

# Lecture 9: Full Calibration of SIR Models

**Radu Balan**

Department of Mathematics, NWC  
University of Maryland, College Park, MD

Version: February 23, 2023

# SIR Model Calibration

Recall the model:

$$\begin{cases} \frac{dS}{dt} = -\beta S \frac{I}{N}, & S(0) \\ \frac{dI}{dt} = \beta S \frac{I}{N} - \alpha I, & I(0) \\ \frac{dR}{dt} = \alpha I, & R(0) \end{cases}$$

with the sub-compartments  $X(t) = (1 - \gamma)R(t)$  for “recovered” and  $Y(t) = \gamma R(t)$  for deaths.

Before making useful predictions (testing), the model has to be calibrated. Last time we analyzed estimators for  $\gamma$  from the time series of *cumulative detected infections*,  $\{V(0), \dots, V(T_{max})\}$ , and the time series of *cumulative deaths*,  $\{Y(0), \dots, Y(T_{max})\}$ . The idea was to minimize over  $\tau \geq 0, \gamma \in [0, 1]$  the norm  $\|Y(\cdot + \tau) - \gamma V\|_p$ , scaled by the number of terms in each sum. Now we integrate these estimators into a scheme that performs full calibration of the SIR model.

# Optimization based on Forward Model Simulation

## Pre-processing steps

**Step 1:**  $t_0$ . First detect the onset of infections, and reset the time origin to match this onset  $t_0$ .

**Step 2:**  $I(t)$ . The time series of cumulative detected infections  $\{V(0), \dots, V(T_{T_{max}})\}$  should be turned into a rate of infections. The daily rate may not be relevant to infection transmissions. Instead use a time window to convert the cumulative count into a rate:

$$I(t) = V(t + t_0 + \tau_0) - V(t + t_0 - \tau_0) \quad , \quad t = 0, 1, 2, \dots, T_{max}$$

where  $\tau_0 > 0$  is chosen so that  $\tau_0$  accounts for average infection period. For Covid-19 it is somewhere between 5 days and 10 days. A possible value is  $\tau_0 = 7$ . Thus  $I(t)$  measures the number of infections in a 2-week period centered around  $t$ .

# Optimization based on Forward Model Simulation

## The Meta Loop

A natural choice for initial conditions is given by:  $S(0) = N$  and  $R(0) = 0$ , where we assumed  $I(0) \ll N$ :

$$\begin{cases} \frac{dS}{dt} = -\beta S \frac{I}{N}, & S(0) = N \\ \frac{dI}{dt} = \beta S \frac{I}{N} - \alpha I, & I(0) > 0 \\ \frac{dR}{dt} = \alpha I, & R(0) = 0 \\ Y = \gamma R \end{cases}$$

At this point, the parameters that need to be estimated are:  $\{\alpha, \beta, \gamma, N\}$ .

The Forward Model based calibration works like this:

1. Construct a search set  $\Omega$  of the “free” parameters  $(\alpha, \beta, N)$
2. For each triple  $(\alpha, \beta, N) \in \Omega$ : (i) run a SIR simulator using the Euler scheme that produces  $(S_{sim}, I_{sim}, R_{sim})$ . (ii) Fit  $\hat{\gamma}$  to match the observed time series  $Y$ . (iii) Compute the value of the objective function  $J(\alpha, \beta, N, \hat{\gamma})$
3. Select  $(\hat{\alpha}, \hat{\beta}, \hat{N})$  that minimize  $J$ .

# Optimization based on Forward Model Simulation

## Details: The objective function

The objective function  $J$  can be chosen to measure a norm of residuals  $I - I_{sim}$  and  $Y - Y_{sim}$ , where  $Y_{sim} = \hat{\gamma}R_{sim}$

To do so, you need to choose and fix two meta-parameters  $p$  and  $\lambda$ . For  $1 \leq p < \infty$  define

$$J(\alpha, \beta, N, \gamma) = \sum_{t=0}^{T_{max}} |I(t) - I_{sim}(t)|^p + \lambda \sum_{t=0}^{T_{max}} |Y(t) - \gamma R_{sim}(t)|^p$$

For  $p = \infty$  define

$$J(\alpha, \beta, N, \gamma) = \max_{0 \leq t \leq T_{max}} |I(t) - I_{sim}(t)| + \lambda \max_{0 \leq t \leq T_{max}} |Y(t) - \gamma R_{sim}(t)|$$

The meta-parameter  $\lambda$  weights one term of error measure (the rate of infections mismatch) against the other term (the number of deaths).

