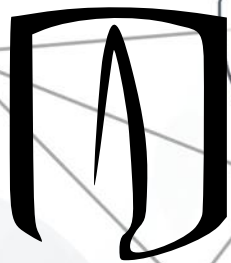


Simulation of different age distributions for the analysis of the aging curve of a population of “*S. cerevisiae*”.

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Universidad de
los Andes



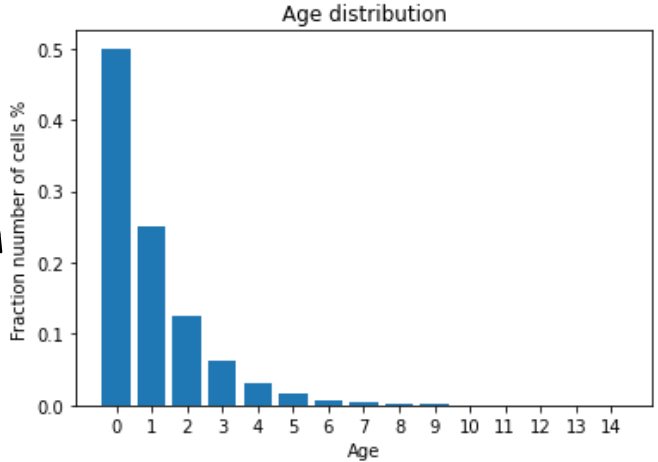
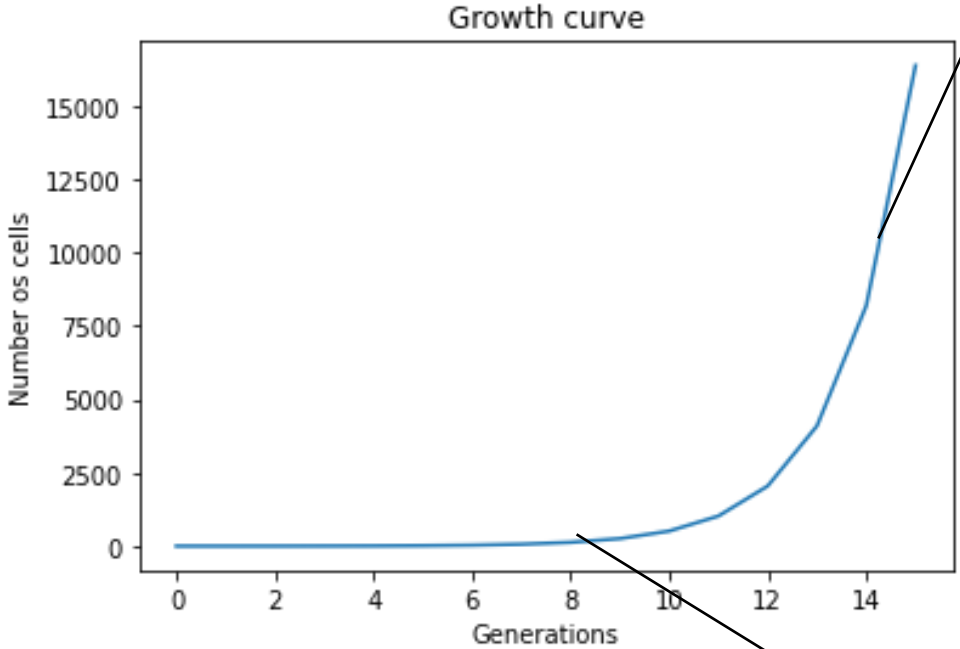
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¹ Laboratorio de Biofísica, Departamento de Física, Universidad de los Andes, Bogotá 111711, Colombia.

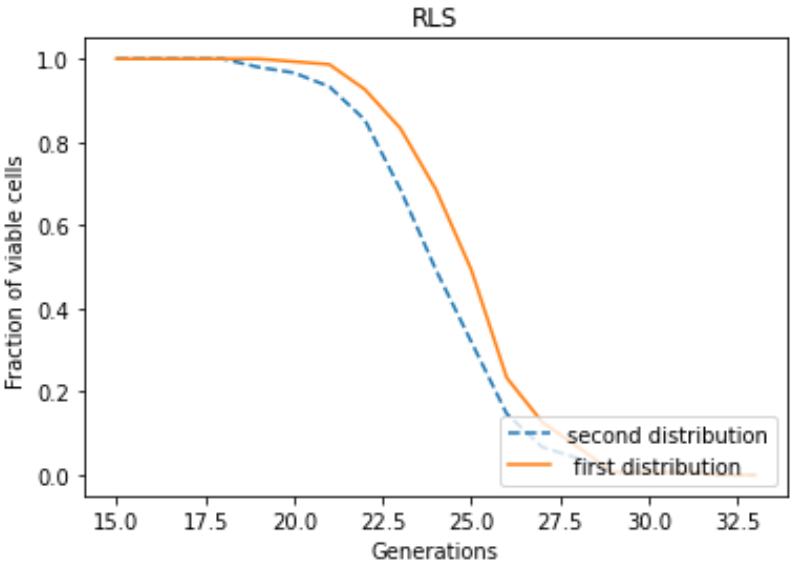
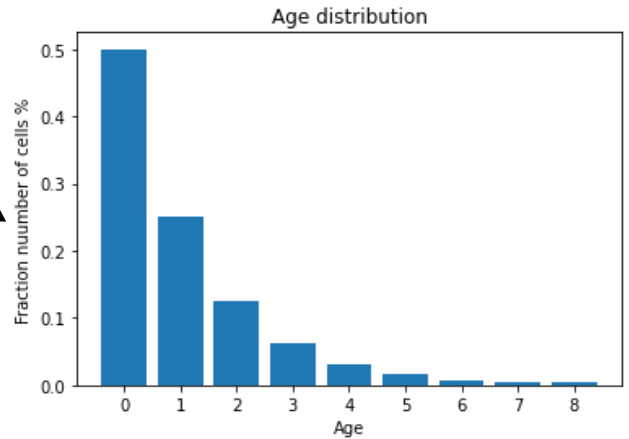
² Proyecto Curricular Licenciatura en Física, Facultad de Ciencias y Educación, Universidad Distrital Francisco José de Caldas, Bogotá 110311, Colombia.

