



# A Linear Nonlocal Model for Outbreak of COVID-19 and Parameter Identification

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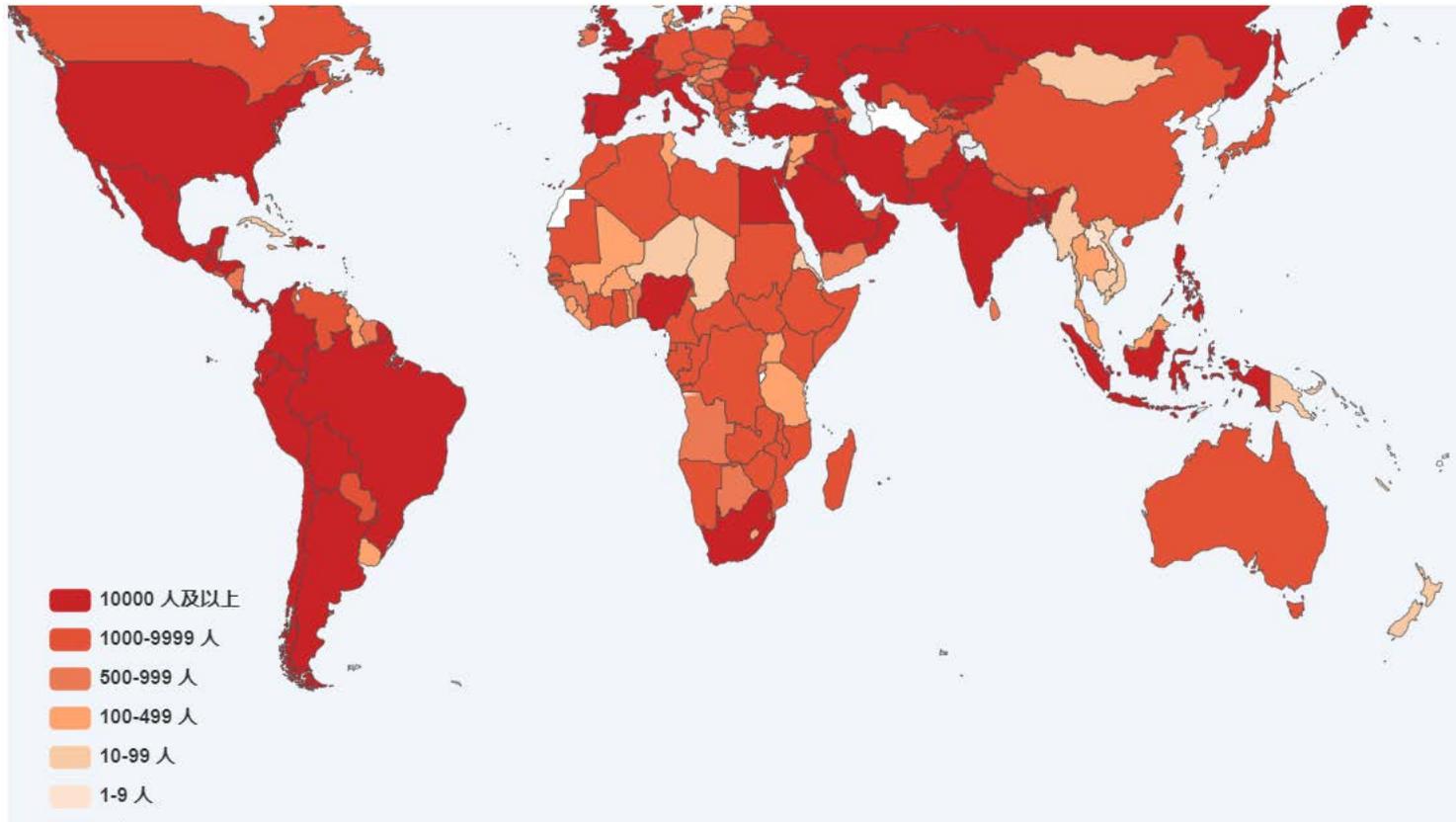
# Contents

- Spread of novel corona-virus in China and World
- Linear Nonlocal Dynamical Models
- Theoretical Analysis for the nonlocal models
- Parameter identification and Prediction based on the public data
- Some References

# Novel Corona-virus (COVID-19)

- Major event in the world

| Confirmed Cases | Death         |
|-----------------|---------------|
| <b>16059531</b> | <b>641840</b> |



# Media Coverage



01月23日  
疫情日报

疫情形势 [查看疫情实时地图 >](#)



今日：确诊830例，死亡25例，治愈34例

昨日：确诊571例，死亡17例

截至2020.01.23 24:00

# Mathematical Models for COVID19

- Why should we establish the “reasonable” mathematical models for COVID?
    - Some information about COVID-19 can not be **observed directly**
    - We want to **predict** the trends of the virus epidemic. Especially when the spread of the virus ends
- 

- We are trying to describe the problems, not from the medical point of view, but from the general transmission mechanism

# Questions

- Is it possible?
- Are there some existing models?
- What kind models are “reasonable” ?
- What is the characteristic of novel corona-virus pneumonia?

# Existing Models and Methods

- SI、 Logistic models (Nonlinear Quadratic models)
  - It is mainly used in epidemiology. The common situation is to explore the risk factors of a disease and predict the probability of a disease according to the risk factors

$$\frac{dI}{dt} = rI \left( 1 - \frac{I}{K} \right), \quad r = \beta K.$$

- SIR、 SIRS、 SEIR Models
  - Ordinary Differential Systems

$$\begin{cases} \frac{dS}{dt} = -\beta IS, & \frac{dE}{dt} = \beta IS - (\alpha + \gamma_1)E, \\ \frac{dI}{dt} = \alpha E - \gamma_2 I, & \frac{dR}{dt} = \gamma_1 E + \gamma_2 I. \end{cases}$$

# Review Analysis for SAR

- Prof. Li Daqian, Prof. Ding Guanghong and their team did the review analysis for SARS, which happened at the end of 2002:
  - Based on SIJR model, which is the simplified model of SEIJR.

## SARS 爆发预测和预警的数学模型研究

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# The Characteristic of COVID-19

- Latent Period
  - The novel corona-virus can spread in latent period
- The Delay of Data
  - The public data will be announced with some delays due to diagnosis

# Some Notations

- Notation as follows

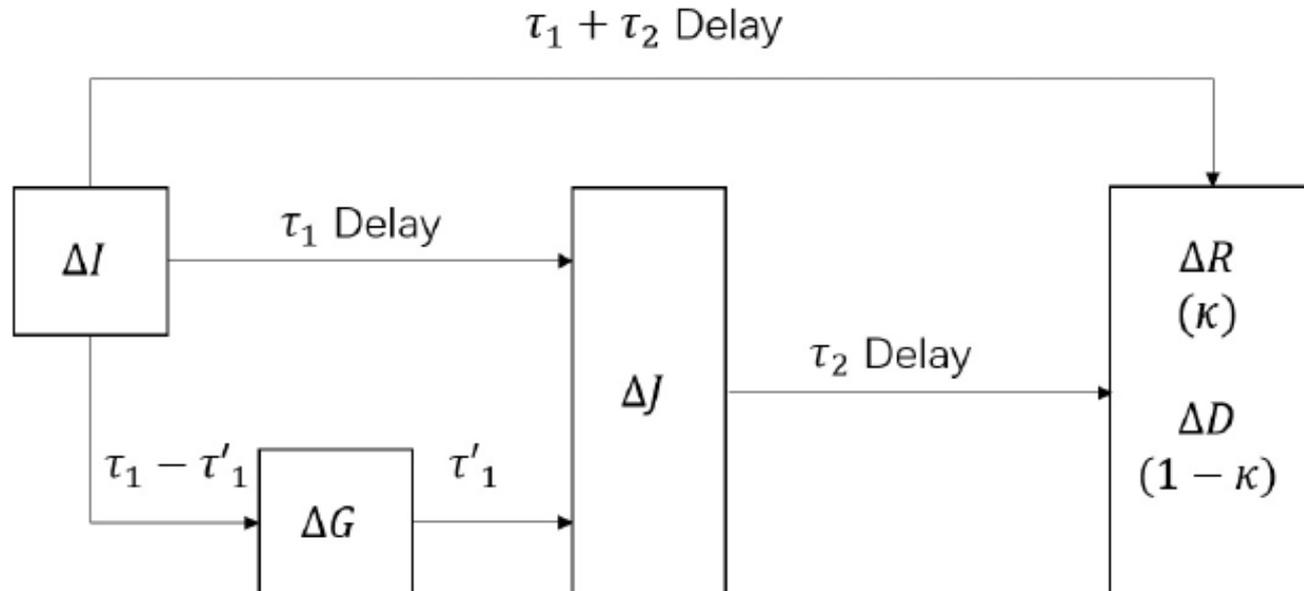
Public Data



- $I(t)$ : cumulative infected people at time  $t$ ;
- $J(t)$ : cumulative confirmed people at time  $t$ ;
- $G(t)$ : currently isolated people who are infected but still in latent period at time  $t$ ;
- $R(t)$ : cumulative cured people at time  $t$ .

