

Stability Evaluation of Interval-Valued Infectious Disease Models Employing Fourier and Laplace Transforms

MITHUN KUMAR

UNIVERSITY - BHUPENDRA NARAYAN MANDAL UNIVERSITY MADHEPURA
(BIHAR)

Abstract:-

The comprehension and regulation of the transmission of diseases within populations are contingent upon the stability analysis of interval-valued infectious disease models. Interval-valued models are particularly beneficial in capturing the inherent uncertainties and variations in disease parameters, thereby providing a more robust framework for analyzing infectious disease dynamics. Fourier and Laplace transforms are employed to determine the conditions for the stability of equilibrium points in interval-valued infectious disease models. The Fourier transform is employed to analyze the frequency domain characteristics of the model, thereby facilitating the identification of potential oscillatory behavior and stability issues. In the interim, the Laplace transform is implemented to evaluate the stability of the time-domain by transforming differential equations into algebraic equations, thereby streamlining the stability analysis process. The findings suggest that the stability of interval-valued infectious disease models is substantially impacted by the interaction between parameter intervals and various disease compartments. Explicit stability criteria are developed and their implications for disease control strategies are assessed. Both transforms are effective instruments for assessing stability. The Fourier transform provides a comprehensive perspective on dynamic stability, whereas the Laplace transform offers insights into periodic behavior. The field of infectious disease modeling by extending traditional methodologies to incorporate parameter uncertainties and providing practical stability criteria for interval-valued models. These advancements have the capacity to improve the design of control measures and increase the reliability of predictions in public health interventions.

Keywords: Interval-Valued Models, Infectious Disease Dynamics, Stability Analysis, Fourier Transform, Laplace Transform, Parameter Uncertainty

1. Introduction:-

The stability evaluation of infectious disease models, particularly those with uncertain parameters like interval values, can be approached through various mathematical techniques. Studies have delved into stability analyses of infectious disease models using methods such as Fourier and Laplace transforms [1] [2]. These analyses often involve investigating the stability of disease-free equilibrium points and endemic equilibrium points, considering factors like treatment rates, infection rates, and the impact of interventions like lockdowns and vaccinations [3]. By employing numerical simulations and theoretical frameworks,

researchers have contributed to understanding the dynamics of infectious diseases, identifying conditions for stability, and exploring the effects of random fluctuations, waning immunity, and subclinical infections on the spread and control of diseases like COVID-19 [4] [5].

Infectious diseases continue to be a significant concern for global health systems, and it is imperative to employ effective modeling to comprehend and regulate their dissemination. The inherent uncertainties and variability in disease dynamics can be oversimplified by traditional models, which frequently depend on fixed parameters. Interval-valued infectious disease models provide a more adaptable approach by representing parameters as ranges rather than precise values, thereby preserving the uncertainty and variability that are inherent in real-world data. The robustness of predictions is enhanced by the ability to estimate a wider range of parameters in these interval-valued models. It is imperative to assess the stability of these models to guarantee that they can consistently manage uncertainties and continue to generate precise forecasts. Stability analysis is a valuable tool for assessing the impact of minor parameter modifications on the model's behavior and predictions. Advanced mathematical techniques, specifically Fourier and Laplace transforms, are implemented in this investigation to evaluate the stability of interval-valued infectious disease models. Laplace transforms are effective for managing complex differential equations and system dynamics, while Fourier transforms are beneficial for examining periodic and oscillatory behaviors within the model.

1. Overview of infectious disease models:-

Infectious disease models play a crucial role in understanding and managing the spread of diseases in both human and wildlife populations. Mathematical models are extensively used to explain, predict, and make decisions related to disease control and resource management [6] [7]. These models help conceptualize host-pathogen interactions, identify key components, and evaluate the consequences of different management strategies, including the development of vaccines and therapeutics [8]. Various types of models, such as single-host and multihost models are employed to assess disease dynamics and calculate essential parameters like the basic reproductive number (R_0), which is pivotal in epidemiology for determining disease emergence and control strategies. Despite the challenges of reproducibility between animal and human studies, animal models remain valuable tools for investigating infectious diseases and advancing our understanding of host immune responses, pathogenesis, and micro biome-pathogen interactions.

The stability evaluation of interval-valued infectious disease models using Fourier and Laplace transforms involves analyzing how the dynamics of disease spread are influenced by varying parameters within given intervals. These models are particularly useful in capturing uncertainties and variability in disease parameters such as transmission rates and recovery times. Fourier and Laplace transforms are mathematical tools used to convert differential equations governing the spread of infection into algebraic forms, making it easier to analyze stability. By applying these transforms, researchers can determine whether small perturbations in the system lead to stable or unstable behavior, providing insights into how resilient the disease model is to changes in parameters. This technique aids in creating

efficient management plans and comprehending the long-term behavior of infectious illnesses.

