

## Lecture 11: Model Analysis and Sensitivity Analysis.

- Suppose,  $Y_j(t_j)$  denote data points at time  $t_j$ .
- Suppose,  $I(t_j)$  denote model predictions at time  $t_j$ .

$$\Rightarrow SSE = \|Y - I\|_e^2 = \sum_{j=1}^n (Y_j(t_j) - I(t_j))^2$$

Fitting a model to data means minimizing  $\|Y - I\|_e$  over free parameters.

### Example:

For the SIR model

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

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

$$S(0) = S_0, I(0) = I_0$$

there are potentially four parameters  
 $\alpha, \beta, S_0, I_0$ .

However, exact knowledge of the data should be used to reduce the number of parameters.

### Model Selection:

$$AIC = n \ln \left( \frac{SSE}{n} \right) + 2K$$

$n \sim$  number of data points

$K \sim$  number of parameters + 1.

Find model that minimizes AIC when  
 $n/K \geq 40$ .

For  $n/k < 40$  use

$$AIC_c = n \ln \left( \frac{SSE}{n} \right) + 2K + \frac{2K(K+1)}{n-K-1}$$

$$\Delta_j = AIC_j - AIC_{\min}$$

Distance from model to minimum explanation.

$0 \leq \Delta_j \leq 2$  support in data.

$4 \leq \Delta_j \leq 7$  little support in data.

$$W_i = \frac{\exp(-\Delta_i/2)}{\sum_j \exp(-\Delta_j/2)} \sim \text{weight of the } i\text{-th model.}$$

$W_i > .9 \Rightarrow$  Robust inferences can be made using the best fitted model.

Example:

$m_1:$

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

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

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

$m_2:$

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

$$\dot{E} = \beta SI - \gamma E$$

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

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

\* Can learn exposed people.

$m_3:$

$$\dot{S} = -\beta S(I + \gamma A)$$

$$\dot{E} = \beta S(I + \gamma A) - (\gamma + \kappa)E$$

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

$$\dot{A} = \kappa E - \nu A$$

$$\dot{R} = \alpha I + \nu A$$

\* Can learn about asymptomatic class.

All the light we cannot see.

## Sensitivity Analysis

$$\dot{y}_i = f_i(y_1, \dots, y_n, t, p)$$

↑  
parameter

How does small changes in  $p$  influence behavior of system.

We view

$$y_i: \underset{\substack{\uparrow \\ \text{time}}}{\mathbb{R}} \times \underset{\substack{\uparrow \\ \text{parameter}}}{\mathbb{R}} \rightarrow \mathbb{R}$$

Define,

$$z_i = \frac{\partial y_i}{\partial p}, \text{ (how does } y \text{ change with } p \text{).}$$

$$\Rightarrow \frac{dz_i}{dt} = \frac{d}{dt} \frac{\partial y_i}{\partial p} = \frac{d}{dp} \frac{dy_i}{dt} = \frac{d}{dp} f_i(y_1(t, p), \dots, y_n(t, p), t, p)$$

$$\Rightarrow \frac{dz_i}{dt} = \frac{\partial f_i}{\partial p} + \sum_{j=1}^n \frac{\partial f_i}{\partial y_j} \frac{\partial y_j}{\partial p}$$

$$\Rightarrow \boxed{\frac{d\vec{z}}{dt} = \nabla_p f + J \cdot \vec{z}}$$

## Initial Conditions:

$$z_i(0) = \lim_{\Delta p \rightarrow 0} \frac{y_i(0, p + \Delta p) - y_i(0, p)}{\Delta p}$$

1. If  $p$  is not an initial condition:

$$y_i(0, p + \Delta p) - y_i(0, p) = 0$$

$$\text{for all } \Delta p \Rightarrow z_i(0) = 0.$$

2. If  $p$  is an initial condition:

$$y_k(0, p + \Delta p) - y_k(0, p) = \Delta p$$

for some  $k$ ; therefore,  $z_k(0) = 1$ ,  $z_i(0) = 0$  for  $i \neq k$ .

