaters even when the cost of punishment exceeds that of being cheated
320 (Frank, 1988; Xiao and Houser, 2005). Considering the effects of reputation or gossip on run-
321 away selection for cooperation and punishment will also be of interest, since reducing an
322 individual's reputation can serve as a cost-free means of punishment (Nakamaru and Kawata,
323 2004). Similarly, it will be worthwhile taking a closer look at conditions and mechanisms that
324 can eventually stop the runaway process investigated here. This could involve cost functions
325 that are decelerating for low investments and accelerating for high investments, diminishing
326 fitness returns from received investments, or an explicit modeling of the availability of re-
327 sources that individuals exchange when they cooperate or punish.

328 The evolutionary framework we have utilized here recognizes three levels of interlocking
329 dynamics, ranging from the demographic dynamics of individuals in a population, to the be-
330 havioral dynamics of cooperation and punishment in the interactions between individuals, and
331 to the psychological dynamics underlying the identification of cheaters. Naturally, psycho-
332 logical dynamics affect behavioral dynamics, which in turn affect demographic dynamics.
333 Conversely, demographic dynamics affect behavioral and psychological dynamics by chang-

334 ing the selection pressures that cause adjustments in the traits governing behavior and psy-
 335 chology. Experimental tools and modeling approaches for studying such feedbacks have
 336 emerged over the past decades and are now increasingly applied to tackling questions in co-
 337 operation research (e.g., de Quervain et al., 2004; Enquist and Ghirlanda, 2005). We hope that
 338 the framework and results put forward here may further inspire and facilitate such studies. In
 339 a similar vein, our approach could be used to address questions raised by evolutionary psy-
 340 chologists who have challenged conjectured adaptive explanations of behavior and
 341 psychological predispositions regarding mate choice, emotion, cheater detection, and the abil-
 342 ity to recognize spatial locations (e.g., Bawkow et al., 1992). While such explanations are
 343 often based on verbal and qualitative reasoning, the approach adopted here allows for formal
 344 and quantitative reasoning.

345 It is our hope that, from a methodological perspective, our evolutionary explanation of
 346 runaway selection for cooperation and strict-and-severe punishment might be no more than a
 347 start. We believe that, more in general, studies of cooperation have much to gain from inves-
 348 tigating models with joint evolution of multiple continuous traits, explicit dynamics for
 349 demography and trait changes, and interpretation of traits in terms of reaction norms for psy-
 350 chological and behavioral processes.

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359 **Appendix**

360 In this appendix we show that cooperation and punishment cannot evolve in well-mixed
 361 populations. For this purpose we investigate the dynamics of a rare variant strategy with fre-
 362 quency $x' \approx 0$, cooperation level c' , and punishment reaction norm p' in the population of a
 363 resident strategy with frequency x , cooperation level c , and punishment reaction norm p ,

$$364 \quad \frac{1}{x'} \frac{dx'}{dt} = (b_0 + cx)(1 - x) - \{d_0 + [p(c') + C_p(p'(c)) + C_c(c')]x\} .$$

365 We assume that the resident population is at its equilibrium frequency $0 \leq \hat{x} \leq 1$, so that
366 $(b_0 + c\hat{x})(1 - \hat{x}) = d_0 + [p(c) + C_p(p(c)) + C_c(c)]\hat{x}$, from which we obtain
367 $\hat{x} = \frac{1}{2c}[\sqrt{l^2 + 4c(b_0 - d_0)} - l]$ with $l = b_0 - c + p(c) + C_p(p(c)) + C_c(c)$. Denoting the variant's
368 per capita growth rate or fitness $(dx'/dt)/x'$ by f' (e.g., Metz et al., 1992), the selection
369 pressures g_c , g_{c_0} , g_{p_0} , and g_s on the resident's adaptive traits c , c_0 , p_0 , and s are given by
370 the derivatives df'/dc' , df'/dc'_0 , df'/dp'_0 , and df'/ds' evaluated at $c' = c$ and $p' = p$ (e.g.,
371 Dieckmann and Law, 1996; Geritz et al., 1997). Using Eqs. (1), (2a), and (3), this gives

$$\begin{aligned}
g_c &= \hat{x}c^{-1}[sc^s c_0^{-s} p(c) - e_c C_c(c)], \\
g_{c_0} &= -\hat{x}sc^s c_0^{-(s+1)} e_p C_p(p(c)), \\
g_{p_0} &= -\hat{x}p_0^{-1} e_p C_p(p(c)), \\
g_s &= \hat{x}c^s c_0^{-s} e_p C_p(p(c)) \ln(c/c_0).
\end{aligned}$$

373 Since g_{p_0} is negative, evolution will always diminish punishment severity p_0 in well-mixed
374 populations. Once p_0 has evolved to 0, selection on c_0 and s ceases: $C_p(0) = 0$ and thus
375 $g_{c_0} = 0$ and $g_s = 0$. The selection pressure on c is negative for $p_0 = 0$, $g_c = -\hat{x}c^{-1}e_c C_c(c)$, so
376 that, driven by the cost of cooperation, the cooperation level c will also evolve to 0.

