

ADVANCED PROJECT I

DESIGN AND ANALYSIS NONLINEAR SYSTEMS- LOTKA VOLTERRA MODEL AND FIREFLY RESPONSE TO THE STIMULUS

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Abstract

In this report, firstly, we introduce dynamical systems, we explore predator-prey models in two dimensions, then we see the behaviour of the model by changing the parameters. We also explore predator-prey models for three dimensions, see the change in behaviour of the model by changing parameters. We also explore the model of the response of fireflies to the stimulus and analyze the behavior of the model in different frequencies and phase differences. In the first part, we introduce the term Dynamical system, give examples of dynamical systems, and get familiar with the basic terms used in the report. In the second part, we analyze the behavior of the model in the presence of one stimulus. In the third part, we analyze the behavior of the model in the presence of two stimuli. Finally, we give a conclusion for our analysis.

This document is the final derivable of the course Advanced Project 1 from Jacobs University master on Data Engineering. The purpose of the course as expressed in the program syllabus is to provide the student with an in-depth understanding and command of one of the data analytics or data management techniques that are represented by the research groups of the faculty of Data Engineering.

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Introduction

The dynamical system in simple words is a system whose state changes over time. The example includes the mathematical model that describes population growth, swinging of a pendulum, or predator-prey model. There are two types of dynamical systems: iterated maps and differential equations. The iterated maps are used to solve problems where time is discrete, whereas the differential equation is used to solve problems where time is continuous.

Now confining our attention to differential equations, differential equations can be divided into various types: Ordinary or Partial and Linear or Nonlinear, etc. In this report, we will deal with Ordinary differential equations and Nonlinear differential equations.

We can understand the idea of a dynamic system, ordinary and nonlinear differential equations by using an example of a population growth model. In any population, the number of individuals changes over time. The change in population can occur due to various reasons, for example, resource availability, competition, and disease. The simplest model describing changes in the population size is the exponential growth model. For simplicity, we will assume that the population does not interact with its environment and does not get affected by environmental changes. Therefore the equation will be

$$\frac{dn(t)}{dt} = rn$$

Where $\frac{dn(t)}{dt}$ is a first degree ordinary differential equation, where r is a growth rate and n in the population size. Figure 1, shows an exponential growth in population.

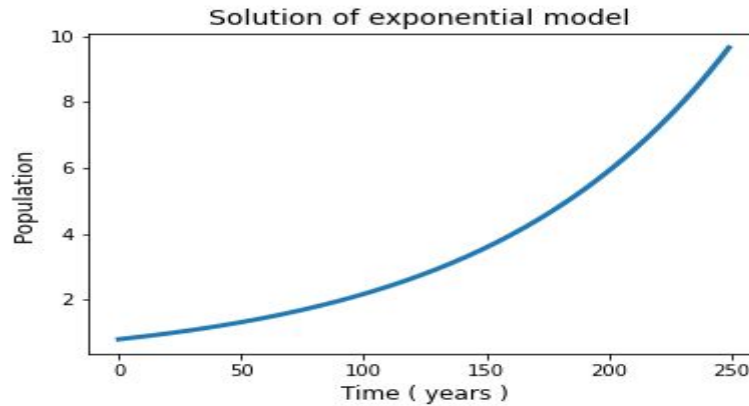


Figure 1: Exponential growth in population.

Predator-Prey Model

We will have a look at the Predator-Prey model, also known as the Lotka-Volterra model. This model is a pair of first order, nonlinear, differential equations used to describe the biological system. To be more precise, the dynamic of the system in which two species interact with each other, one being the predator and another being the prey. The model can be described as follows:

To design a model, we assume that there are two species - Rabbits and Foxes, where fox is a predator and rabbit is a prey, and fox feed on rabbits. Let the population of prey be $x(t)$ and the population of predator be $y(t)$. If there are no predators, then the population of prey will grow at the natural rate, let us consider a constant **a** as a parameter for the population growth of prey. [4] The equation is given by:

$$dx/dt = ax \Rightarrow x(t) = x_0 e^{at}$$

If there is no prey, then the population of the predator grows at the natural rate, let us consider a constant **b** as a parameter for the population growth of the predator. [4] The equation is given by:

$$dy/dt = -by \Rightarrow y(t) = y_0 e^{-bt}$$

When the population of prey and predator interact with each other, the predator will feed upon the prey, therefore, the population of predator and prey will change dependently, the population of predator increases when it feeds upon the prey and the population of prey decreases at the rate proportional to the frequency of interaction of prey and predator i.e xy .

Therefore, the resulting model is given by:

$$dx/dt = ax - pxy$$

$$dy/dt = qxy - by$$

Before using the ordinary differential equation, we will have a closer look at position equilibrium. The point x_0 is called an equilibrium point of $x' = f(x)$ if $f(x_0) = 0$. If $x(0) = x_0$ then $x(t) = x_0$ for all times. In our case, equilibrium occurs when growth rate is equal to 0, or there is no change in the growth rate over time. [4] For our model, we get two fixed points $(0, 0)$ and $\left(\frac{b}{q}, \frac{a}{p}\right)$.

