

# Epidemiological Modelling of Coronavirus Infection in Quarantined Population

## A Outline of Project for MATH371 Wi2020

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# Outline

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# Objectives

- ▶ to explore the SIRD model and its stability points
- ▶ to develop methodology to estimate parameters for data without assumption of population
- ▶ to estimate parameters of the SIR model given COVID-19 epidemiological data from Wuhan of China
- ▶ to explore SIRD models that embed quarantining of population, and age distribution

# COVID-19

The COVID-19 pandemic is caused by a novel virus named SARS-CoV-2 that found its first confirmed case in December 2019. By March 2020, 780,000 cases of COVID-19 have been confirmed around the world, and it is still developing.

# SIRD Model

The fraction of individuals susceptible to contract COVID-19 in a population,  $\frac{S}{N} = s$ , fraction of infected individuals,  $\frac{I}{N} = i$ , fraction of removed individuals,  $\frac{R}{N} = r$  and fraction of dead,  $\frac{D}{N} = d$  change in dynamic described in the following system of differential equations:

$$\begin{aligned}\dot{s} &= -\beta si & s(0) &= s_0 \\ \dot{i} &= \beta si - (\alpha + \mu)i & i(0) &= i_0 \\ \dot{r} &= \alpha i & r(0) &= r_0 \\ \dot{d} &= \mu i & d(0) &= 0\end{aligned}\tag{1}$$

where:

$\beta$  = infection rate;  $\frac{1}{\beta}$  = average days to be infected

$\alpha$  = recovery rate;  $\frac{1}{\alpha}$  = average days to recover

$\mu$  = death rate

$R_0$  = basic reproduction number,  $\frac{\beta}{\alpha + \mu}$

# Stability Analysis of the SIRD

Taking out  $d = 1 - s - i - d$ , the system in (1) can be rewritten as

$$\dot{\mathbf{x}} = \mathbf{F}(\mathbf{x})$$

where  $\mathbf{x}^T = [s \ i \ r]$ .

$\dot{\mathbf{x}}^T = 0$  at steady state. This implies  $\mathbf{x}^T = \hat{\mathbf{x}}^T = [s_{ss} \ 0 \ r_{ss}]$

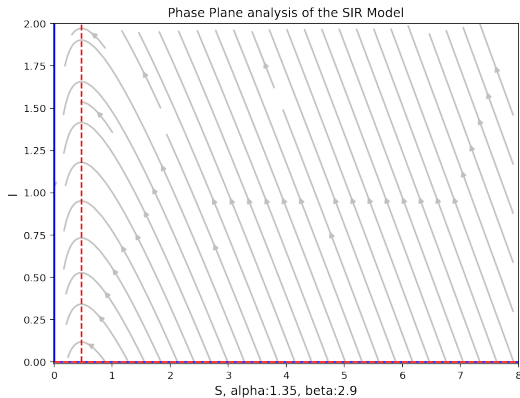
where  $s_{ss}$  and  $r_{ss}$  are  $s$  and  $r$  at steady state, respectively.

The linearized system around steady state has Jacobian matrix

$$J\hat{\mathbf{x}} = \begin{pmatrix} 0 & \beta s_{ss} & 0 \\ 0 & \beta s_{ss} - (\alpha + \mu) & 0 \\ 0 & \alpha & 0 \end{pmatrix} \quad (2)$$

and eigenvalues  $\lambda_1, \lambda_2 = 0, \lambda_3 = bs - (\alpha + \mu) < 0$  since  $s_{ss}$  is less than  $s$  when  $i$  is maximum

# Phase Plane Analysis



**Figure:** Phase plane plot obtained from Jay Newby 2020 lecture materials  
The phase plane plot reveals that  $i = 0, s = 0$  is a stable point.

## SIRD with Quarantine

The SIRD model with quarantine adds a compartment  $\dot{q} = \frac{Q}{N}$ , where  $Q$  represents the quarantined infected individuals. Infected individuals are quarantined at rate  $\gamma$ , recover at rate  $\alpha$ , and perish at rate  $\mu$ .

$$\begin{aligned} \dot{s} &= -\beta si & s(0) &= s_0 \\ \dot{i} &= \beta si - (\alpha + \mu + \gamma)i & i(0) &= i_0 \\ \dot{r} &= \alpha(i + q) & r(0) &= r_0 \\ \dot{q} &= \gamma i - (\mu + \alpha)q & q(0) &= 0 \\ \dot{d} &= \mu(i + q) & d(0) &= 0 \end{aligned} \tag{3}$$

Replacing  $i$  with  $i + q$  in the linearized system of the SIRD in Eq(2) yields the same stable point  $(s, i) = (0, 0)$ .