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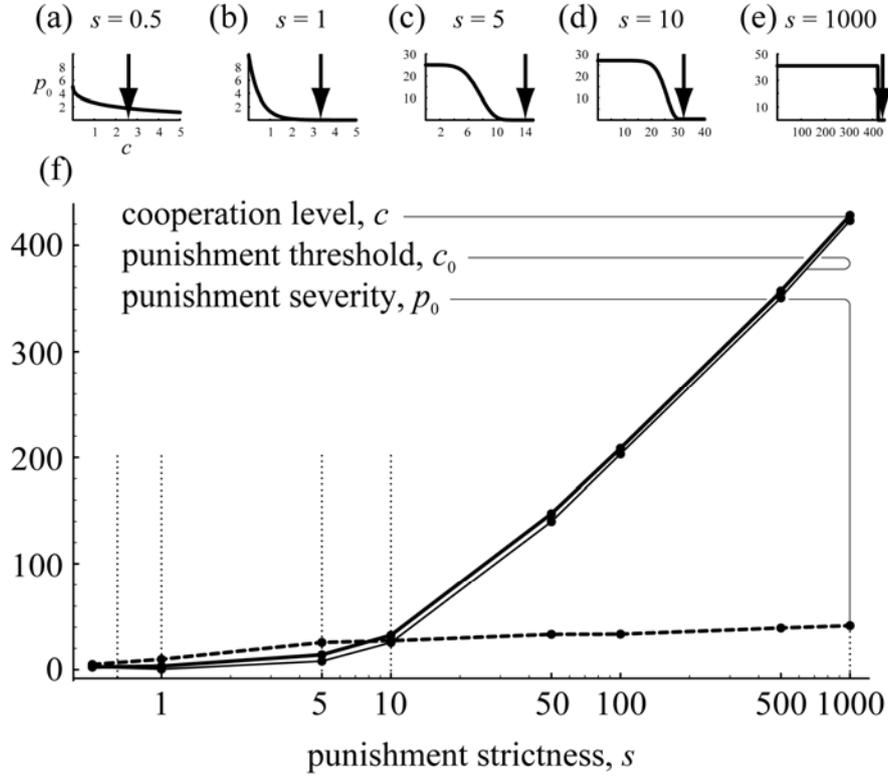
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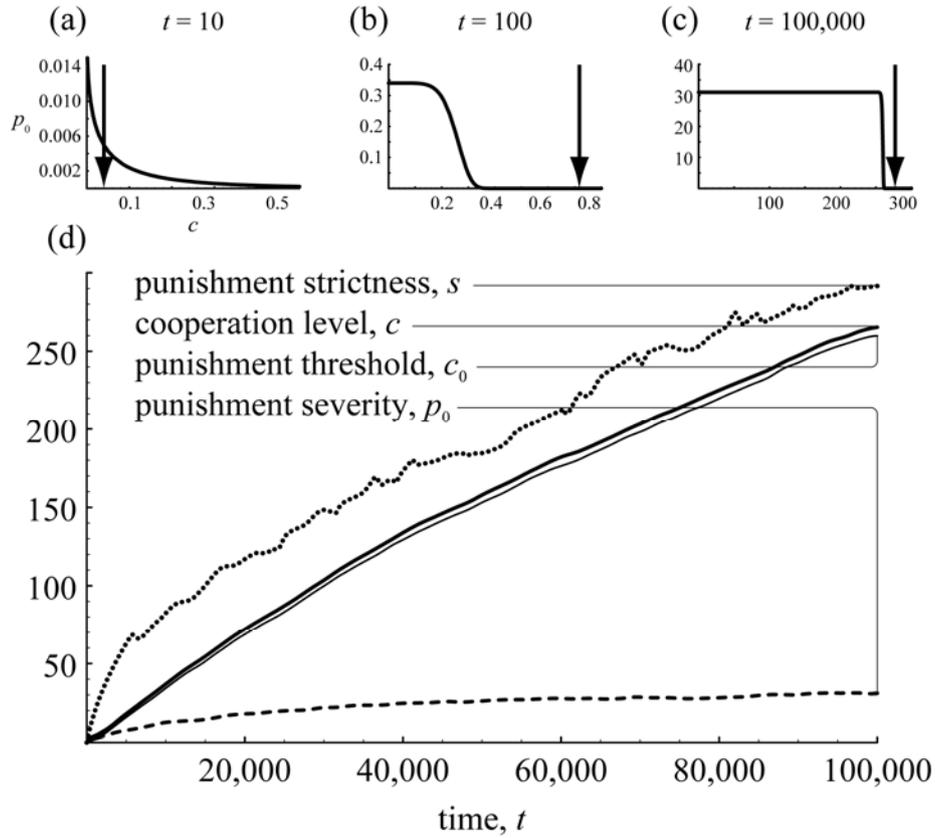
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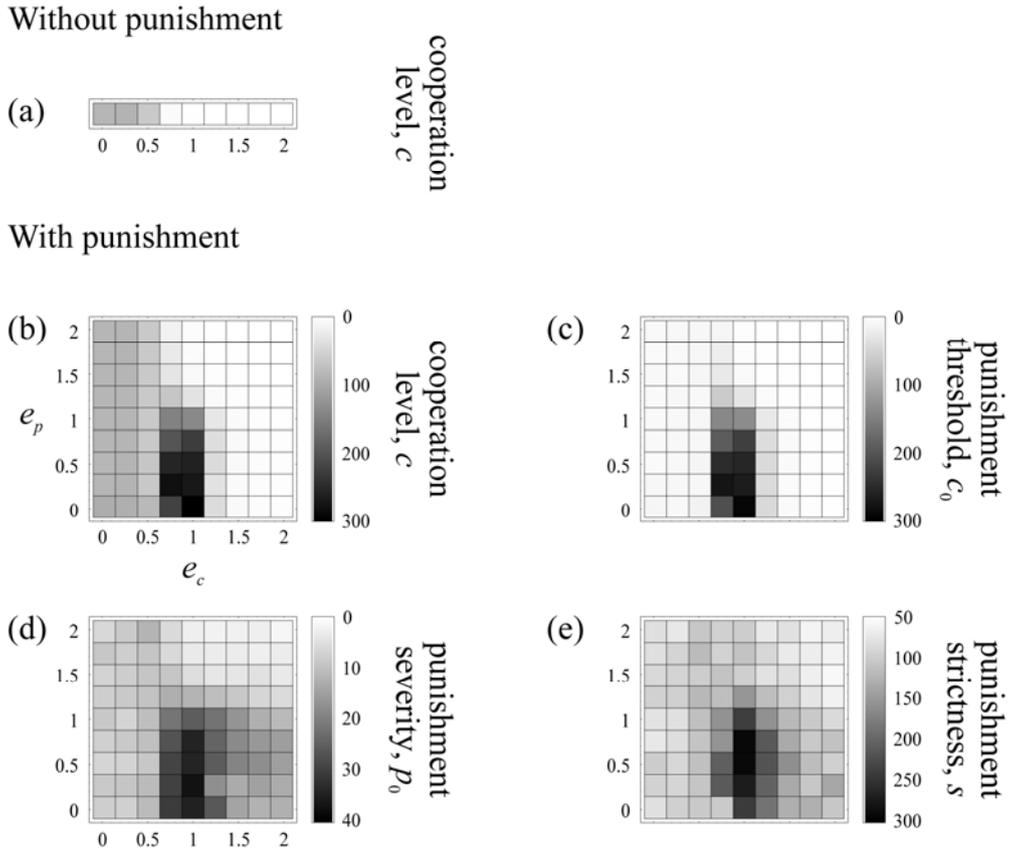
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502 **Figure 1.** Joint evolution of cooperation level c , punishment threshold c_0 , and punishment
503 severity p_0 , when punishment strictness s is kept fixed. Panels (a) to (e) show the average
504 evolved punishment reaction norms (continuous curves) and corresponding average evolved
505 cooperation levels (vertical arrows) at time $t = 100,000$ for five different fixed values of pun-
506 ishment strictness s (0.5, 1, 5, 10, and 1000). Panel (f) shows the average evolved values of
507 c (thick continuous curve), c_0 (thin continuous curve), and p_0 (dashed curve) as functions of
508 s (varying along the horizontal axis). All results are averaged over fifty model runs in the lat-
509 tice-structured population. Runaway selection for cooperation and punishment accelerates
510 with punishment strictness, leading to much elevated cooperation levels (for comparison:
511 when punishment severity is kept fixed at $p_0 = 0$, the average cooperation level equilibrates
512 at merely $c \approx 1.6$). The initial values of $c = 0$ and $p_0 = 0$ are chosen so as to highlight the
513 bootstrapping of cooperation and punishment, i.e., their gradual and steady evolution in popu-
514 lations in which cooperation and punishment are entirely absent initially. The initial value of
515 $c_0 = 10^{-5} > c$ means that all individuals are initially recognized as defectors. The initial fre-
516 quency of empty sites is 50%. The punishment reaction norm is described by Eq. (2a). Other
517 parameters: $b_0 = 2$, $d_0 = 1$, $a_c = 0.2$, $e_c = 1$, $a_p = 0.3$, $e_p = 0.5$, $m = 0.01$, and $\sigma_c = 1$.