Usually near these fixed points, we analyze the nature of the fixed points. Once the fixed point is identified, the next step is to determine if the fixed point is stable or unstable, this is known as stability analysis. We analyze the type of fixed point by first computing the Jacobian matrix.

$$J = \begin{bmatrix} a - py & -px \\ qy & qx - b \end{bmatrix}$$

For the fixed point $(0,0)$, the Jacobian is $J(0,0) = \begin{bmatrix} a & 0 \\ 0 & -b \end{bmatrix}$ and the linearization is $\frac{du}{dt} = au$ and $\frac{dv}{dt} = -bv$. So near a fixed point $(0,0)$ the number of rabbits increases and the population of foxes decreases. The fixed point is a saddle point.

Similarly for the fixed point $\left(\frac{b}{q}, \frac{a}{p}\right)$, the Jacobian is $J\left(\frac{b}{q}, \frac{a}{p}\right)$ has linearization $\frac{du}{dt} = -\frac{bp}{q}v$ and $\frac{dv}{dt} = -\frac{aq}{p}u$, whose eigenvectors are $\lambda^2 + ab = 0 \Rightarrow \lambda = \pm i\sqrt{ab}$. They are imaginary numbers. We have learned that purely imaginary eigenvalues in dynamical systems can

have several possible behaviours. However, in our model the fixed point is stable. The linearization therefore has a stable center.

Integrating the ODE

Now we use the above model to integrate the ODE over time. When we plot $x(t)$ and $y(t)$ individually over time, we see that the population curve is periodic and the variation of the predator population $y(t)$ slightly lags behind the population of the prey $x(t)$. The value set for $a=1$, $b=0.1$, $p=1.5$, $q=0.75$.

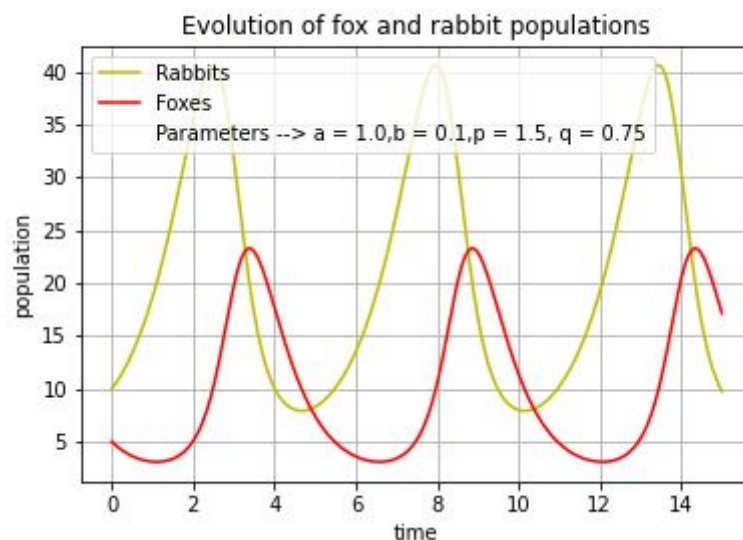


Figure 2: Evolution of fox and rabbit population.

Now we plot $x(t)$ and $y(t)$ individually over time by keeping the parameter a , b , and q constant but changing the parameter p . In figure 3, we set $p=0.5$, we see that the population curve is still periodic and the variation of the predator population $y(t)$ does not lag like figure 2. In figure 4, we set up $p=0.3$, we see that the population $x(t)$ and $y(t)$ has less curves. In figure 5, we set up $p=2.0$, where the population curve is periodic and the variation of the predator population $y(t)$ is slightly lagging behind.