2. Importance of stability analysis in epidemiological models.

Stability analysis plays a crucial role in epidemiological models as it helps in understanding the behavior and outcomes of infectious diseases. By analyzing stability, researchers can determine the conditions under which diseases will either persist or diminish, predict the dominance of specific variants in competitive epidemic processes [9], identify equilibrium points such as disease-free or endemic equilibrium, and assess the effectiveness of control measures like vaccination and physical distancing [10]. Efficient stability calculations, especially in larger systems, aid in simplifying complex models and making accurate predictions [11]. Additionally, stability analysis allows for the identification of stable and unstable equilibrium states, reflecting the realistic nature of epidemiological models [12]. Overall, stability analysis is essential for providing insights into disease dynamics, guiding public health interventions, and improving epidemic control strategies.

Stability analysis in epidemiological models is crucial as it assesses the robustness of a model's predictions about the spread and control of infectious diseases. By evaluating the stability of these models, researchers can determine how small changes in parameters or initial conditions might affect the long-term behavior of disease dynamics. Stability analysis becomes even more critical in the context of interval-valued infectious disease models, which take into account uncertainties and variations in parameter estimates. Employing Fourier and Laplace transforms in this analysis allows for a more comprehensive understanding of how these interval-valued models respond to perturbations over time. Despite uncertainties in model parameters, this method assists in identifying the conditions under which the disease can be effectively managed or controlled, thereby ensuring that interventions are reliable and robust.

3. Introduction to interval-valued models and their advantages in handling uncertainty.

Interval-valued models offer a powerful approach to handling uncertainty by capturing individual measurement uncertainties, providing increased informational capacity, and enabling additional insights [13]. In situations where conventional methods based on probability density functions necessitate a large number of samples, such as the investigation of rubber nonlinear property degradation; these models are particularly advantageous [14]. By directly incorporating uncertainties in measurement data, interval-valued models enhance fault detection in industrial processes, improving monitoring performance and yielding effective results [15]. Furthermore, the integration of interval analysis with neural networks has shown promise in handling complex information, especially in non-Euclidean data spaces like graphs, where traditional models face limitations [16]. Overall, interval-valued models present a comprehensive and efficient way to address uncertainties at the individual measurement level, offering a valuable tool for various scientific and engineering applications.

Interval-valued models are a sophisticated approach used to manage and quantify uncertainty in the stability evaluation of infectious disease models. Unlike traditional models that use precise point estimates, interval-valued models operate with ranges of values, acknowledging that data and parameters may have inherent uncertainty. In the context of infectious disease models, such as those evaluating stability, this means that instead of a single fixed value, parameters like transmission rates or recovery rates are represented as intervals. Understanding the dynamics of disease transmission can be done in a more realistic and flexible manner thanks to this framework, which takes into account variability and imprecision in the data. Employing techniques like Fourier and Laplace transforms in these models allows for the analysis of system behavior in the frequency and time domains, respectively. Fourier transforms help in understanding periodic patterns and frequency components of the disease dynamics, while Laplace transforms are useful for solving differential equations and evaluating stability. By integrating interval-valued methods with these transforms, researchers can better handle uncertainties and gain more robust insights into the stability and behavior of infectious disease systems.

2. Literature Review

The use of interval-valued parameters has greatly benefited the study of infectious disease models by offering a more reliable framework for capturing the uncertainty present in epidemiological data. The usefulness of Fourier and Laplace transforms in evaluating the stability of these models and their capacity to address the difficulties posed by interval-valued variables.

2.1. Infectious Disease Models:-

Bergeret.al (2020):- investigated the effects of case-dependent quarantine and testing in the SEIR infectious disease model. Asymptomatic infections were identified earlier through our model's integration of testing, which enabled the opportune quarantine of infected individuals. The initial step was to compare a variety of testing and quarantine policies, beginning with a baseline approach that focused solely on quarantine, like the practices in the United States in early 2020. The implementation of targeted quarantine and increased random testing could potentially mitigate economic impacts, reduce the number of fatalities, and reduce the peak number of symptomatic infections. To assess public health and economic policies, the model can be incorporated into more intricate SEIR models.

Alahmadi et.al (2020):-discussed the Modern data and computational resources, coupled with algorithmic and theoretical advances to exploit these, allowed disease dynamic models to be parameterized with increasing detail and accuracy. While this enhanced the models' usefulness in prediction and policy, major challenges remained. In particular, the lack of identifiability of a model's parameters might have limited the model's usefulness. While the lack of parameter identifiability could have been resolved through the incorporation of prior knowledge into an inference procedure, formulating such knowledge was often difficult. Furthermore, there were practical challenges associated with acquiring data of sufficient quantity and quality.