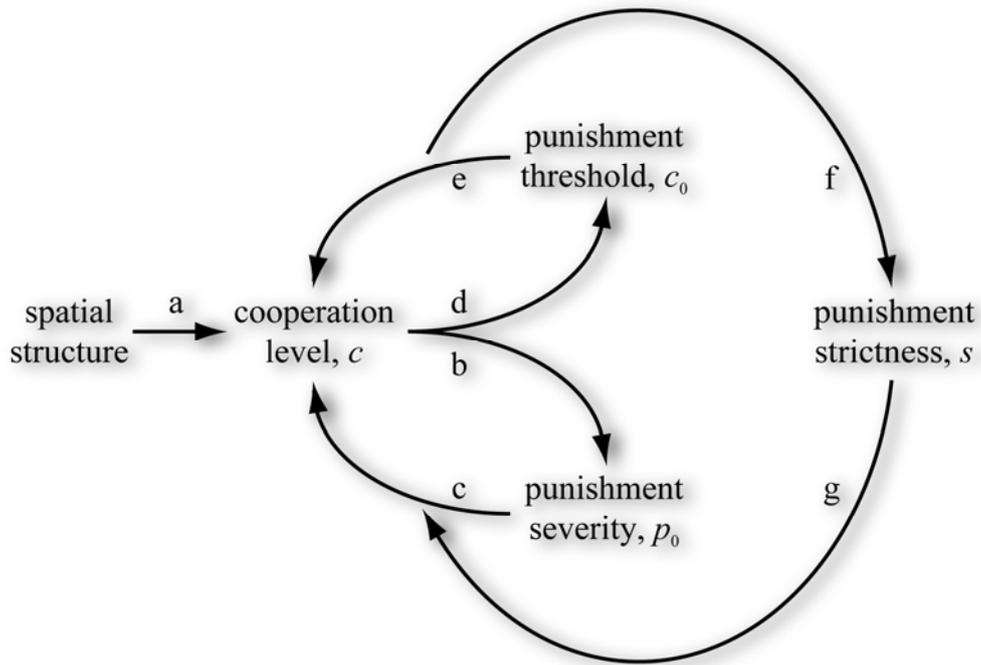
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519 **Figure 2.** Joint evolution of cooperation level c , punishment threshold c_0 , punishment sever-
 520 ity p_0 , and punishment strictness s . Panels (a) to (c) show the average evolved punishment
 521 reaction norms (continuous curves) and corresponding average cooperation levels (vertical
 522 arrows) at times $t = 10, 100$, and $100,000$. Panel (d) shows the evolutionary dynamics of c
 523 (thick continuous curve), c_0 (thin continuous curve), p_0 (dashed curve), and s (dotted
 524 curve). The initial value of $s = 0.01$ implies an essentially flat reaction norm. Other paramete-
 525 rs and settings are as in Fig. 1, with the addition of $\sigma_s = 10$.



526

527 **Figure 3.** Effects of cooperation and punishment costs on the joint evolution of cooperation
 528 level c , punishment threshold c_0 , punishment severity p_0 , and punishment strictness s .
 529 When the exponent e_c (e_p) is small, equal to 1, or large, costs for cooperation (punishment)
 530 are decelerating, linear, or accelerating. Decelerating (accelerating) costs imply that high lev-
 531 els of cooperation or punishment are relatively cheap (expensive). Panel (a) shows the
 532 average evolved cooperation level c as a function of e_c when punishment is absent (i.e.,
 533 when p_0 is fixed at 0). Panels (b) to (e), respectively, show the average evolved values of c ,
 534 c_0 , p_0 , and s as functions of e_c (varying along the horizontal axes) and e_p (varying along
 535 the vertical axes). Other parameters and settings are as in Fig. 2.



536

537 **Figure 4.** Schematic summary of positive feedbacks resulting in runaway selection for coop-
 538 eration and strict-and-severe punishment.