

Analysis of a modified SEIRS compartmental model for COVID-19

Lavanya Shri S A¹, Bhavvikumar Patel², Mahesh K Banavar¹, Cihan Tepedelenlioglu², Andreas Spanias², Stephanie Schuckers¹

¹Department of Electrical and Computer Engineering (ECE),
Clarkson University, Potsdam, NY, USA.

²SenSIP Center, School of Electrical, Computer, and Energy Engineering, Arizona State
University, AZ, USA.



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- Disease modeling uses mathematical models to predict the spread and severity of diseases
- Captures intervention measures and anticipates the number of people impacted
- Models can be used to conduct epidemiological research, studying patterns of disease transmission and factors influencing its spread
- Compartmental models provide one way to do this

Well Known Compartmental models for disease modeling:

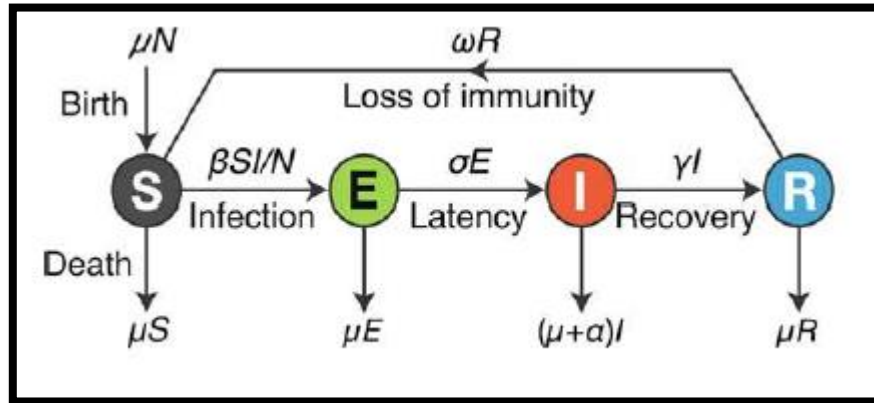
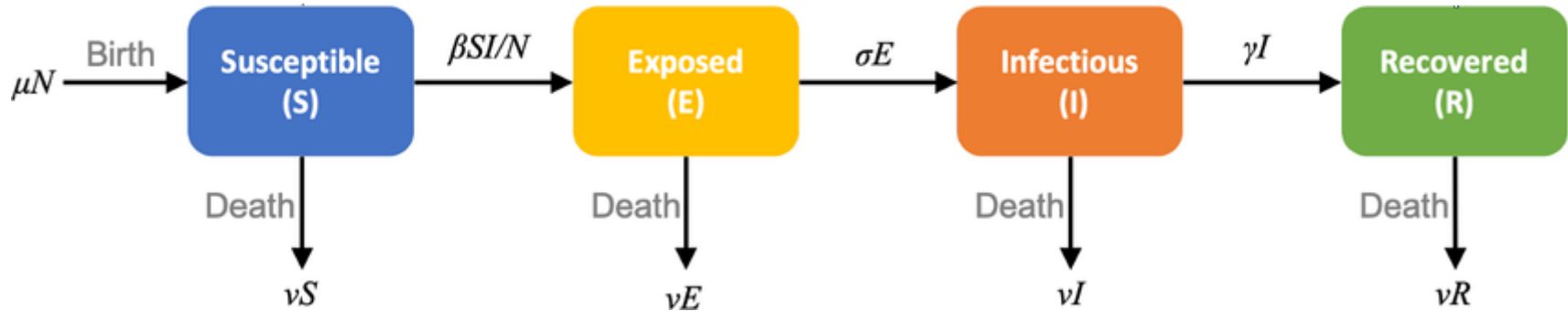
1. SI Model (Susceptible-Infected)
2. SIR Model (Susceptible-Infected- Removed)
3. SEIR Model (Susceptible-Exposed-Infected- Removed)
4. SEIRS Model (Susceptible-Exposed-Infected- Recovered-Susceptible)