# Assumptions

The SIR family of models assume:

- ▶ instantaneous infection, recovery of individuals
- ▶ no delay in any effect, including death
- ▶  $i$  has reached a peak and has started decreasing

which are not captured in the publicly available data.

# Motivation of Methods

The SIR models do not provide a way to decide  $N$ , value of which often distorts total error, and requires assumption by the researcher.

In the following methods,  $N$  is included in the parameter estimation without loss of system behaviour. Specifically, the SIR model is fitted on the data yielding estimated  $\hat{\alpha}, \hat{\mu}, \hat{N}$ .  $\beta$  is assumed 1 since it's linked in the change in time,  $dt$ .

# Minimisation Problem

To estimate parameters from data from regions that **have reached** the maximum number of confirmed cases ( $i$ ), we solve the following problem:

$$\begin{aligned} & \underset{\alpha, \beta=1, \mu, N}{\text{minimize}} && \left( \frac{I_{\max} - \widehat{I}_{\max}}{I_{\max}} \right)^2 + \frac{1}{T} \sum_{t=1}^T \left( \frac{R_t - \widehat{R}_t}{R_{\max}} \right)^2 + \left( \frac{D_t - \widehat{D}_t}{D_{\max}} \right)^2 \\ & \text{subject to} && 0 \leq \frac{\alpha + \mu}{\beta} \leq \left( \frac{N - (I + R + D)}{N} \right) \Big|_{I=I_{\max}} \end{aligned} \quad (4)$$

# Time Scale of Estimated Data

Using Euler's method, a differential form  $\dot{s} = -\beta s i$  can be expressed as

$$s_{t+1} = s_t + dt(-\beta s_t i_t)$$

Which we consider as

$$s_{t+1} = s_t + \beta dt(-s_t i_t)$$

Thus, a change in  $\beta$  leads to a change in the timescale. In Eq(4) we assumed  $\beta = 1$ .

# The Time-scale constant

The estimated data has time scale  $t'$ . It is scaled to match the timescale used by the data in  $t$  time scale. We obtain

$$k = \frac{t_m}{t'_m} \quad (5)$$

where  $t_m$  is the time at which  $i(t)$  is maximum and  $t'_m$  is the time when  $\hat{i}(t')$  is maximum.

Eq(5) implies  $t' = \frac{t}{k}$ . Thus, to match timescales,  $\hat{s}(t') = \hat{s}(\frac{t}{k})$ . In addition, since  $\beta = 1$  in Eq(4), the estimated parameters are scaled i.e.

$$\alpha = \frac{\hat{\alpha}}{k}, \beta = \frac{1}{k}, \mu = \frac{\hat{\mu}}{k}, N = \hat{N}$$

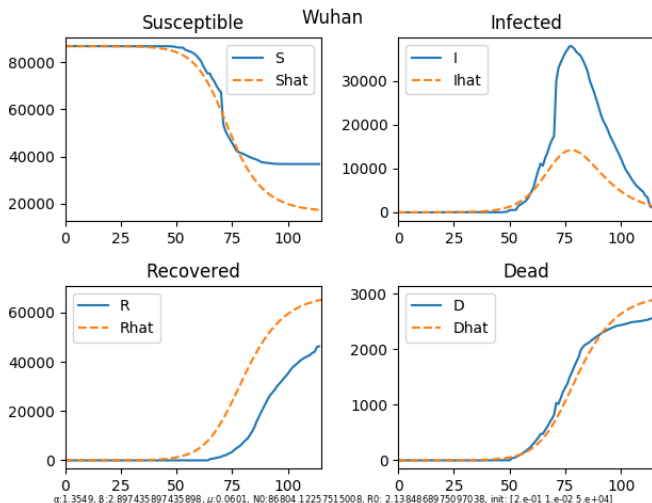
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## Result & Discussion

- ▶ solution to the SIR model is unique by the Picard-Lindelöf theorem ( $\mathbf{F}$  is Lipschitz continuous),
- ▶ the estimation methods avoids assumption of the population  $N$ .
- ▶ the methods are not robust to different initial guess in  $N$  fed to the solver (shown in Fig 2, 3).
- ▶ the SIRDQ model gives better estimates The estimated  $R_0$ 's

are larger than the values in the literature (1.4 to 6.49<sup>[1]</sup>). This may be because the number of people infected and recovered quickly (e.g. in 2 days), which is shown by the model, are not reported in the data.

