

The Impact of Vaccine Resistance on Long-Term Epidemic Outcomes

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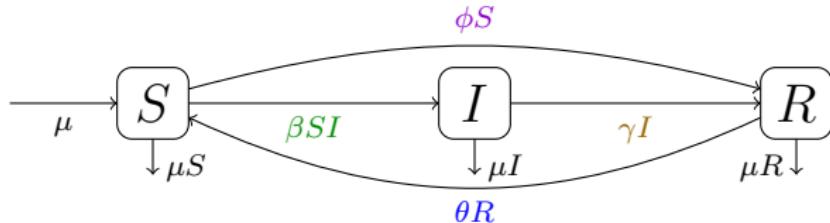
(REUF group at AIM, Summer 2022, 2023)

UNL MathBio Seminar, Oct 3, 2024.

Overview

- ▶ Uniform vaccination can reduce the basic reproduction number of a disease from $\mathcal{R}_0 > 1$ to \mathcal{R}_v .
 - If $\mathcal{R}_v < 1$, the epidemic ends.
- ▶ A significant fraction of the population might be unwilling or unable to be vaccinated. This can help the disease become endemic.
- ▶ The unwilling fraction can change over time.
 - It should generally increase as the disease prevalence decreases.
- ▶ How does this influence epidemic outcomes?

Standard SIR with Vaccination and Loss of Immunity



- ▶ The rate constant ϕ can be reduced to reflect vaccine hesitancy.
 - But even if ϕ is small, everyone gets vaccinated eventually.
- ▶ If everyone gets boosters, the rate constant θ is reduced.
- ▶ Our experience with COVID-19 shows:
 - Many people never get vaccinated.
 - Most people stop getting boosters.

Outline

Model Formulation

General Results

When Vaccination Acceptance is Monotone Increasing

Vaccine Acceptance Peaks at $Y = Y_m$

Conclusions

Model Formulation

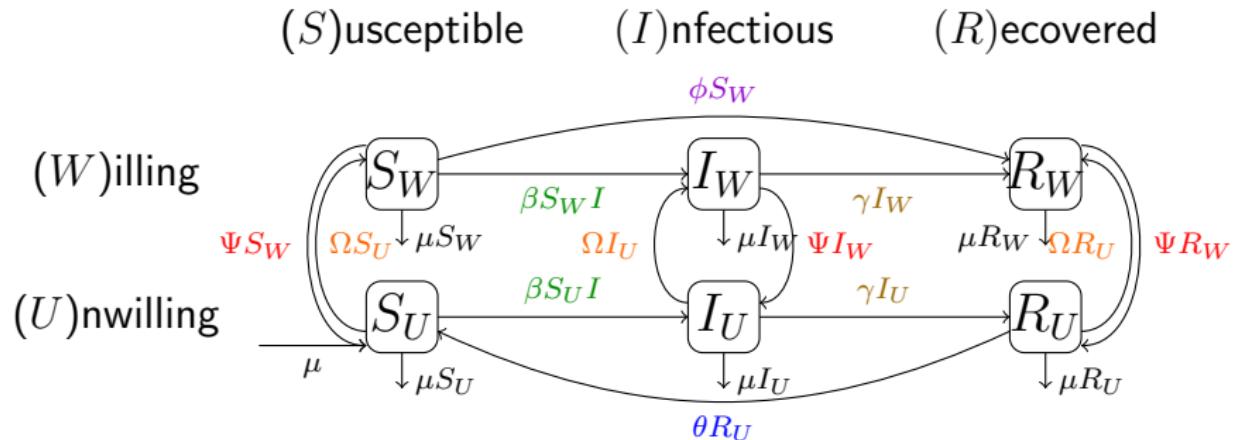
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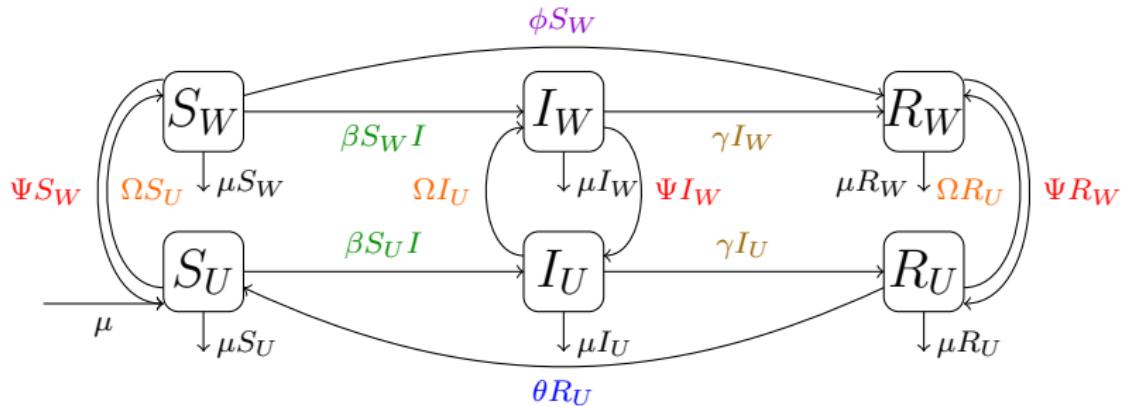
Conclusions

