Selective advantage for multicellular replicative strategies: A two-cell example

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This paper develops a quasispecies model where cells can adopt a two-cell survival strategy. Within this strategy, pairs of cells join together, at which point one of the cells sacrifices its own replicative ability for the sake of the other cell. We develop a simplified model for the evolutionary dynamics of this process, allowing us to solve for the steady-state using standard approaches from quasispecies theory. We find that our model exhibits two distinct regimes of behavior: At low concentrations of limiting resource, the two-cell strategy outcompetes the single-cell survival strategy, while at high concentrations of limiting resource, the single-cell survival strategy dominates. The single-cell survival strategy becomes disadvantageous at low concentrations of limiting resource because the energetic costs of maintaining reproductive and metabolic pathways approach, and may even exceed, the rate of energy production, leaving little excess energy for the purposes of replicating a new cell. However, if the rate of energy production exceeds the energetic costs of maintaining metabolic pathways, then the excess energy, if shared among several cells, can pay for the reproductive costs of a single cell, leaving energy to replicate a new cell. Associated with the two solution regimes of our model is a localization to delocalization transition over the portion of the genome coding for the multicell strategy, analogous to the error catastrophe in standard quasispecies models. The existence of such a transition indicates that multicellularity can emerge because natural selection does not act on specific cells, but rather on replicative strategies. Within this framework, individual cells become the means by which replicative strategies are propagated. Such a framework is therefore consistent with the concept that natural selection does not act on individuals, but rather on populations.

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One of the most interesting questions under investigation in evolutionary biology is the emergence of cooperation and multicellularity in biological systems and (references therein). While the emergence of certain types of cooperative behavior, such as division of labor, is reasonably well understood phenomena, the evolution of multicellular organisms is a more difficult question.

With division of labor, a group of cells can more efficiently metabolize environmental resources than if they worked alone, and so it is in each cell’s replicative interest to cooperate with other cells. In the case of multicellular organisms, however, certain cells forgo their ability to replicate, so that other cells in the organism can survive and reproduce. This is clearly against the replicative interests of the non-replicating cells, a situation that makes the strategy prone to defections. Indeed, defection from a multicellular survival strategy, otherwise known as cancer, is a common phenomenon in multicellular organisms.

Nevertheless, in certain environments, there must exist selective pressures driving the emergence of multicellular organisms. Perhaps one of the clearest demonstrations of such selective pressures is the existence of the organism Dictyostelium discoideum, commonly known as a cellular slime mold. The slime mold has been the focus of considerable research (it is an NIH model organism), because it lives at the border between unicellular and multicellular life: When conditions are favorable, the slime mold exists as a collection of free-living, single-celled organisms. However, when the slime mold cells are stressed, say by depletion of some necessary resource, they respond by coalescing into a differentiated, multicellular organism.

When conditions improve, the slime mold reproduces by sporulation.

One of the interesting features of the slime mold is that, during the differentiation process, some cells inevitably forgo replication for the sake of the multicellular structure. In this Letter, we attempt to elucidate the selective pressures driving this behavior by considering a highly simplified model motivated by the slime mold life cycle, one which we believe illustrates the underlying principles involved in the emergence of multicellularity. We emphasize, however, that this Letter does not consider the evolutionary dynamics modeling how such behavior could have emerged in the first place. We should also emphasize that, although our model is motivated by the slime mold life cycle, it is believed that the ability to engage in multicellular behavior is ubiquitous amongst single-celled organisms, and may even characterize the organization of bacterial biofilms.

For our model, we consider a population of organisms whose genomes consist of three distinct genes (or more appropriately, genome regions): (1) A reproduction region, denoted $\sigma_R$, coding for all the various cellular machinery involved in the growth and reproduction of the organism. (2) A metabolism region, denoting $\sigma_M$, coding for all the various cellular machinery involved in procuring food from the environment, and metabolizing it to release the energy required for various cellular processes (as in the metabolism of glucose and the storage of the energy into ATP). (3) A multicellular region, coding for the machinery necessary to implement the two-cell survival strategy. Among the various machinery required to...
implement the two-cell survival strategy is a switch that causes one of the cells to shut off its reproductive pathways, and to devote itself to metabolizing food from the environment for the sake of the other cell. This part of the genome is denoted by $\sigma_S$.

The full cellular genome is denoted by $\sigma = \sigma_R \sigma_M \sigma_S$. We assume that there exist master sequences, $\sigma_R,0, \sigma_M,0,$ and $\sigma_S,0$, corresponding to gene sequences coding for the appropriate enzymes necessary for the proper functioning of the various systems. In this single-fitness-peak approximation, any mutation to these master sequences leads to the loss in function of the corresponding system.

