Coevolutionary Dynamics: From Finite to Infinite Populations

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Traditionally, frequency dependent evolutionary dynamics is described by deterministic replicator dynamics assuming implicitly infinite population sizes. Only recently have stochastic processes been introduced to study evolutionary dynamics in finite populations. However, the relationship between deterministic and stochastic approaches remained unclear. Here we solve this problem by explicitly considering large populations. In particular, we identify different microscopic stochastic processes that lead to the standard or the adjusted replicator dynamics. Moreover, differences on the individual level can lead to qualitatively different dynamics in asymmetric conflicts and, depending on the population size, can even invert the direction of the evolutionary process.

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Darwinian evolution represents an intrinsically frequency dependent process. The fitness or reproductive output of an individual is not only linked to environmental conditions but also tightly coupled to the type and frequency of its competitors. Evolutionary game theory [1, 2, 3, 4] has become also tightly coupled to the type and frequency of its competitors on the fixation probability [10]. Originally, the Moran process was formulated in a frequency independent setting where the fitness of an individual is genetically determined and remains unaffected by interactions with other individuals as in Ref. [11]. Only recently, the frequency dependent approach of evolutionary game theory and the Moran process have been successfully combined in order to investigate the evolutionary dynamics in finite populations [12, 13]. The fitness of an individual now comprises two components: the frequency independent baseline fitness which is associated with genetic predisposition and the frequency dependent contribution from interactions with other members of the population.

Thus, evolutionary dynamics can be described by a continuous deterministic replicator equation or by a stochastic microscopic description of a birth-death process such as the Moran process. So far, the relation and transition between the two approaches remained unclear. We show that different replicator dynamics are associated with different microscopic processes and moreover that the dynamics derived for infinite populations may undergo qualitative changes in finite populations.

If every individual interacts with a representative sample of the population, the average payoff of A and B individuals will be determined by the fraction of coplayers of both types. Excluding self interactions, this leads to the payoffs

\[ \pi_i^A = \frac{a (i - 1) + b (N - i)}{N - 1} \]

\[ \pi_i^B = \frac{c i + d (N - i - 1)}{N - 1} \]

where \( i \) is the number of A individuals and \( N \) is the population size.

The effective reproductive fitness of an individual of type \( k \) is then given by \( 1 - w + w \pi_k^i \) where \( w \) determines the relative contributions of the baseline fitness, which is conveniently normalized to one, and the payoffs result from interactions with other individuals [12]. Thus, \( w \) corresponds to the strength of selection acting on the game under consideration. Note that any \( w < 1 \) can be mapped to a system with a different payoff matrix and \( w = 1 [14] \). For the Moran process, the probability that the number of A individuals increases from \( i \) to \( i + 1 \) is

\[ T^+(i) = \frac{1 - w + w \pi_i^A i}{1 - w + w \langle \pi_i \rangle N} \frac{N - i}{N}, \]

where \( \langle \pi_i \rangle \) is the average payoff of the population.
whereas it decreases from $i$ to $i-1$ with probability
\[
T^-(i) = \frac{1 - w + w \pi^B_i}{1 + w} \frac{N-i}{N},
\]
(4)
\[
\langle \pi_i \rangle = \frac{\pi_A^i + \pi^B_i}{N} (N-i) / N \text{ is the average payoff in the population.}
\]
Note that the selection mechanism in the Moran process requires perfect global information on the current state of the population. This is a very strong requirement and in many situations undesirable. Therefore, we propose an alternative mechanism if
\[
\Delta \pi_{\text{max}} = \max \{ \pi_A^i, \pi^B_i \} (N-i) / N \text{ is the maximum possible payoff difference and}
\]
\[
0 < w \leq 1 \text{ measures the strength of selection. Note that in contrast to the Moran process, the local update is invariant under linear rescaling of the payoff matrix. The transition matrix for the number of A individuals in this process $i$ is then given by}
\]
\[
T^+(i) = \left( \frac{1}{w} \frac{\pi_A^i - \pi^B_i \Delta \pi_{\text{max}}}{\pi_A^i + \pi^B_i} \right) \frac{i}{N} \frac{N-i}{N},
\]
\[
T^-(i) = \left( \frac{1}{w} \frac{\pi^B_i - \pi_A^i \Delta \pi_{\text{max}}}{\pi_A^i + \pi^B_i} \right) \frac{i}{N} \frac{N-i}{N}.
\]
(6)
In both process, the number of A individuals remains constant with probability $T^0(i) = 1 - T^+(i) - T^-(i)$. Further, the states $i = 0$ and $i = N$ are absorbing.

We can directly compute the probability that a single A individual fixes in the population, $\phi_A$. In general, this probability is given by
\[
\phi_A = \frac{1}{1 + \sum_{k=1}^{N-1} \prod_{i=1}^{N} T^+(i)}.
\]
(7)
In the limit of weak selection, $w \ll 1$, the fixation probabilities can be computed analytically. The fixation probability for the Moran process is higher than for the local update mechanism if $\Delta \pi_{\text{max}} > 2$.