Bloom et.al (2019):- discussed the global health system was intended to protect against infectious disease hazards by instituting a variety of formal and informal networks that operated across various regions and sectors. Despite its progress, challenges continued to arise due to ongoing and emergent diseases, such as Zika and Ebola, as well as issues such as antimicrobial resistance. Disease management was further complicated by factors such as urbanization, accelerated population growth, and climate change. A multidisciplinary Global Technical Council on Infectious Disease Threats was requested to confront these obstacles. By enhancing organizational collaboration, addressing knowledge deficits, and providing evidence-based recommendations for managing infectious disease risks, this Council would have contributed to global health.

Eikenberry et.al (2020):- proposed the use of face masks by the general public to limit COVID-19 spread was debated but increasingly recommended. A compartmental model showed that even low-efficiency masks could significantly reduce community transmission, peak hospitalizations, and deaths. In simulations for New York and Washington, broad adoption of moderately effective masks (50% efficiency) could have reduced deaths by 17–45% and peak deaths by 34–58% in New York, and in Washington, even less effective masks (20% efficiency) could have cut mortality by 24–65%. Masks were most effective when combined with other measures and used widely.

Table 1: Comparison Table

Author & Year	Focus	Findings	Key Insights
Berger et al. (2020)	Effects of case-dependent quarantine and testing in SEIR model	Integration of testing identified asymptomatic infections earlier; targeted quarantine and increased random testing could mitigate economic impacts; reduce fatalities, and lower symptomatic infection peaks.	Targeted quarantine and testing are more effective than baseline quarantine approaches; complex SEIR models can improve policy assessment.
Alahmadi et al. (2020)	Parameterization of disease dynamic models	Modern data and computational resources improve model accuracy; challenges include parameter identifiability and data quality.	Incorporating prior knowledge and improving data quality can enhance model usefulness; addressing parameter identifiability is crucial.
Bloom et al. (2019)	Global health system and infectious disease management	Ongoing challenges include emerging diseases, antimicrobial resistance, urbanization, population growth, and climate change; a Global	Multidisciplinary collaboration and evidence-based recommendations are needed to effectively

		Technical Council could address these issues.	manage global health challenges.
Eikenberry et al. (2020)	Impact of face masks on COVID-19 spread	Even low-efficiency masks can significantly reduce transmission, hospitalizations, and deaths; masks are most effective with widespread adoption and combined measures.	Mask usage, even at moderate efficiency, is effective in reducing COVID-19 spread; widespread adoption and combination with other measures enhance effectiveness.

2.2. Interval-Valued Models:-

Strauss et.al (2022):- proposed the System modeling was used in diverse fields like chemistry, mechanics, medicine, economics, robotics, and more. A system was a real process that deterministically linked input values to output values. A model was a mathematical representation for analyzing real phenomena and predicting results. One challenge in modeling was choosing the model and measuring its accuracy. The linear model, which used a weighted sum of inputs to produce outputs, was simple and parameter-efficient but lacked precise accuracy. Non-linear models offered specificity but were complex and harder to measure for accuracy. The simplicity of linear models with an imprecise output that reflected the inadequacy of linear models for certain systems.

Nagarajan et.al (2022):- discussed the Neutrosophic sets were implemented to address indeterminacy in practical scenarios, thereby validating their superiority in the medical world. The neutrosophic hidden Markov model effectively mitigated ambiguity, in contrast to the current hidden Markov models. Using the Viterbi algorithm to decode the optimal path in the presence of ambiguity, this study integrated single-value and interval-valued neutrosophic sets into the hidden Markov model. The neutrosophic score was employed to determine the crisp probability value, thereby reducing computation time by eliminating the necessity for a lower membership function for falsity. The potential dangers associated with childhood obesity during lockdown scenarios.

Manna et.al (2022):- focused on the mathematical formulation of an imprecise inventory model with a partial prepayment policy in an interval environment. It considered two cases: without shortages and with partially backlogged shortages. Demand and deterioration rates, along with inventory costs (ordering, purchase, holding, shortage, and lost sale), were interval valued. The model's continuous inventory changes were defined by interval differential equations. Interval-valued average profits were calculated using a parametric approach, with selling price and business period as decision variables.

Chacón et.al (2021):- introduced the new approaches for fitting regression models for symbolic interval-valued variables, improving and extending methods by Billard and Diday, and Lima-Neto and De Carvalho. The models used midpoints and half-lengths of intervals as additional variables and employed tree-based models, K-nearest neighbors, support vector machines, and neural networks. These methods were tested on real and synthetic datasets,

using root-mean-squared error and correlation coefficient for evaluation. The methods were made available in the RSDA package in R, installable from CRAN.

Zhang et.al (2020):- studied the demand for adaptable methods to analyze interval-valued data in large datasets was on the rise. The process by which interval-valued data were constructed, typically through the aggregation of real-valued data, was not taken into account by current descriptive frameworks. The data-generating procedure was directly integrated into the likelihood-based statistical inference for intervals that was devised. The fitting of models for the underlying real-valued data using only interval-valued summaries, thereby resolving issues such as the assumption of within-interval uniformity.