Typical choices:  $p = 1, 2, \infty$  and  $\lambda = 1$ .

# The Calibration Algorithm for SIR Models

## Algorithm (Meta-Algorithm for SIR Calibration)

*Inputs:* Time series  $\{V(0), \dots, V(T)\}$ ,  $\{Y(0), \dots, Y(T)\}$ . *Parameters:*  $V_{min}$  (default,  $V_{min} = 5$ ),  $\tau_0$  (default,  $\tau_0 = 7$ ),  $p \in [1, \infty]$  (default,  $p = 2$ ),  $\lambda > 0$  (default,  $\lambda = 1$ ), Search set  $\Omega$ .

- 1 Detect the onset of the infection  $t_0$  as the first time so that  $V(t_0) \geq V_{min}$ , reset the time origin, and create the time series of infection rates  $I(t) = V(t + t_0 + \tau_0) - V(t + t_0 - \tau_0)$ ,  $0 \leq t \leq T_{max}$ .
- 2 For each  $(\alpha, \beta, N) \in \Omega$  repeat:
  - 1 Simulate a SIR model with parameters  $(\alpha, \beta, N)$  and obtain **daily** time series  $(S_{sim}, I_{sim}, R_{sim})$ .
  - 2 Solve  $\hat{\gamma} = \operatorname{argmin}_{\gamma} \|Y - \gamma R_{sim}\|_p$ .
  - 3 compute the objective function  $J$ .
- 3 Determine the minimum and the minimizer of  $J$ .

*Outputs:* Estimated  $\hat{\alpha}, \hat{\beta}, \hat{N}, \hat{\gamma}$  and minimum value of the objective function  $J_{min}$ .

# Numerical results (1)

Analysis of DC data for 2020

$V_{min} = 5$ ,  $\tau_0 = 7$  [days].

Onset  $t_0 = 55$  (Monday: March 16, 2020),

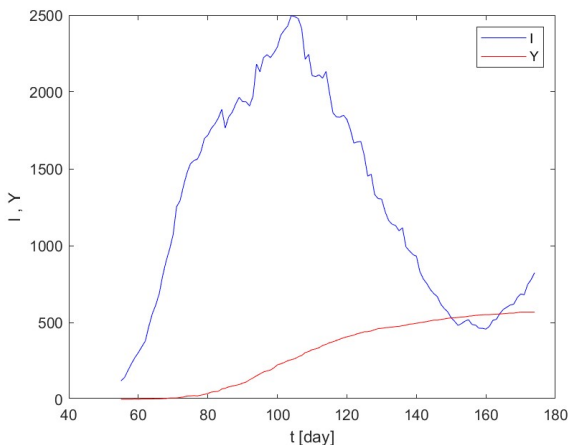
$T_{max} = 119$ .

Notable dates:

Local dips: day=152 (Sunday: 6/21/2020),  
day=159 (Sunday: 6/28/2020).

Memorial weekend: Sunday May 24, 2020: day = 124 .

4th of July: Saturday July 4, 2020: day = 165.



