Evolutionary rescue

Remus Stana¹, Yoav Ram¹, and Daniel Weissman²

¹School of Zoology, Tel Aviv University, Tel Aviv-Yafo, Israel ²Department of Physics, Emory University, Atlanta, United States of America

August 20, 2022

Abstract

1 Introduction

The question of what role does chromosomal instability (CIN) play in the emergence of cancer has been studied extensively in the past decades [1, 2, 3, 4, 5, 6]. One hypothesis is that CIN facilitates tumorgenesis by accelerating the removal of tumor suppression genes (TSG) and subsequent appearance of cancer. The deletion of tumor suppression genes can happen in two ways (see Figure 1): two point mutations deleting both alleles of the TSG or one point mutation and one chromosomal loss event. The case of two chromosomal loss events is taken into consideration because the fitness cost would be very large and the cell would die.

Populations adapted to a certain environment are vulnerable to change in the environment which might cause extinction of the population. Adaptation is a race against time as the population size decreases in the new environment. Evolutionary rescue is the process where the population acquires mutations which increases the fitness of the population in the new environment such that the extinction in averted. We are interested in how likely is a population to adapt to the changing environment through evolution. Examples of such changes include climate change, invasive species or the onset of drug therapies. There exists three possible ways for a population to survive environmental degradation: migration to a new habitat similar to the one before the onset of environmental changes [7], adaptation by phenotypic plasticity which involves no changes in the genotype [8, 9] and adaptation through genetic mutations [10, 11, 12].

Majority of studies assume that the fitness of the wildtype and mutant are time homogeneous. An exception is [13] fitness of wildtype and mutant are taken to be functions of time.

2 Model

We model evolutionary rescue of cancer cells through an euploidy with the help of multitype branching processes: we have three populations of cancer cells, wildtype, an euploidy and mutant cells which proliferate and die with rates λ_w , λ_a , λ_m and μ_w , μ_a , μ_m , respectively. Wildtype cells can acquire an euploidy at rate u while an euploidy and wildtype cells can mutate with rate v (se Figure 2). The probability that a single mutant cell will survive is given by [14]:

$$p_m = \begin{cases} \frac{\lambda_m - \mu_m}{\lambda_m}, & \text{if } \lambda_m > \mu_m \\ 0, & \text{else.} \end{cases}$$
 (1)

If the population originally consists of a single aneuploidy cell, then the probability that the population will survive is given by the following quadratic equation:

$$1 - p_a = \frac{\mu_a}{\lambda_a + \mu_a + v} + \frac{v}{\lambda_a + \mu_a + v} (1 - p_m) + \frac{\lambda_a}{\lambda_a + \mu_a + v} (1 - p_a)^2,$$
 (2)

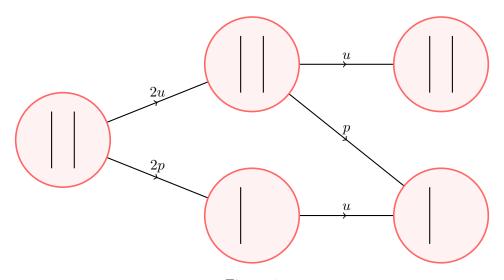


Figure 1

whose smallest solution is the survival probability:

$$p_a = \frac{\lambda_a - \mu_a - v + \sqrt{(\lambda_a - \mu_a - v)^2 + 4\lambda_a v p_m}}{2\lambda_a}.$$
 (3)

Analogously, the survival probability of a population consisting initially of one wildtype cell satisfies the quadratic equation:

$$1 - p_w = \frac{\mu_w}{\lambda_w + \mu_w + u + v} + \frac{u}{\lambda_w + \mu_w + u + v} (1 - p_a) + \frac{\lambda_w}{\lambda_w + \mu_w + u + v} (1 - p_w)^2 + \frac{v}{\lambda_w + \mu_w + u + v} (1 - p_m),$$

$$(4)$$

where the smallest solution gives the survival probability:

$$p_{w} = \frac{\lambda_{w} - \mu_{w} - u - v + \sqrt{(\lambda_{w} - \mu_{w} - u - v)^{2} + 4\lambda_{w} (up_{a} + vp_{m})}}{2\lambda_{w}}.$$
 (5)