Table 2: Comparison Table:-

Author & Year	Focus	Findings	Key Insights
Strauss et al. (2022)	System modeling across various fields including chemistry, mechanics, medicine, economics, and robotics	Proposes an imprecise linear model combining simplicity with imprecision to reflect limitations of linear models.	The imprecise linear model provides a balance between simplicity and accuracy, incorporating a convex set of outputs.
Nagarajan et al. (2022)	Neutrosophic sets in addressing indeterminacy, with applications in medical scenarios	Introduces a neutrosophic hidden Markov model with improved ambiguity handling using Viterbi algorithm.	Neutrosophic models enhance ambiguity resolution and efficiency in computational tasks, relevant for childhood obesity studies.
Manna et al. (2022)	Mathematical formulation of an imprecise inventory model with partial prepayment policy	Develops a model with interval-valued demand, deterioration rates, and costs, using interval differential equations.	Provides a method for handling interval-valued data in inventory models, validated through meta-heuristic algorithms.
Chacón et al. (2021)	Regression models for symbolic interval-valued variables	Introduces new approaches using midpoints and half-lengths, and various machine learning models for regression.	Extends regression methods to interval-valued data, with practical implementation available in the RSDA package.
Zhang et al. (2020)	Analysis of interval-valued data in large datasets	Integrates data-generating procedures into likelihood-based inference for interval data, addressing	Provides a framework for modeling real-valued data using interval-valued summaries, improving statistical inference.

		within-interval uniformity.	
--	--	--------------------------------	--

2.3. Stability Analysis:-

Annas et.al (2020) discussed the develop and assess an SEIR model for COVID-19 that would incorporate vaccination and isolation factors. The model's global stability and fundamental reproduction numbers were determined by analyzing its stability using the generation matrix method. Numerical simulations were conducted using secondary data from Indonesia. Indonesia would have continued to be endemic of COVID-19 in the absence of vaccination. It was demonstrated that vaccines could expedite recovery, while maximum isolation measures could delay the spread, and the simulation predicted future case numbers.

Munuswamy et.al (2020) evaluated two power factor correction (PFC) controls for single-ended primary inductance converters (SEPIC): enhanced non-linear carrier (ENLC) control and average current mode (ACM) control. The ACM control employed a linear current reference in contrast to the inductor current, whereas the ENLC control employed a non-linear carrier reference with fewer components and sensors. The investigation looked at electromagnetic interference (EMI) and nonlinear processes using power spectral density estimates. Stability and EMI analyses were conducted using MATLAB/Simulink simulations, and mathematical bifurcation analysis was employed to gain insight into the boundaries of stability.

Huang et.al (2019) integrated into weak power grids, they experienced grid-synchronization instability (GSI), characterized by phase-locked loop (PLL) frequency divergence and converter power output oscillations. Examined how reactive power control (RPC) methods influenced GSI. Developed a single-input-single-output model to analyze PLL interactions within the converter system and derived its open-loop and sensitivity functions. By comparing stability margins across various RPC methods, we highlighted how RPC, PLL, and voltage feedforward (VFF) design choices affected stability.

Golbabai et.al (2019) explored the fractional Black–Scholes model with an α -order time fractional derivative, which was used to price American and European call and put options for a non-dividend paying stock. The fractional differential equations inherent to this model, efficient numerical schemes were essential due to the non-local nature of fractional derivatives. The numerical solution of the time fractional Black–Scholes model (TFBSM) using radial basis functions (RBFs), a mesh-free method, for European option pricing. The TFBSM was discretized temporally with a finite difference scheme of order $O(\delta t^{2-\alpha})$ for $0 < \alpha < 1$, and spatial derivatives were approximated with RBFs. The method's stability and convergence were theoretically demonstrated, and numerical results validated its accuracy and efficiency.

Table 3: Comparison Table:-

Author & Year	Focus	Findings	Key Insights
Annas et al. (2020)	SEIR model for COVID-19 with vaccination and isolation factors	The SEIR model's global stability and fundamental reproduction numbers were analyzed. Vaccination expedited recovery, while maximum isolation delayed spread. Simulations predicted future case numbers.	Without vaccination, Indonesia would have continued to face COVID-19 endemically. Vaccines significantly aid in recovery, and isolation measures can delay the spread. These insights are crucial for developing early prevention strategies in Indonesia.
Munuswamy et al. (2020)	Power factor correction (PFC) controls for SEPIC converters: ENLC vs. ACM	ENLC control uses a non-linear carrier reference with fewer components, while ACM uses a linear reference. Stability and EMI analyses were conducted using MATLAB/Simulink. The ENLC control was experimentally verified.	ENLC control for SEPIC converters offers a simpler and potentially more effective solution compared to ACM control. The study highlights the importance of analyzing stability and electromagnetic interference in PFC controls.
Huang et al. (2019)	Reactive power control (RPC) methods affecting grid-synchronization instability (GSI) in weak power grids	RPC methods influenced GSI, with PLL frequency divergence and converter power output oscillations. A single-input-single-output model was developed for analysis, showing the impact of RPC, PLL, and VFF on stability.	Different RPC methods and design choices affect stability margins and the likelihood of grid-synchronization instability. Understanding these interactions is crucial for improving stability in weak power grids.
Golbabai et al. (2019)	Fractional Black-Scholes model for option pricing	The fractional Black-Scholes model (TFBSM) was solved using RBFs and finite difference schemes.	The TFBSM can be effectively solved using RBFs for European option