GRAPHICAL ABSTRACT

2



1



ABSTRACT

Newly developed microfluidic devices (“Mother Machines”) have improved data gathering for the study of aging in unicellular models, and thereby the understanding of this process. Each device has different features that cause them to have certain advantages or disadvantages. At the University of the Andes a new microfluidic device was developed that uses the Slipstreaming effect to trap the cells. This has the advantage of not using mechanical pressure to trap the cells, but as it starts with a mixed age population it does not guarantee that the cells studied are virgin.

One of the basic outputs in these studies is the aging curve, which shows how the fraction of viable cells varies with respect to time. From this it can be deduced how fast or slow the population ages. For devices where it is not possible to work with virgin cells the age distribution is assumed, but changes in this distribution could affect the analysis of the data.

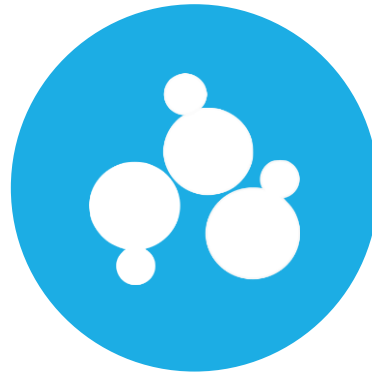
Therefore, the present work seeks to carry out a series of simulations to find the different age distributions that could be present and determine the corresponding changes in the aging curve. We propose two population growth models, synchronous and asynchronous. For each model we will start with the possible age distributions and determine the various curves that can be obtained and then compare these computational results with the experimental data to propose a better interpretation of the data obtained from Mother Machine devices.

INTRODUCTION



AGING

Changes in the function of the cells over time which can trigger different kinds of diseases



YEAST CELLS

It is a model organism because it is a very simple eukaryotic cell and has replicative aging

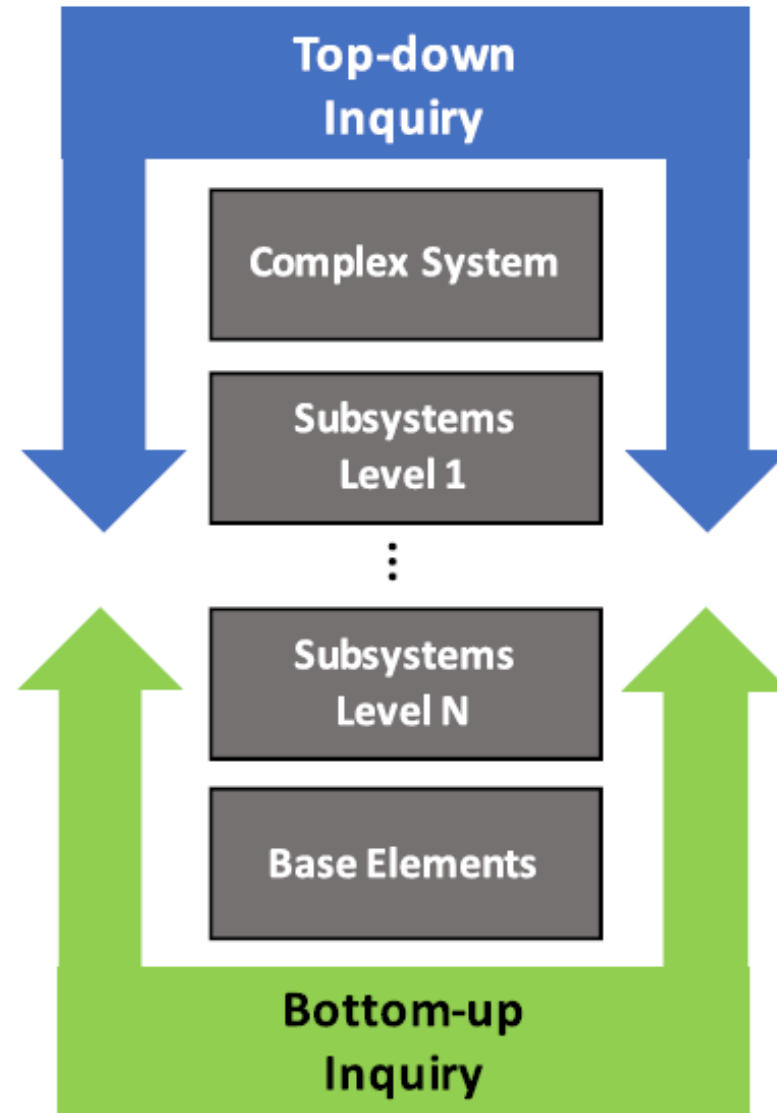


MICROFLUIDIC DEVICES

Our devices work via a laminar flow and which detaches the daughter cells from the mother cells

SYSTEMS BIOLOGY OF CELLS AGING

With the help of computational modeling and quantitative analysis of complex biological systems, it takes a different approach to research, providing another point of view to the large number of experimental results that are obtained, giving a map or a route of the different networks of aging (Song et al., 2018)

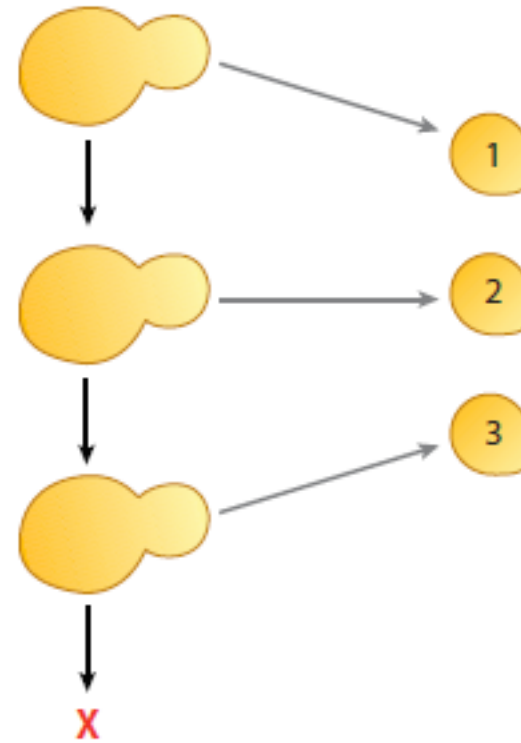


Outline of the two ways by which aging can be analyzed. Image taken from : The Systems Biology of Single-Cell Aging