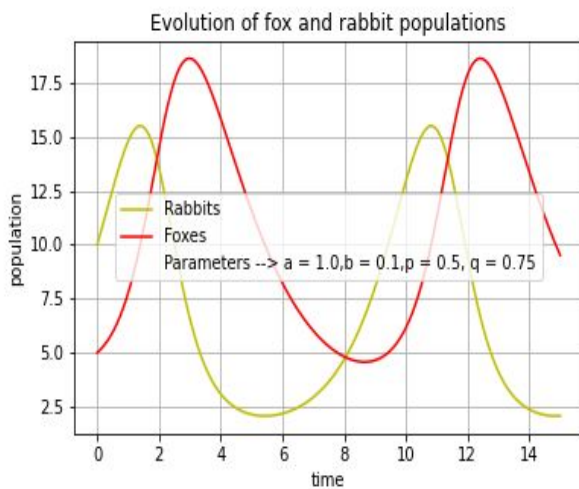


Figure 3: setting $p=0.5$

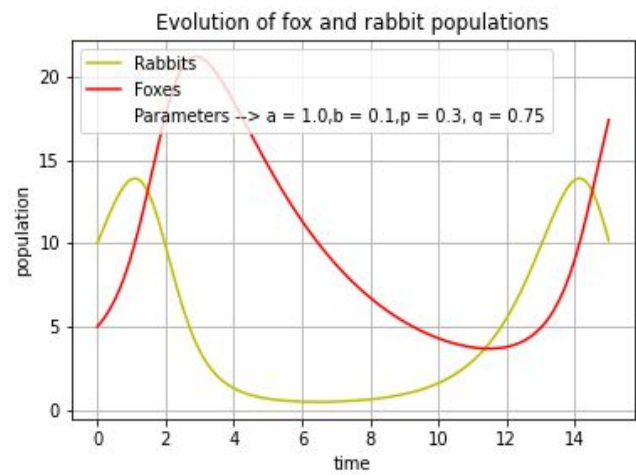


Figure 4: setting $p=0.3$

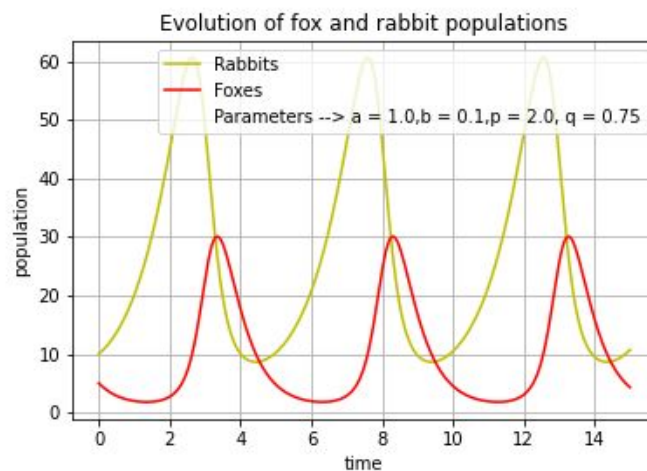


Figure 5: setting $p=2.0$

Direction field and trajectories in the phase plane

Now we use the above model to see the trajectories in a phase plane for different starting points between two fixed points. In figure 6, the graph shows five out of an infinite number of possible limit cycles, the graph also shows that by changing either the predator or the prey population, we have an unintuitive effect. [4] For example, in

order to reduce the number of rabbits, the number of foxes is increased, but this can increase the population of rabbits in the long, depending on the time.

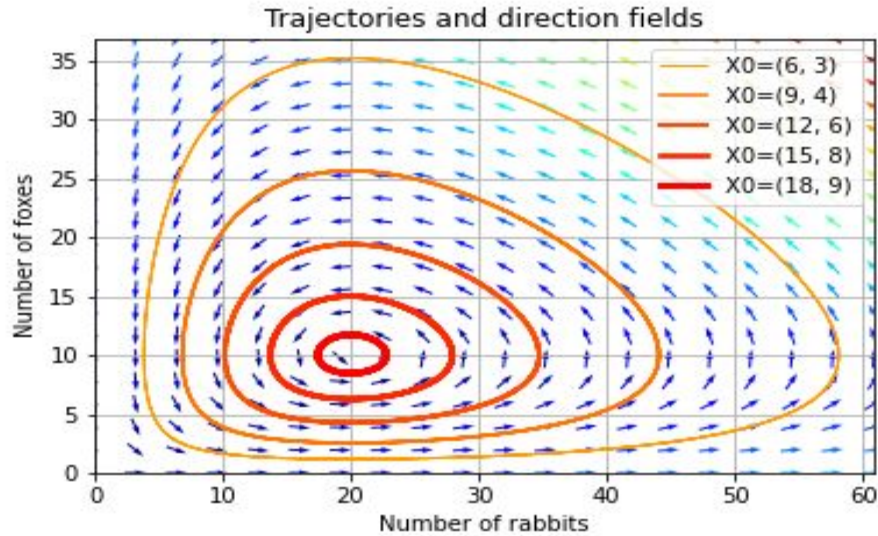


Figure 6: Trajectories and direction field of predator and prey interaction.

Three-Species Model

Now we extend the model from two species to the three species model. In the three species model, one species is at the bottom of the food chain, one species is at the middle and one is at the top of the. For example Lion, fox and rabbit. Let these species be x , y , and z . The model is given by:

$$dx/dt = ax - bxy$$

$$dy/dt = -py + qxy - eyz$$

$$dz/dt = -fz + gyz$$

Now we use the above model to integrate the ODE over time. When we plot $x(t)$, $y(t)$ and $z(t)$ individually over time, we see that the population curve is periodic for all $x(t)$, $y(t)$, and $z(t)$. Population of foxes is $y(t)$ has a near to constant rise and flow of the curve, however for population curve rabbit and lion, $x(t)$ and $y(t)$ respectively is rising towards the end of the graph. The value set for $a=1$, $b=0.1$, $p=1.5$, $q=0.75$, $e=0.5$, $f=0.75$, $g=0.1$.

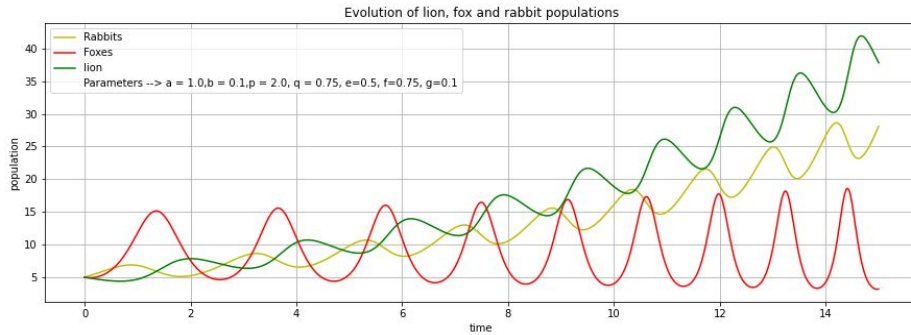


Figure 7: Population evolution of lion, fox and rabbit.

Now we plot $x(t)$, $y(t)$ and $z(t)$ individually over time by keeping the parameter a , b , p , e , and g constant but changing the parameter f . In figure 3, we set $f=0.3$, we see that the population curve is not periodic anymore. In figure 9, we set up $f=1.0$, we see that the population curve of lions $z(t)$ and the rabbits $x(t)$ has a tiny period whereas $y(t)$ does not show any change. In figure 10, we set up $f=2.0$, where the population curve is periodic and the variation of the population $y(t)$ is slightly lagging behind.