We will focus on the SEIRS model

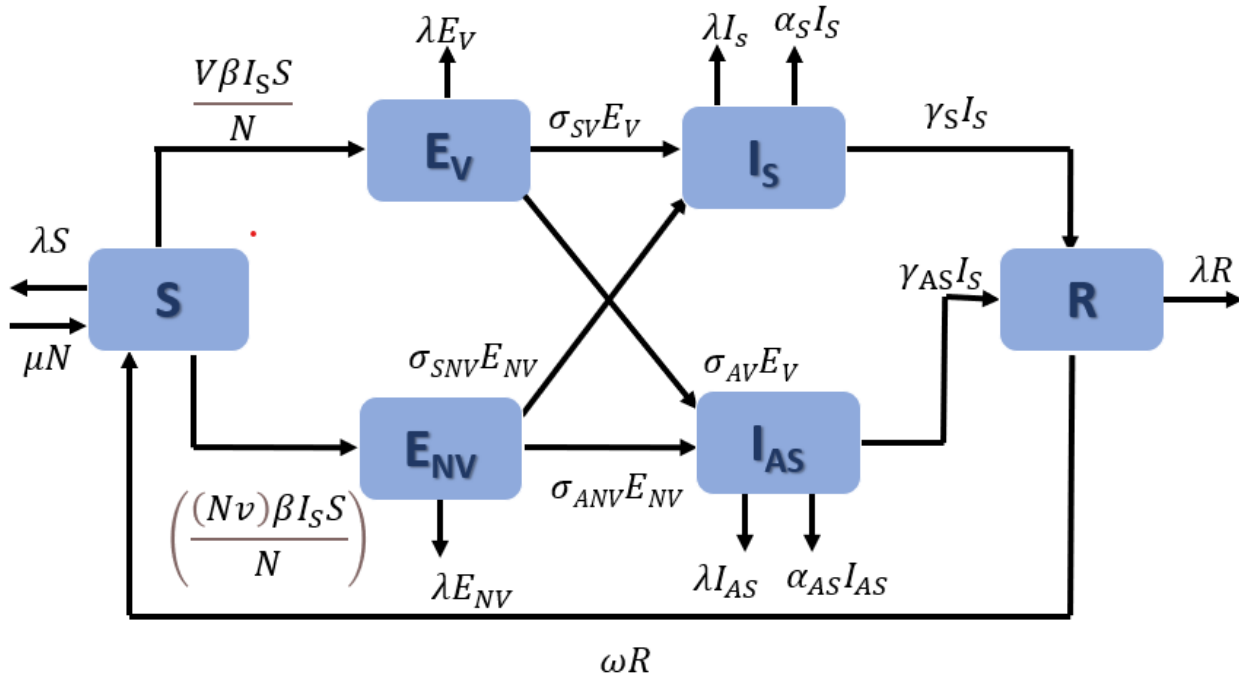
Definitions: S, E, I, R

1. S - Susceptible: Individuals who are at risk of contracting the disease.
2. E - Exposed: Individuals who have been exposed to the pathogen but are not yet infectious.
3. I - Infected: Individuals who are infected and capable of transmitting the disease to susceptible individuals.
4. R - Recovered: Individuals who have recovered from the disease and are assumed to be immune or no longer able to transmit the disease,
Removed – Die from the community.

SEIR and SEIRS Model



O. N. Bjørnstad, K. Shea, M. Krzywinski, and N. Altman, "The seirs model for infectious disease dynamics." Nature methods, vol. 17, no. 6, pp. 557–559, 2020



- λ is the natural death rate.
- μ is the natural birth rate.
- β is the contact rate.
- V is the fraction of the population vaccinated against COVID-19.
- NV is the fraction without the vaccine so $V+NV = 1$.
- α_S, α_{AS} are the death rate due to COVID-19.
- γ_S, γ_{AS} : the recovery rates affect symptomatic and asymptomatic infected persons, respectively.
- $\sigma_{SV}, \sigma_{SNV}, \sigma_{AV},$ and σ_{ANV} are the rates of transition from exposed to infected.
- ω is the fraction of recovered population who lose immunity and return to susceptible.

$$\frac{ds}{dt} = \mu N + \omega R - \lambda S - \left(\frac{V\beta I_S S}{N} \right) - \left(\frac{(NV)\beta I_S S}{N} \right)$$

$$\frac{dI_S}{dt} = \sigma_{SV} E_V + \sigma_{SNV} E_{NV} - (\lambda + \alpha_S + \gamma_S) I_S$$