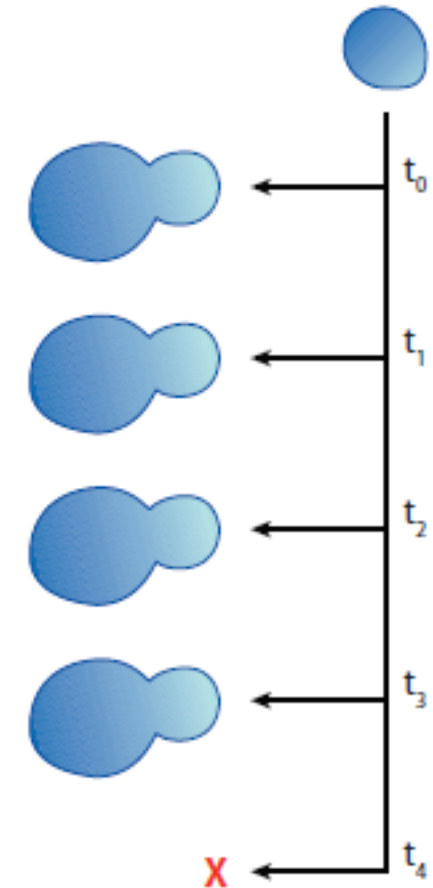
REPLICATIVE AGING

According to Steinkraus et al,(2008) the replicative aging can be defined as “the cumulative number of mitotic divisions a cell can undergo” (p.31).

Replicative life span =
number of mitotic divisions



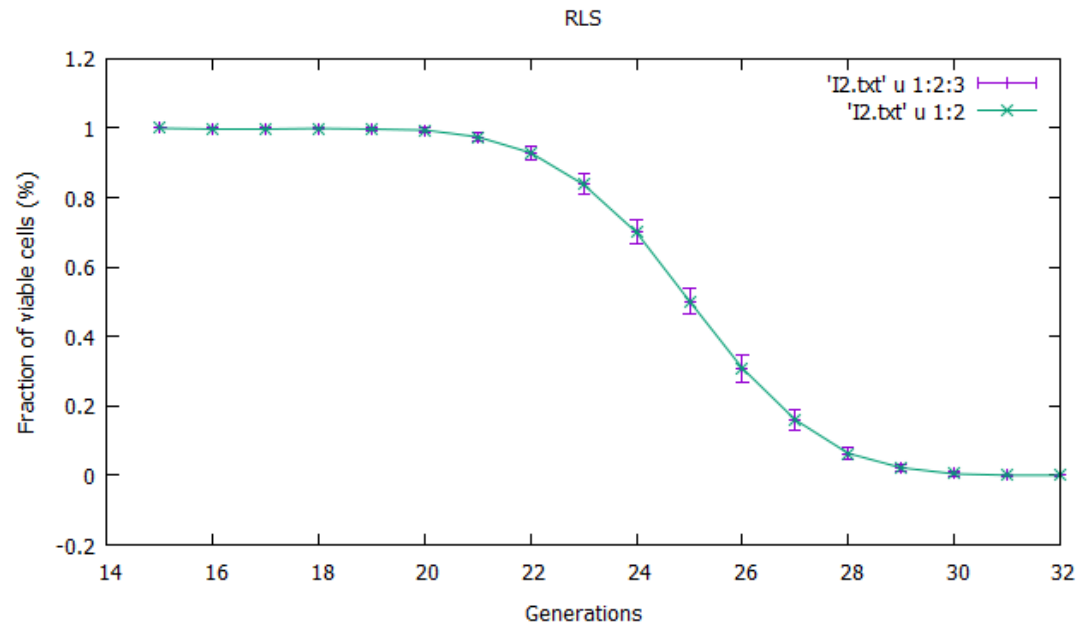
Chronological life span =
days viable in
postreplicative state



*Representation of replicative and chronological aging of yeast cell.
Image taken from: Replicative Aging in Yeast: The Means to the End*

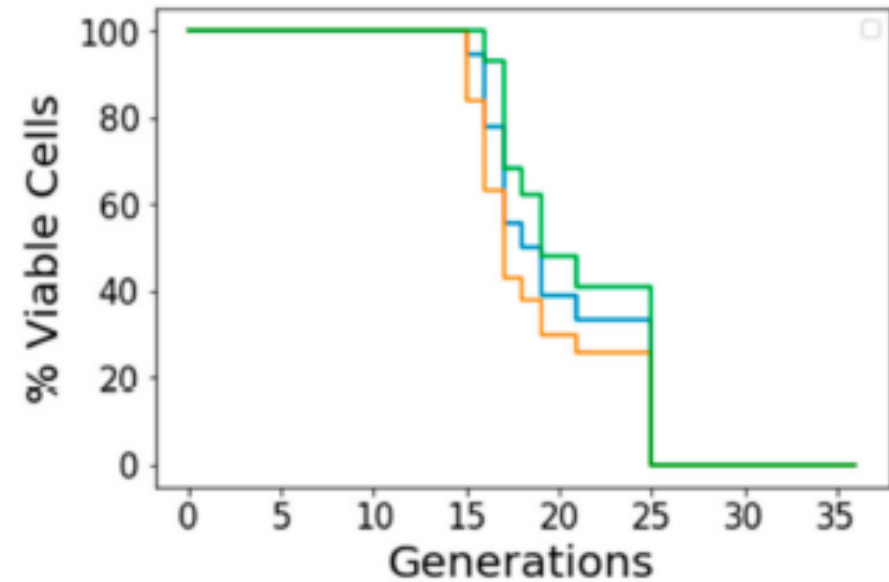
AGING CURVE

Simulation



The replicative life span is simulated using parameters obtained from a population of cells.