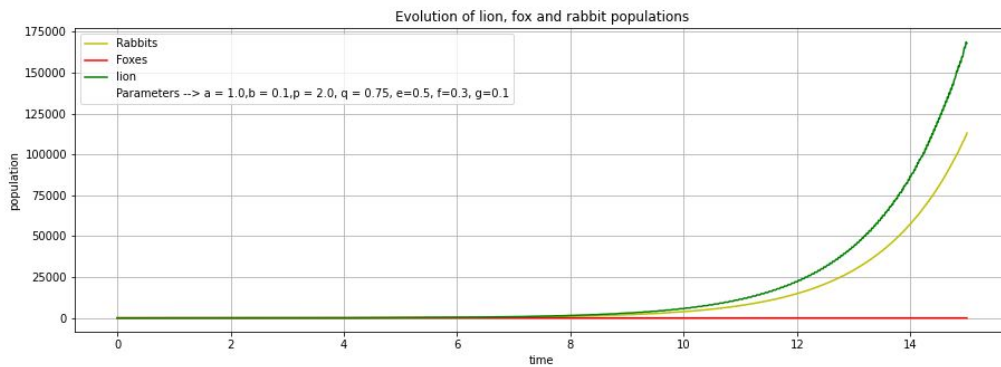


Figure 8: setting parameter $f=0.3$

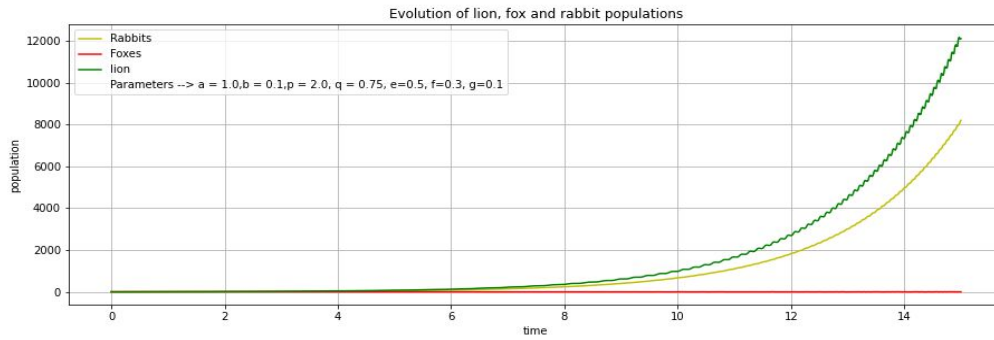


Figure 9: setting parameter $f = 1.0$

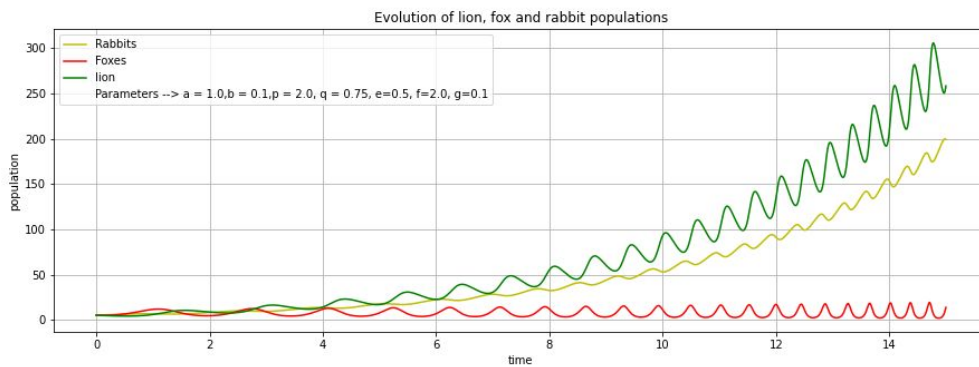


Figure 10: setting parameter $f = 2.0$

Direction field and trajectories in the three species model

Now we use the above model to see the trajectories for three species models for different starting points between 3 fixed points. In figure 11, the graph shows the direction field of the population of $x(t)$, $y(t)$ and $z(t)$. The graph also shows that by changing either the predator or the prey population, we have an unintuitive effect.

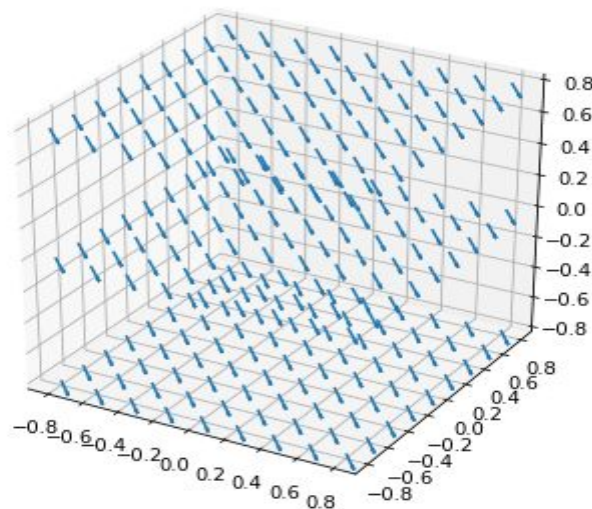


Figure 11: Direction field and trajectories in three species model

Synchronization in Fireflies

Synchronization - the operation or activity of two or more things at the same time or rate, often occurs in nature. For example, neurons in our brain, cardiac pacemaker cells, blinking of fireflies in unison.