# Dynamical System with Delays

- The demonstration of the dynamic system



# Frame Diagram

$f_2$  : the transition probabilities from infection to illness onset

$f_4$  : the transition probabilities from infection to hospitalization

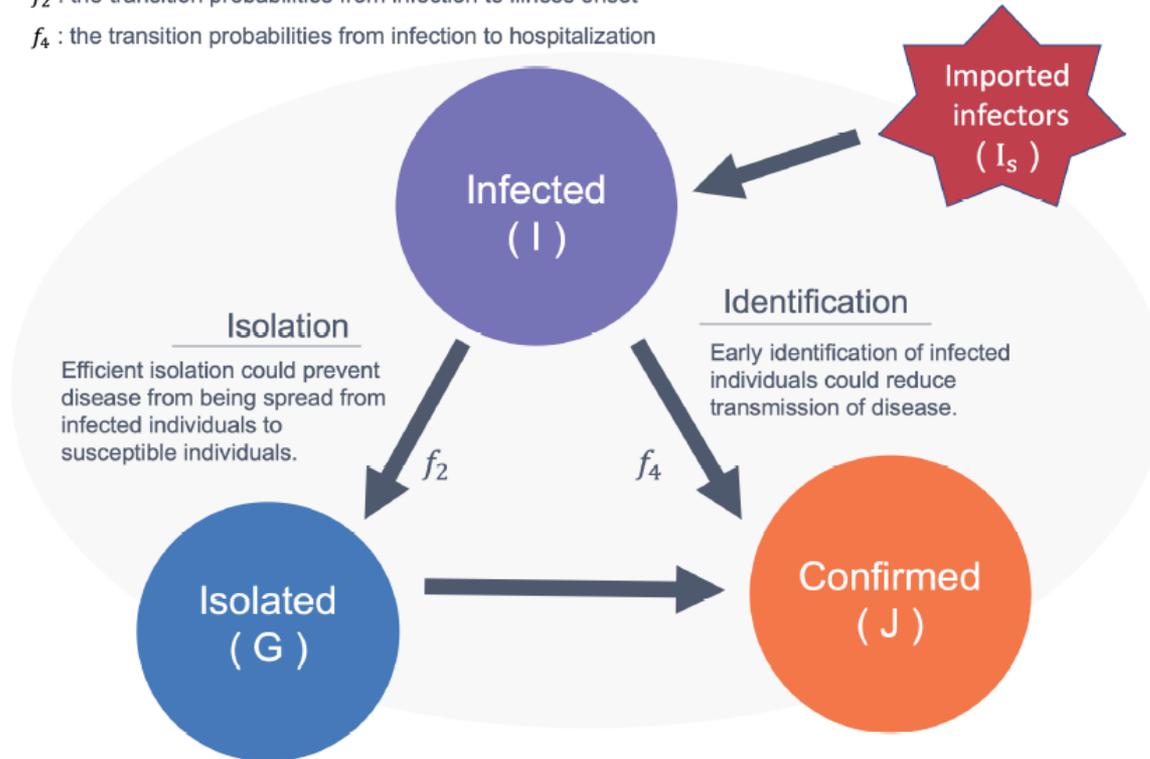


Figure 1: Sketch map of the model.

# The Source of Infection

- In the modeling, the most important thing is to consider the source of infection

$$I_0(t) = I(t) - J(t) - G(t)$$

- We focus on the source of infection and establish the dynamical model

# The infected cases $I(t)$

- The changes of the infected cases is proportional to the source of infection

$$I(t + \Delta t) - I(t) = \beta I_0(t) \Delta t$$



$$\frac{dI}{dt} = \beta I_0(t)$$

- $\beta$  defined as spread rate
  - $\beta$ , is defined by the average amount of people becoming infected by this person per unit time

# The diagnosed cases $J(t)$

- The changes of the diagnosed cases is proportional to the source of infected cases in the latent period

$$J(t + \Delta t) - J(t) = \left\{ \gamma \int_0^{\tau_1} h_1(t') \beta(I(t - t') - J(t - t') - G(t - t')) dt' \right\} \Delta t$$

  $\frac{dJ}{dt} = \gamma \int_0^{\tau_1} h_1(t') \beta(I(t - t') - J(t - t') - G(t - t')) dt'$

- The coefficient  $\gamma$  is the morbidity
- *The kernel  $h_1$  is a distribution*

# The isolated cases $G(t)$

- The changes of the isolated cases depends on two factors:
  - The source of infection, where  $\ell$  is the rate of isolation;
  - The infected cases, which have been isolated. The term with delay represents the effect

$$G(t+\Delta t)-G(t) = \left\{ l(I(t) - J(t) - G(t)) - l \int_0^{\tau_1'} h_2(t')(I(t-t') - J(t-t') - G(t-t'))dt' \right\} \Delta t$$

  $\frac{dG}{dt} = l(I(t) - J(t) - G(t)) - l \int_0^{\tau_1'} h_2(t')(I(t-t') - J(t-t') - G(t-t'))dt'$

# The cured cases $R(t)$

- Once infected, it needs to go through the incubation period of  $\tau_1$  day and the treatment period of  $\tau_2$  days to end the treatment

$$R(t + \Delta t) - R(t) = \left\{ \kappa \int_0^{\tau_1 + \tau_2} h_3(t') \beta (I(t - t') - J(t - t') - G(t - t')) dt' \right\} \Delta t$$



$$\frac{dR}{dt} = \kappa \int_0^{\tau_1 + \tau_2} h_3(t') \beta (I(t - t') - J(t - t') - G(t - t')) dt'.$$

# The kernel functions $h_i(t)$ , $i = 1, 2, 3$

- The kernel functions are the probability distributions of delay days with the compact supports and satisfy:

$$\int_0^{\tau_1} h_1(t') dt' = 1$$

$$\int_0^{\tau_1 + \tau_2} h_3(t') dt' = 1$$

$$\int_0^{\tau_1'} h_2(t') dt' < 1$$

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We can take :  $\delta$  function or uniform distribution or cut-off Gauss-like distribution..

$$h_i(t) = c_i e^{-d_i t^2}$$

# TDD-NCP Model

- Time Delay Dynamic System

- $I(t)$ : cumulative infected people at time  $t$ ;
- $J(t)$ : cumulative confirmed people at time  $t$ ;
- $G(t)$ : currently isolated people who are infected but still in latent period at time  $t$ ;
- $R(t)$ : cumulative cured people at time  $t$ .

$$\frac{dI}{dt} = \beta(I(t) - J(t) - G(t)),$$

$$\frac{dJ}{dt} = \gamma\beta \int_0^{\tau_1} h_1(t')(I(t-t') - J(t-t') - G(t-t'))dt',$$

$$\frac{dG}{dt} = l(I(t) - J(t) - G(t)) - l \int_0^{\tau_1'} h_2(t')(I(t-t') - J(t-t') - G(t-t'))dt',$$

$$\frac{dR}{dt} = \kappa \int_0^{\tau_1 + \tau_2} h_3(t')\beta(I(t-t') - J(t-t') - G(t-t'))dt'$$

# TDD-NCP Model (version 1)

- First version

$$\frac{dI}{dt} = \beta(I(t) - J(t) - G(t)),$$

$$\frac{dJ}{dt} = \gamma \int_0^t h_1(t - \tau_1, t') \beta(I(t') - J(t') - G(t')) dt',$$

$$\frac{dG}{dt} = \ell(I(t) - J(t) - G(t)) - \int_0^t h_2(t - \tau'_1, t') G(t') dt',$$

$$\frac{dR}{dt} = \kappa \int_0^t h_3(t - \tau_1 - \tau_2, t') \beta(I(t') - J(t') - G(t')) dt'.$$

# Features of time delay model

- Linear equations
  - With integral terms (nonlocal terms)
- The meaning of parameters are clear

# Shortage of the linear model

- If the spread of the virus last for a little bit long time, the prediction results are not so reasonable.
- 
- We are considering two improvements:
    - Locally, use the linear model and try to do the prediction in a short time.
    - Add some nonlinear terms in our models.