Experimental






Experimental aging curve . Image taken from: Slipstreaming Mother Machine: A Microfluidic Device for Single-Cell Dynamic Imaging of Yeas, Durán et al. (2021)

SLIPSTREAMING MOTHER MACHINE



Article

Slipstreaming Mother Machine: A Microfluidic Device for Single-Cell Dynamic Imaging of Yeast

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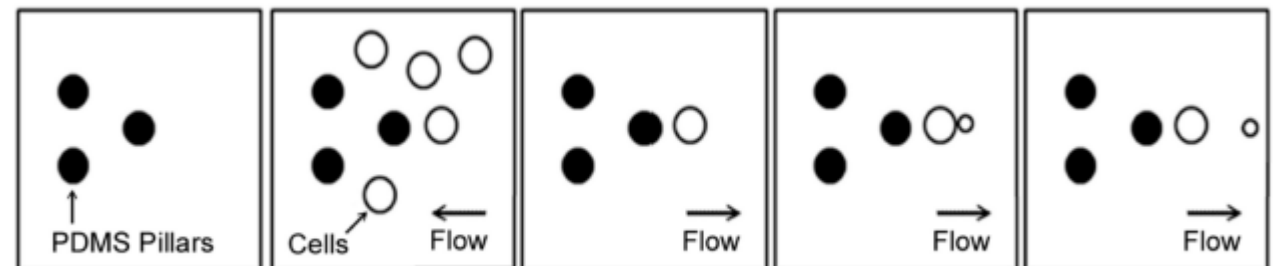
² Centro de Microelectrónica, Departamento de Ingeniería Eléctrica y Electrónica, Universidad de los Andes, CMUA, Bogotá 111711, Colombia; ca.hernandez11@u

³ Proyecto Curricular Licenciatura en Física, Facultad de José de Caldas, Bogotá 110311, Colombia

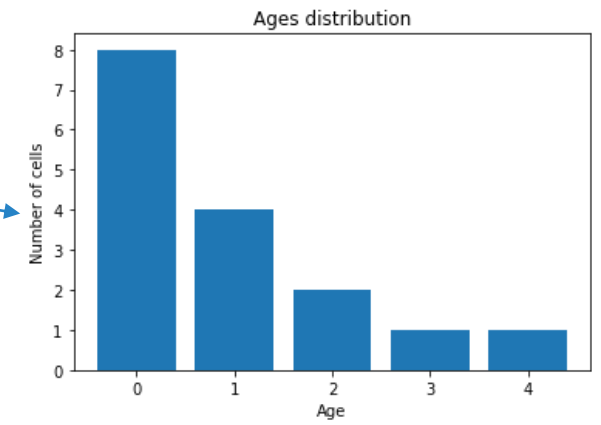
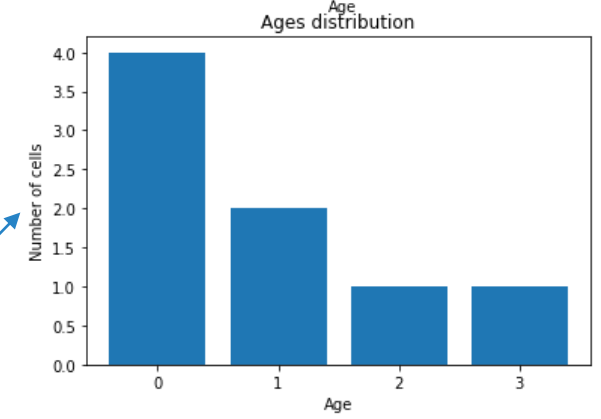
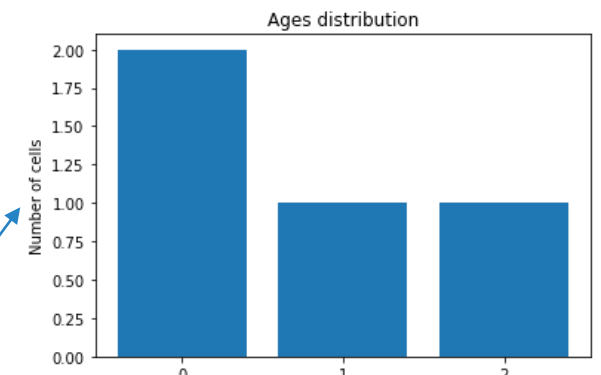
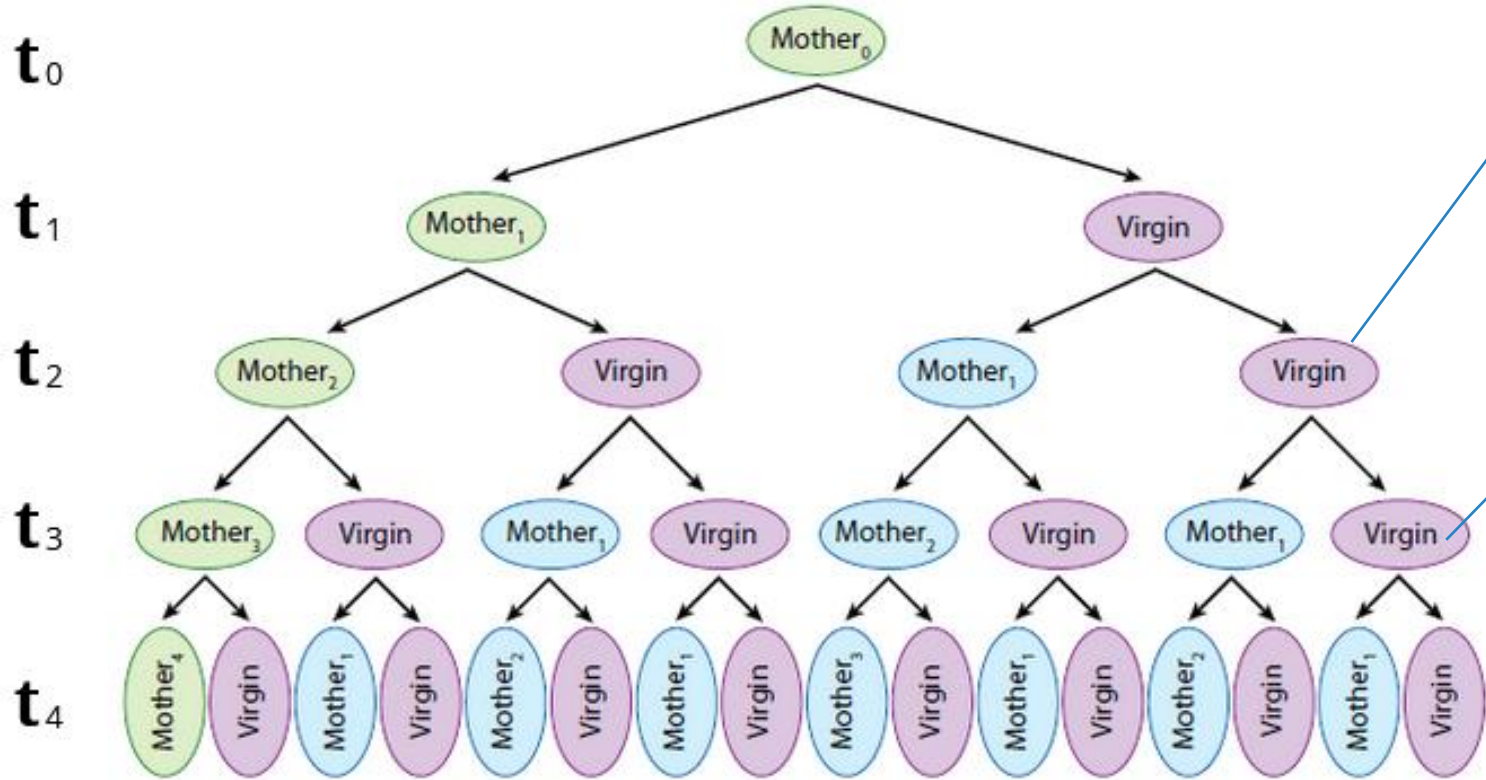
* Correspondence: da-duran@uniandes.edu.co (D.C.D.)
Tel.: +57-1-3394949 (ext. 5179 (COL)) (J.M.P.)

