

Mathematical Modeling of Population Growth Using Differential Equations

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Abstract- Population growth is a central topic in mathematical biology and ecological modeling. This paper analyzes population dynamics using exponential, logistic, and extended nonlinear models. The study formulates first-order differential equations to describe how populations evolve over time under different environmental constraints. Analytical solutions are derived for classical models, while advanced extensions such as harvesting effects, Allee thresholds, time-dependent growth rates, delayed feedback systems, and predator-prey interactions are also introduced. Results show that exponential models describe ideal conditions, while logistic and nonlinear models provide realistic and stable long-term population behavior.

Keywords: Logistic Growth, Exponential Growth, Carrying Capacity, Stability Analysis, Ecological Systems, Population Biology.

I. INTRODUCTION

Mathematical modeling has become an essential tool for understanding complex natural and social systems. Among these, population growth is one of the most widely studied phenomena in applied mathematics and theoretical biology. Population dynamics describe how the number of individuals in a species changes over time due to birth rates, death rates, and environmental influences.

Differential equations provide a powerful framework for modeling these changes because they describe the rate at which a system evolves continuously. If P represents the population at time t , then the general population growth model can be expressed as:

$$\frac{dP}{dt} = f(P, t)$$

This equation describes how population changes as a function of both time and population size.

The simplest case assumes that the growth rate is proportional to the current population:

$$\frac{dP}{dt} = rP$$

where r is the intrinsic growth rate. The solution to this equation leads to exponential growth:

$$P(t) = P_0 e^{rt}$$

However, this model assumes unlimited resources, which is unrealistic in real ecosystems.

To introduce environmental constraints, the logistic growth model is used:

$$\frac{dP}{dt} = rP \left(1 - \frac{P}{K} \right)$$

where K is the carrying capacity of the environment. This model captures population saturation due to limited resources.

In more realistic ecological settings, additional effects such as migration can be included:

$$\frac{dP}{dt} = rP \left(1 - \frac{P}{K} \right) + I - E$$

where I represents immigration and E represents emigration.

Population dynamics are also influenced by interactions between species. A basic predator-prey interaction can be represented as:

$$\frac{dP}{dt} = aP - bPQ$$

$$\frac{dQ}{dt} = -cQ + dPQ$$

where is prey population and is predator population. In real-world ecosystems, population growth is influenced by multiple interacting factors including limited resources, environmental resistance, species interaction, migration, and sudden environmental changes. These complexities require nonlinear differential equations for accurate modeling.

Furthermore, population models are not restricted to ecological systems alone. They are extensively used in epidemiology to study disease spread, in economics to analyze human population trends, and in environmental science to evaluate sustainability of natural resources.

This paper focuses on developing a structured understanding of population growth models starting from basic exponential growth and progressing toward advanced nonlinear systems. It also highlights the importance of stability analysis and real-world applications in interpreting model behavior.

II. MATHEMATICAL FOUNDATIONS

2.1 Continuous Population Assumption

Let be a continuous function representing population size.

The instantaneous growth rate is:

$$\frac{dP}{dt}$$

This represents net change per unit time.

2.2 General Growth Equation

$$\frac{dP}{dt} = f(P, t)$$

If environment is time-independent:

$$\frac{dP}{dt} = f(P)$$

This simplifies analysis into autonomous systems.

2.3 Equilibrium Concept

Equilibrium occurs when:

$$\frac{dP}{dt} = 0$$

These points determine long-term system behavior.

III. EXPONENTIAL GROWTH MODEL

3.1 Derivation

Assume growth rate proportional to population:

$$\frac{dP}{dt} = rP$$

3.2 Solution Steps

Separate variables:

$$\frac{dP}{P} = r dt$$

Integrate:

$$\ln P = rt + C$$

Exponentiate:

$$P(t) = P_0 e^{rt}$$

3.3 Growth Interpretation

- If : unlimited growth
- If : exponential decay
- No environmental restriction included

3.4 Limitation

Fails for long-term systems due to infinite growth assumption.

IV. LOGISTIC GROWTH MODEL

The logistic growth model is one of the most important models in population dynamics because it introduces environmental constraints into population growth. Unlike exponential growth, it explains how real populations stabilize over time due to limited resources.

4.1 Motivation

In real biological systems, populations cannot grow indefinitely. Every ecosystem has a limited amount of resources such as food, water, habitat space, and energy. Additionally, competition among individuals increases as population size grows.

These limitations create a natural resistance to growth. Initially, when population size is small, resources are abundant and growth is fast. However, as population increases, competition intensifies, leading to a slowdown in growth. This motivates the development

of the logistic model, which incorporates environmental resistance into mathematical form.

4.2 Model Equation

The logistic growth model is defined by the differential equation:

$$\frac{dP}{dt} = rP \left(1 - \frac{P}{K} \right)$$

where:

- $P(t)$ = population at time t
- r = intrinsic growth rate
- K = carrying capacity of the environment

This equation modifies exponential growth by introducing a limiting factor $(1 - P/K)$, which reduces growth as population approaches K .