# Some Remarks on Data

- It should be noted that :
  - In the public data by the government, only the information about the diagnosed cases and the cured cases are provided.
  - $I(t)$  and  $G(t)$  usually can not be obtained directly.

# Sparsity assumptions

- Due to that the public data can only provide limited information, we have to assume that:
  - The parameters are constants or piecewise constants.

# Forward Problems

- Given  $\{\beta, l, \gamma, \kappa, \tau_1, \tau_1', \tau_2\}$
- Find the solution of

$$\frac{dI}{dt} = \beta(I(t) - J(t) - G(t)),$$

$$\frac{dJ}{dt} = \gamma\beta \int_0^{\tau_1} h_1(t')(I(t-t') - J(t-t') - G(t-t'))dt',$$

$$\frac{dG}{dt} = l(I(t) - J(t) - G(t)) - l \int_0^{\tau_1'} h_2(t')(I(t-t') - J(t-t') - G(t-t'))dt',$$

$$\frac{dR}{dt} = \kappa \int_0^{\tau_1 + \tau_2} h_3(t')\beta(I(t-t') - J(t-t') - G(t-t'))dt'$$

$$I(t) = I_0(t), \quad t \leq 0$$

$$J(t) = J_0(t), \quad t \leq 0$$

$$G(t_0) = G_0(t) \quad t \leq 0$$

$$R(t) = R_0(t), \quad t \leq 0.$$

# Theoretical Analysis

- We have the equation of source infection  $I_0(t)$ .

$$\frac{dI_0}{dt}(t) = \beta \left( I_0(t) - \int_0^{\tau_1} h_1(t') I_0(t-t') dt' \right) - l \left( I_0(t) - \int_0^{\tau_2} h_2(t') I_0(t-t') dt' \right)$$

$$I_0(t) = \tilde{g}(t), \quad t \in (-\tau_1 - \tau_2, 0]$$

# Dynamical Analysis for the equation of source of infection

- Linear time delay dynamical systems

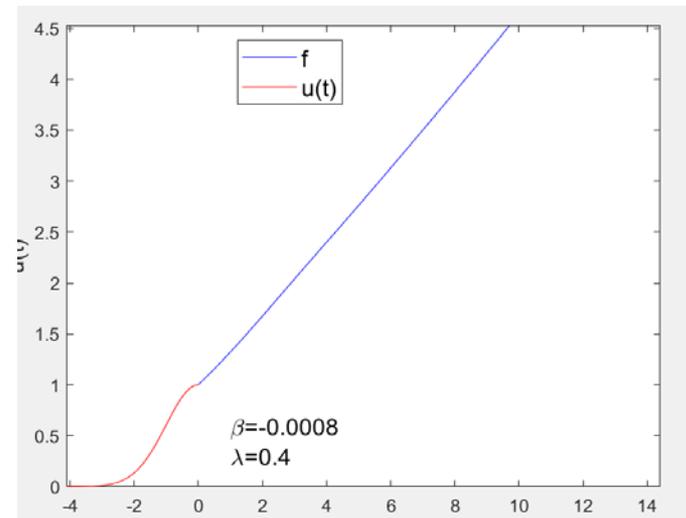
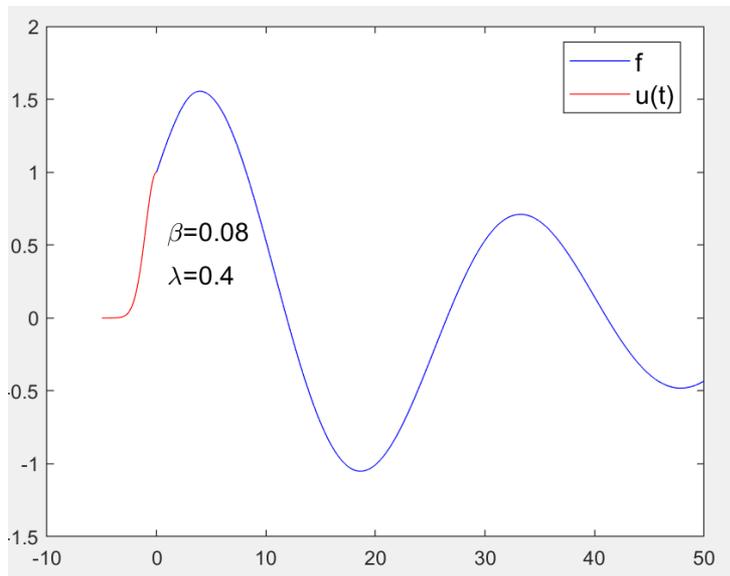
$$\frac{dI_0}{dt}(t) = (\beta - l)I_0(t) + \int_0^{\tau_1 + \tau_2} H(t')I_0(t - t')dt'$$

$$H(t) = h_1(t) - h_2(t)$$

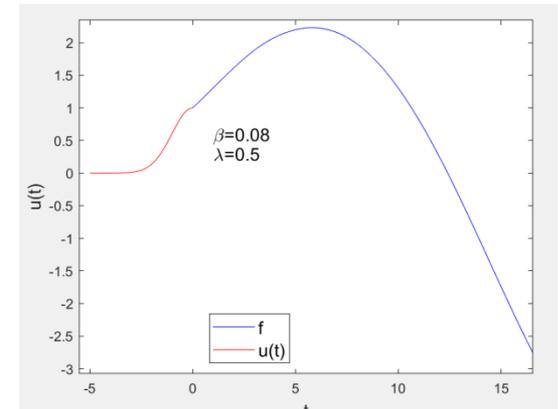
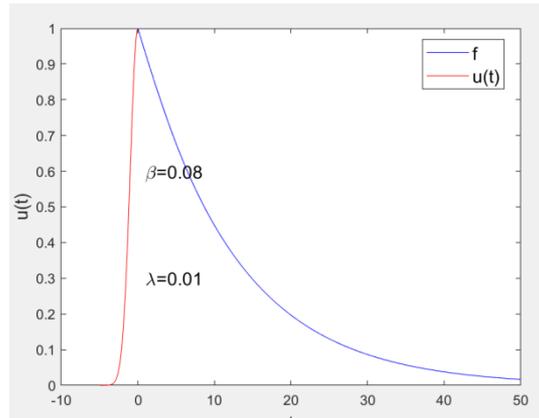
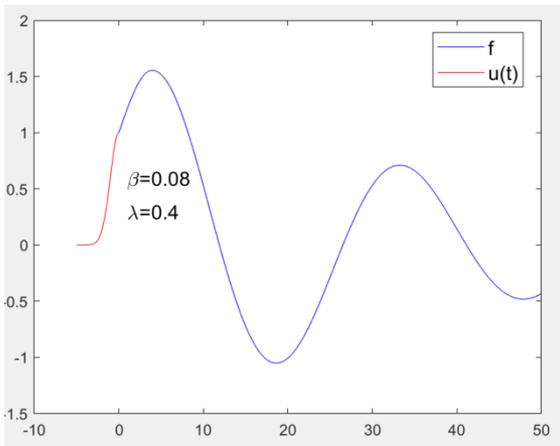
# Simplified System

- Simplified system

$$\frac{du}{dt} = -\beta u + \lambda \frac{1}{T} \int_{t-T}^t (u(t) - u(t')) dt'$$
$$u(s) = f(s), \quad x \in (-T, 0)$$



# Changes with respect to $\lambda$



# Conditions for the validity of the model

- The models make sense

$$\begin{aligned} \frac{dI}{dt} &\geq 0, \\ \frac{dJ}{dt} &\geq 0, \\ \frac{dR}{dt} &\geq 0. \end{aligned}$$

Conditions for  
validity of model



Conditions for  
Stopping



The first  $T$ ,  
such that  
 $I_0(T) = 0$  and  
 $I_0(t) > 0$  for  $t$   
 $< T$

- The conditions for the validity of equation
- $T$  the stopping time of epidemic situation

$$I_0(t) > 0, \quad t \in (0, T)$$

$$I_0(T) = 0$$

# Inverse Problems and Prediction

- From the public data, determine the parameters in the model
  - Reconstruct the spread rate  $\beta$ , isolated rate  $l$
- By the parameters, which are reconstructed from the public data, we can do the numerical simulation and predict the development of the epidemic situation.
  - Especially, “stopping time  $T$ ” of the epidemics.