Oscillators are a system which has some kind of periodic movement and shows steady state behaviour over a period of time [1]. In nature some systems become synchronized through the oscillation phenomena. Oscillator is a huge area of mathematics, therefore, we will only describe phase oscillators in this report.

Entertainment of fireflies

Fireflies are the soft bodied beetles commonly known as lightning bugs for their use of bioluminescence during twilight to attract mates. When the fireflies begin to imitate light. The fireflies blink at their own pace and frequency, however there comes a time when thousands of fireflies start flashing on and off all at once. Most of the fireflies can modify their natural frequencies to match up with the fireflies around it. Here, every firefly is trying to dynamically synchronize its frequency with that of every other firefly. This synchronization phenomenon has been studied by many scholars. We are taking a simplified model from the chapter on Firefly from the book "*Nonlinear Dynamics and Chaos*" -by Steven H. Strogatz. Here in the model

In this chapter, Strogatz examines a model where firefly responds to rhythm of a flashing stimuli. Let us assume that the phase of the firefly's flashing is given by $\theta(t)$ where $\theta=0$ is the instant when the firefly emits the flash. We also assume that θ is 2π -periodic. Therefore, $\theta = 2\pi n$. Then its natural frequency -the frequency at which it blinks without the presence of a stimuli be $\theta = \omega$. Similarly, now let us introduce a stimulus with 2π -periodic phase $\Theta(t)$ where $\Theta=0$ is the instant of the stimulus flashing, and $\Theta = \Omega$ is its frequency.

As we know that, firefly will attempt to synchronize with the stimulus, if the stimulus flashes after the firefly then firefly will attempt to slow. Similarly, if the stimulus is flashing before the firefly then the firefly will attempt to speed up to synchronize with the stimulus. We can describe this model with an equation as

$$\dot{\Theta} = \omega + A \sin(\Theta - \theta) \quad (1)$$

Where A is a measure of the ability of firefly to change its frequency in response to a stimulus, and $(\Theta - \theta)$ is a phase difference between the stimulus and the firefly. In order to have a better picture of a above equation (1), we plotted a graph by giving a initial value for ω , Ω and A . Figure 12, shows the curve for phases of firefly and the stimulus.

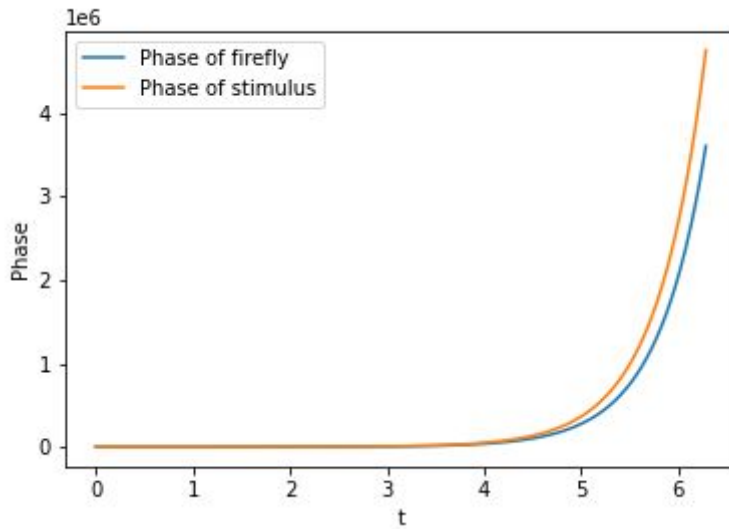


Figure 12: Phase of firefly and stimulus.

From this equation we can imply that if $0 < \Theta - \theta < \pi$, which means stimulus is ahead of the firefly, meaning $\dot{\Theta} > \omega$, therefore the firefly speeds up. Similarly, if $-\pi < \Theta - \theta < 0$, meaning $\dot{\Theta} < \omega$, and the firefly slows down.

Now that we have a model, we analyze the firefly's response to the stimulus. When the entertainment occurs, the difference in phase between the stimulus and the

firefly ϕ approaches a fixed constant. This can show various behaviour in the system : phase synchronization, phase locked, phase drift. Using equation,

$$\dot{\phi} = \Omega - \omega - A \sin(\phi) \quad (2)$$

If the constant is 0, that means that the firefly has synchronized with the stimulus and the firefly and stimulus are flashing together. We can show this phenomenon by solving equation (2) and plotting the graph for ϕ vs time, where time is in the range of 0 to 2π splitted into 1000, and the initial value for ω and Ω is 11.5564 Hz. Figure 13 shows the phase synchronization.

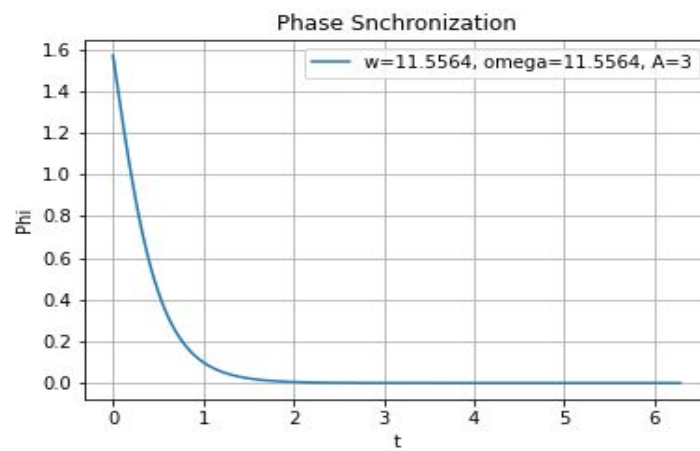


Figure 13: Phase Synchronization

Similarly, if the constant is not 0 then it can show two different behaviours. The firefly can either be in a phase locked to the stimulus, meaning that firefly and the stimulus have the same instantaneous frequency, but the firefly will always flash behind or ahead by a fixed amount. Figure 14 shows Phase locked behaviour.