The stochastic process can be formulated in terms of the master equation
\[
P^{\tau+1}(i) - P^\tau(i) = P^\tau(i-1)T^+(i-1) - P^\tau(i)T^-(i) + P^\tau(i+1)T^-(i+1) - P^\tau(i)T^+(i)
\]
where $P^\tau(i)$ is the probability that the system is in state $i$ at time $\tau$. Introducing the notation $x = i/N$, $t = \tau/N$ and the probability density $\rho(x, t) = N P^\tau(i)$ yields
\[
\rho(x, t + N^{-1}) - \rho(x, t) = \rho(x - N^{-1}, t) T^+(x - N^{-1}) + \rho(x + N^{-1}, t) T^-(x + N^{-1}) - \rho(x, t) T^-(x) - \rho(x, t) T^+(x).
\]

For $N \gg 1$, the probability densities and the transition probabilities are expanded in a Taylor series at $x$ and $t$. Neglecting higher order terms in $N^{-1}$ we get
\[
\frac{d}{dt} \rho(x, t) = -\frac{d}{dx} \left[ a(x) \rho(x, t) \right] + \frac{1}{2} \frac{d^2}{dx^2} \left[ b^2(x) \rho(x, t) \right],
\]
(8)
with $a(x) = T^+(x) - T^-(x)$ and $b(x) = \sqrt{\frac{1}{N} \left[ T^+(x) + T^-(x) \right]}$. Note that, for large but finite $N$, this equation has the form of a Fokker-Planck equation. Since the internal noise is not correlated in time as subsequent update steps are independent, the Itô calculus can be applied to derive a Langevin equation
\[
\dot{x} = a(x) + b(x) \xi,
\]
(9)
where $\xi$ is uncorrelated Gaussian noise and $b(x) = 0$ for $x = 0$ and $x = 1$. The multiplicative noise term leaves the absorbing nature of the boundaries unaffected. Qualitatively similar results are obtained by introducing noise in the payoff matrix. In contrast, to account for stochastic effects through additive noise is problematic because of the boundaries of $x$.

For $N \to \infty$, the diffusion term $b(x)$ vanishes with $\frac{1}{\sqrt{N}}$ and a deterministic equation is obtained. For the Moran process, this yields
\[
\dot{x} = \lim_{N \to \infty} \left( \frac{\pi_A^i - \pi^B_i \Delta \pi_{\text{max}}}{\Gamma + \langle \pi_i \rangle} \frac{i}{N} \frac{N-i}{N} \right) = x \frac{\pi_A(x) - \langle \pi(x) \rangle}{\Gamma + \langle \pi(x) \rangle}.
\]
(10)
where $\pi_A(x) = x a + (1-x)b$, $\pi^B(x) = x c + (1-x)d$, and $\langle \pi(x) \rangle = x \pi_A(x) + (1-x)\pi^B(x)$. $\Gamma = \frac{1-w}{\Delta \pi_{\text{max}}}$ is essentially the baseline fitness. Equation (10) is the adjusted replicator dynamics introduced in Ref. 13; an alternative derivation for imitation dynamics is given in Ref. 23.

For the local update mechanism, we find
\[
\dot{x} = \lim_{N \to \infty} \left( \frac{\pi_A^i - \pi^B_i \Delta \pi_{\text{max}}}{\kappa x \langle \pi_A(x) - \langle \pi(x) \rangle \rangle} \right) = \kappa x \pi_A(x) - \langle \pi(x) \rangle,
\]
(11)
where $\kappa = \frac{1-w}{\Delta \pi_{\text{max}}}$ is a constant factor influencing the timescale only. Equation (11) is the standard replicator equation and represents the traditional approach to evolutionary dynamics in infinite populations. The difference between the two dynamics amounts to a dynamic rescaling of time which leaves the fixed points unchanged. Nevertheless, the differences in the microscopic updating can give rise to substantial differences in the macroscopic dynamics. To illustrate this, let us consider two famous examples: the Prisoner’s Dilemma and Dawkins Battle of the Sexes.

The Prisoner’s Dilemma describes the problem of cooperation where two players simultaneously decide whether to cooperate (C) or defect (D). Cooperation incurs costs $c$ but produces a benefit $b$ to the other player ($b > c$) whereas defection
neither costs nor benefits anyone. In this game defection is dominant because defectors are better off no matter what the other player does. Thus, rational players end up with nothing rather than earning the more favorable payoff for mutual cooperation — hence the dilemma. The Prisoner’s Dilemma is determined by the payoff matrix

\[
\begin{array}{c|cc}
 & D & C \\
-\hline
D & c & b + c \\
C & 0 & b
\end{array}
\] (12)

where we added c in order to avoid negative payoffs. Figure 1 compares the evolution from a state with a small fraction of defectors into the evolutionary end state with defectors only. The standard replicator dynamics for the Prisoner’s Dilemma is given by \( \dot{x} = w x (1-x) \), where \( x \) is the fraction of defectors. The solution yields

\[
x(t) = x_0 \left[ x_0 + (1-x_0) e^{-w t / (b+c)} \right]^{-1}, \quad (13)
\]

with \( x(0) = x_0 \) and \( x(t \to \infty) = x^* = 1 \). The adjusted replicator dynamics can be solved only numerically for the given payoff matrix and general \( w \). For \( w > 0 \), Fig. 1 illustrates that the convergence to \( x^* = 1 \) is faster for the adjusted replicator dynamics as compared to the standard replicator dynamics.