4.3 Biological Meaning (Theoretical Interpretation)

The logistic model consists of two important biological components:

- rP : represents natural reproduction or intrinsic growth of the population under ideal conditions.
- $1-(P/K)$: represents environmental resistance, which increases as population approaches the carrying capacity.

This structure reflects a feedback mechanism where population growth slows automatically when resources become scarce, making the model biologically realistic (Verma, 2017).

4.4 Solution Derivation

The solution of the logistic differential equation is obtained using separation of variables:

$$P(t) = \frac{K}{1 + Ae^{-rt}}$$

where:

$$A = \frac{K - P_0}{P_0}$$

and P_0 is the initial population.

This solution shows that population growth follows an S-shaped (sigmoid) curve, starting slowly, increasing rapidly, and finally stabilizing.

4.5 Theoretical Behavior Analysis

The logistic model exhibits three distinct growth phases:

- Initial phase: When $P \ll K$, the term $1 - P/K \approx 1$, so growth behaves almost exponentially.
- Intermediate phase: As population increases, competition for resources becomes significant, causing growth to slow down.
- Final phase: The population approaches a stable equilibrium at carrying capacity K , where growth becomes zero.

This behavior represents a self-regulating system, where internal feedback controls population size naturally.

4.6 Equilibrium Points and Their Meaning

Equilibrium points are obtained by setting:

$$\frac{dP}{dt} = 0$$

This gives:

$$P = 0, \quad P = K$$

The equilibrium at $P=0$ represents extinction, while $P=K$ represents a stable sustainable population level supported by the environment.

These equilibrium points help in understanding long-term system behavior.

4.7 Stability Analysis (Theoretical Insight)

Stability describes whether small disturbances in population will grow or disappear over time.

For the logistic model:

- $P=0$ is unstable equilibrium, meaning any small population will grow away from zero if resources are available.
- $P=K$ is stable equilibrium, meaning population naturally returns to carrying capacity after small disturbances.

This stability property makes the logistic model highly realistic for ecological systems.

4.8 Theoretical Significance of Logistic Model

The logistic model is significant because it introduces the concept of self-regulation in biological systems. Unlike exponential growth, it demonstrates that

natural systems tend to stabilize rather than grow indefinitely.

It also forms the foundation for more advanced models such as:

- Harvesting models
- Competition models
- Predator-prey systems
- Delayed population models

Thus, the logistic equation is a cornerstone in mathematical biology and nonlinear dynamics.

4.9 Summary Insight

The logistic growth model successfully bridges the gap between ideal mathematical growth and real ecological systems. It explains how populations grow rapidly under favorable conditions but eventually stabilize due to environmental limitations and resource constraints.

V. STABILITY THEORY

5.1 Linearization

Let:

$$\frac{dP}{dt} = f(P)$$

Stability depends on:

$$J = \frac{df}{dP}$$

5.2 Stability Rules

- : stable equilibrium
- : unstable equilibrium

VI. EXTENDED MATHEMATICAL MODELS

6.1 Harvesting Model

$$\frac{dP}{dt} = rP - h$$

Solution:

$$P(t) = \left(P_0 - \frac{h}{r}\right)e^{rt} + \frac{h}{r}$$

Interpretation:

Constant removal reduces long-term population.

6.2 Logistic Model with Harvesting

$$\frac{dP}{dt} = rP \left(1 - \frac{P}{K}\right) - h$$

Used in fisheries and wildlife management.

6.3 Allee Effect Model

$$\frac{dP}{dt} = rP \left(\frac{P}{A} - 1\right) \left(1 - \frac{P}{K}\right)$$

Behavior:

- Below: extinction
- Above: growth

6.4 Time-Dependent Growth Rate

$$\frac{dP}{dt} = r(t)P$$

Where:

$$r(t) = r_0 e^{-kt}$$

Solution:

$$P(t) = P_0 \exp\left(\frac{r_0}{k}(1 - e^{-kt})\right)$$

6.5 Delayed Logistic System

$$\frac{dP}{dt} = rP(t) \left(1 - \frac{P(t-\tau)}{K}\right)$$

Importance:

Introduces biological delay effects (maturity time).

6.6 Predator-Prey Model (Lotka-Volterra)

$$\frac{dP}{dt} = rP - aPQ$$

$$\frac{dQ}{dt} = -sQ + bPQ$$

Interpretation:

- : prey
- : predator

6.7 Nonlinear Interaction Effects

System becomes oscillatory due to interaction feedback loops.

6.8 Normalized Logistic Model

Let:

$$x = \frac{P}{K}$$

Then:

$$\frac{dx}{dt} = rx(1 - x)$$

VII. PHASE PLANE ANALYSIS AND SYSTEM DYNAMICS

Phase plane analysis is used to understand how population systems behave when more than one variable is involved. Instead of solving only time-based equations, this method studies system behavior using graphical interpretation.