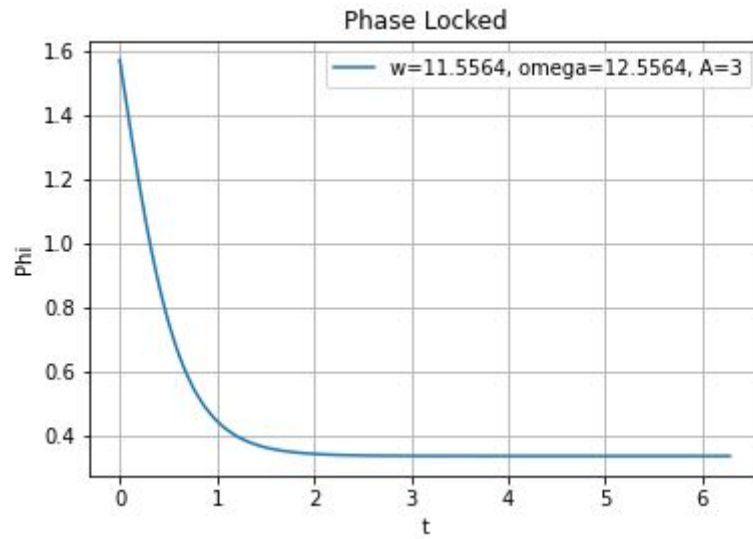


Figure 14: Phase locked.

Finally, if the constant is not 0 and if the frequency of the stimulus is too high or too low then firefly will struggle to match the frequency of the stimulus, and thus entrainment will not occur, this behaviour is known as phase drift. Figure 15 shows the phase drift behaviour.

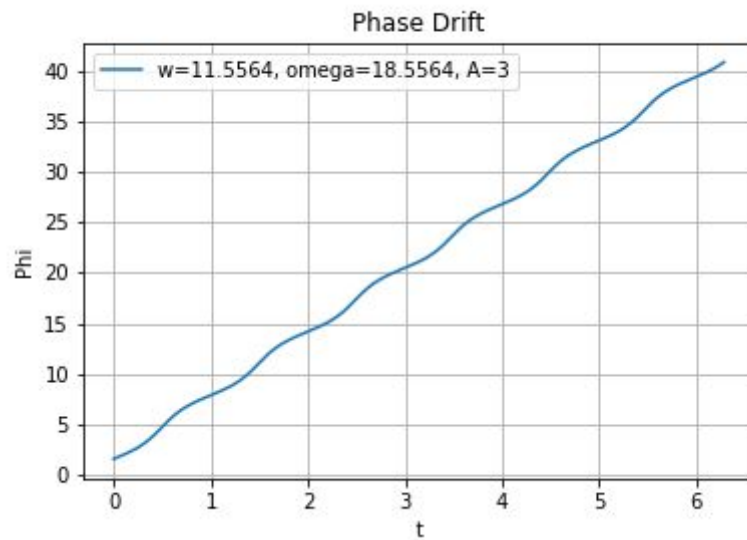


Figure 15: Phase drift

Fixed Point and Stability Analysis

In previous figures (3,4,5) we saw the phase difference trajectories of different behaviour in the system. Now, to analyze the stability of these behaviours, we can examine the stability of the fixed points. Fixed points are defined by $f(x)=0$, it is a point which remains stagnant when the system changes along with the time. We can examine the stability by plotting a graph for $\dot{\phi}$ versus ϕ using equation (2). For each type of behaviour, there are different types of stability of the fixed points. For figure 16, stable fixed point is at $\phi = 0$ which corresponds to phase synchronization, for figure 17, stable fixed point is at $\phi = c$ where c is a real constant, which corresponds to phase locked, and finally in figure 18 there are no stable fixed points.