**Dawkins Battle of the Sexes** [25,26] is a cyclic game referring to asymmetric conflicts in parental care: If males (\( \sigma \)) are philanderers (\( B_\sigma \)), it pays for females (\( \varphi \)) to be coy (\( A_\varphi \)), insisting on a long courtship period to make males invest more in the offspring. However, once most males are faithful (\( A_\sigma \)), fast females are favored (\( B_\varphi \)) avoiding the costs of courtship. Subsequently, the male investment into the offspring is no longer justified, philanderers are again favored (\( B_\sigma \)), and the cycle continues. This is characterized by the payoff matrix

\[
\begin{array}{c|cc}
 & A_\varphi & B_\varphi \\
-\hline
A_\sigma & (+1, -1) & (-1, +1) \\
B_\sigma & (-1, +1) & (+1, -1)
\end{array}
\] (14)

where the first element is the payoff of the males and the second element is the payoff of the females. [22,28]. Note that this game is also called ’Matching Pennies’. [22,28].

The dynamics of Dawkins Battle of the Sexes is qualitatively different for the adjusted and the standard replicator dynamics [22]. Comparisons of the deterministic dynamics to the corresponding stochastic process in finite populations reveal further interesting differences. For the standard replicator dynamics

\[
H = -x(1-x)y(1-y),
\]

where \( x \) is the fraction of faithful males and \( y \) the fraction of coy females, is a constant of motion which measures the distance from the Nash equilibrium \( q = (1/2, 1/2) \). Consequently, \( q \) is a neutrally stable fixed point surrounded by periodic orbits. Figure 1 shows that this is a spurious result valid only in the limit \( N = N_\sigma = N_\varphi \to \infty \). For any finite \( N \), \( H \to 0 \) holds and every trajectory spirals away from \( q \). Note that in this case \( H \) is approximated by

\[
\langle \Delta H \rangle = \langle H(t+1) - H(t) \rangle,
\]

which is calculated for a single stochastic update step and averaged over many realizations. In contrast, the adjusted replicator dynamics leads to \( H \leq 0 \), where equality holds only at the Nash equilibrium \( q \). \( H \) represents a Lyapunov function with \( q \) as the unique and asymptotically stable interior fixed point. Interestingly, for finite \( N \), i.e. for the Moran process, the drift towards \( q \) changes sign: Below a critical population size \( N_c(w) \), \( q \) is unstable and the trajectories spiral towards the boundary. However, for \( N > N_c(w) \), \( q \) becomes stable, cf. Fig. 2. Similar qualitative changes based on population sizes have recently been reported in Refs. [13,29].

**In conclusion**, we presented a mathematically consistent transition from descriptions of the microscopic processes relevant for individual based simulations over stochastic approximations of the dynamics in finite populations to a deterministic mean field theory governed by replicator dynamics. In particular, we have shown that the intrinsic stochasticity arising in finite populations can be captured by a Langevin term in the replicator dynamics. This leads naturally to absorbing bound-
FIG. 2: Battle of the sexes: (a) Average drift in a finite population for different population size $N$ and fixed $w = 0.3$ for the local update mechanism (\(\Delta H\)) and the Moran process (\(\phi\)). For $\Delta H < 0$, the system spirals to the Nash equilibrium $\phi$; for $\Delta H > 0$, it spirals towards the absorbing boundaries. For the local update process, $\Delta H \to 0$ with $1/N$ (see inset), which hints at the constant of motion in the limit $N \to \infty$ (see text). For the Moran process, the system drifts towards the boundaries below a critical population size $N_c$ (marked by the arrow) but approaches $\phi$ for $N > N_c$. (b) The critical population size $N_c$ (where $\Delta H \approx 0$) decreases as a function of the selection pressure $w$ ($N = N_0 = N_2$, averages over $10^3$ realizations).

aries in the resulting stochastic differential equation. Both, the microscopic processes and their stochastic approximation converge with $1/N$ to the solution of the corresponding replicator equation.

The frequency dependent Moran process [12] is intimately connected to the adjusted replicator dynamics [3]. Conversely, the proposed local update rule corresponds to a finite population equivalent of the standard replicator dynamics [4]. While the qualitative dynamics of the two approaches is the same, quantitative differences vanish only upon non-linear rescaling of time. However, for interacting (sub)populations, such rescaling is no longer possible and can give rise to entirely different qualitative dynamics. In particular, for Dawkins Battle of the Sexes, we have shown that, for the Moran process, the stability of the mixed Nash equilibrium depends on the population size such that it becomes unstable below a critical size and no mixed population can persist.

John Maynard Smith asked whether the adjusted or the standard replicator dynamics is more appropriate to describe evolutionary changes [3]. Here we have shown that this is fully determined by the underlying microscopic processes.

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