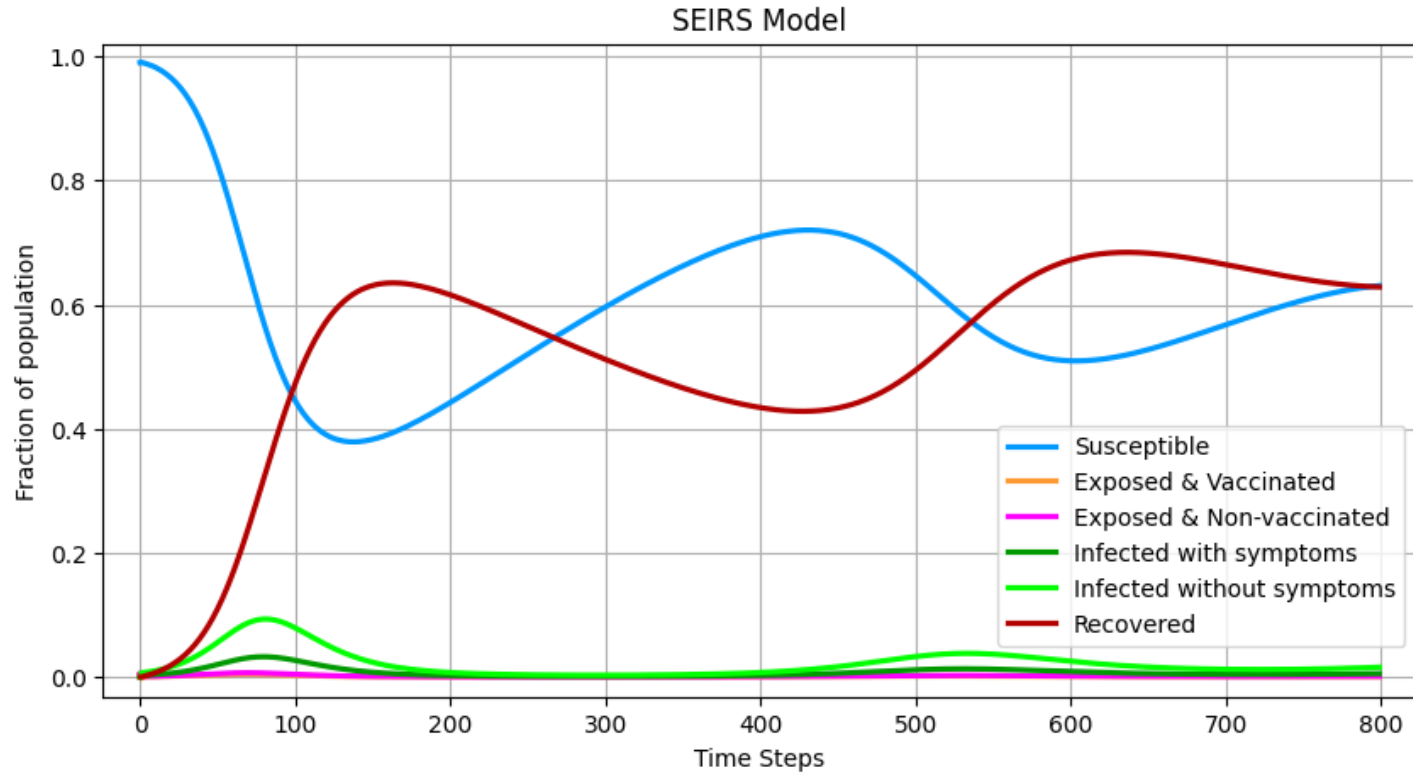
$$\frac{dE_V}{dt} = \left(\frac{V\beta I_S S}{N} \right) - (\lambda + \sigma_{SV} + \sigma_{AV}) E_V$$

$$\frac{dI_{AS}}{dt} = \sigma_{AV} E_V + \sigma_{ANV} E_{NV} - (\lambda + \alpha_{AS} + \gamma_{AS}) I_{AS}$$

$$\frac{dE_{NV}}{dt} = \left(\frac{(NV)\beta I_S S}{N} \right) - (\lambda + \sigma_{SNV} + \sigma_{ANV}) E_{NV}$$

