

Lecture 10: Next Generation Approach

The Jacobian can be too difficult to analyze directly sometimes. The idea of the next generation approach is to track infections.

Idea:

1. split equation into infected and non-infected components:

$$\dot{x}_i = f_i(x, y) \quad i = 1, \dots, n$$

$$\dot{y}_j = g_j(x, y) \quad j = 1, \dots, n$$

$x \sim$ infected compartments

$y \sim$ uninfected compartments

2. Split infected compartment into new infections minus old infections

$$\dot{x}_i = \mathcal{F}_i(x, y) - \mathcal{V}_i(x, y)$$

$$\dot{y}_j = g_j(x, y)$$

a) $\mathcal{F}_i(0, y) = 0, \mathcal{V}_i(0, y) = 0$

b) $\mathcal{F}_i(x, y) \geq 0$

c) $\mathcal{V}_i(x, y) \leq 0$ when $x_i = 0$

d) $\sum_i \mathcal{V}_i(x, y) \geq 0$

$$1. \dot{S} = \Lambda - \beta IS - \nu S$$

$$\dot{E} = \beta IS - \gamma E - \nu E$$

$$\dot{I} = \gamma E - \alpha I - \nu I$$

$$\dot{R} = \alpha I - \nu R$$

↓

$$\dot{E} = \beta IS - \gamma E - \nu E$$

$$\dot{I} = \gamma E - \alpha I - \nu I$$

Infected compartments

$$\dot{S} = \Lambda - \beta IS - \nu S$$

$$\dot{R} = \alpha I - \nu R$$

Uninfected compartments

$$\mathcal{F}_1(E, I, S, R) = \beta IS$$

$$\mathcal{F}_2(E, I, S, R) = \gamma E$$

$$\mathcal{V}_1(E, I, S, R) = -\gamma E - \nu E$$

$$\mathcal{V}_2(E, I, S, R) = -\alpha I - \nu I$$

All properties are satisfied in this case.

3. There exists a disease free equilibrium $E_0 = (0, y_0)$ such that all initial conditions of the form $(0, y)$ approach $(0, y_0)$ as $t \rightarrow \infty$.

Disease free equilibrium for SEIR is $(0, 0, \Lambda/\mu, 0)$.

If we have an initial condition of the form $(0, 0, S_0, 0)$ then

$$\dot{E} = 0$$

$$\dot{I} = 0$$

$$\dot{S} = \Lambda - \mu S$$

$$\dot{R} = 0$$

$$\Rightarrow \lim_{t \rightarrow \infty} S(t) = \Lambda/\mu$$

4. By construction if we linearize near the disease free equilibrium:

$$\dot{x}_j = f_j(x, y) - g_j(x, y)$$

$$\dot{y}_j = g_j(x, y)$$

$$\Rightarrow \dot{x}_j = \sum_{i=1}^n \left(\frac{\partial f_i}{\partial x_j} \Big|_{(0, y_0)} x_i - \frac{\partial g_i}{\partial x_j} \Big|_{(0, y_0)} x_i \right)$$

where by property (a) the other partial derivatives vanish.

$$\text{Letting } F = \frac{\partial f_i}{\partial x_j}, V = \frac{\partial g_i}{\partial x_j}$$

we obtain

$$\dot{\vec{x}} = (F - V)\vec{x}$$

near $(0, y_0)$.

For the SEIR model we obtain

$$-\frac{\partial \dot{E}}{\partial E} \Big|_{(0, y_0)} = -\gamma - \mu = -\gamma - \mu$$

$$-\frac{\partial \dot{E}}{\partial I} \Big|_{(0, y_0)} = \beta S \Big|_{S = \Lambda/\mu} = \frac{\beta \Lambda}{\mu}$$

$$-\frac{\partial \dot{I}}{\partial E} = \gamma$$

$$-\frac{\partial \dot{I}}{\partial I} = -\alpha - \mu$$

$$\Rightarrow F = \begin{bmatrix} 0 & \beta \Lambda / \mu \\ \gamma & 0 \end{bmatrix}$$

$$V = \begin{bmatrix} \gamma + \mu & 0 \\ 0 & \alpha + \mu \end{bmatrix}$$

5. The next generation matrix is defined as

$$K = FV^{-1}$$

and

$$R_0 = \rho(FV^{-1}),$$

i.e. the maximum of the absolute values of the eigenvalues of FV^{-1} .

$$F \cdot V^{-1} = \begin{bmatrix} 0 & \beta \Delta y \\ \gamma & 0 \end{bmatrix} \begin{bmatrix} 1/\mu \\ 0 \end{bmatrix}$$

$$= \begin{bmatrix} 0 & \beta \Delta y / \mu \\ \gamma/\mu & 0 \end{bmatrix}$$

$$\Rightarrow \lambda_{1,2} = \pm \sqrt{\frac{\beta \Delta y}{\mu(\lambda + \mu)(\gamma + \mu)}}$$

The next generation reproduction number is

$$R_0 = \sqrt{\frac{\beta \Delta y}{\mu(\lambda + \mu)(\gamma + \mu)}}$$

Why does this work??