Abstract: The yeast *Saccharomyces cerevisiae* is one of aging and other phenomena such as division studied with the use of microfluidic devices to keep However. all of the existing devices trap cells mech

This has the advantage of not using mechanical pressure to trap the cells



HOW COULD THE AGE DISTRIBUTION AFFECT THE AGING CURVE?



RESULTS



MODELS

Implementation of two models developed in the Python programming language



GROWTH CURVE

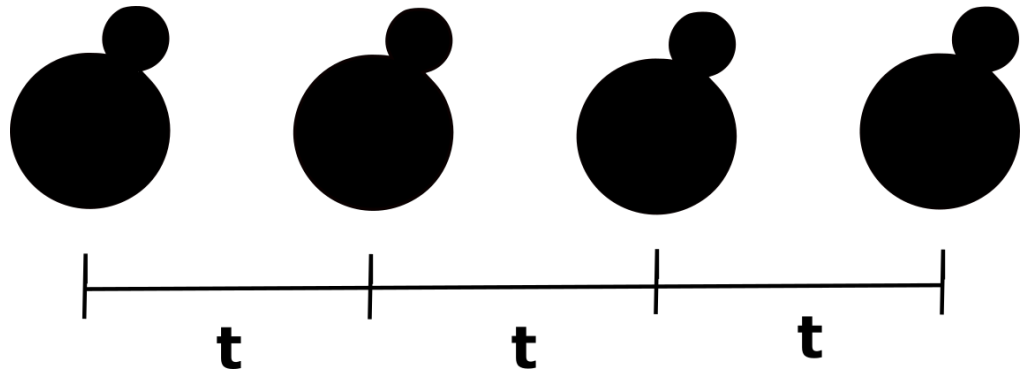
A code that simulates the exponential and stationary phases of yeast population growth.



AGE DISTRIBUTION

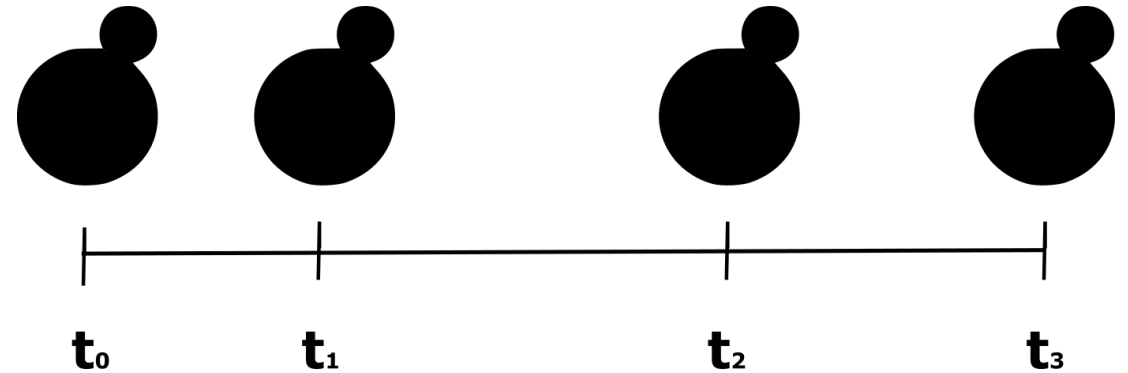
A code that simulates the distributions of ages, according to the division times, as a function of time.