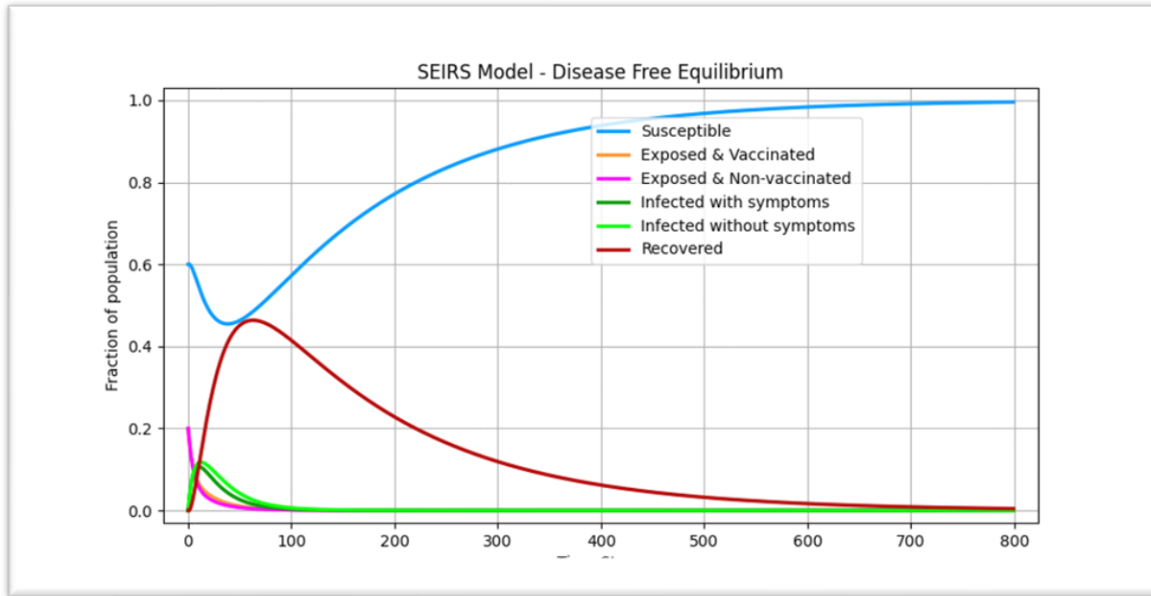
$$\frac{dR}{dt} = \gamma_S I_S + \gamma_{AS} I_{AS} - (\lambda + \omega) R$$

SEIRS Model Time Evolution



Definition: Equilibrium Points

1. An equilibrium point is a state where the system variables (S, E, I, R) remain constant over time.
2. For the modified SEIRS model, we discuss two equilibrium points:
 - Disease-free equilibrium (DFE)
 - Endemic equilibrium (EE)
3. To solve for equilibrium points, we set the derivatives of the compartment values (with respect to time) to zero.



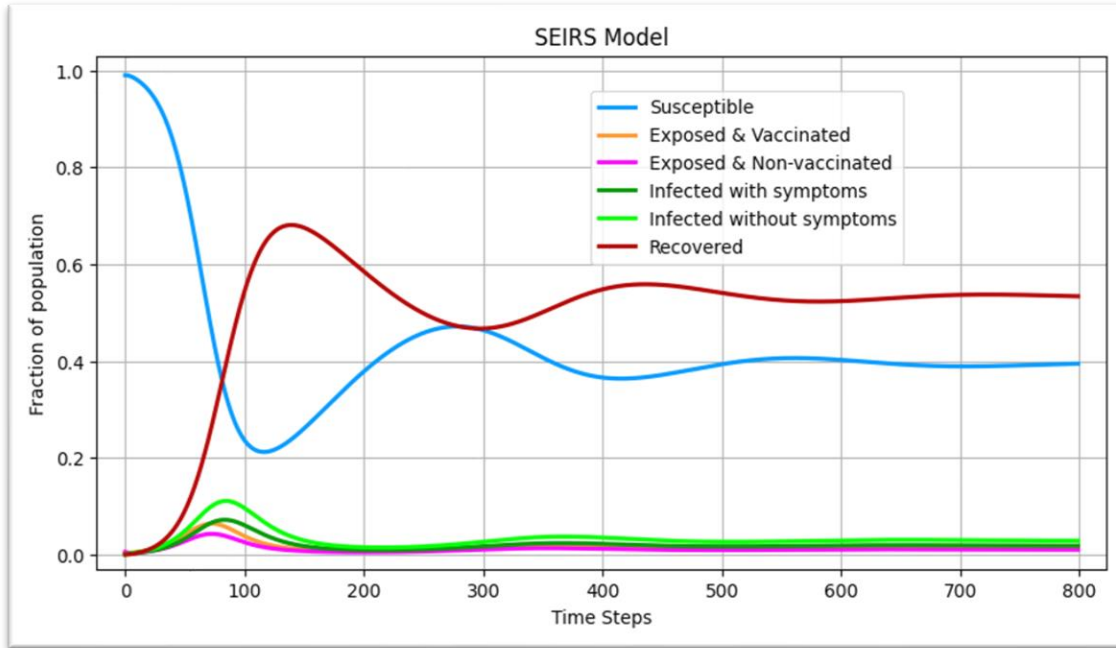
Condition : $I_S = I_{AS} = 0$ (At equilibrium)