SIR-WU Model Assumptions



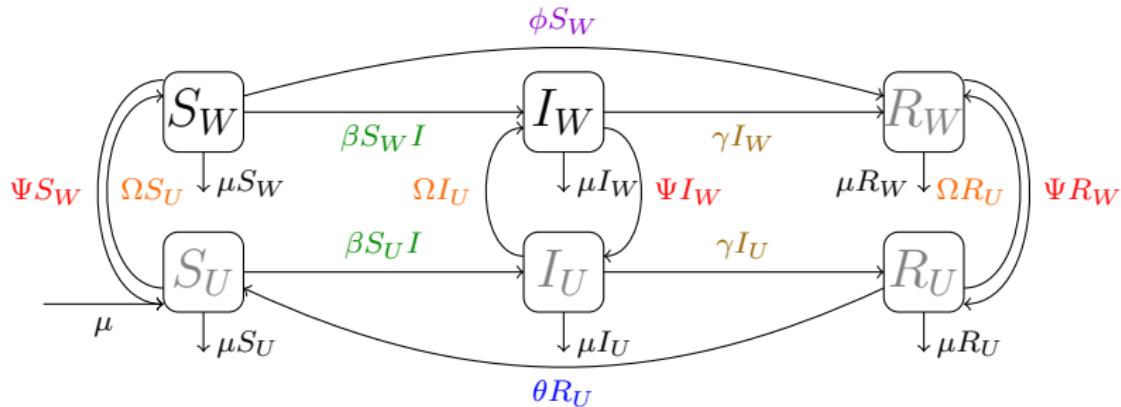
- ▶ Newborns are initially unable to be vaccinated.
- ▶ Willing Susceptibles are vaccinated at rate ϕS_W .
- ▶ Unwilling Recovereds lose immunity at rate θR_U .
- ▶ Unwilling become Willing with rate coefficient $\Omega(I)$.
- ▶ Willing become Unwilling with rate coefficient $\Psi(I)$.

Simplifying Assumptions



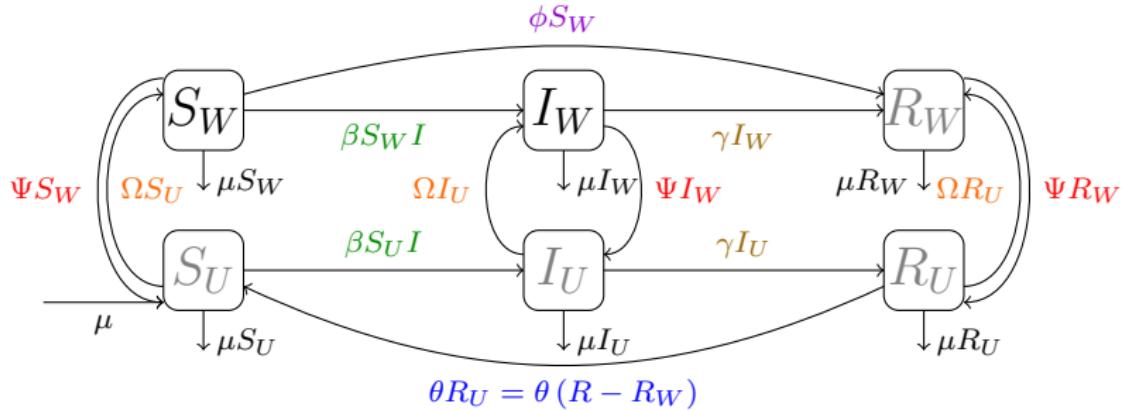
1. No disease-induced mortality.
 2. Infection confers total immunity for 100% of patients.
 3. Willing Recovereds remain recovered through regular boosters.
- These assumptions make disease eradication easier.
- A more realistic model would add unnecessary complications.