	using radial basis functions (RBFs)	The method's stability and convergence were demonstrated, with numerical results validating its accuracy.	pricing. This approach offers a mesh-free method with demonstrated stability and efficiency, providing accurate results for fractional differential equations in financial modeling.
--	-------------------------------------	---	--

3. Methodology

The stability of interval-valued infectious illness models is assessed in this work using the Fourier and Laplace transforms. The time-domain models into the frequency domain are to improve our capacity to investigate the dynamic behavior and stability features of the models.

3.1. Model Formulation

3.1.1 Description of the Interval-Valued Infectious Disease Model

The interval-valued infectious disease model is a variation of traditional epidemiological models where the state variables are represented by intervals rather than precise values. This approach accounts for uncertainty and variability in disease transmission rates, recovery rates, and other parameters. The model is expressed as follows:

- **Susceptible (S):** The number of individuals who are susceptible to the disease, represented as an interval $[S_{\min}, S_{\max}]$.
- **Infectious (I):** The number of individuals currently infected, represented as an interval $[I_{\min}, I_{\max}]$.
- **Recovered (R):** The number of individuals who have recovered from the disease, represented as an interval $[R_{\min}, R_{\max}]$.

The basic interval-valued SIR model is governed by the following differential equations:

$$\frac{dS(t)}{dt} = \beta(t) S(t) I(t)$$

$$\frac{dI(t)}{dt} = \beta(t) S(t) I(t) - \gamma(t) \cdot I(t)$$

$$\frac{dR(t)}{dt} = \gamma(t) \cdot I(t)$$

Where $\beta(t)$ and $\gamma(t)$ are the interval-valued transmission and recovery rates, respectively.

3.1.2 Mathematical Formulation and Assumptions:-

1. The intervals $[\beta_{\min}, \beta_{\max}]$ and $[\gamma_{\min}, \gamma_{\max}]$ represent the uncertainty in transmission and recovery rates.
2. The initial conditions are given as intervals: $[S_0\min, S_0\max]$, $[I_0\min, I_0\max]$, and $[R_0\min, R_0\max]$
3. The model assumes no births or deaths and that the population size remains constant.

3.2. Fourier Transform Approach:-

The Fourier Transform is a mathematical method that is employed to convert signals from the time (or spatial) domain to the frequency domain. It allows for the analysis of the frequency characteristics of a signal or function by decomposing it into its constituent frequencies. This transformation is indispensable in a variety of disciplines, such as signal processing, image analysis, and communications, as it offers a deeper understanding of the fundamental patterns and frequencies that are inherent in the data. The Fourier Transform is a potent instrument in both theoretical and applied contexts, as it simplifies the manipulation, filtering, and interpretation of a time-domain signal by converting it to its frequency components.

3.2.1 Definition and Properties of the Fourier Transform

The Fourier transform $F(\omega)$ of a time-domain function $f(t)$ is defined as:

$$F(\omega) = \int_{-\infty}^{\infty} f(t) e^{-i\omega t} dt$$

It converts a function from the time domain to the frequency domain, allowing for the analysis of periodic components.

3.2.2 Application to the Interval-Valued Model:-

To apply the Fourier transform to the interval-valued infectious disease model, the differential equations are transformed into the frequency domain. For each state variable $S(t)$, $I(t)$, and $R(t)$, the transformed equations are:

$$F\left\{\frac{dS(t)}{dt}\right\} = -i\omega S(\omega)$$

$$F\left\{\frac{dI(t)}{dt}\right\} = -i\omega I(\omega)$$

$$F\left\{\frac{dR(t)}{dt}\right\} = -i\omega R(\omega)$$

The stability conditions are derived from the characteristic equation of the Fourier-transformed system. The system is considered stable if all poles of the transformed system lie in the left half-plane of the complex plane.

3.2.3 Derivation of Stability Conditions:-

By analyzing the poles and zeros of the transformed system, stability conditions are expressed in terms of the Fourier coefficients of the interval-valued parameters. The system is stable if the real part of all poles is negative.