3 First case: $4\lambda_a v p_m < (\lambda_a - \mu_a - v)^2$

We assume that $\mu_w > \lambda_w$ and $\mu_a < \lambda_a$ and, as a result, we rewrite (3) and (12) as:

$$p_{a} = \frac{\lambda_{a} - \mu_{a} - v}{2\lambda_{a}} \left(1 + \sqrt{1 + \frac{4\lambda_{a}vp_{m}}{(\lambda_{a} - \mu_{a} - v)^{2}}} \right),$$

$$p_{w} = \frac{\lambda_{w} - \mu_{w} - u - v}{2\lambda_{w}} \left(1 - \sqrt{1 + \frac{4\lambda_{w}(vp_{m} + up_{a})}{(\lambda_{w} - \mu_{w} - u - v)^{2}}} \right).$$

Making use of the Taylor series expansions:

$$\left(1 + \frac{4\lambda_a v p_m}{(\lambda_a - \mu_a - v)^2}\right)^{\frac{1}{2}} = 1 + \frac{2\lambda_a v p_m}{(\lambda_a - \mu_a - v)^2} + \cdots$$

$$\left(1 + \frac{4\lambda_w (v p_m + u p_a)}{(\lambda_w - \mu_w - u - v)^2}\right)^{\frac{1}{2}} = 1 + \frac{2\lambda_w (v p_m + u p_a)}{(\lambda_a - \mu_a - u - v)^2} + \cdots$$

we obtain the following approximation for the survival probability of a population consisting of a single individual wildtype cell:

$$p_w \approx \frac{vp_m + up_a}{\lambda_a - \mu_a - u - v} \tag{6}$$

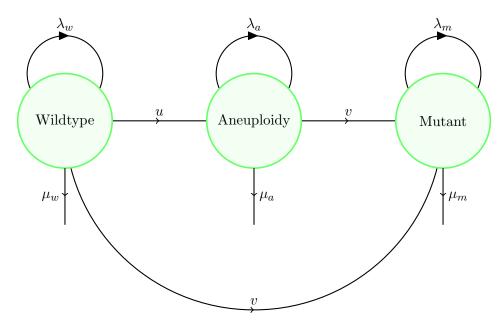


Figure 2: Diagram of the evolutionary rescue model where the population of cancer cells is subdivided into three populations of cancer cells, wildtype, an uploidy and mutant cells which proliferate and die with rates λ_w , λ_a , λ_m and μ_w , μ_a , μ_m , respectively. Wildtype cells can acquire an uploidy at rate u while an uploidy and wildtype cells can mutate with rate v.

$$\approx -\frac{1}{\lambda_{w} - \mu_{w} - u - v} \left[\frac{v \left(\lambda_{a} - \mu_{a} - u\right)}{\lambda_{a}} + \frac{uv \left(\lambda_{m} - \mu_{m}\right)}{\lambda_{m} \left(\lambda_{a} - \mu_{a} - u\right)} + \frac{v \left(\lambda_{m} - \mu_{m}\right)}{\lambda_{m}} \right]$$

$$\approx -\frac{1}{\lambda_{w} - \mu_{w} - u - v} \left[\frac{v \left(\lambda_{a} - \mu_{a}\right)}{\lambda_{a}} + \frac{uv \left(\lambda_{m} - \mu_{m}\right)}{\lambda_{m} \left(\lambda_{a} - \mu_{a}\right)} + \frac{v \left(\lambda_{m} - \mu_{m}\right)}{\lambda_{m}} \right], \tag{7}$$

where in the last line we have used the fact that $u, v \ll 1$. Using the notational convention

$$\Delta_i = \lambda_i - \mu_i, \tag{8}$$

we write (12) as

$$p_w = -\frac{1}{\Delta_w} \left(\frac{v\Delta_a}{\lambda_a} + \frac{uv\Delta_m}{\lambda_m \Delta_a} + \frac{v\Delta_m}{\lambda_m} \right). \tag{9}$$

Given an initial population consisting of N wildtype cancer cells, the probability that the population will survive is given by:

$$p_{est} = 1 - (1 - p_w)^N \approx 1 - e^{-Np_w} = 1 - \exp\left[\frac{N}{\Delta_w} \left(\frac{v\Delta_a}{\lambda_a} + \frac{uv\Delta_m}{\lambda_m\Delta_a} + \frac{v\Delta_m}{\lambda_m}\right)\right],\tag{10}$$

which we plot in Figure 4 as a function of N and in Figure 3 as a function of λ_w .