Initial Formulation



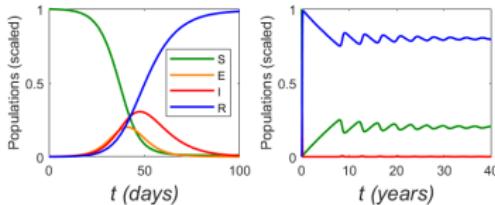
- ▶ Population total is constant ($N = 1$); we need 5 differential equations for the 6 variables.
- ▶ $S + I + R = 1$, so we can use S and I as dependent variables.
- ▶ $W + U = 1$, so we can use W as a dependent variable.
- ▶ We use S_W and I_W as the remaining dependent variables.

Initial Formulation



$$\begin{aligned}
 \frac{dI}{dt} &= -\mu I + \beta SI - \gamma I \\
 \frac{dI_W}{dt} &= -\mu I_W + \beta S_W I - \gamma I_W + \Omega(I)(I - I_W) - \Psi(I)I_W \\
 \frac{dS}{dt} &= \mu - \mu S - \beta SI - \phi S_W + \theta ([1 - S - I] - R_W) \\
 \frac{dS_W}{dt} &= -\mu S_W - \beta S_W I - \phi S_W + \Omega(I)(S - S_W) - \Psi(I)S_W \\
 \frac{dW}{dt} &= -\mu W + \Omega(I)(1 - W) - \Psi(I)W
 \end{aligned}$$

Two Time Scales in Disease Models



- ▶ **The fast time scale (days) shows the epidemic phase.**
 - Infectious population fractions are significant.
 - Plots on the fast time scale show no clue to endemic behavior.
 - Demographic changes (birth, natural death, etc) are negligible.
- ▶ **The slow time scale (years) shows the long-term behavior.**
 - Infectious population fractions are very small.
 - On the slow scale, the epidemic behavior appears at $t = 0$.
 - Both demographic and disease processes are important.

Scaling for the Endemic Phase

- ▶ Time is referred to the demographic time scale $1/\mu$.
- ▶ Infectious populations are rescaled via $I = \epsilon Y$, $I_W = \epsilon Z$.
 - *This is necessary so that all variables are $O(1)$ as $t \rightarrow \infty$.*
- ▶ Slow process parameters are referred to a slow time scale.
 - Loss of immunity and behavior changes are slow.
 - $1/\mu$ is the principal slow time scale (≈ 80 years).

$$h = \frac{\theta}{\mu}, \quad \omega = \frac{\Omega}{\mu}, \quad \Sigma = \frac{\Omega + \Psi}{\mu}.$$

- ▶ Fast process parameters are referred to a fast time scale.
 - Infection and vaccination are fast.
 - $1/(\gamma + \mu)$ is the principal fast time scale (≈ 2 weeks).

$$\mathcal{R}_0 = \frac{\beta}{\gamma + \mu}, \quad v = \frac{\phi}{\gamma + \mu}, \quad \epsilon = \frac{\mu}{\gamma + \mu} \ll 1.$$

Final Formulation

$$\begin{aligned}\epsilon Y' &= (\mathcal{R}_0 S - 1) Y, \\ S' &= \bar{h}(1 - S) - P - hW - \mathcal{R}_0 S Y + O(\epsilon), \\ \epsilon v^{-1} P' &= \omega(Y) S - P + O(\epsilon), \\ W' &= \omega(Y) - \bar{\Sigma}(Y) W,\end{aligned}$$
$$P = vS_W, \quad \bar{h} = h + 1, \quad \bar{\Sigma} = \Sigma + 1.$$