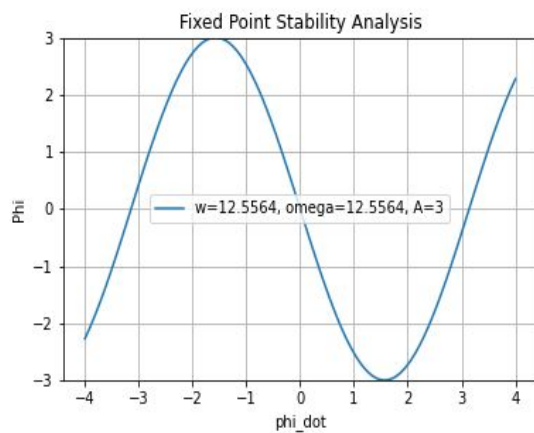


Figure 16: Phase Synchronization

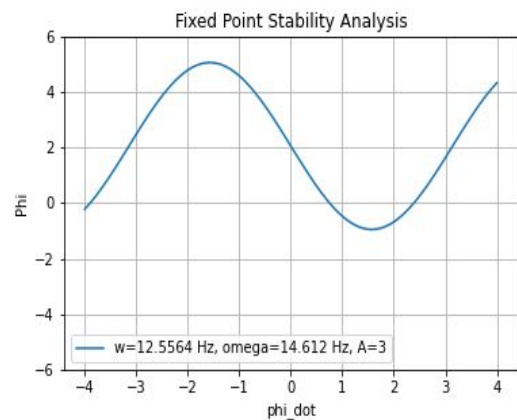


Figure 17: Phase Locked

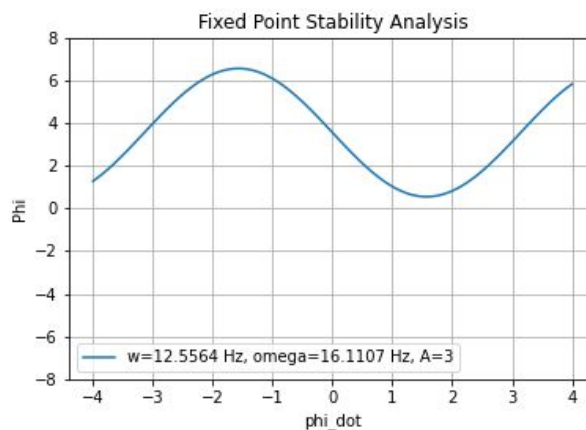


Figure 18: Phase Drift

In order for entertainment to occur , ϕ should move along with c as time move along with the infinity, here c can be any real number including 0, therefore, for the simplicity we consider $c=0$, which means fixed point , $\phi' = 0$. From equation (2), we can write,

$$0 = \Omega - \omega - A \sin(\phi) \quad (3)$$

$$\sin(\phi) = \Omega - \omega / A \quad (4)$$

We know that $-1 \leq \sin(\phi) \leq 1$, therefore equation 4 will be,

$$-1 \leq \Omega - \omega / A \leq 1$$

$$-A \leq \Omega - \omega \leq A$$

$$\omega - A \leq \Omega \leq \omega + A \quad (5)$$

Therefore, from equation (5) we can say that for the entertainment to occur omega should be between $\omega - A$ and $\omega + A$.

We also found out that parameter A also affects the behaviour of the system. According to the plot, larger values of A cause phase synchronization, shown in figure 19, whereas smaller values cause phase drift and phase locking shown in figure 20 and figure 21 respectively.

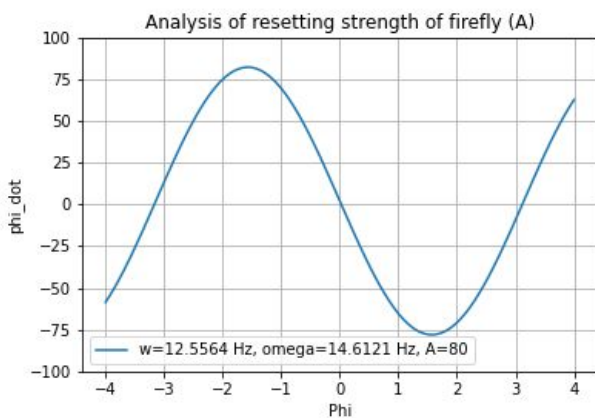


Figure 19: Phase Synchronization

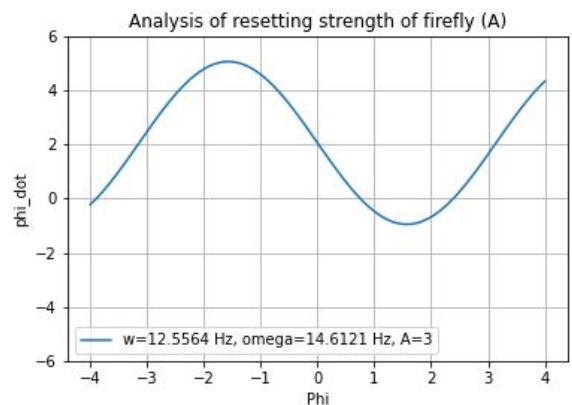


Figure 20: Phase Locked

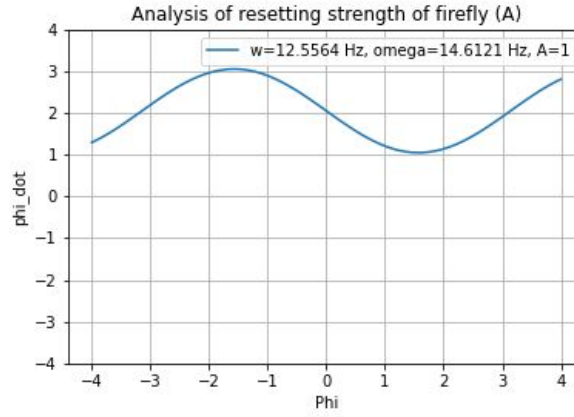


Figure 21: Phase Drift

Firefly synchronization with two stimuli

So far, we were working with a simple model which had one firefly and one stimulus. We can call this a single stimulus case. But how will the fireflies react to the two different stimuli - stimulus 1 and stimulus 2 represented by Θ_1 and Θ_2 respectively, meaning two different frequencies? We can explore this case by building another model. The model will be the same as the equation (1). For both the models we will use the same parameters for phase of firefly but two different phases for two different stimuli. Let the frequencies of stimulus 1 $\Theta_1 = \Omega_1$ and the frequency of stimulus 2 $\Theta_2 = \Omega_2$. Let the phase difference between stimulus 1 and firefly and phase difference between stimulus 2 and firefly be $\phi_1 = \Theta_1 - \theta$ and $\phi_2 = \Theta_2 - \theta$ respectively. Let us suppose a model that predicts the phase of the firefly change be