SYNCHRONOUS MODEL



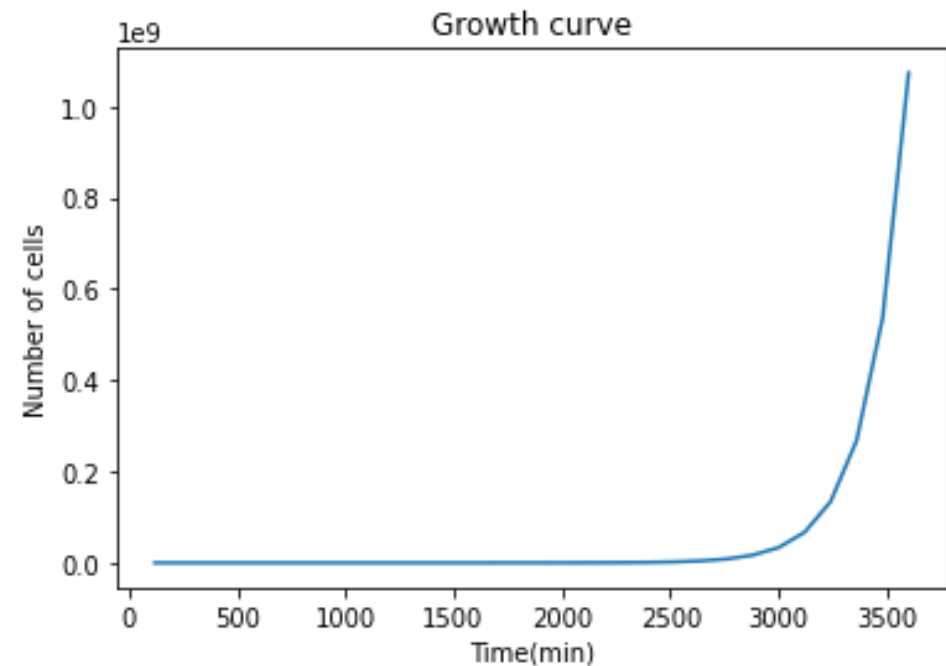
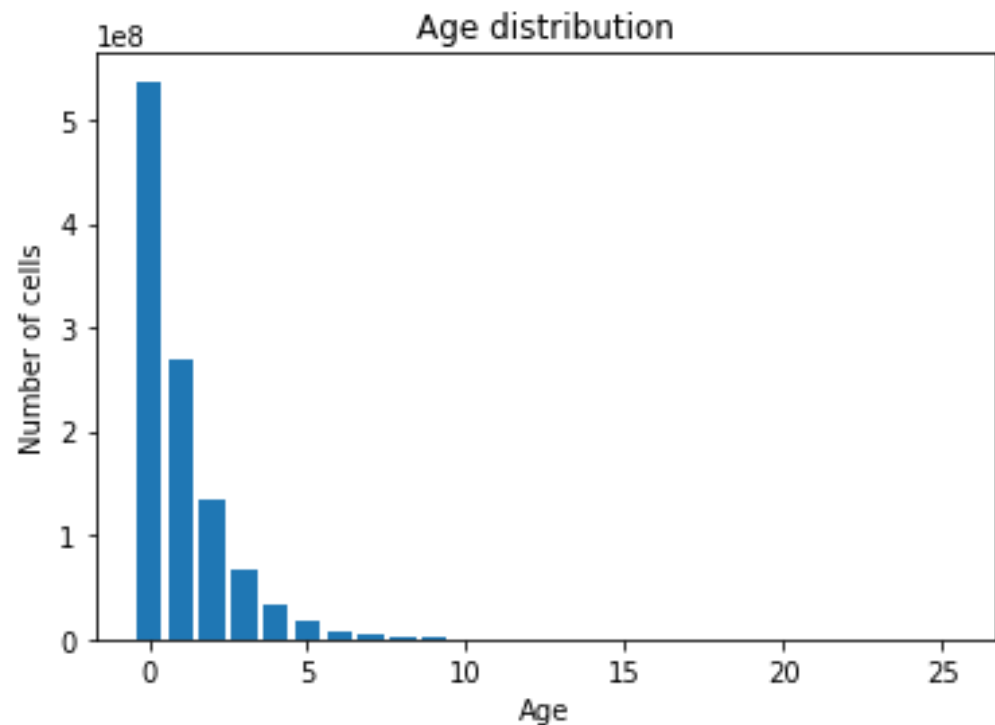
- The count begins with one single cell
- All the cells are replicated at the same time
- The age is defined as the number of daughter cells that each cell has had
- In this model the death times are deterministic

ASYNCHRONOUS MODEL



- This model includes the fact that cells do not divide at the same time
- We obtained the distribution of replication times from experimental data by fitting a normal distribution
- A replication time from that distribution it is assigned at each division
- We ignored mother-daughter correlations in division times