Base Parameters and Values:

- ▶ $\mathcal{R}_0 = 4$: infectiousness is between influenza and Covid
- ▶ $v = 0.5$: vaccination takes a mean of one month
- ▶ $h = 10$: loss of immunity has a mean of 8 years
- ▶ $\Sigma = 8$: opinion change has a mean of 10 years
(can be taken as constant)

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Disease-Free Equilibrium (DFE)

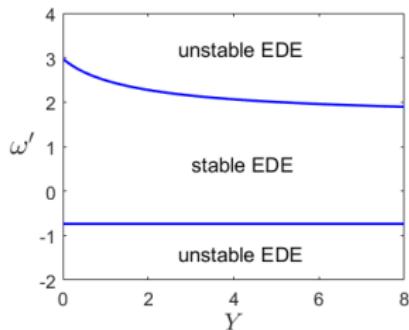
1. There is a unique disease-free equilibrium having susceptible fraction $S^0 < 1$.
2. There is a critical value $\omega_{cr} > 0$ such that the DFE is stable if and only if $\omega(0) > \omega_{cr}$.

Interpretation

- ▶ If the infectious population drops to 0, very few Unwilling will become Willing.
 - Then $\omega(0) < \omega_{cr}$.
- ▶ **Optional vaccination means the disease will persist.**

Endemic Disease Equilibrium (EDE) ($\mathcal{R}_0 > 1$)

1. EDEs are solutions of $\omega(Y) = \omega_{cr} \left(1 - \frac{\mathcal{R}_0 Y}{(\mathcal{R}_0 - 1) h} \right)$.
 - $\omega(0) < \omega_{cr}$, $\omega' > 0$ ($\forall Y$): guarantees a unique EDE.
2. There exist functions $\Upsilon_1(Y) > 0$ and $\Upsilon_2(Y) > 0$ such that an EDE is stable if and only if $-\Upsilon_2(Y) < \omega' < \Upsilon_1(Y)$.



- **Extreme sensitivity of attitude to prevalence (large $|\omega'|$) is destabilizing.**

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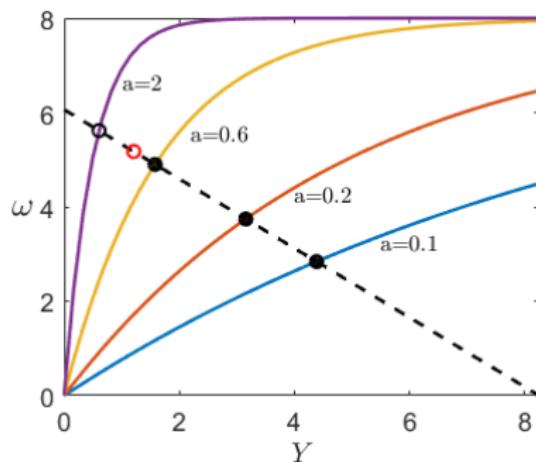
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EDE for monotone ω (more disease \rightarrow more vaccination)

- If $\omega(0) < \omega_{cr}$, then \exists a unique EDE:

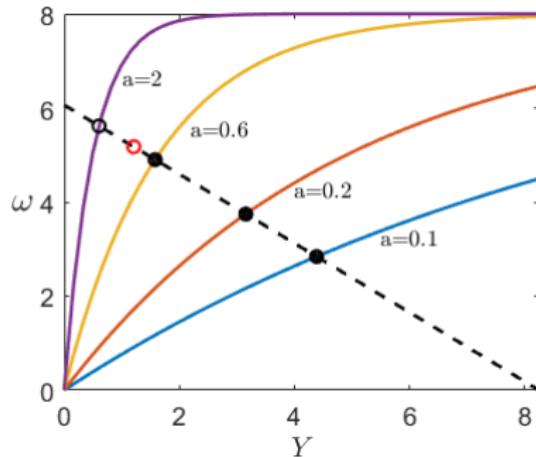
$$0 < Y^* < (\mathcal{R}_0 - 1)(h + 1).$$

Example: $\omega = \Sigma (1 - e^{-aY})$



EDE for monotone ω (more disease \rightarrow more vaccination)