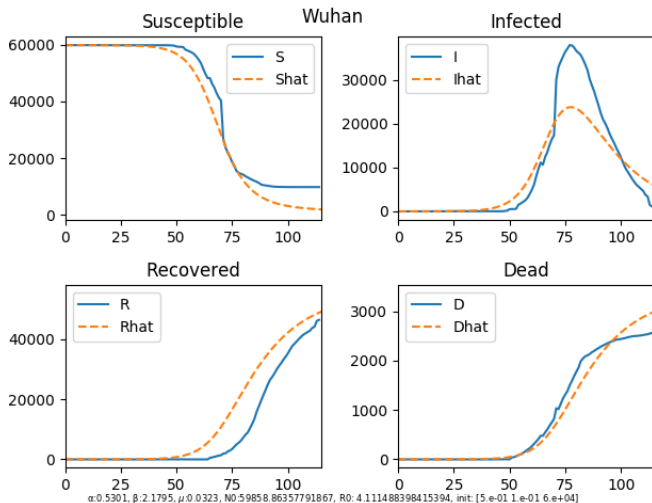
# SIRD Fit 1



**Figure:** Estimated  $s, i, r, d$  from Wuhan epidemiological data.  $R_0 \approx 2.138$

Using initial guess  $N_0 = 5 \times 10^4$ , the estimated  $\alpha \approx 1.35, \beta \approx 2.9, \mu \approx 0.06, N_0 \approx 9 \times 10^4$

# SIRD Fit 2



**Figure:** Estimated  $s, i, r, d$  from Wuhan epidemiological data.  $R_0 \approx 4.111$

Using initial guess  $N_0 = 6 \times 10^4$ , the estimated  $\alpha \approx 0.53, \beta \approx 2.18, \mu \approx 0.03, N_0 \approx 6 \times 10^4$



# SIRDQ Fit 1

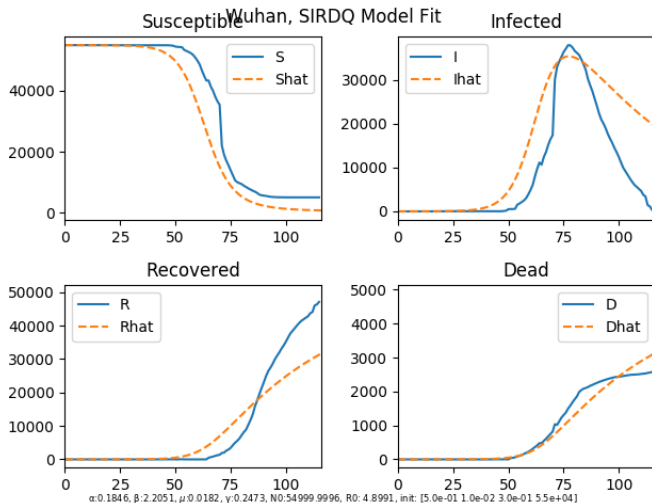


Figure: Estimated  $s, i, r, d$  from Wuhan epidemiological data.  $R_0 \approx 4.9$

Using initial guess  $N_0 = 5.5 \times 10^4$ , the estimated  $\alpha \approx 0.18, \beta \approx 2.2, \mu \approx 0.01, \gamma \approx 0.25, N_0 \approx 6 \times 10^4$

## Appendix and Reference

1. Liu, Y., Gayle A A., Wilder-Smith, A., Rocklov, J. (2020). The reproductive number of COVID-19 is higher compared to SARS coronavirus. *Journal of Travel Medicine*, 4-10.
2. The code and data used for estimation is published here:  
[Github/NatashaTing/covid19-modelling](https://github.com/NatashaTing/covid19-modelling).