We want to improve the accuracy of our approximation by taking into consideration the second term of the Taylor series expansion:

$$\left(1 + \frac{4\lambda_a v p_m}{(\lambda_a - \mu_a - v)^2}\right)^{\frac{1}{2}} = 1 + \frac{2\lambda_a v p_m}{(\lambda_a - \mu_a - v)^2} - \frac{(\lambda_a v p_m)^2}{4(\lambda_a - \mu_a - v)^4} + \cdots,$$

which gives us the following approximation for p_a :

$$p_{a} = \frac{\lambda_{a} - \mu_{a} - v}{\lambda_{a}} + \frac{vp_{m}}{\lambda_{a} - \mu_{a} - v} - \frac{\lambda_{a} (vp_{m})^{2}}{8 (\lambda_{a} - \mu_{a} - v)^{3}}$$
(11)

From which we deduce that:

$$p_{w} \approx -\frac{1}{\lambda_{w} - \mu_{w} - u - v} \left[\frac{v \left(\lambda_{a} - \mu_{a} - u\right)}{\lambda_{a}} + \frac{uv \left(\lambda_{m} - \mu_{m}\right)}{\lambda_{m} \left(\lambda_{a} - \mu_{a} - u\right)} + \frac{v \left(\lambda_{m} - \mu_{m}\right)}{\lambda_{m}} - \frac{uv^{2} \lambda_{a} \left(\lambda_{m} - \mu_{m}\right)^{2}}{8\lambda_{m}^{2} \left(\lambda_{a} - \mu_{a} - v\right)^{3}} \right]$$

$$\approx -\frac{1}{\lambda_w - \mu_w - u - v} \left[\frac{v \left(\lambda_a - \mu_a\right)}{\lambda_a} + \frac{uv \left(\lambda_m - \mu_m\right)}{\lambda_m \left(\lambda_a - \mu_a\right)} + \frac{v \left(\lambda_m - \mu_m\right)}{\lambda_m} - \frac{uv^2 \lambda_a \left(\lambda_m - \mu_m\right)^2}{8\lambda_m^2 \left(\lambda_a - \mu_a\right)^3} \right]. \tag{12}$$

Using the notations described in (8) we write the above equation as:

$$p_w = -\frac{1}{\Delta_w} \left(\frac{v\Delta_a}{\lambda_a} + \frac{uv\Delta_m}{\lambda_m\Delta_a} + \frac{v\Delta_m}{\lambda_m} - \frac{uv^2\lambda_a\Delta_m^2}{8\lambda_m^2\Delta_a^3} \right). \tag{13}$$

Given an initial population consisting of N wildtype cancer cells, the probability that the population will survive is given by:

$$p_{est} = 1 - (1 - p_w)^N \approx 1 - e^{-Np_w} = 1 - \exp\left[\frac{N}{\Delta_w} \left(\frac{v\Delta_a}{\lambda_a} + \frac{uv\Delta_m}{\lambda_m\Delta_a} + \frac{v\Delta_m}{\lambda_m}\right)\right].$$
(14)

4 Second case: $4\lambda_a v p_m > (\lambda_a - \mu_a - v)^2$

If we assume that $4\lambda_a v p_m > (\lambda_a - \mu_a - v)^2$ then we write:

$$p_a = \frac{\lambda_a - \mu_a - v + 2\sqrt{\lambda_a v p_m} \left(1 + \frac{(\lambda_a - \mu_a - v)^2}{4\lambda_a v p_m}\right)^{\frac{1}{2}}}{2\lambda_a}$$

$$(15)$$

and using the following Taylor series expansion:

$$\left(1 + \frac{(\lambda_a - \mu_a - v)^2}{4\lambda_a v p_m}\right)^{\frac{1}{2}} = 1 + \frac{(\lambda_a - \mu_a - v)^2}{8\lambda_a v p_m} + \cdots$$

we obtain:

$$p_{a} \approx \frac{\lambda_{a} - \mu_{a} - v + 2\sqrt{\lambda_{a}vp_{m}} \left[1 + \frac{(\lambda_{a} - \mu_{a} - v)^{2}}{8\lambda_{a}vp_{m}}\right]}{2\lambda_{a}}$$

$$= \frac{\lambda_{a} - \mu_{a} - v + 2\sqrt{\lambda_{a}vp_{m}} + \frac{(\lambda_{a} - \mu_{a} - v)^{2}}{4\sqrt{\lambda_{a}vp_{m}}}}{2\lambda_{a}}$$

$$\approx \sqrt{\frac{v(\lambda_{m} - \mu_{m})}{\lambda_{a}\lambda_{m}}}$$

In the last line of the above we have used the following inequalities:

$$\lambda_a - \mu_a, v, \frac{(\lambda_a - \mu_a - v)^2}{4\sqrt{\lambda_a v p_m}} \ll 1.$$