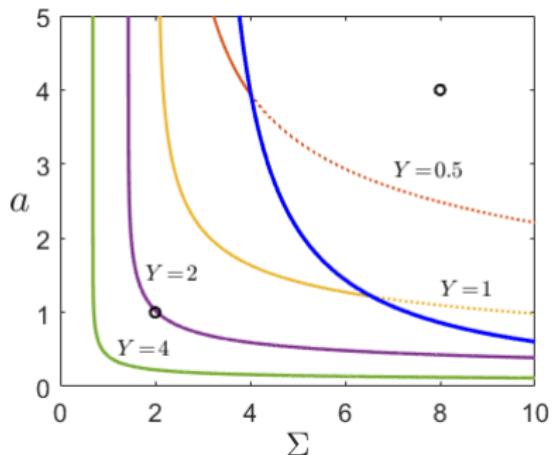
- ▶ The EDE is unstable if $\omega'(Y^*) > \Upsilon_1(Y^*)$.
 - $\omega = \Sigma (1 - e^{-aY})$: Stable for smaller a .



EDE Stability for monotone ω (more disease \rightarrow more vaccination)

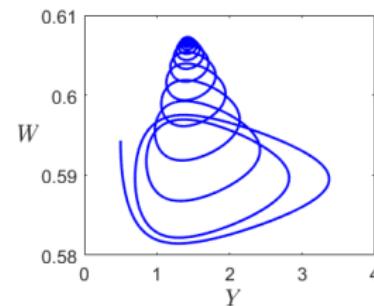
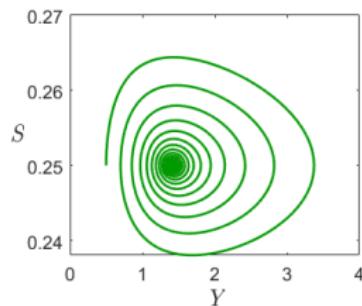
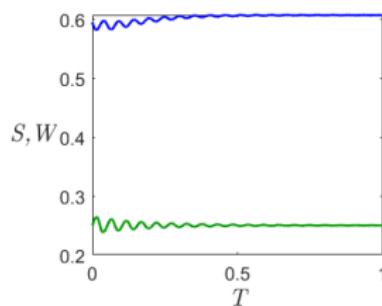
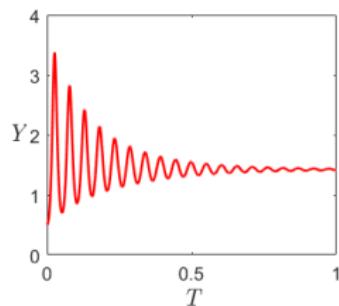
Example: $\omega = \Sigma (1 - e^{-aY})$

- ▶ Σ is the mean number of opinion changes per lifetime.
- ▶ a is the sensitivity of the switch to vaccination.



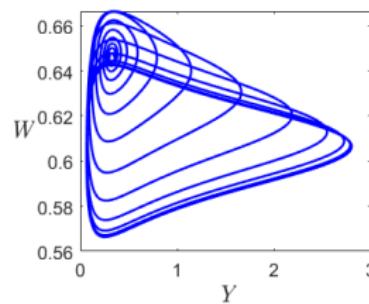
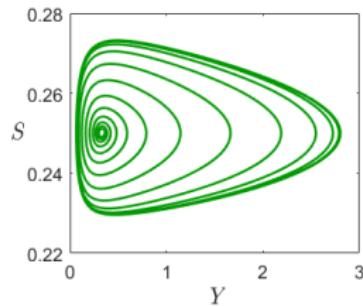
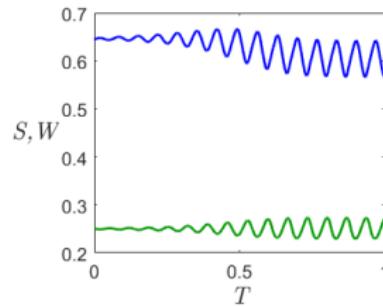
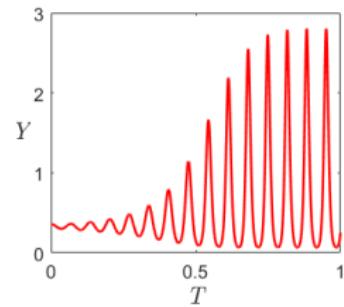
Numerical Solutions: Example 1, Stable Solution

