

Reaction-diffusion spatial modeling of COVID-19 in Chicago

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- ① The COVID Problem
- ② Project Context and Objectives
- ③ Current Work
- ④ Future Goals



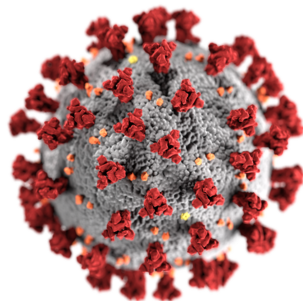
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Identified in Wuhan, China in December 2019.
Caused by the severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2).

In the United States alone, we currently have

- **37,768,911** total cases
- **626,833** total deaths



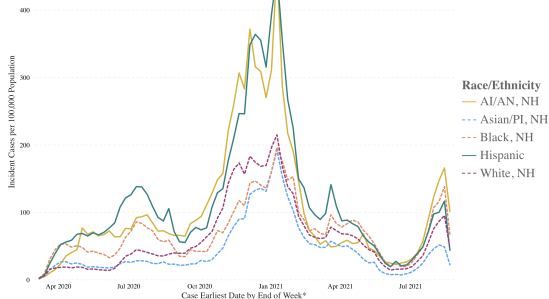


Transmission by exposure to infectious respiratory fluids:

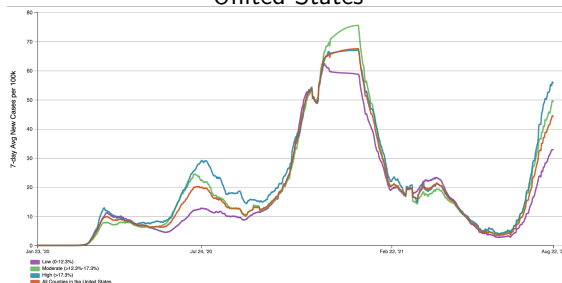
1. Inhalation of virus
2. Deposition of virus on exposed mucus membranes
3. Touching mucous membranes with soiled hands contaminated with virus



COVID-19 weekly cases per 100,000 population by race/ethnicity, United States



COVID-19 weekly case rate per 100,000 population by percentage of county population in poverty, United States





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Why make a mathematical study of COVID-19?

An epidemic model provides . . .

- a convenient summary of the data
- insight into the underlying processes of the disease spread
- a testing ground for assessing control procedures



Models of infectious diseases are usually variations on the **Kermack-McKendrick model** (1927).



Usual assumptions:

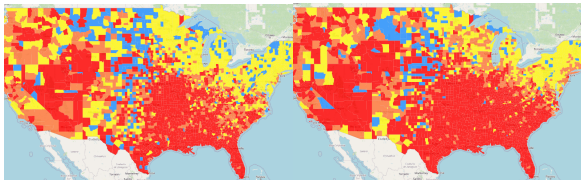
- Population is *homogeneous*
- Transmission is spatially independent

Project Context and Objectives



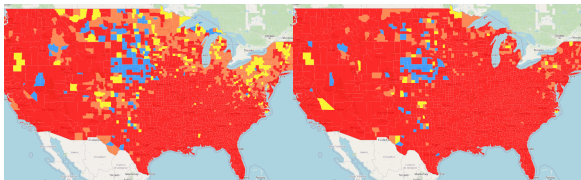
These assumptions don't match reality!

Level of community transmission by county



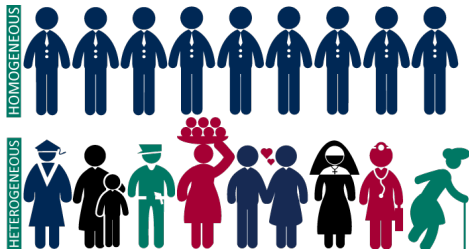
(a) 7/22/2021

(b) 7/29/2021



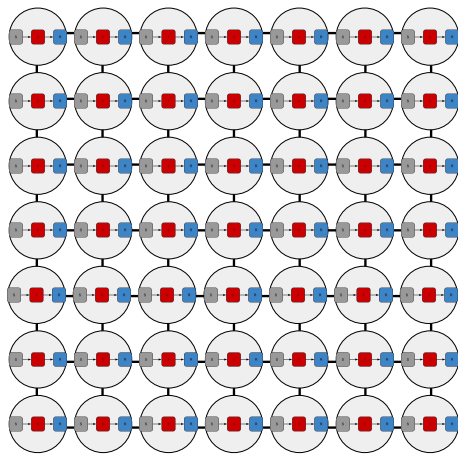
(c) 8/8/2021

(d) 8/21/2021



Idea: Use reaction-diffusion to build a spatially explicit model.

- The data clearly shows a diffusive pattern
- Spatial dependence can approximate demographic differences
- Spatially dependent data exists for many scales and regions



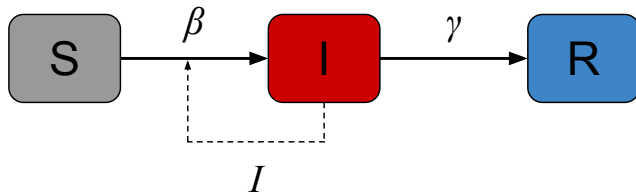


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Original motivation from Summer 2020 IPRO with Prof. Robert Ellis and Mr. David Eads (NPR).
Expanded over Summer 2021 with Prof. Chun Liu and Prof. Yiwei Wang.