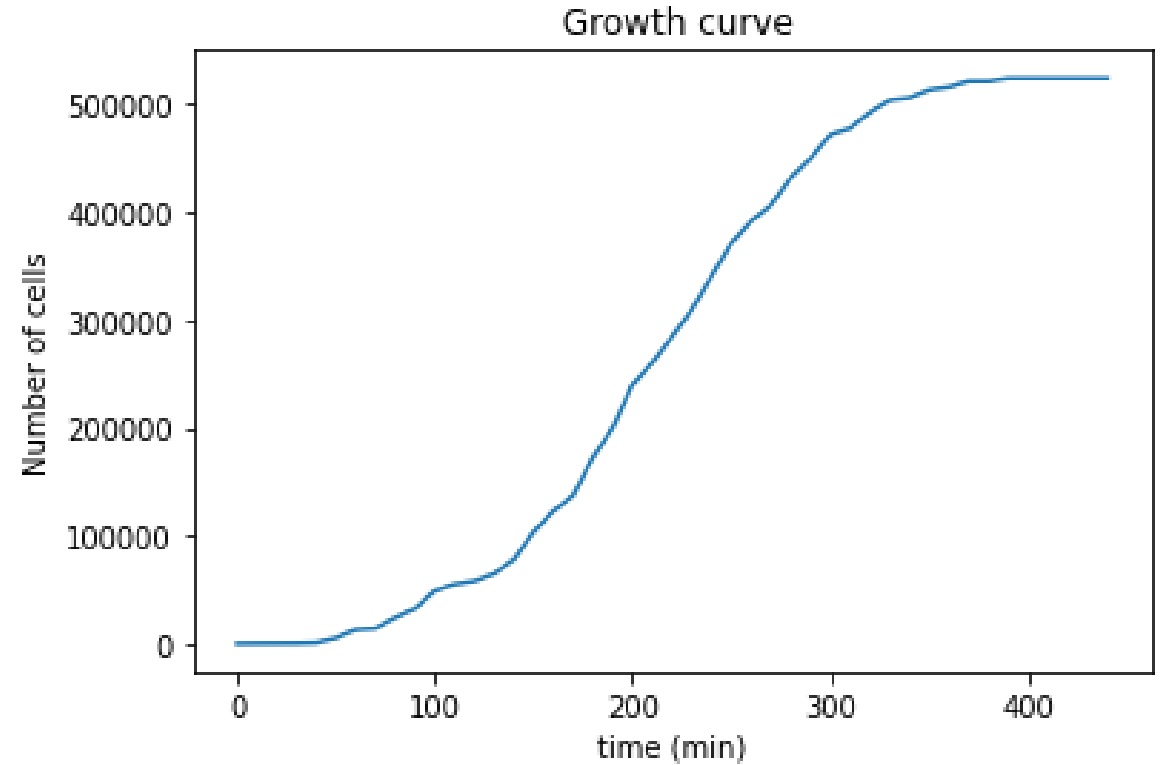
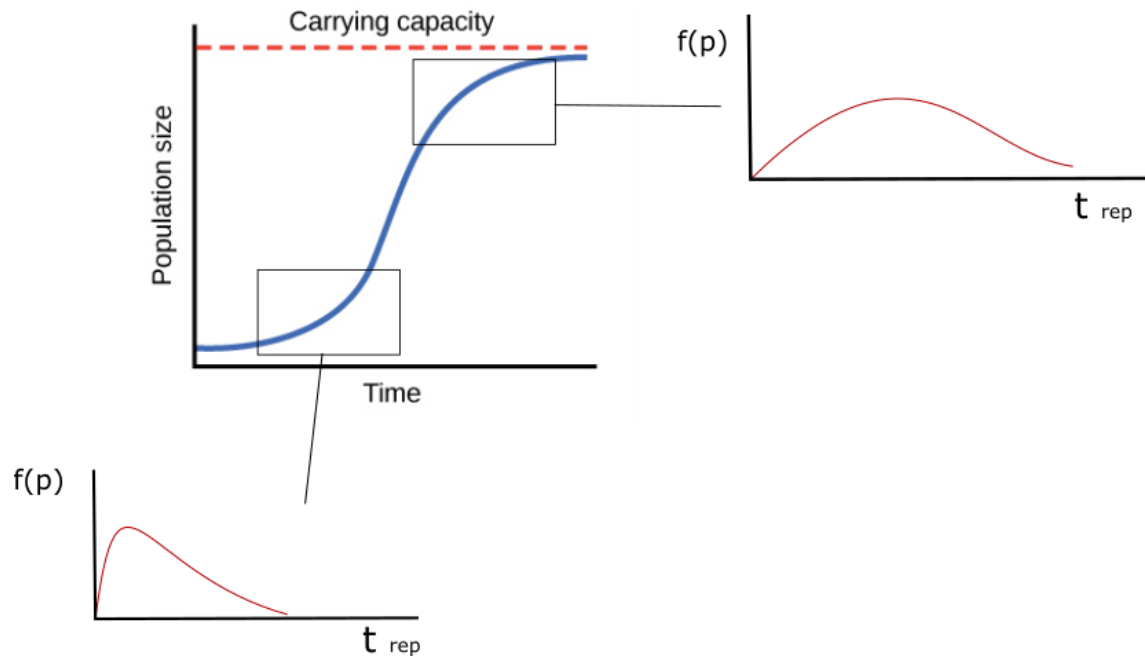
SYNCHRONOUS WITH CONSTANT REPLICATION TIMES



Cells	536870864	268435436	134217720	67108861	33554431	16777216	8388608	4194304	2097152	1048576	524288	262144	131072	65536	32768	16384	8192	4096	2048	1024	512	256	128	64	32	16
Age	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25

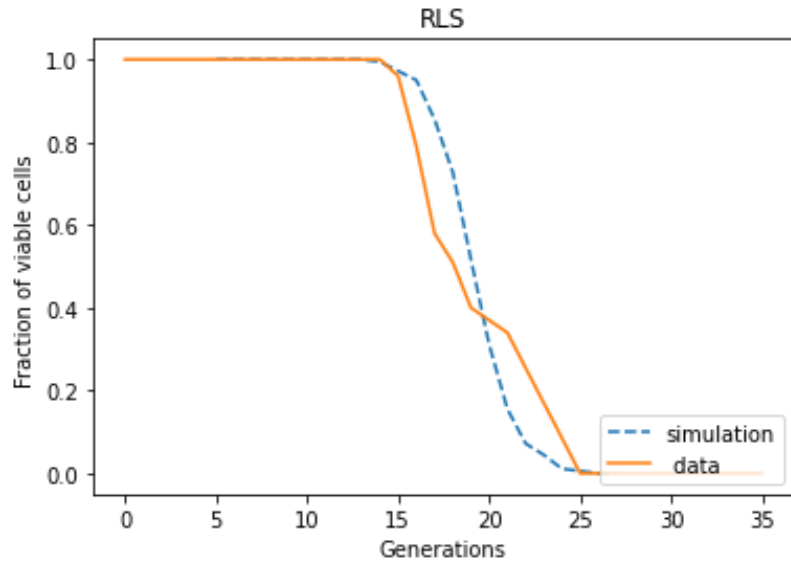
In this case we suppose that the cells died at the age of 25