$$R_0 = 4, \ h = 10, \ v = 0.5, \ \Sigma = 4, \ a = 1$$



Numerical Solutions: Example 2, Limit Cycle

$$R_0 = 4, \ h = 10, \ v = 0.5, \ \Sigma = 8, \ a = 2$$



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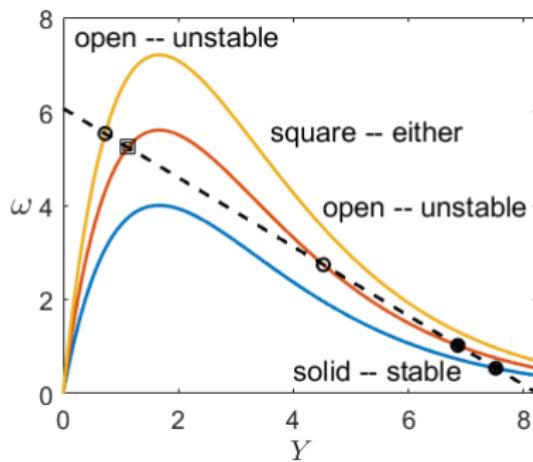
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Non-monotone ω (high disease \rightarrow waning confidence)

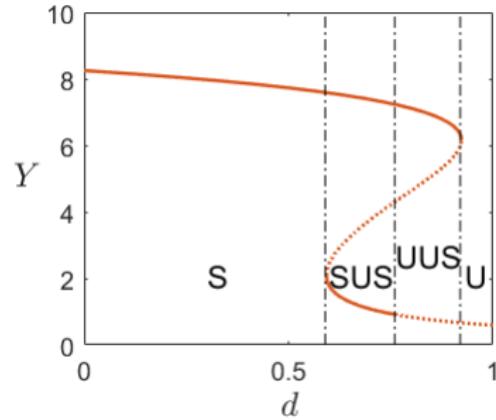
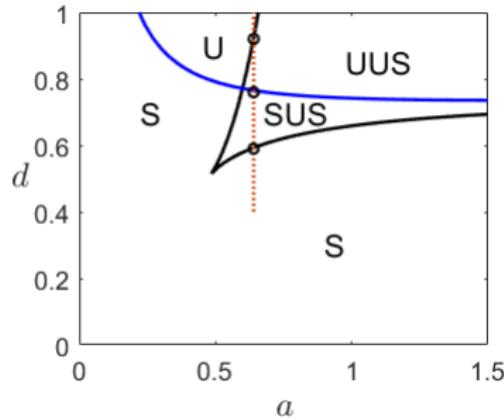
Example: $\omega = \sum daY e^{-aY}$



- ▶ Small d : large stable solution
- ▶ Moderate d : small maybe, middle unstable, large stable
- ▶ Large d : small unstable solution

Non-monotone ω (high disease \rightarrow waning confidence)

Example: $\omega = \sum daY e^{-aY}$



- ▶ Above/right of black curve: three solutions.
 - Middle unstable, largest stable.
- ▶ Above blue curve: smallest Y is unstable.
 - There is a stable limit cycle.

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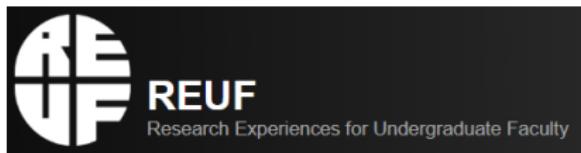
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- ▶ We proposed a disease spread model that includes flexible ideology on vaccination, where **the transition of the opinion is a function of the disease spread**.
- ▶ There is a unique endemic disease equilibrium for monotone increasing willingness, with a result that may be stable or unstable.
- ▶ By tuning the parameters carefully, we were able to observe limit cycles in the phase plane, that is, **oscillatory endemic states**.
- ▶ Multiple equilibria are possible with a willingness function that increases to a peak and then decreases.
 - There is a large Y stable equilibrium and either a small Y stable equilibrium or a small Y limit cycle.

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