# Inversion Scheme

- By the cumulative confirmed cases  $J_{\text{obs.}}$  and the cumulative cured cases  $R_{\text{obs.}}$

reconstruct  $\theta = (\beta, \kappa)$  :

$$\min_{\theta} \|J(\theta; t) - J_{\text{obs.}}\|_2$$

$$\min_{\theta, \kappa} \|R(\theta, \kappa; t) - R_{\text{obs.}}\|_2.$$

# Questions

- Whether the mathematical model can well describe the spread of the novel corona-virus?
- Well-posedness of the model?
- Uniqueness of the reconstruction of the parameters ?
- The prediction results are “trustable” ?

# Linear integral differential equations

- Existence and Uniqueness (Nonlocal)

$$\frac{dI}{dt} = \beta(I(t) - J(t) - G(t)),$$

$$\frac{dJ}{dt} = \gamma\beta \int_0^{\tau_1} h_1(t')(I(t-t') - J(t-t') - G(t-t'))dt',$$

$$\frac{dG}{dt} = l(I(t) - J(t) - G(t)) - l \int_0^{\tau_1'} h_2(t')(I(t-t') - J(t-t') - G(t-t'))dt',$$

$$\frac{dR}{dt} = \kappa \int_0^{\tau_1 + \tau_2} h_3(t')\beta(I(t-t') - J(t-t') - G(t-t'))dt'$$

$$I(t) = I_0(t), \quad J(t) = J_0(t), \quad G(t) = G_0(t), \quad R(t) = R_0(t), \quad t \leq 0.$$

# Formulation of the inverse problems

- For TDD-NCP model

$$\begin{aligned}\frac{dI}{dt} &= \beta(I(t) - J(t) - G(t)), \\ \frac{dJ}{dt} &= \gamma\beta \int_0^{\tau_1} h_1(t')(I(t-t') - J(t-t') - G(t-t'))dt', \\ \frac{dG}{dt} &= l(I(t) - J(t) - G(t)) - l \int_0^{\tau_1'} h_2(t')(I(t-t') - J(t-t') - G(t-t'))dt', \\ \frac{dR}{dt} &= \kappa \int_0^{\tau_1+\tau_2} h_3(t')\beta(I(t-t') - J(t-t') - G(t-t'))dt'\end{aligned}$$

$$I(t) = I_0(t), \quad J(t) = J_0(t), \quad G(t) = G_0(t), \quad t \leq 0.$$

- Given other data, we reconstruct  $\beta, l$  from the measurements of  $J(t)$

# Inverse Problems for the time delay systems

- Inverse coefficient problem (parameters  $\beta, l$ )
  - Uniqueness
  - Stability (with respect to the errors)
  - Algorithms

# Based on $\beta, l$ , Predict the development

- Solve

$$\frac{dI_0}{dt}(t) = (\beta - l)I_0(t) + \int_0^{\tau_1 + \tau_2} H(t')I_0(t - t')dt'$$

$$I_0(t) = g(t), \quad t \in (-\tau_1 - \tau_2, 0]$$

$H(t) = h_1(t) - h_2(t)$



- Get the first stopping time  $T$ , which satisfies:

$$I_0(t) > 0, \quad t < T, \quad \text{and} \quad I_0(T) = 0$$

If such  $T$  does not exist, it means the epidemic will not stop until all people are infected.

# Inverse Problems for time delay dynamical system

- Inverse Source Problems (Find the Virus super Disseminator from Data)
  - Uniqueness
  - Algorithms

# Case Study I

- By the public data from Jan. 23<sup>rd</sup> to Feb. 1<sup>st</sup>, we reconstruct spread rate  $\beta$  and isolate rate  $l$ :

中国科学：数学 2020年 第50卷 第3期：1~8

SCIENTIA SINICA Mathematica

论 文



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## 基于一类时滞动力学系统对新型冠状病毒肺炎疫情的建模和预测

严阅<sup>1</sup>, 陈瑜<sup>1</sup>, 刘可伋<sup>1,2</sup>, 罗心悦<sup>1</sup>, 许伯熹<sup>1</sup>, 江渝<sup>1</sup>, 程晋<sup>3\*</sup>

# Reduce parameters (Sparse assumption)

- Based on public statistics

表 1 参数值

| $\gamma$ | $\kappa$ | $\tau_1$ | $\tau'_1$ | $\tau_2$ |
|----------|----------|----------|-----------|----------|
| 0.99     | 0.97     | 7        | 4         | 12       |