A cell for which $\sigma = \sigma_R,0 \sigma_M,0 \sigma_S,0$ replicates via a two-cell strategy, whereby it seeks out and joins with another cell with an identical genome. The pathways encoded within $\sigma_S,0$ cause one of the cells to shut off its reproductive pathways, and to devote its metabolic efforts to sustaining the other cell (a possible algorithm that the switch could implement is to instruct a cell to shut off its reproductive pathways if the reproductive pathways of the other cell is on, and to turn on its reproductive pathways if the reproductive pathways of the other cell is off. The only two stable solutions to this algorithm are where one of the cells has its reproductive pathways on, while the other cell has its reproductive pathways off. Presumably, although the two cells join with both of their reproductive pathways on, random fluctuations will break the symmetry and lead to collapse into an equilibrium state).

A cell for which $\sigma = \sigma_R,0 \sigma_M,0 \sigma_S$, $\sigma_S \neq \sigma_S,0$, replicates independently of the other cells. It is assumed that all other genotypes, with faulty copies of either reproductive and metabolic pathways, do not replicate at all.

The cells metabolize a single external resource, which provides both the energy and the raw materials for all the cells’ needs. If we let the basic unit of energy be the amount of energy released by metabolism of a set quantity of resource, then up to a conversion factor it is possible to measure all energy and accumulation changes in terms of the resource itself. Of course, because only that quantity of resource that has been metabolized has provided the cell with energy and raw materials, our basic measurement unit becomes the quantity of metabolized resource.

It is assumed that resource is metabolized by each cell via a two-step process: (1) A binding step, whereby the resource binds to certain receptors, which then pass on the resource for metabolism. (2) A metabolism step, whereby the resource bound the receptors is then metabolized. Assuming each of the steps is an elementary reaction, we obtain a metabolism rate $r(c)$ of the Michaelis-Menten form $\alpha c/(1 + \beta c)$, where $c$ denotes the concentration of resource in the environment. Note that this form of the metabolism rate has the property that it reaches a maximal value as the concentration of external resource becomes infinite. This makes sense, since a cell cannot metabolize an external resource at arbitrarily high rates. It should be noted, however, that our expression for $r(c)$ is not the only one that exhibits this saturation property, but it is one of the simplest expressions possible.

In order to replicate a cell, the various cellular systems must be replicated. Each system has an associated build cost (measured in units of metabolized resource). Thus, if $\rho_R, \rho_M,$ and $\rho_S$ denote the build costs of the reproductive, metabolic, and two-cell pathways, respectively, then the total cost required to build a new cell replicating via the single-cell strategy is given by $\rho_R + \rho_M$, while the total cost required to build a new cell replicating via the two-cell strategy is given by $\rho_R + \rho_M + \rho_S$.

In addition to the build costs for the various systems, each system has an associated fixed cost, corresponding to the energy and resources required to maintain system function. These fixed costs arise because the various components of the cellular systems have intrinsic decay rates (protein degradation, auto-hydrolysis of mRNAs, etc.), and in the case of switches that have to respond to changes in the external environment or the internal states of the cell, there is a minimal rate of energy consumption associated with measuring ambient conditions.

There is also an operating cost associated with each subsystem, corresponding to energy and resource costs associated with carrying out a given system task. For example, the replication machinery consumes energy in order to process a certain amount of metabolized resource toward the construction of a new cell. The metabolic pathways require energy to break-down the external resource (in chemistry, such costs are known as activation barriers).

Let $\omega_R$ denote the cost of replication per unit of metabolized resource incorporated into a new cell, and let $\omega_M$ denote the cost of metabolizing one unit of resource. Then for a cell replicating via the single-cell strategy, the total amount of resource that must be metabolized is given by, $\rho \equiv (1 + \omega_M)(\rho_R + \rho_M)$. The net rate of energy production is given by $(1 - \omega_M)r(c)$. Since replication and metabolism consume energy at a rate given by $\dot{\rho} \equiv \dot{\rho}_R + \dot{\rho}_M$, the net rate of energy accumulation is given by, $(1 - \omega_M)\alpha c/(1 + \beta c) - \dot{\rho}_R - \dot{\rho}_M$. The replication time is therefore given by,

$$\tau_{rep} = \frac{\rho}{(1 - \omega_M)r(c) - \dot{\rho}} \quad (1)$$

yielding a first-order growth-rate constant of

$$\kappa_1(c) = \frac{1}{\tau_{rep}} = \frac{(1 - \omega_M)r(c) - \dot{\rho}}{\rho} \quad (2)$$

For the two-cell replication strategy, the cell that is replicating in the cell-pair must accumulate a total of $(1 + \omega_M)(\rho_R + \rho_M + \rho_S) = \rho + \Delta \rho$ of metabolized resource. The net rate of energy production from both cells is given by $2(1 - \omega_M)r(c)$. Since only one of the
cells in the replicating cell-pair has active reproductive pathways, the total energy consumption rate is given by

\[ 2(\rho - \Delta \dot{\rho}) = \rho_R + 2\rho_M + 2\rho_S, \]

where \( \Delta \dot{\rho} \equiv (1/2)(\rho_R - 2\rho_S) \).