Example:

1. SIR:

$$\dot{S} = -\beta SI, \quad S(0) = S_0,$$

$$\dot{I} = \beta SI - \alpha I, \quad I(0) = I_0,$$

$$\dot{R} = \alpha I, \quad R(0) = 0,$$

- Sensitivity of  $\beta$ :

$$z_1 = \frac{dS}{d\beta}, \quad z_2 = \frac{dI}{d\beta}, \quad z_3 = \frac{dR}{d\beta}$$

$$\left. \begin{aligned} \dot{z}_1 &= -SI - \beta I z_1 - \beta S z_2 \\ \dot{z}_2 &= SI + \beta I z_1 + (\beta S - \alpha) z_2 \\ \dot{z}_3 &= \alpha z_2 \end{aligned} \right\} \begin{array}{l} S \text{ and } I \text{ are computed} \\ \text{from known solutions.} \end{array}$$

$$z_1 = z_2 = z_3 = 0.$$

- Sensitivity of  $\alpha$ :

$$z_1 = \frac{dS}{d\alpha}, \quad z_2 = \frac{dI}{d\alpha}, \quad z_3 = \frac{dR}{d\alpha}$$

$$\dot{z}_1 = -\beta I z_1 - \beta S z_2$$

$$\dot{z}_2 = -I + \beta I z_1 + (\beta S - \alpha) z_2$$

$$\dot{z}_3 = I + \alpha z_2$$

$$z_1 = z_2 = z_3 = 0.$$

Numerical Computation for  $\beta$ :

$$z_1(t + \Delta t) = z_1(t) - (S(t)I(t) + \beta I(t)z_1(t) + \beta S(t)z_2(t))\Delta t$$

$$z_2(t + \Delta t) = z_2(t) + (S(t)I(t) + \beta I(t)z_1(t) + (\beta S(t) - \alpha)z_2(t))\Delta t$$

$$z_3(t + \Delta t) = z_3(t) + \alpha z_2(t)\Delta t$$

$$z_1(0) = z_2(0) = z_3(0) = 0.$$

## 2. SEIR:

$$\begin{aligned}\dot{S} &= -\beta SI & S(0) &= S_0 \\ \dot{E} &= \beta SI - \gamma E & I(0) &= I_0 \\ \dot{I} &= \gamma E - \alpha I & E(0) &= E_0 \leftarrow \text{parameter.} \\ \dot{R} &= \alpha I & R(0) &= 0\end{aligned}$$

### - Sensitivity of $\gamma$ :

$$z_1 = \frac{dS}{d\gamma}, \quad z_2 = \frac{dE}{d\gamma}, \quad z_3 = \frac{dI}{d\gamma}, \quad z_4 = \frac{dR}{d\gamma}$$

$$\Rightarrow \dot{z}_1 = -\beta I z_1 - \beta S z_3$$

$$\dot{z}_2 = -E + \beta I z_1 - \gamma z_2 + \beta S z_3, \quad z_1(0) = z_2(0) = z_3(0) = z_4(0) = 0$$

$$\dot{z}_3 = E + \gamma z_2 - \alpha z_3$$

$$\dot{z}_4 = \alpha z_3$$

### - Sensitivity of $E_0$ :

$$\dot{z}_1 = -\beta I z_1 - \beta S z_3$$

$$\dot{z}_2 = \beta I z_1 - \gamma z_2$$

$$\dot{z}_3 = \gamma z_2 - \alpha z_3$$

$$\dot{z}_4 = \alpha z_3$$

$$z_1(0) = z_3(0) = z_4(0) = 0$$

$$z_2(0) = 1.$$

## Elasticity of Parameters

Sensitivity of quantity  $Q$  with respect to parameter  $p$ :

$$\lambda_Q^p = \frac{\partial Q}{\partial p}$$

Elasticity of quantity  $Q$  with respect to parameter  $p$ :

$$E_Q^p = \frac{\partial Q}{\partial p} \frac{p}{Q} = \frac{\% \Delta Q}{\% \Delta p}$$

Example:

- SIR:

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

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

$$\Rightarrow \frac{\partial R_0}{\partial \beta} \cdot \frac{\beta}{R_0} = \frac{N}{\alpha} \cdot \frac{\beta}{\beta N / \alpha} = 1 \Rightarrow 100\% \text{ change in } R_0 \text{ with } 100\% \text{ change in } \beta.$$

$$\Rightarrow \frac{\partial R_0}{\partial \alpha} \cdot \frac{\alpha}{R_0} = -\frac{\beta N}{\alpha^2} \cdot \frac{\alpha}{\beta N / \alpha} = -1 \Rightarrow 100\% \text{ decrease in } R_0 \text{ with } 100\% \text{ change in } \alpha.$$