# Spread rate and Isolate rate

- Different areas

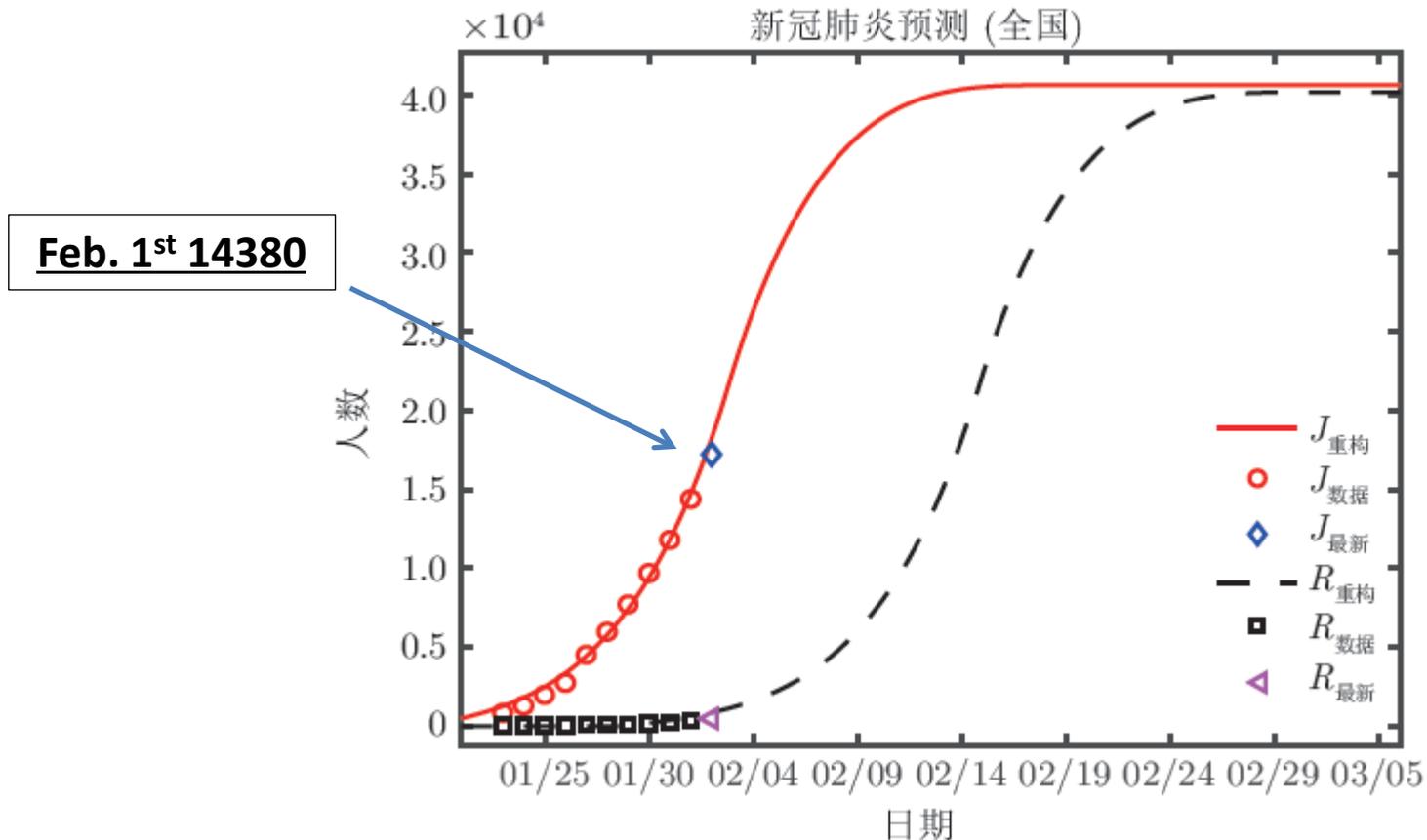
表 2 参数的估计值

| 地区                   | 传染率 $\beta$ | 隔离率 $\ell$ |
|----------------------|-------------|------------|
| <b>Whole country</b> | 0.2320      | 0.4202     |
| Wuhan                | 0.1957      | 0.5500     |
| Shanghai             | 0.2113      | 0.5500     |
| Jiangsu              | 0.2581      | 0.5500     |

# Prediction (Whole country)

- Whole country

March 13th **81004**

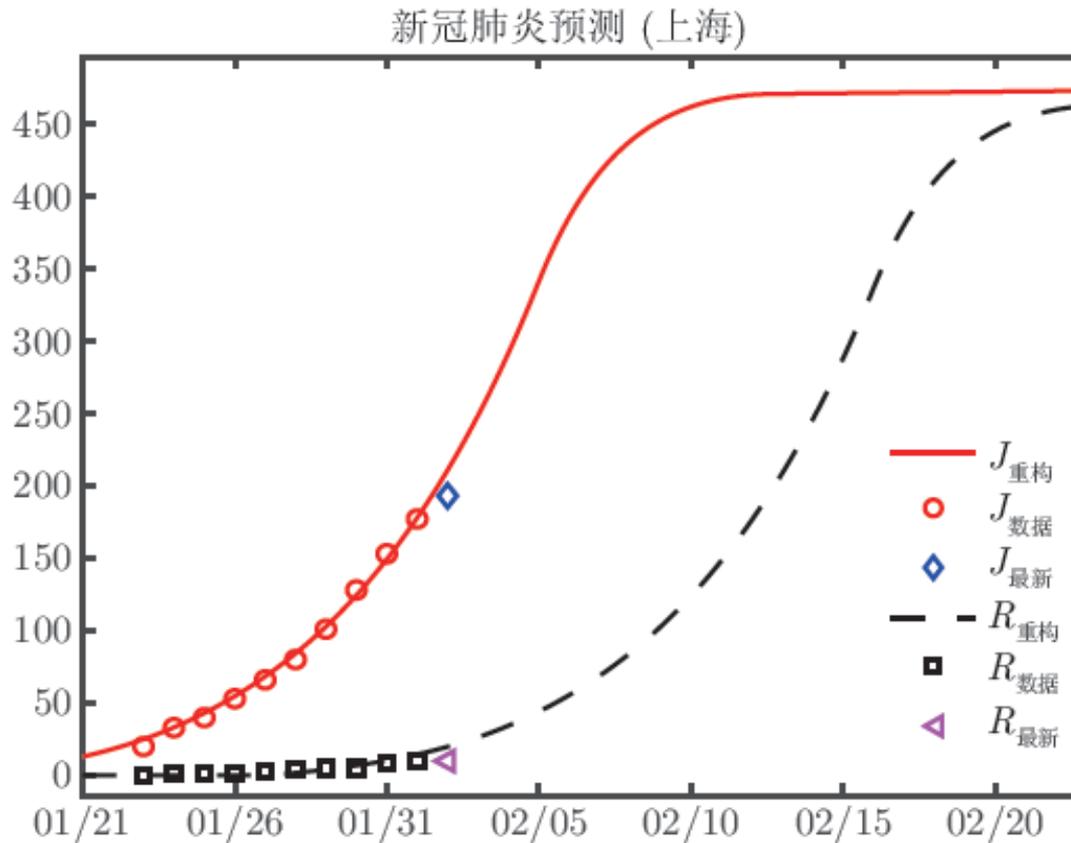


# Prediction (Shanghai)

- Shanghai

Strengthening of control measures !