A general two-variable population system is written as:

$$\frac{dP}{dt} = f(P, Q), \quad \frac{dQ}{dt} = g(P, Q)$$

In predator-prey systems, the model is:

$$\frac{dP}{dt} = rP - aPQ$$

$$\frac{dQ}{dt} = -sQ + bPQ$$

These equations describe how two species interact, where one is the prey and the other is the predator. Equilibrium points are found by setting:

$$\frac{dP}{dt} = 0, \quad \frac{dQ}{dt} = 0$$

These points help identify stable or unstable population conditions.

VIII. REAL-WORLD APPLICATIONS

Mathematical models are widely used in real-world systems to understand population behavior and resource management.

8.1 Ecological Population Growth

The most common model used in ecology is the logistic equation:

$$\frac{dP}{dt} = rP \left(1 - \frac{P}{K}\right)$$

This model explains how populations grow quickly at first and then stabilize due to limited resources.

8.2 Disease Spread Modeling

Population models are also used in epidemiology. A basic infection model is:

$$\frac{dI}{dt} = \beta SI - \gamma I$$

This helps predict how diseases spread and decline over time.

8.3 Resource Management

Population equations are used to control natural resources like fish and forests. A harvesting model is:

$$\frac{dP}{dt} = rP - h$$

This helps ensure sustainable resource usage.

IX. LIMITATIONS OF THE MODEL

Although population models are useful, they have limitations.

These models assume smooth and continuous growth:

$$\frac{dP}{dt} = f(P)$$

But real populations change unpredictably due to external factors.

They also assume constant parameters like growth rate:

$$r = \text{constant}$$

In reality, growth rates change due to environment, climate, and resources.

Migration effects are also not fully included:

$$\frac{dP}{dt} = rP + I - E$$

where I is immigration and E is emigration.

X. CONCLUSION

Population growth is a dynamic process that reflects the interaction between living organisms and their environment. This study shows that differential equation models provide a structured and effective way to understand how populations change over time. The simplest models explain growth under ideal conditions, but they are limited because they do not consider real-world constraints such as limited food, space, and competition. More advanced models introduce environmental limitations and explain how populations naturally slow down and eventually stabilize when resources become scarce. These models are more realistic and help describe how ecosystems maintain balance over time. Further extensions of population models include factors such as harvesting, migration, time delays, species interactions, and environmental changes, all of which make the system more complex and closer to real biological situations. Stability analysis helps in understanding whether a population will persist, decline, or stabilize at a certain level, which is important for ecological planning and resource management. Additionally, interaction-based models show that when multiple species are involved, population behavior can become cyclical or fluctuating rather than steady. These mathematical approaches are not only useful in ecology but are also widely applied in areas such as disease modeling, environmental studies, and socio-economic systems. Overall, the study highlights that while simple models provide basic insight, realistic population behavior can only be understood through more advanced and interconnected mathematical formulations.

REFERENCES

- [1] Allen, L. J. S. (2007). *An introduction to mathematical biology*. Pearson.
- [2] Bacaër, N. (2011). *A short history of mathematical population dynamics*. Springer.
- [3] Boyce, W. E., & DiPrima, R. C. (2017). *Elementary differential equations and boundary value problems*. Wiley.
- [4] Brauer, F., Castillo-Chavez, C., & Feng, Z. (2019). *Mathematical models in epidemiology and ecology*. Springer.
- [5] Edelstein-Keshet, L. (1988). *Mathematical models in biology*. McGraw-Hill.
- [6] Edelstein-Keshet, L. (2005). *Mathematical models in biology*. SIAM.
- [7] Edelstein-Keshet, L., & Edelstein-Keshet, L. (2013). *Mathematical models in biology*. SIAM.
- [8] Freedman, H. I. (1980). *Deterministic mathematical models in population ecology*. Marcel Dekker.
- [9] Hirsch, M. W., Smale, S., & Devaney, R. L. (2013). *Differential equations, dynamical systems, and an introduction to chaos*. Academic Press.
- [10] Kaplan, D., & Glass, L. (1995). *Understanding nonlinear dynamics*. Springer.
- [11] Kot, M. (2001). *Elements of mathematical ecology*. Cambridge University Press.
- [12] Logan, J. D. (2015). *Applied mathematics*. Wiley.
- [13] May, R. M. (1976). *Simple mathematical models with very complicated dynamics*. Nature.
- [14] Murray, J. D. (1989). *Mathematical biology*. Springer.
- [15] Murray, J. D. (2002). *Mathematical biology I: An introduction*. Springer.
- [16] Murray, J. D. (2003). *Mathematical biology II: Spatial models and biomedical applications*. Springer.
- [17] Perko, L. (2013). *Differential equations and dynamical systems*. Springer.
- [18] Smith, H. L. (2011). *An introduction to delay differential equations with applications to the life sciences*. Springer.
- [19] Strogatz, S. H. (2018). *Nonlinear dynamics and chaos*. CRC Press.
- [20] Thieme, H. R. (2003). *Mathematics in population biology*. Princeton University Press.
- [21] Verma, A.K. (2017). *A Handbook of Zoology*. Shri Balaji Publications, Muzaffarnagar. 1- 648. pp.