Start by defining the zero-dimension (ODE) model:

1. Start with a population of *susceptibles* (S)
2. Some of the population may become *infected* (I) upon emergence of the virus
3. Infected individuals interact with susceptibles at rate β to draw new members into I
4. A fraction of I recover or die at a rate γ moving to the *removed* (R) population



This represents processes that occur in a “well mixed” situation.



The population model at the PDE level is an autonomous diffusion with a source:

$$\begin{aligned}S_t &= \nabla(\mathfrak{D}_S \nabla S) - \beta SI, \\I_t &= \nabla(\mathfrak{D}_I \nabla I) + \beta SI - \gamma I, \\R_t &= \gamma I.\end{aligned}$$

Assumptions:

- The R population has immunity.
- Diffusion performs all relevant spreading.
- Arriving infected individuals form local hotspots.
- Disease progression is the only transition mechanism.



Next we identify the parameters β and γ at the ODE level:

Idea: Minimize the distance between the model time series and the observed time series.

$$\mathcal{N} = \sum_i^{t_{\text{fit}}^{\text{end}}} \left(|C_{\text{num}}(t_i) - C_{\text{obs}}(t_i)|^2 + |D_{\text{num}}(t_i) - D_{\text{obs}}(t_i)|^2 \right)$$

where the index i identifies a point in the time series.

Optimize the parameters to reproduce:

- the reported *total* number of infected cases $C(t) = I(t) + R(t)$
- the *total* number of deceased $D(t) = R(t)$

ODE Results

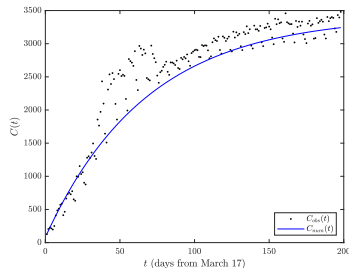
Data were obtained from the Chicago Data Portal.

Optimal ODE model parameters obtained from 1,000 optimizations (implemented in MATLAB).

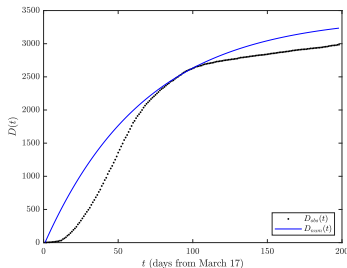
		Median (interquartile range)	Initial value
Population	N	2,695,598	
Initial population	(I_0, R_0)	(127, 2)	
Transmission rate, $S \rightarrow I$ [per day]	β	0.38206(0.38204-0.38209)	$c \in U[0, 1]$
Transition rate, $I \rightarrow R$ [per day]	γ	0.39656(0.39654-0.39659)	$c \in U[0.25, 0.75]$

Note the transmission rate β must be divided by N when used in the ODE model.

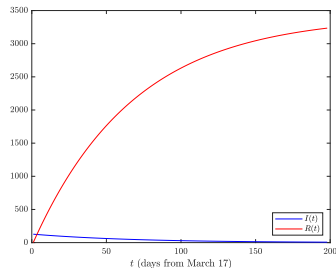
ODE model for Chicago with fitting to official data from March 17, 2020 ($t = t_{\text{init}} = 1$) to September 30, 2020 ($t = t_{\text{fit}}^{\text{end}} = 198$)



(e) Confirmed cases $C(t) = I(t) + R(t)$



(f) Number of deaths $D(t)$



(g) Infected $I(t)$ and Recovered $R(t)$



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Observations on ODE results

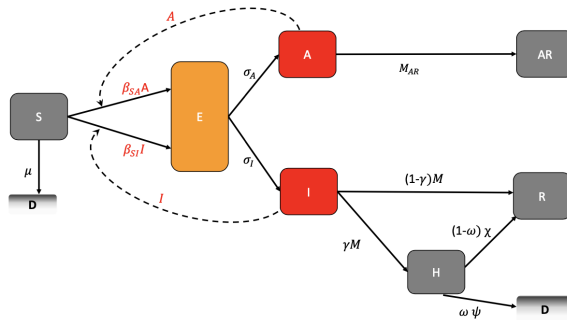
- The model reproduces reasonably well the number of cases.
- The model fails to capture the characteristic S-shape in the number of deaths.
- Time-dependent control measures are not accounted for.



Transition to PDE model (proof of concept)

- The transition rate γ is an “individual” parameter, so the value is preserved at the PDE level.
- The transmission rate β depends on the interaction between individuals, so this must be modified.
- The geometry (shape and density distribution) of Chicago must be defined (Census data).
- Suitable initial infected distributions must be seeded.

Improve the model



Source: Kevrekidis et al.

Redefine the equations and perform the same calculations.