Results: Values of I_S and I_{AS} become zero over time.

Summary:

- ❑ DFE indicates the disease is eradicated from the population.
- ❑ The entire population moves to the susceptible compartment.

➤ Disease-free Equilibrium converges to $(1,0,0,0,0,0)$



Condition : $I_S > 0, I_{AS} > 0$ (At equilibrium)

Results: Values of I_S and I_{AS} become non-zero constant over time.

Summary:

- EE indicates the disease is present among the population.
- The disease spread remains constant.

➤ Endemic Equilibrium converges to $(S^*, E_V^*, E_{NV}^*, I_S^*, I_{AS}^*, R^*)$

$$S^* = \frac{ABCN}{\sigma_{SV}(V)\beta B + \sigma_{SNV}(NV)\beta A}$$

$$E_V^* = \frac{\left(\frac{(V)BC}{\sigma_{SV}(V)B + \sigma_{SNV}(NV)A}\right) \left\{\frac{\lambda ABCN}{\beta} - \mu N(\sigma_{SV}(V)B + \sigma_{SNV}(NV)A)\right\}}{\frac{\omega}{(\lambda + \omega)} \left\{\gamma_S(\sigma_{SV}(V)B + \sigma_{SNV}(NV)A) + \frac{\gamma_{AS}C}{D}(\sigma_{AV}(V)B + \sigma_{ANV}(NV)A)\right\} - ABC((V) + (NV))}$$

$$E_{NV}^* = \frac{\left(\frac{(NV)AC}{\sigma_{SV}(V)B + \sigma_{SNV}(NV)A}\right) \left\{\frac{\lambda ABCN}{\beta} - \mu N(\sigma_{SV}(V)B + \sigma_{SNV}(NV)A)\right\}}{\frac{\omega}{(\lambda + \omega)} \left\{\gamma_S(\sigma_{SV}(V)B + \sigma_{SNV}(NV)A) + \frac{\gamma_{AS}C}{D}(\sigma_{AV}(V)B + \sigma_{ANV}(NV)A)\right\} - ABC((V) + (NV))}$$