As a result, we have from (6) the probability of rescue of a population starting from one wildtype individual:

$$p_{w} \approx \frac{1}{\lambda_{a} - \mu_{a} - u - v} \left[v \frac{\lambda_{m} - \mu_{m}}{\lambda_{m}} + u \sqrt{\frac{v (\lambda_{m} - \mu_{m})}{\lambda_{a} \lambda_{m}}} \right]$$

$$= \sqrt{\frac{\lambda_{m} - \mu_{m}}{\lambda_{m}}} \frac{\sqrt{v}}{\lambda_{a} - \mu_{a} - u - v} \left[\frac{u}{\sqrt{\lambda_{a}}} + \sqrt{\frac{v (\lambda_{m} - \mu_{m})}{\lambda_{m}}} \right]$$

$$= \sqrt{\frac{\Delta_{m}}{\lambda_{m}}} \frac{\sqrt{v}}{\Delta_{a} - u - v} \left[\frac{u}{\sqrt{\lambda_{a}}} + \sqrt{\frac{v \Delta_{m}}{\lambda_{m}}} \right],$$

where in the last line we have used the notations defined in (8).

Given an initial population consisting of N wildtype cancer cells, the probability that the population will survive is given by:

$$p_{est} = 1 - (1 - p_w)^N \approx 1 - e^{-Np_w} = 1 - \exp\left[-N\sqrt{\frac{\Delta_m}{\lambda_m}} \frac{\sqrt{v}}{\Delta_a - u - v} \left(\frac{u}{\sqrt{\lambda_a}} + \sqrt{\frac{v\Delta_m}{\lambda_m}}\right)\right].$$

5 Standing genetic variation

Until now we have assumed that the initial population of cells consisted entirely of wildtype cells. We modify this assumtion such that the initial population includes a fraction f of cells with an euploidy. The probability of evolutionary rescue by the cells with an euploidy from the initial population is:

$$p_{old} = 1 - (1 - p_a)^{fN} \approx 1 - e^{-fNp_a}.$$

The total probability of evolutionary rescue is given by:

$$p_{total} = p_{new} + (1 - p_{new}) p_{old}$$

= 1 - e^{-[(1-f)p_w+fp_a]N}. (16)

The fraction of the cases in which the population is rescued by the standing genetic variation is given by:

$$F(f) = \frac{1 - e^{-fNp_a}}{1 - e^{-[(1-f)p_w + fp_a]N}}.$$

We let $F = \frac{1}{2}$ and we obtain:

$$f^* \approx \frac{p_w}{p_w + p_a},$$

wher we have used the expansion $e^x \approx 1 + x$. We plot F and f^* in Figure 5.

6 Discussion

References

- [1] Franziska Michor, Yoh Iwasa, Bert Vogelstein, Christoph Lengauer, and Martin A Nowak. Can chromosomal instability initiate tumorigenesis? In *Seminars in cancer biology*, volume 15, pages 43–49. Elsevier, 2005.
- [2] J Ye Christine, Sarah Regan, Guo Liu, Sarah Alemara, and Henry H Heng. Understanding aneuploidy in cancer through the lens of system inheritance, fuzzy inheritance and emergence of new genome systems. *Molecular cytogenetics*, 11(1):1–13, 2018.
- [3] Martin A Nowak, Natalia L Komarova, Anirvan Sengupta, Prasad V Jallepalli, Ie-Ming Shih, Bert Vogelstein, and Christoph Lengauer. The role of chromosomal instability in tumor initiation. *Proceedings of the National Academy of Sciences*, 99(25):16226–16231, 2002.
- [4] Norman Pavelka, Giulia Rancati, and Rong Li. Dr Jekyll and Mr Hyde: role of aneuploidy in cellular adaptation and cancer. Current opinion in cell biology, 22(6):809–815, 2010.
- [5] Natalia L Komarova, Anirvan Sengupta, and Martin A Nowak. Mutation—selection networks of cancer initiation: tumor suppressor genes and chromosomal instability. *Journal of theoretical biology*, 223(4):433–450, 2003.
- [6] Jin Zhu, Hung-Ji Tsai, Molly R Gordon, and Rong Li. Cellular stress associated with aneuploidy. *Developmental cell*, 44(4):420–431, 2018.
- [7] Christina A Cobbold and Remus Stana. Should I stay or should I go: partially sedentary populations can outperform fully dispersing populations in response to climate-induced range shifts. *Bulletin of Mathematical Biology*, 82(2):1–21, 2020.
- [8] Oana Carja and Joshua B Plotkin. Evolutionary rescue through partly heritable phenotypic variability. *Genetics*, 211(3):977–988, 2019.
- [9] Oana Carja and Joshua B Plotkin. The evolutionary advantage of heritable phenotypic heterogeneity. Scientific reports, 7(1):1–12, 2017.