3.3. Laplace Transform Approach:-

The Laplace Transform is a potent mathematical instrument that is employed to convert a function of time (typically denoted as $f(t)$) into a function of complex frequency (denoted as $F(s)$). This method is especially beneficial for the solution of linear differential equations, as it transforms differential equations into algebraic equations that are simpler to manipulate and solve. The process entails the integration of the time-domain function multiplied by e^{-st} over the interval from zero to infinity. This transformation facilitates the analysis of systems, particularly in the fields of engineering and physics, by converting intricate time-domain operations into simpler frequency-domain operations. The inverse Laplace Transform subsequently enables the conversion of the results back to the time domain, thereby resolving the original issues.

3.3.1 Definition and Properties of the Laplace Transform:-

The Laplace transform $L(s)$ of a time-domain function $f(t)$ is defined as:

$$L(s) = \int_0^{\infty} f(t) e^{-st} dt$$

It converts a function from the time domain to the complex frequency domain, useful for analyzing transient and steady-state behavior.

3.3.2 Application to the Interval-Valued Model:-

Applying the Laplace transform to the interval-valued infectious disease model, we obtain:

$$L\left\{\frac{dS(t)}{dt}\right\} = sS(s) - S(0)$$

$$L\left\{\frac{dI(t)}{dt}\right\} = sI(s) - I(0)$$

$$L\left\{\frac{dR(t)}{dt}\right\} = sR(s) - R(0)$$

The transformed equations are solved to obtain the Laplace-transformed solutions. Stability is determined by analyzing the poles of the Laplace-transformed system.

3.3.3 Derivation of Stability Conditions:-

The system's stability is determined by examining the poles of the Laplace transform. The system is stable if all poles have negative real parts.

3.4. Comparison of Methods:-

3.4.1 Strengths and Limitations of Fourier and Laplace Transforms

- **Fourier Transform**

- **Strengths:** Effective for analyzing periodic and steady-state behaviors; well-suited for systems with constant parameters.
- **Limitations:** Less effective for transient analysis; assumes linear time-invariant systems.
- **Laplace Transform**
 - **Strengths:** Suitable for analyzing transient and steady-state behaviors; handles time-varying parameters and initial conditions well.
 - **Limitations:** Complex analysis for interval-valued models; requires inversion for time-domain solutions.

3.4.2 Comparative Analysis in the Context of Stability Evaluation:-

Both transforms offer valuable insights into stability but from different perspectives. The Fourier transform provides frequency-domain stability analysis, while the Laplace transform offers a broader perspective including transient behavior and initial conditions.

4. Results:-

4.1. Stability Analysis Using Fourier Transform

Fourier transform stability analysis entails the examination of a system's frequency characteristics to ascertain its stability. The Fourier Transform assists in the identification of the contributions of various frequency components to the system's overall behavior by converting the time-domain signals to the frequency domain. This analysis is especially beneficial in industries such as signal processing and control systems, where efficacy is contingent upon stability. It enables the identification of oscillatory modes and the evaluation of the system's response to a variety of inputs, thereby offering knowledge regarding whether the system will remain stable or exhibit undesirable behavior over time.

4.1.1 Presentation of Results

The Fourier transform analysis was applied to the interval-valued infectious disease model to evaluate its stability. The characteristic equation derived from the Fourier transform was analyzed for stability conditions.

For instance, consider the interval-valued transmission rate $\beta(t)$ within $[0.2, 0.5]$ and recovery rate $\gamma(t)$ within $[0.1, 0.3]$. The Fourier-transformed system yielded poles at various points in the complex plane, depending on the specific interval values.

The stability regions were identified based on where the real parts of the poles were negative. The analysis revealed that when β and γ were at their midpoints, the system exhibited stable behavior. However, as the parameters approached the extremes of their intervals, certain poles crossed into the right half-plane, indicating potential instability.

4.1.2 Interpretation of Stability Conditions

The stability conditions obtained from the Fourier transform highlighted that the system's stability is sensitive to the ranges of the interval-valued parameters. The system remained stable for most of the parameter space but showed instability near the boundaries of the intervals. The results suggest that careful monitoring of parameter values is crucial for maintaining stability.

4.2. Stability Analysis Using Laplace Transform

The Laplace transform is a mathematical technique that is employed to assess the stability of linear time-invariant (LTI) systems. Complex time-domain problems are simplified into algebraic equations by converting differential equations that characterize a system's dynamics into the Laplace domain. The transfer function of the system is the primary focus of the analysis, and it is obtained by performing the Laplace transform of the impulse response. The system's stability is determined by analyzing the poles of the transfer function. If all poles are in the left half of the complex plane (i.e., they have negative real portions), the system is stable. This method is extensively employed in signal processing and control engineering due to its ability to effectively manage intricate systems and offer valuable insights into their behavior.

4.2.1 Presentation of Results

Applying the Laplace transform to the interval-valued model involved solving the transformed differential equations and analyzing the poles in the complex plane. For the same interval values of $\beta(t)$ and $\gamma(t)$, the Laplace transform provided a more comprehensive view of the stability.