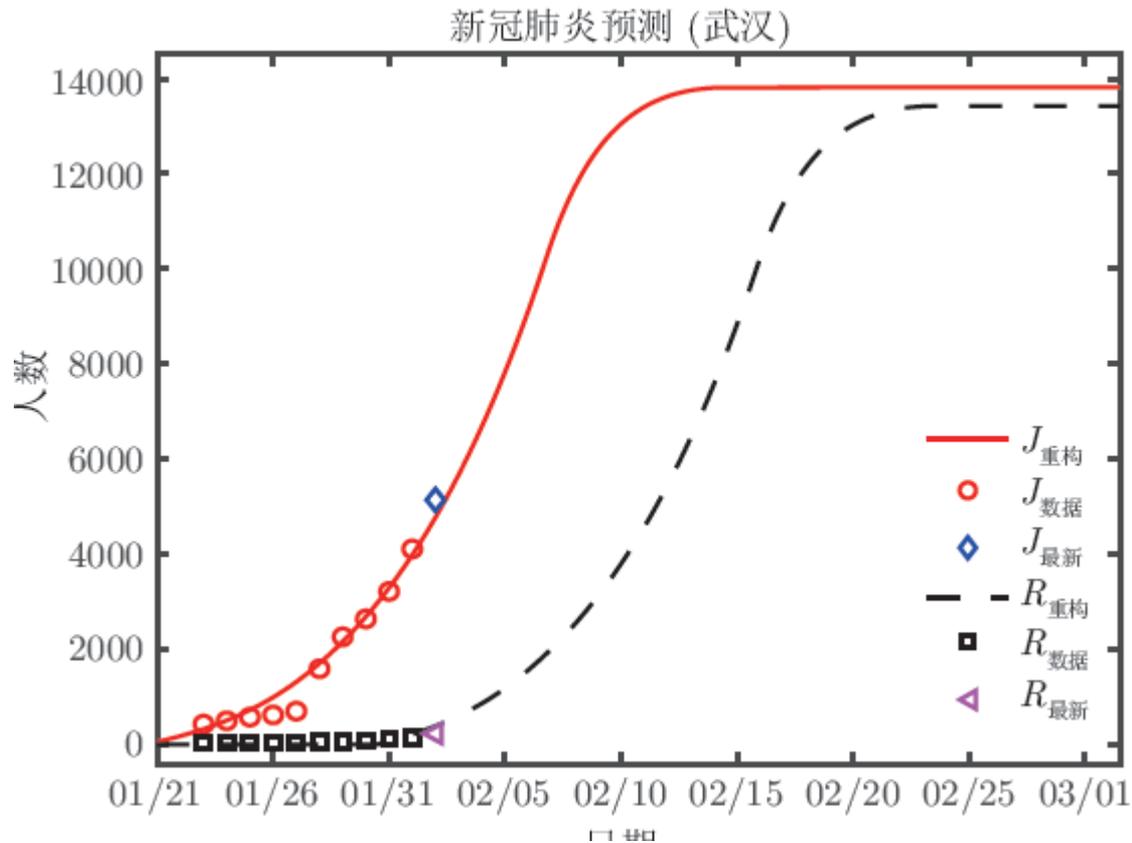
March 13<sup>th</sup> 346



# Prediction (Wuhan)

- Wuhan

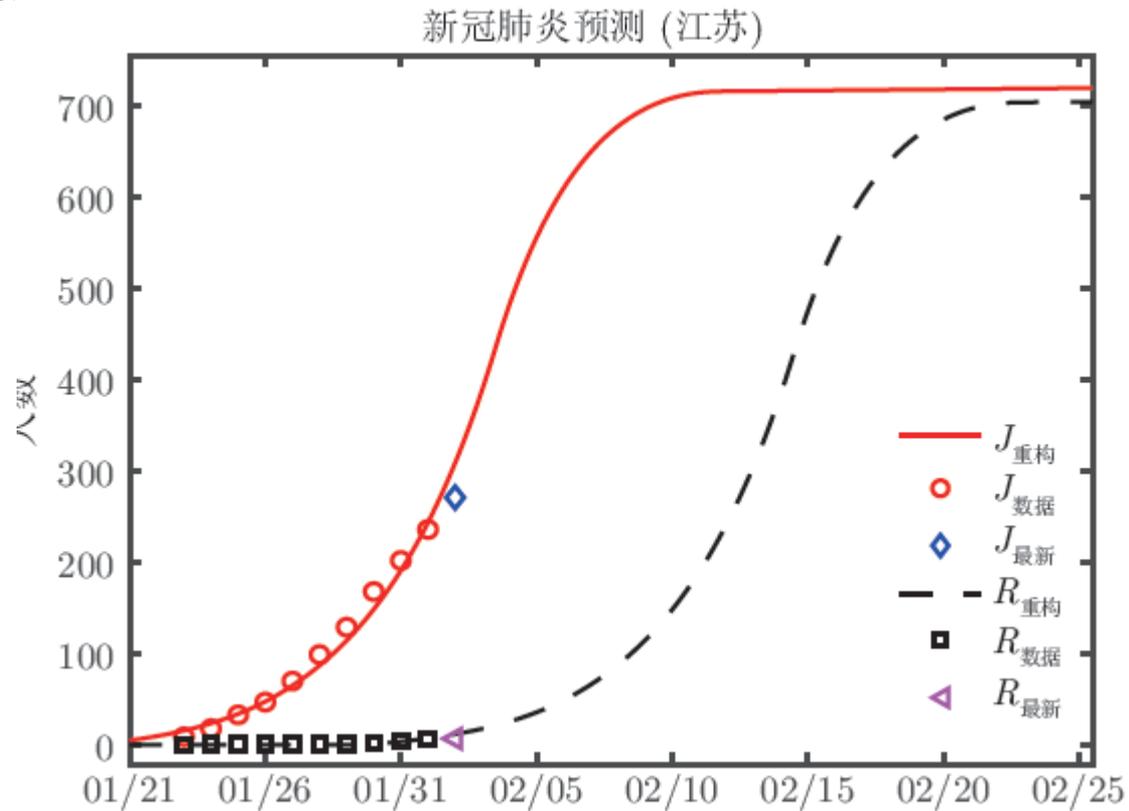
March 13<sup>th</sup> **49991**



# Prediction (Jiangsu)

- Jiangsu

March 13<sup>th</sup> 631



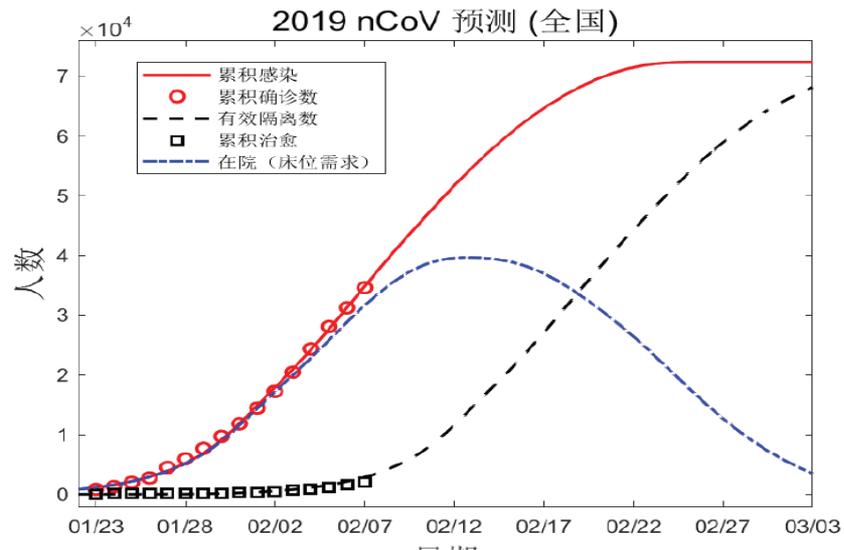
There is a certain gap between the prediction result and the actual facts

- The model is not good?
- The method is not correct?
- Data is not correct?

# Research Report

- Prediction for China

基于时滞动力学系统新冠肺炎传播模型的若干预测分析



# Lab. Report

- Feb. 9<sup>th</sup>, 2020

上海市现代应用数学重点实验室研究报告 Research Report Series of SKLCAM (2020 年第一期)

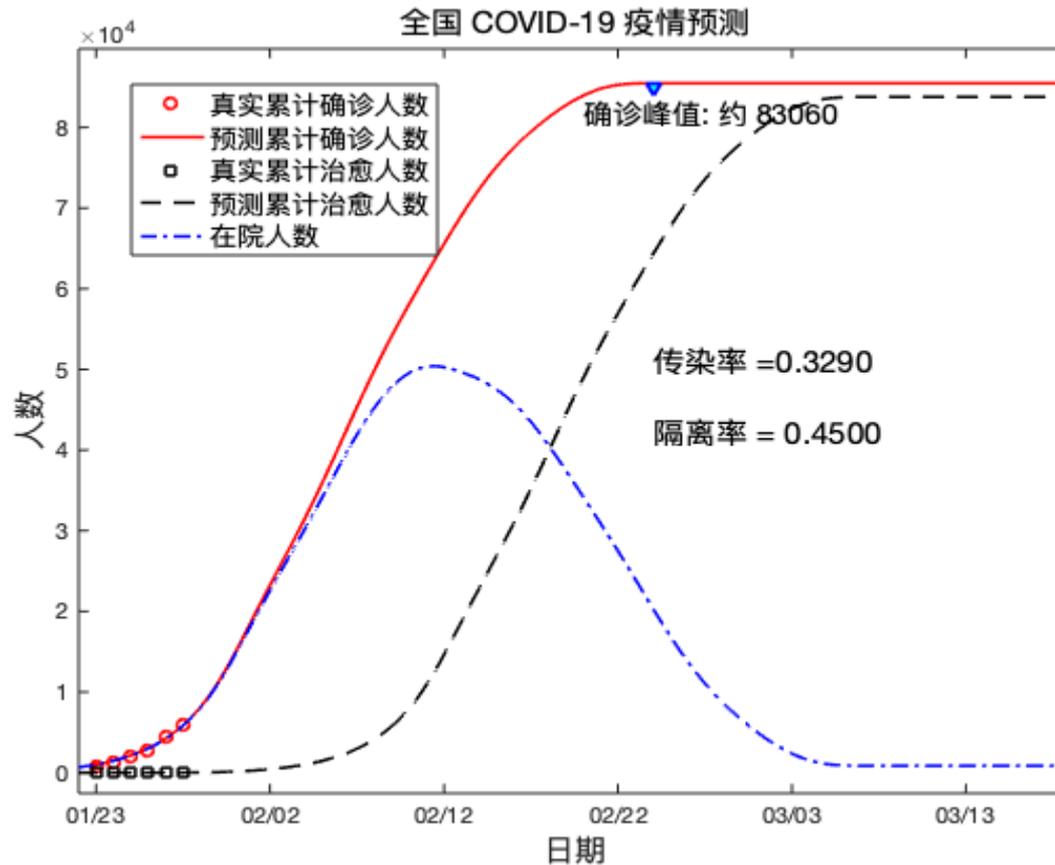
发布时间：2020-02-09 阅读次数：421



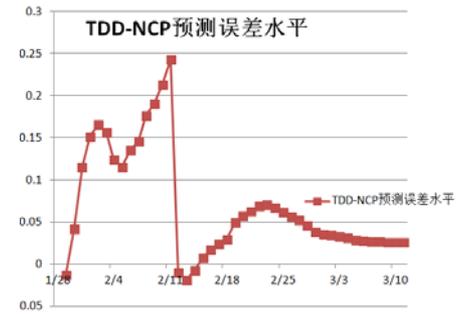
上海市现代应用数学重点实验室研究报告  
Research Report Series of SKLCAM