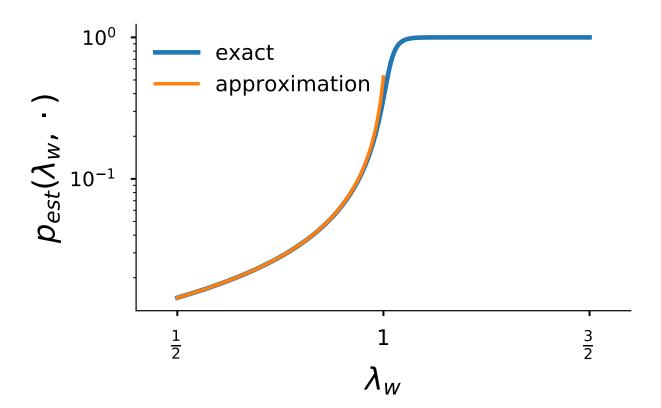


Figure 3: Plot of the probability of survival of a population as a function of the proliferation rate of the wildtype cells. The blue line represents the exact solution (12) and the orange line line represents the approximation (9). Here the population initially consists of N wildtype cells and for the simulations we have chosen the following parameters: N = 75, $\lambda_a = 1 + 10^{-2}$, $\lambda_m = 1 + 10^{-3}$, $\mu_w = 1$, $\mu_a = 1$, $\mu_m = 1$.

- [10] Hildegard Uecker, Sarah P Otto, and Joachim Hermisson. Evolutionary rescue in structured populations. *The American Naturalist*, 183(1):E17–E35, 2014.
- [11] Hildegard Uecker and Joachim Hermisson. The role of recombination in evolutionary rescue. *Genetics*, 202(2):721–732, 2016.
- [12] Hildegard Uecker and Joachim Hermisson. On the fixation process of a beneficial mutation in a variable environment. *Genetics*, 188(4):915–930, 2011.
- [13] Loïc Marrec and Anne-Florence Bitbol. Adapt or perish: Evolutionary rescue in a gradually deteriorating environment. *Genetics*, 216(2):573–583, 2020.
- [14] Linda JS Allen. An introduction to stochastic processes with applications to biology. CRC press, 2010.

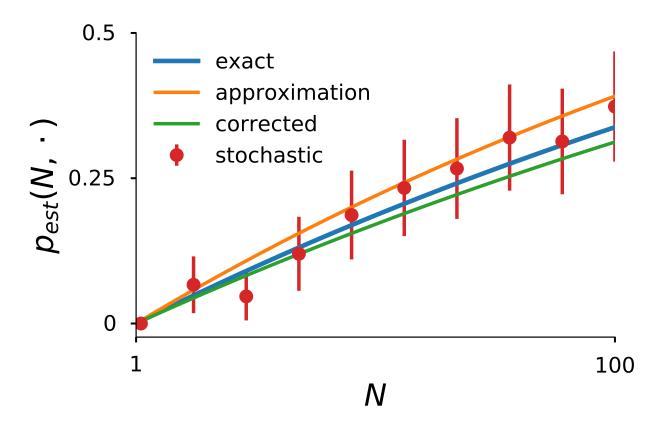


Figure 4: Plot of the probability of survival of a population as a function of the initial population size of wildtype cells. The blue line represents the exact solution (12), the orange line line represents the approximation (9), the green line represents the first order correction (13) and the red dots represents stochastic simulations. For the simulations we have chosen the following parameters: $\lambda_w = 1 - 10^{-2}$, $\lambda_a = 1 + 10^{-2}$, $\lambda_m = 1 + 10^{-3}$, $\mu_w = 1$, $\mu_a = 1$, $\mu_m = 1$.

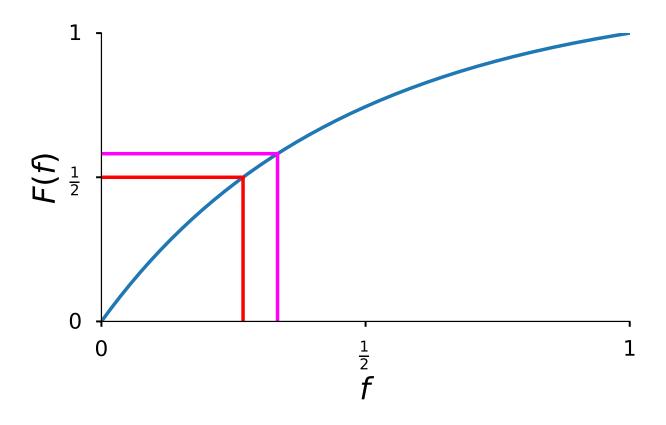


Figure 5