Therefore, the replication time is given by,

\[ \tau_{rep} = \frac{1}{2} \left( 1 - \omega_M \right) r(c) - \dot{\rho} + \Delta \dot{\rho} \tag{3} \]

yielding a first-order growth-rate constant of

\[ \kappa_2(c) = \frac{2(1 - \omega_M) r(c) - \dot{\rho} + \Delta \dot{\rho}}{\rho + \Delta \rho} \tag{4} \]

We should note that we are implicitly assuming in this derivation that the amount of time it takes for two cells to find each other and combine is negligible compared to the replication time. We are also assuming that the costs associated with transporting metabolized resource from one cell to another is negligible. Finally, we are also assuming that the reproductive pathways can process the metabolized resource as fast as it is produced.

We let \( n_1 \) denote the number of organisms with the single-cell genome. Because we are neglecting the time it takes for two organisms replicating via the two-cell strategy to find each other and to combine, we may assume that all such cells exist in the two-cell state. We therefore define \( n_2 \) to be the number of such cell-pairs in the system. Then define the total population of cells

\[ n = n_1 + 2n_2, \]

and population fractions \( x_1 = n_1/n \) and \( x_2 = 1 - x_1 = 2n_2/n \).

We also assume that cells may generate mutated daughter cells as a result of point-mutations during replication. For simplicity, we assume that replication of the master sequences \( \sigma_{M,0} \) and \( \sigma_{S,0} \) is error-free, so that we do not need to consider cells with faulty reproduction or metabolic pathways (this situation can be created by assuming that the portions of the genomes coding for reproduction and metabolism are short, so the probability of mutations occurring in these regions is negligible).

However, we assume that the per-base replication error probability in \( \sigma_S \) is given by \( \epsilon \). We let \( L \) denote the length of \( \sigma_S \), and define \( \mu = L \epsilon \). We then consider the infinite sequence length limit, while holding \( \mu \) constant. In this limit, the probability of correctly replicating \( \sigma_S \) is given by \( p_S = e^{-\mu} \). We then have,

\[ \frac{dx_1}{dt} = (\kappa_1(c) - \bar{\kappa}(t))x_1 + \frac{1}{2}\kappa_2(c)(1 - p_S)x_2 \tag{5} \]

\[ \frac{dx_2}{dt} = \frac{1}{2}\kappa_2(c)p_S - \bar{\kappa}(t)x_2 \tag{6} \]

\[ \frac{dn}{dt} = \bar{\kappa}(t)n \tag{7} \]

where \( \bar{\kappa}(t) = \kappa_1(c)x_1 + \frac{1}{2}\kappa_2(c)x_2 \).

The above population fractions will evolve to a steady-state, whose properties we can readily determine: The condition that \( dx_2/dt = 0 \) at steady-state implies that either \( x_2 = 0 \) or \( \bar{\kappa}(t = \infty) = \frac{1}{2}\kappa_2(c)p_S \). If \( x_2 = 0 \), then \( dx_1/dt = 0 \) implies that \( \bar{\kappa}(t = \infty) = \kappa_1(c) \).

For a steady-state to be stable to perturbations, we must have \( \bar{\kappa}(t = \infty) \geq \kappa_1(c) \), \( \frac{1}{2}\kappa_2(c)p_S \). Therefore, at steady-state we have,

\[ \bar{\kappa}(t = \infty) = \max \left\{ \frac{1}{2}\kappa_2(c)p_S, \kappa_1(c) \right\}. \]

Using the formulas for \( (1/2)\kappa_2(c)p_S \) and \( \kappa_1(c) \), and assuming that \( \Delta \rho > 0 \), we have that

\[ \frac{1}{2}\kappa_2(c)p_S > \kappa_1(c), \]

\[ x_2 > 0, \quad \text{if} \ 0 \leq r(c) < r(c) \]

\[ \frac{1}{2}\kappa_2(c)p_S < \kappa_1(c), \quad x_2 = 0, \quad \text{if} \ r(c) > r(c) \]

where,

\[ r(c) = \frac{\dot{\rho}}{1 - \omega_M} \frac{1 - p_S}{1 - p_S} \frac{\Delta \dot{\rho} - \rho}{\rho + \Delta \rho} \tag{10} \]