# Choose suitable Data

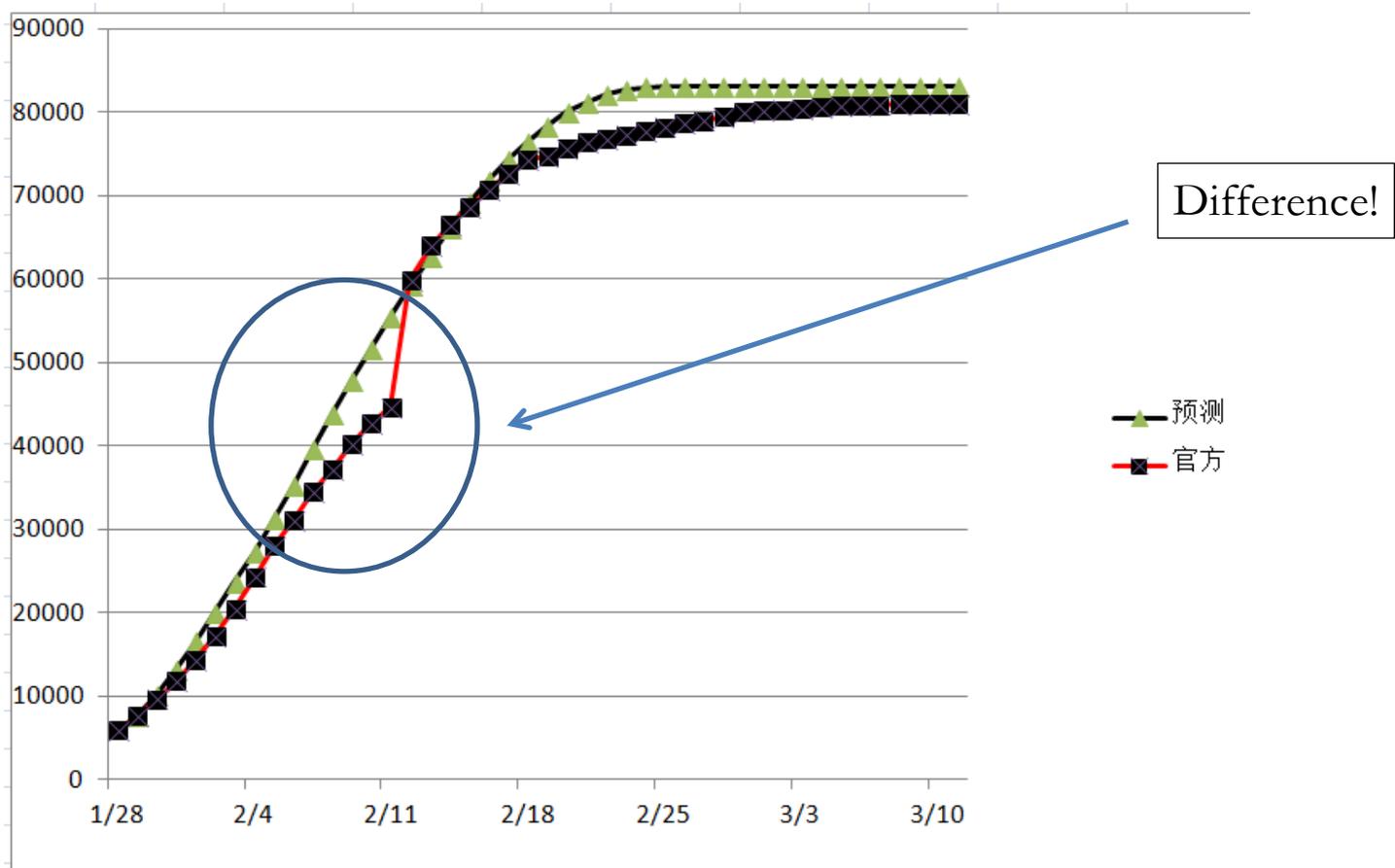
- Data from Jan. 23<sup>rd</sup> to Jan. 28<sup>th</sup>, 2020



# Prediction



- Public Data and our simulation Results

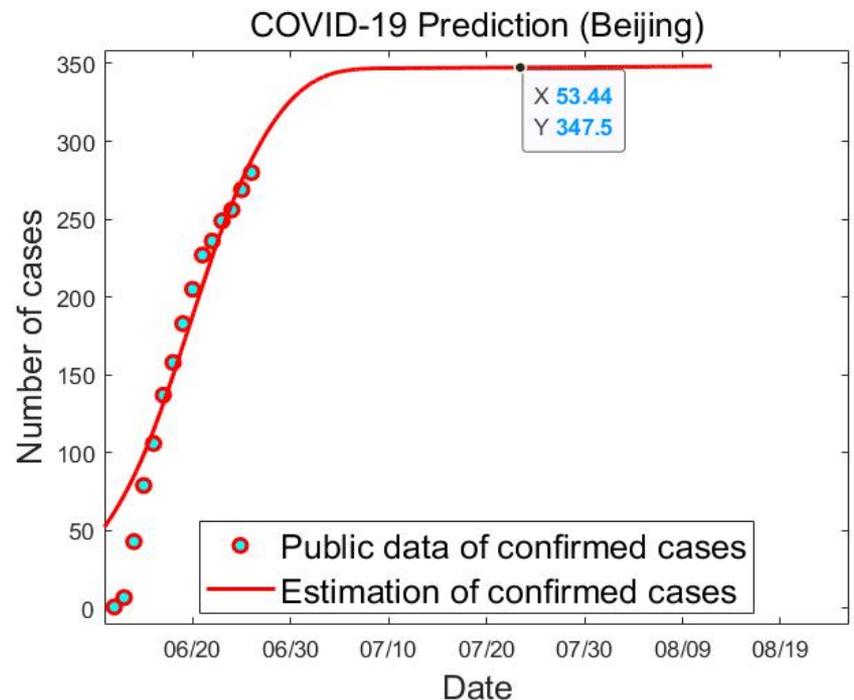


# Problems in simulation

- Credibility of data ?
  - Data at the early stage. Data in US...
- The reconstruction results heavily depend on the data we choose!
  - Some data is distorted.
- Prediction: Whole Country > Hubei > Other cities
  - Model in the sense of statistical average