$$\dot{\theta} = \omega + A \sin(\phi_1) + A \sin(\phi_2) \quad (6)$$

We can call equation (6), *multiple stimuli models*. We know that $\phi_1 = \Theta_1 - \theta$ and $\phi_2 = \Theta_2 - \theta$, therefore we can write equation (6) as

$$\dot{\phi}_1 = \Omega_1 - \omega - A \sin(\phi_1) - A \sin(\phi_2) \quad (7)$$

$$\dot{\phi}_2 = \Omega_2 - \omega - A \sin(\phi_1) - A \sin(\phi_2) \quad (8)$$

Suppose frequencies for stimulus 1 and 2 both are same, then $\Omega_1 = \Omega_2$. As a result $\phi_1 = \phi_2$. This will be the case where we will not have different types of stimuli, the model will act as a single stimulus model.

However, if the frequencies for stimulus 1 and 2 both are not same, then Ω_1 is not equal to Ω_2 . In this case, when the firefly will begin to change its frequency to match that of stimulus 1, phase difference ϕ_2 will increase or decrease, but then phase difference ϕ_2 will also increase or decrease along with the ϕ_2 . Therefore, the trajectories of ϕ_1 and ϕ_2 will be uniformly increasing and decreasing. Thus the firefly will never synchronize with the stimuli, but instead will cause phase drift with both the stimuli.

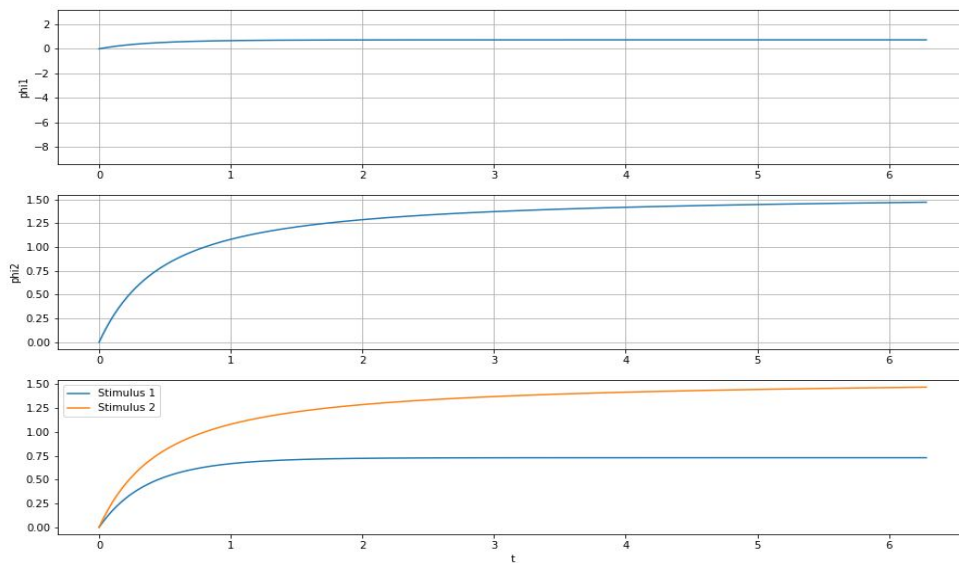


Figure 22: Firefly behaviour with two stimuli

Conclusion

We tried to understand non-linear dynamics by studying the lotka volterra model. First, we studied the lotka volterra model in two dimensions, we calculated the fixed points of the system and then analyzed the behaviour of the fixed point. Similarly, we studied the lotka volterra model in three dimensions and analysed the behaviour of the model by changing parameters. Secondly, we tried to understand firefly synchronization in presence of the stimulus. In the model which had only one stimulus, showed three different kinds of behaviour - Phase synchronization, Phase locked and Phase drift. Phase synchronization only occurs when frequency of the firefly and the stimulus is the same, otherwise either the firefly and stimulus and in phase locked state or phase drift state. We also tried to analyze behaviour in the change in the resetting strength (A) of the firefly. We found out that large resetting strength may result in phase synchronization, however, less A will cause either phase lock or there will be no entertainment at all. Finally we tried to see the behaviour of the firefly when two stimuli are present. From the model we found out that in presence of two stimuli, if the frequencies of both the stimuli is same then synchronization can occur, otherwise they will not synchronize.

References

- [1] The Oscillator Principle of Nature- A simple Observation
- [2] Runyeon, Hope. "Firefly Synchronization." (2006).
- [3] Strogatz, S. H. Nonlinear Dynamics and Chaos, with Applications to Physics, Biology, Chemistry, and Engineering. Reading, MA: Addison-Wesley, 1994.
- [4] Canale, Raymond P. "An analysis of models describing predator-prey interaction." *Biotechnology and Bioengineering* 12.3 (1970): 353-378.

Code Repository

All the code implemented and utilized during the execution of the process described in this report is available at the GitHub repository:

https://github.com/sumansanyukta/Modelling-and-Analysis-of-Complex-System/blob/master/firefly_synchronization.ipynb