Let $m(F-V) = \max \text{ real part of eigenvalues of } F-V$.

FV^{-1} - has nonnegative entries $\Rightarrow R_0$ is an eigenvalue of FV^{-1} .

Theorem:

$m(F-V) < 0$ if and only if $\rho(FV^{-1}) < 1$.

$m(F-V) > 0$ if and only if $\rho(FV^{-1}) > 1$.

proof (Idem):

(totally wrong):

$$\begin{array}{l|l} F-V < 0 & F-V > 0 \\ \Rightarrow F < V & F > V \\ \Rightarrow F \cdot V^{-1} < 1 & F \cdot V^{-1} > 1 \end{array}$$

Example:

$$\dot{S} = -\beta SI$$

$$\dot{I} = \beta SI - \alpha I$$

$$\dot{R} = \alpha I$$

Infected compartments

$$\dot{I} = \beta SI - \alpha I$$

Disease free state!

$$S = N$$

$$I = 0$$

$$R = 0$$

$$\frac{\partial \dot{I}}{\partial I} = \beta S - \alpha$$

$$\left. \frac{\partial \dot{I}}{\partial I} \right|_{S=N} = \beta N - \alpha$$

$$\Rightarrow F = \beta N$$

$$V = \alpha$$

$$\Rightarrow F \cdot V^{-1} = \frac{\beta N}{\alpha}$$

Jacobian at $(N, 0, 0)$

$$= \begin{bmatrix} 0 & -\beta N & 0 \\ 0 & \beta N - \alpha & 0 \\ 0 & \alpha & 0 \end{bmatrix}$$

\Rightarrow Eigenvalues are $\beta N - \alpha, 0, 0$.

$$\Rightarrow R_0 = \frac{\beta N}{\alpha}$$

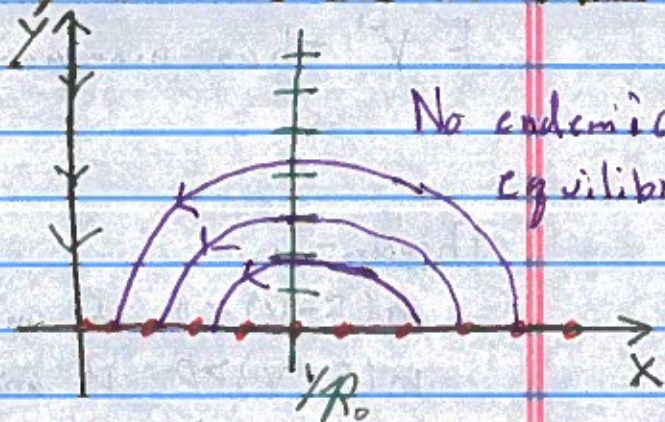
What is going on with 0 eigenvalues?
Reduce to dimensionless form!

$$\dot{x} = -R_0 x y$$

$$\dot{y} = R_0 x y - y$$

$x=0, x=1/R_0$ are nullclines

$y=0$ is a line of fixed points.



Numerical Computation.

Suppose a matrix A has only positive eigenvalues. How can we compute the largest eigenvector??

Let $\{\vec{v}_1, \dots, \vec{v}_n\}$ be an eigenbasis with eigenvalues $\lambda_1, \dots, \lambda_n$ ordered so that $\lambda_1 > \lambda_2 > \lambda_3 > \dots > \lambda_n$.

$$\vec{x} = c_1 \vec{v}_1 + \dots + c_n \vec{v}_n$$

$$\Rightarrow A\vec{x} = c_1 \lambda_1 \vec{v}_1 + \dots + c_n \lambda_n \vec{v}_n$$

$$A^2 \vec{x} = c_1 \lambda_1^2 \vec{v}_1 + \dots + c_n \lambda_n^2 \vec{v}_n$$

⋮

$$A^m \vec{x} = c_1 \lambda_1^m \vec{v}_1 + \dots + c_n \lambda_n^m \vec{v}_n$$

$$= c_1 \lambda_1^m \left(\vec{v}_1 + \frac{\lambda_2^m}{\lambda_1^m} \vec{v}_2 + \dots + c_n \frac{\lambda_n^m}{\lambda_1^m} \vec{v}_n \right)$$

$$\approx c_1 \lambda_1^m \vec{v}_1$$

Therefore,

$$\|A^m \vec{x}\| \approx c_1 \lambda_1^m$$

$$\Rightarrow \vec{v}_1 \approx \frac{A^m \vec{x}}{\|A^m \vec{x}\|}$$

To find λ_1 , note that

$$\lambda_1 = \frac{\vec{v}_1^T A \vec{v}_1}{\|\vec{v}_1\|} = \vec{v}_1^T A \vec{v}_1$$