# Numerical results (2)

Analysis of DC data for 2020.  $p = 1$

Results:

$$\hat{\alpha} = 0.12, \hat{\beta} = 0.21,$$

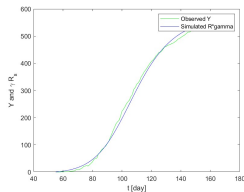
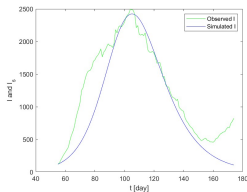
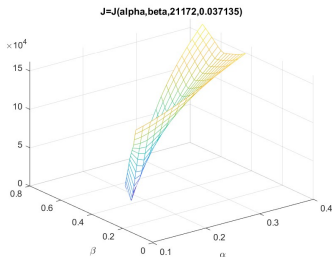
$$\hat{R}_0 = 1.75,$$

$$\hat{N} = 21172,$$

$$\frac{\hat{N}}{\text{CityPopulation}} = 3\%,$$

$$\text{CityPopulation} = 705749.$$

$$\hat{\gamma} = 3.7\%.$$





# Numerical results (3)

Analysis of DC data for 2020.  $p = 2$

Results:

$$\hat{\alpha} = 0.12,$$

$$\hat{\beta} = 0.216,$$

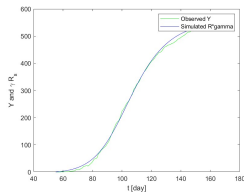
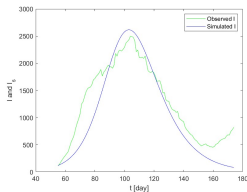
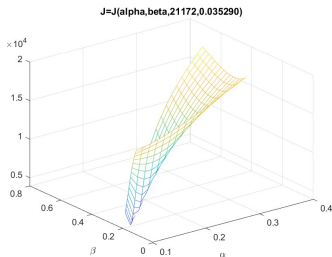
$$\hat{R}_0 = 1.8,$$

$$\hat{N} = 21172,$$

$$\frac{\hat{N}}{\text{CityPopulation}} = 3\%,$$

$$\text{CityPopulation} = 705749.$$

$$\hat{\gamma} = 3.5\%.$$



# Numerical results (4)

Analysis of DC data for 2020.  $p = \infty$

Results:

$$\hat{\alpha} = 0.14,$$

$$\hat{\beta} = 0.238,$$

$$\hat{R}_0 = 1.7,$$

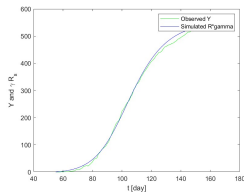
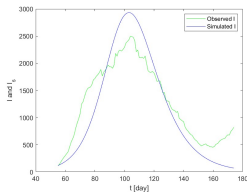
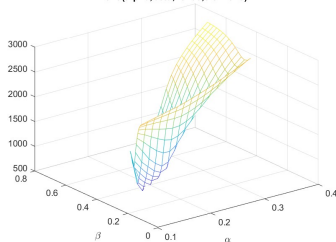
$$\hat{N} = 28230,$$

$$\frac{\hat{N}}{\text{CityPopulation}} = 4\%,$$

$$\text{CityPopulation} = 705749.$$

$$\hat{\gamma} = 2.8\%.$$

$J=J(\alpha,\beta,28230,0.027929)$



## Next Model: SEIR

The Susceptible-Exposed-Infected-Removed (SEIR) model is obtained from SIR by introducing a compartment between Susceptible and Infected of population that has been exposed to the virus but are not yet contagious:

$$\left\{ \begin{array}{l} \frac{dS}{dt} = -\beta \frac{SI}{N}, \quad S(0) \\ \frac{dE}{dt} = \beta \frac{SI}{N} - \delta E, \quad E(0) \\ \frac{dI}{dt} = \delta E - \alpha I, \quad I(0) \\ \frac{dR}{dt} = \alpha I, \quad R(0) \end{array} \right. \quad (\text{SEIR Model})$$

where  $\delta \geq 0$  is the rate of transition from exposed to infected. Its reciprocal  $1/\delta$  represents the *average incubation period*.

If data is selected from the onset of infections, a natural initial condition is:  $R(0) = N \gg I(0)$ ,  $R(0) = 0$ . The initial exposed population  $E(0)$  may be set to  $I(0)$ , or can be fine tuned to fit the data. Assumin  $E(0) = I(0)$  is known, the parameters that need to be calibrated are:  $\alpha, \beta, \gamma, \delta, N$ .