Let \( z_{1,l} \) denote the fraction of the population whose genome \( \sigma_{M,0} \sigma_{S,0} \) is such that \( D_M(\sigma_S, \sigma_{S,0}) = l \), where \( l > 0 \). Then, using similar techniques to those found in [3], it is possible to show that,

\[ \frac{dz_{1,l}}{dt} = \frac{1}{2}\kappa_2(c)x_2 \mu \frac{1}{l} e^{-\mu} + \kappa_1(c) e^{-\mu} \sum_{l' = 0}^{l-1} \frac{\mu}{l'!} z_{1,l'-l} - \bar{\kappa}(t)z_{1,l} \tag{11} \]

Defining the localization length \( \langle l \rangle \) via,

\[ \langle l \rangle_S = \sum_{l=1}^{\infty} \langle l \rangle z_{1,l} \tag{12} \]

then at steady-state,

\[ \langle l \rangle_S = \mu \frac{\kappa(\infty)}{\kappa(\infty) - \kappa_1(c)} \tag{13} \]

which is finite as long as \( \bar{\kappa}(t = \infty) = \frac{1}{2}\kappa_2(c)p_S > \kappa_1(c) \), and \( \infty \) otherwise.

In other words, once the selective advantage for replicating via the two-cell survival strategy disappears, the
portion of the genome coding for this strategy undergoes a localization to delocalization transition, analogous to the error catastrophe (it is also similar to a phenomenon known as “survival of the flattest”).

Figure 2 shows the various solution regimes as a function of \( r(c) \) and \( p_S \). Note that, for a given value of \( p_S \), there exists a low-concentration regime where the fraction of cells adopting the two-cell strategy is positive. In this regime, there is a selective advantage for a genome to maintain a functional copy of the multicell switch \( \sigma_{S,0} \). At a critical concentration given by \( r(c) = r(c)_c \), resources are sufficiently plentiful that it becomes disadvantageous to instruct a cell to sacrifice its own reproductive ability for the sake of the other one. The reason for this is that, although the average fixed cost per cell is lower with the two-cell strategy, the cost of having to replicate the strategy outweighs the savings in fixed costs when resources are plentiful. Thus, once \( r(c) > r(c)_c \), the fraction of cells adopting the two-cell strategy disappears, and the population consists entirely of cells replicating via the single-cell strategy.

If \( r(c) > r(c)_c \) for \( p_S = 1 \), then varying \( p_S \) at this concentration will never lead to a selective advantage for maintaining a two-cell survival strategy. If \( r(c) < r(c)_c \) for \( p_S = 1 \), but \( r(c) > r(c)_c \) for \( p_S = 0 \), then for sufficiently large \( p_S \) there will exist a finite fraction of the cells which replicate via the two-cell strategy. As \( p_S \) drops below some critical value, denoted \( p_{S,\text{crit}} \), the probability of incorrectly duplicating the strategy becomes sufficiently large that the fraction of cells replicating via the two-cell strategy disappears. This concentration regime is interesting because it corresponds to a regime where replicating via the two-cell strategy is actually the advantageous one, but it might not be observed because of replication errors.

Finally, once \( r(c) \) drops below \( r(c)_c|_{p_S=0} \), then \( \kappa_1(c) < 0 \), so as long as \( \kappa_2(c)p_S > 0 \), there will exist a selective advantage for maintaining the two-cell strategy in the population. Due to mutation, this will also lead to the maintenance of the single-cell strategy, although this strategy is not self-sustaining in the population.

If, due to saturation, \( r(\infty) \) is finite, then one possibility is that the parameters of our model are such that \( r(\infty) < r(c)_c \) at a given \( p_S \). Then for this value of \( p_S \), there will exist a selective advantage for the two-cell strategy no matter what the external concentration of resource (the cells cannot metabolize the resources sufficiently fast to eliminate the selective advantage for multicellularity).

The results of our model show that natural selection does not act on individual cells, but rather on the survival strategy as encoded for in \( \sigma_{S,0} \). Individual cells then are more properly viewed as vehicles by which the multicell strategy is passed on to the next generation. When food resources become limited (or when the cells cannot rapidly metabolize the food resources present), the effective growth rate of the multicell strategy is competitive with the total growth rate of the single-cell strategies, resulting in its preservation in the population. Essentially, it becomes advantageous (from the point of view of the strategy) for several cells to pool their resources together for the purposes of replicating a single cell. When food becomes more plentiful, or when the rate of replication errors reaches a threshold value, the selective advantage for retaining the strategy disappears, and delocalization occurs over the corresponding region of the genome.

A potentially interesting avenue of future research is to determine whether there exist natural bounds on the possible multicellular replicative strategies, and whether it is possible, using thermodynamics and information theory, to connect these natural bounds to basic physicochemical properties of the constituent reaction networks.

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