ASYNCHRONOUS WITH INCREASING REPLICATION TIMES

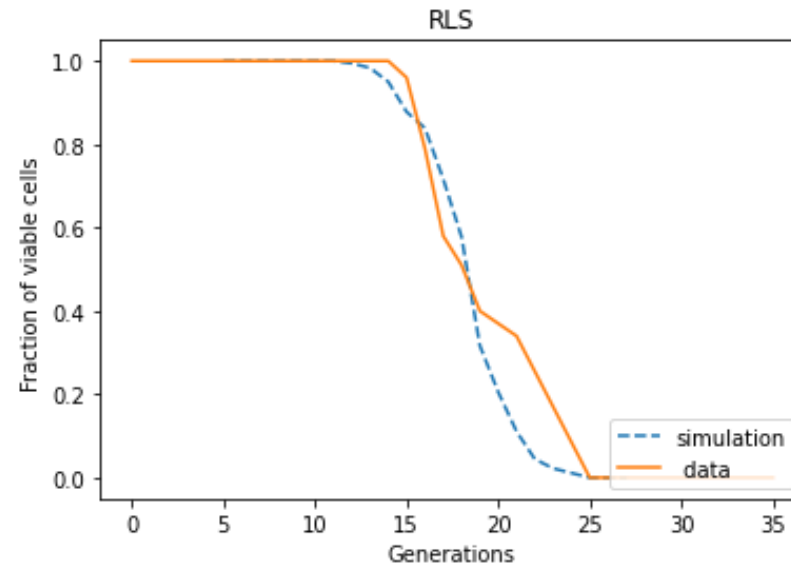


Cells	262144	131072	65536	32768	16384	8192	4096	2048	1024	512	256	128	64	32	16	8	4	2	1	1	0	0	0	0	0	0
Age	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25

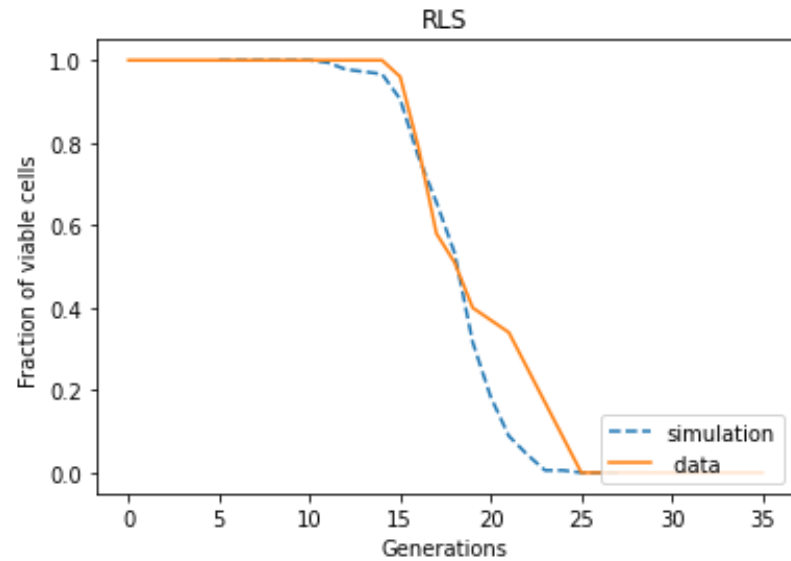
COMPARING SIMULATED RLS CURVES



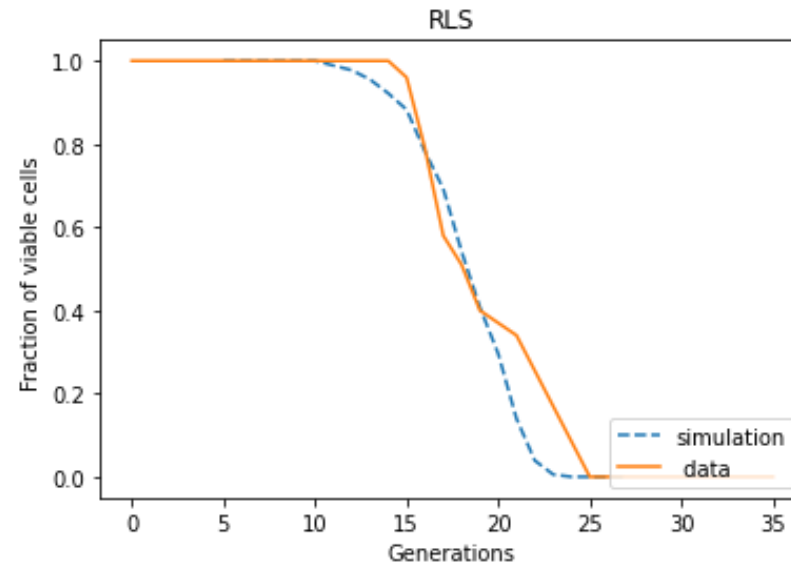
Virgin, $g=0$
medianRLS=19,3



$g=20$
medianRLS=18,1



$g=10$
medianRLS=18.3



$g=30$
medianRLS=18,2

We compare the RLS resulting from age distributions at different generations (0,10,20,30)
experimental data taken from: Durán et al. (2021)

DISCUSSION

- In the first approximation of the synchronous model, it cannot simulate the logistic behavior of the growth of a yeast cell population.
- Now, the being death is determined at a certain age, we plan to determine a distribution of death times to be added to the simulation.
- We want to adjust the asynchronous model with the experimental data to adjust the time at which the stationary phase is reached and thus improve the simulation.
- By having different age distributions in different generations, the RLS curves change, but they fit the experimental data.

CONCLUSIONS

We were able to determine an age distribution for the two models.

The initial model was adjusted to include death and fit the replicative times to more closely resemble the growth behavior of a yeast cell population.

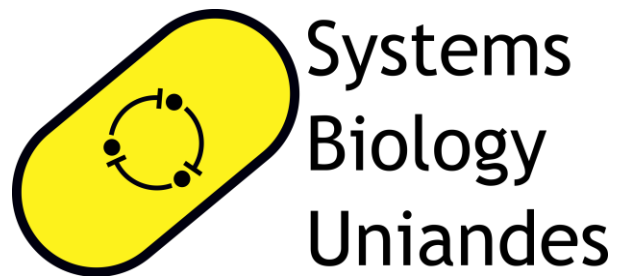
We have a model that simulates the growth of the population and gives the time in which a certain distribution may occur.

When considering a possible distribution of initial ages, in a first analysis, the RLS curve is affected, causing it to give a lower RLS value than expected.

ACKNOWLEDGEMENTS

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