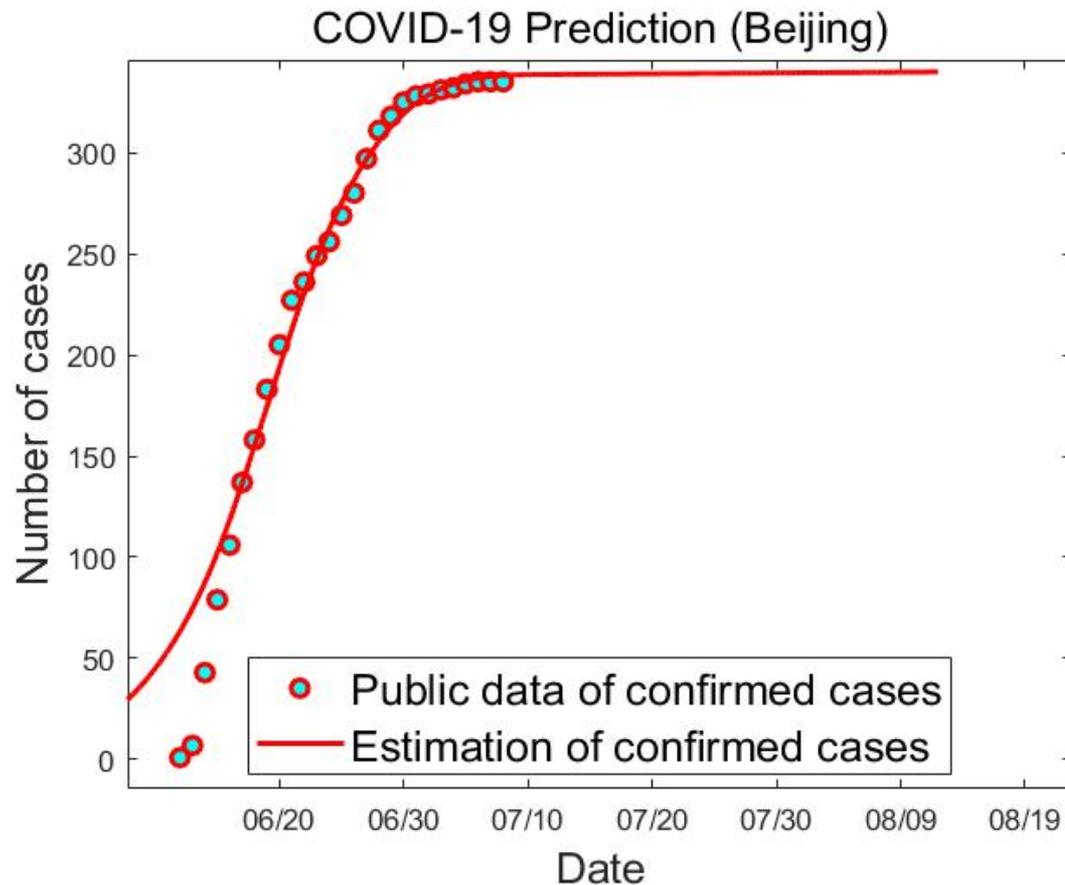
# Case Study II

- Beijing,
  - June 11–14,
  - Confirmed Cases 79
  - $\beta=0.20$
  - $I=0.354$
  
  - Prediction 350



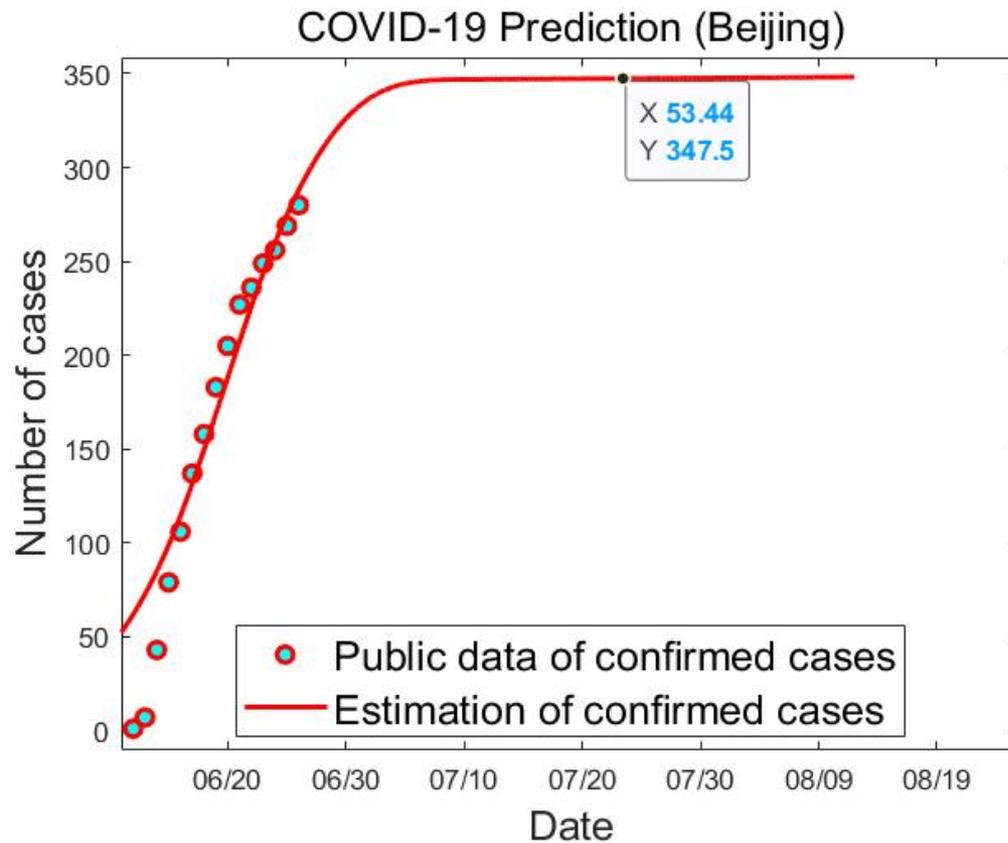
# Review

- By all public data



# Prediction

- By partial data



# Fudan-CCDC Model (Wenbin CHEN)

- Continuous case

$$\frac{dI}{dt} = \beta I_0(t),$$

$$\frac{dJ}{dt} = \beta \int_{-\infty}^t f_4(t-s) I_0(s) ds,$$

$$\frac{dG}{dt} = \ell I_0(t) - \ell \int_{-\infty}^t f_4(t-s) I_0(s) ds.$$

- Discrete case

$$I(t+1) = I(t) + \beta I_0(t),$$

$$J(t+1) = J(t) + \beta \sum_{s \leq t} f_4(t-s) I_0(s),$$

$$G(t+1) = G(t) + \ell I_0(t) - \ell \sum_{s < t} f_4(t-s) I_0(s).$$

# Choose the kernels

- Choose the kernels based on Data from CCDC

$$f_2(t) = \frac{0.5977}{t} e^{-1.105(\ln(t)-1.417)^2}$$

$$f_3(t) = 0.005559t^{1.641} e^{-0.002105t^{2.641}}$$

$$f_4(t) = f_2 * f_3(t) = 0.06244e^{-\left(\frac{t-10.87}{5.378}\right)^2} + 0.03322e^{-\left(\frac{t-15.97}{23.8}\right)^2}.$$

# Ongoing works

- Improving the time delay models
- Tracking epidemic data across countries
- Study the uniqueness and stability of the inverse problems

# Research Results from our team

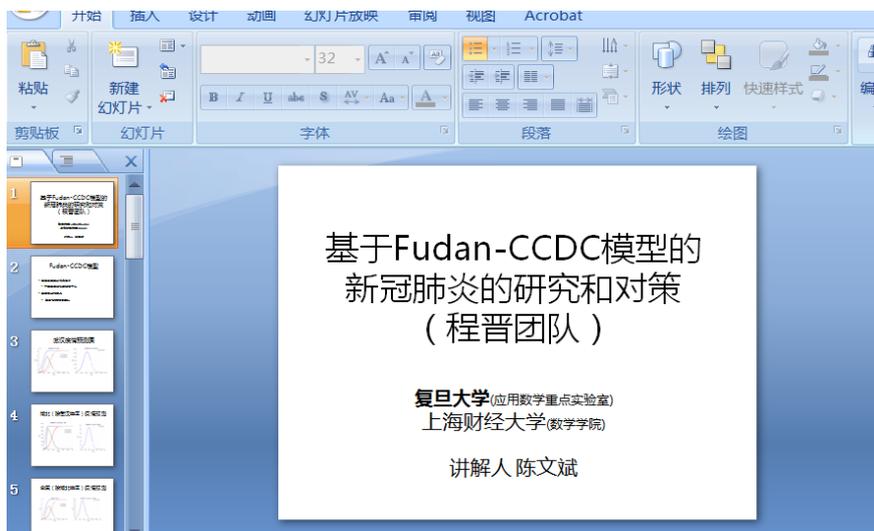
- **Internal Reports :**
  - Cheng Jin, Chen Wenbin, Report to Shanghai Government on Feb. 16<sup>th</sup>, 2020.
  - Report of the novel coronavirus disease (COVID-19) epidemic situation, with Wuhan Health Center Information Center, Wei Ning health Artificial Intelligence Laboratory, School of Mathematical Sciences, Fudan University

# Presentation on Feb. 16<sup>th</sup>, 2020

- Presentation in Shanghai

## 一些政策建议

- 阶梯状逐步放开流动性，流动性的影响可能会延迟出现
- 密切跟踪数据，对隔离出站人员持续定期跟踪（14天）。
- 严防高危地区和人员的追踪，发现病例，马上在一定范围内隔离排查，实行打地鼠挖洞策略
- 建议调整医保慢性病人的取药政策，减少就诊次数
- 建议对幼儿园区、高校等重点区域依然保持严密防守
- 建议错峰出行
- 建议对海外回国人员进行密切关注，防止海外二次输入



One of our suggestions is to pay attention to overseas personnel to prevent secondary input from overseas

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# Meaning of Models

- In strict sense, all models are "wrong"
- The models are meaningful
  - Give the rough estimates and prediction
  - See "the light at the end of the tunnel"

Thanks!

Welcome the comments!

All world unite to fight the epidemic