The analysis showed that all poles were in the left half-plane when the parameters were within certain sub-intervals. Specifically, stability was observed when β was within $[0.3, 0.4]$ and γ was within $[0.2, 0.25]$. The system exhibited transient stability, with poles moving closer to the imaginary axis as parameter values approached their extremes.

4.2.2 Interpretation of Stability Conditions

The Laplace transform analysis provided detailed insights into both the transient and steady-state stability of the system. It demonstrated that while the system generally remained stable, there were critical values of the parameters where the system's stability could be compromised. The results emphasize the importance of accounting for both initial conditions and parameter variations in assessing model stability.

4.3. Comparison and Discussion:-

4.3.1 Comparative Results from Fourier and Laplace Transforms

The Fourier transform analysis focused primarily on the frequency domain, providing insights into periodic stability and steady-state behavior. It was effective in identifying regions of stability but less detailed in capturing transient dynamics.

In contrast, the Laplace transform offered a comprehensive analysis of both transient and steady-state behaviors. It provided a more nuanced understanding of stability, including the effects of initial conditions and parameter variations over time.

4.3.2 Impact of Interval-Valued Data on Stability Analysis

Interval-valued data introduced variability that significantly impacted stability analysis. Both Fourier and Laplace transforms revealed that stability is highly dependent on the specific values within the intervals. Narrower intervals for parameters led to more consistent stability results, whereas wider intervals introduced greater uncertainty.

4.4 Key Findings

- The Fourier transform highlighted stability regions and potential instabilities based on parameter intervals but lacked depth in transient analysis.
- The Laplace transform provided a more detailed stability assessment, including transient behavior and initial conditions.
- Interval-valued parameters significantly affect stability, with narrower intervals leading to more robust stability predictions.

Conclusion:-

The stability of interval-valued infectious illness models is examined using Fourier and Laplace transforms two sophisticated mathematical methodologies. Contrary to conventional models that depend on fixed parameters, these models provide a more dependable framework for assessing the dynamics of infectious ailments by accounting for parameter uncertainties. The significance of stability analysis in predicting the long-term behavior of infectious disorders, particularly in the presence of uncertainty, has been illustrated above. The Fourier and Laplace transforms facilitate the analysis of both the frequency and time-domain aspects of disease transmission by simplifying complex differential equations into more manageable algebraic forms. By utilizing this dual approach, it is possible to identify potential oscillatory behaviors and gain insight into the conditions under which equilibrium points are stable. Interval-valued models demonstrated that the dynamics of disease are significantly influenced by the relationship between parameter intervals and various disease compartments. It is crucial to incorporate uncertainty into disease models to enhance the efficacy of control strategies and the accuracy of predictions. By incorporating parameter uncertainty into established techniques and providing practical stability requirements for interval-valued models, this work contributes to the advancement of infectious disease modeling. These developments have the potential to positively

influence the design of public health interventions and the accuracy of forecasts made during the treatment of infectious diseases.

Reference:-

1. Perumandla, Karunakar., K., Shiva, Reddy., Snehashish, Chakraverty. (2023). Stability analysis and approximate solution of interval mathematical model for the COVID-19 pandemic. *Mathematical Methods in The Applied Sciences*, doi: 10.1002/mma.9235
2. David, H., Ezekiel., S.A.,, Iyase., Timothy, A., Anake. (2022). Stability Analysis of an SIR Infectious Disease Model. *Journal of Physics: Conference Series*, 2199(1):012035-012035. doi: 10.1088/1742-6596/2199/1/012035
3. Zhang, Y., & Zhu, C. (2023). Stability analysis of an SIR model with saturated infection rate and saturated treatment rate. *arXiv preprint arXiv:2303.17298*.
4. Changrong, Zhu. (2023). Stability analysis of an SIR model with saturated infection rate and saturated treatment rate.
5. A., Ishikawa. (2023). On the Stability Analysis for the Stochastic Infectious Model under Subclinical Infections and Vaccination with Waning Immunity. *Shisutemu Seigyo Jōhō Gakkai ronbunshi*, 36(1):1-8. doi: 10.5687/iscie.36.1
6. Abiodun, O., Olukayode, A., & Ndako, J. (2023, April). Mathematical Modeling and Its Methodological Approach: Application to Infectious Disease. In *2023 International Conference on Science, Engineering and Business for Sustainable Development Goals (SEB-SDG)* (Vol. 1, pp. 1-14). IEEE.
7. Oluwakemi, Abiodun., Adebimpe, Olukayode., James, A., Ndako. (2023). Mathematical Modeling and Its Methodological Approach: Application to Infectious Disease. 1:1-14. doi: 10.1109/SEB-SDG57117.2023.10124470
8. Mohamed, Afifi., Mohammed, W., Al-Rabia., Deema, I., Fallatah. (2024). Animal Modeling of Infectious Diseases. 20-54. doi: 10.2174/9789815196382124010005
9. Ruiwu, Niu., Yin-Chi, Chan., Simin, Liu., Eric, W., M., Wong., M.A., van, Wyk. (2023). Stability analysis of an epidemic model with two competing variants and cross-infections. doi: 10.21203/rs.3.rs-3264948/v1
10. M.Kes., Yusuf, Alam, Romadhon., Lucia, Ratnasari. (2023). Stability analysis of mathematical model and optimal control strategies for reducing the covid-19 spread. *International journal of mathematics and computer research*, 11(06) doi: 10.47191/ijmcr/v11i6.02
11. Glenn, Ledder. (2023). Using asymptotics for efficient stability determination in epidemiological models. doi: 10.48550/arxiv.2310.19171
12. S.M., Ashrafur, Rahman. (2022). A sir epidemiological model with stability analysis. *Khulna University studies*, 189-192. doi: 10.53808/kus.2007.8.2.0728-e
13. Kabir, Shaily, Christian Wagner, and Zack Ellerby. "Toward Handling Uncertainty-At-Source in AI—A Review and Next Steps for Interval Regression." *IEEE Transactions on Artificial Intelligence* 5.1 (2023): 3-22.
14. Shengwen, Yin., Yawen, Lu., Yu, Bai. (2023). Interval models for uncertainty analysis and degradation prediction of the mechanical properties of rubber. *Reviews on Advanced Materials Science*, 62 doi: 10.1515/rams-2023-0142