Endemic Equilibrium Equations

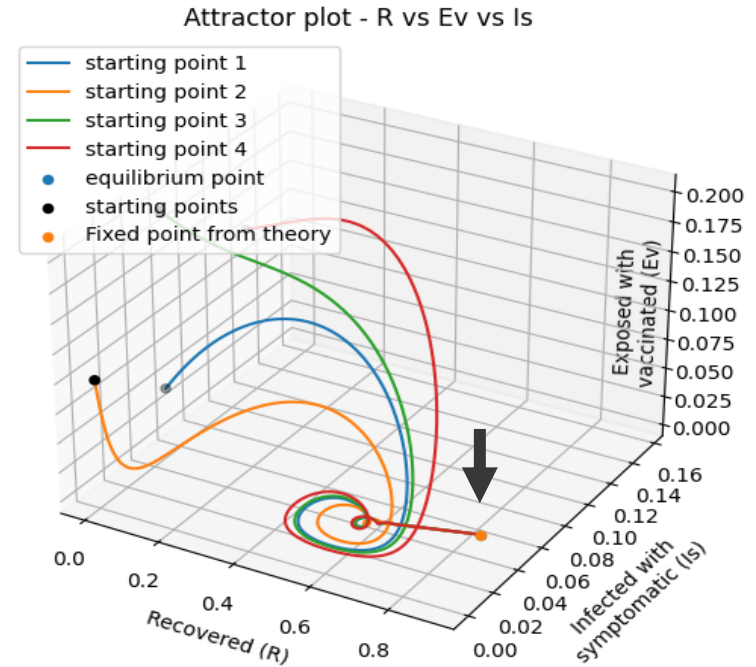
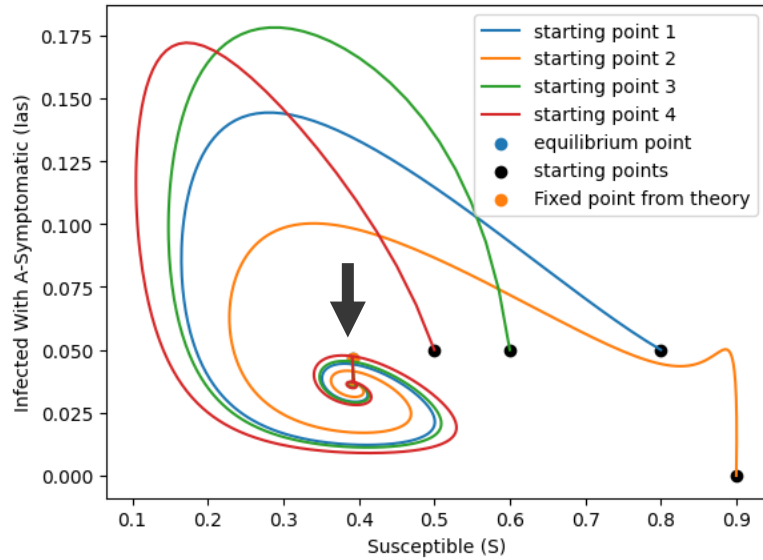
$$I_S^* = \frac{\frac{\lambda ABCN}{\beta} - \mu N(\sigma_{SV}(V)B + \sigma_{SNV}(NV)A)}{\frac{\omega}{(\lambda + \omega)} \left\{ \gamma_S(\sigma_{SV}(V)B + \sigma_{SNV}(NV)A) + \frac{\gamma_{AS}C}{D}(\sigma_{AV}(V)B + \sigma_{ANV}(NV)A) \right\} - ABC((V) + (NV))}$$

$$I_{AS}^* = \frac{\frac{\left(\frac{C(\sigma_{AV}(V)B + \sigma_{ANV}(NV)A)}{D(\sigma_{SV}(V)B + \sigma_{SNV}(NV)A)} \right) \left\{ \frac{\lambda ABCN}{\beta} - \mu N(\sigma_{SV}(V)B + \sigma_{SNV}(NV)A) \right\}}{\frac{\omega}{(\lambda + \omega)} \left\{ \gamma_S(\sigma_{SV}(V)B + \sigma_{SNV}(NV)A) + \frac{\gamma_{AS}C}{D}(\sigma_{AV}(V)B + \sigma_{ANV}(NV)A) \right\} - ABC((V) + (NV))}$$

$$R^* = \frac{\left(\left(\frac{\gamma_S}{(\lambda + \omega)} \right) + \left(\frac{\gamma_{AS}}{(\lambda + \omega)} \right) \frac{C(\sigma_{AV}(V)B + \sigma_{ANV}(NV)A)}{D(\sigma_{SV}(V)B + \sigma_{SNV}(NV)A)} \right) \left\{ \frac{\lambda ABCN}{\beta} - \mu N(\sigma_{SV}(V)B + \sigma_{SNV}(NV)A) \right\}}{\frac{\omega}{(\lambda + \omega)} \left\{ \gamma_S(\sigma_{SV}(V)B + \sigma_{SNV}(NV)A) + \frac{\gamma_{AS}C}{D}(\sigma_{AV}(V)B + \sigma_{ANV}(NV)A) \right\} - ABC((V) + (NV))}$$

- Alternate visualization where states are compared as they evolve over time
- Attractor plots can capture the long-term behavior of a system, showing the states or trajectories that a system tends to converge to
- Attractor plots can assist us in identifying the equilibrium points of the system

Example: Attractor Plots



Conclusion

- Developed a compartmental model for COVID-19 by considering the re-infection and vaccination.
- Analysed the equilibrium points with disease-free and endemic conditions.
- Visualization of SEIRS model using attractor plots.

Future Work

- Will work on the stability analysis for the modified SEIRS model.

Thank you!