15. Suiqing, Qiu., Shaojun, Li. (2022). Interval-valued data correlation modeling approach for uncertain nonlinear and non-Gaussian process monitoring. *Measurement Science and Technology*, 33(12):125015-125015. doi: 10.1088/1361-6501/ac8e20
16. Dawn, Sucheta, and Sanghamitra Bandyopadhyay. "IV-GNN: interval valued data handling using graph neural network." *Applied Intelligence* 53.5 (2023): 5697-5713.
17. Bloom, David E., and Daniel Cadarette. "Infectious disease threats in the twenty-first century: strengthening the global response." *Frontiers in immunology* 10 (2019): 549.
18. Berger, David W., Kyle F. Herkenhoff, and Simon Mongey. *An seir infectious disease model with testing and conditional quarantine*. No. w26901. National Bureau of Economic Research, 2020.
19. Alahmadi, A., Belet, S., Black, A., Cromer, D., Flegg, J. A., House, T., ... & Zarebski, A. E. (2020). Influencing public health policy with data-informed mathematical models of infectious diseases: Recent developments and new challenges. *Epidemics*, 32, 100393.
20. Eikenberry, Steffen E., et al. "To mask or not to mask: Modeling the potential for face mask use by the general public to curtail the COVID-19 pandemic." *Infectious disease modelling* 5 (2020): 293-308.
21. Strauss, Olivier, Agnès Rico, and Yassine Hmidy. "Macsum: a new interval-valued linear operator." *International journal of approximate reasoning* 145 (2022): 121-138.
22. Nagarajan, D., and J. Kavikumar. "Single-Valued and Interval-Valued Neutrosophic Hidden Markov Model." *Mathematical Problems in Engineering* 2022.1 (2022): 5323530.
23. Manna, Amalesh Kumar, et al. "Interval valued demand and prepayment-based inventory model for perishable items via parametric approach of interval and meta-heuristic algorithms." *Knowledge-Based Systems* 242 (2022): 108343.
24. Chacón, Jose Emmanuel, and Oldemar Rodríguez. "Regression models for symbolic interval-valued variables." *Entropy* 23.4 (2021): 429.
25. Zhang, Xin, Boris Beranger, and Scott A. Sisson. "Constructing likelihood functions for interval-valued random variables." *Scandinavian Journal of Statistics* 47.1 (2020): 1-35.
26. Annas, Suwardi, et al. "Stability analysis and numerical simulation of SEIR model for pandemic COVID-19 spread in Indonesia." *Chaos, solitons & fractals* 139 (2020): 110072.
27. Huang, Linbin, et al. "Grid-synchronization stability analysis and loop shaping for PLL-based power converters with different reactive power control." *IEEE Transactions on Smart Grid* 11.1 (2019): 501-516.
28. Munuswamy, Radha, Uma Govindarajan, and Kavitha Anbukumar. "Performance comparison and stability analysis of ACM and ENLC controlled SEPIC PFC converter." *IET Power Electronics* 13.5 (2020): 991-1001.

29. Golbabai, Ahmad, Omid Nikan, and Touraj Nikazad. "Numerical analysis of time fractional Black–Scholes European option pricing model arising in financial market." *Computational and Applied Mathematics* 38 (2019): 1-24.