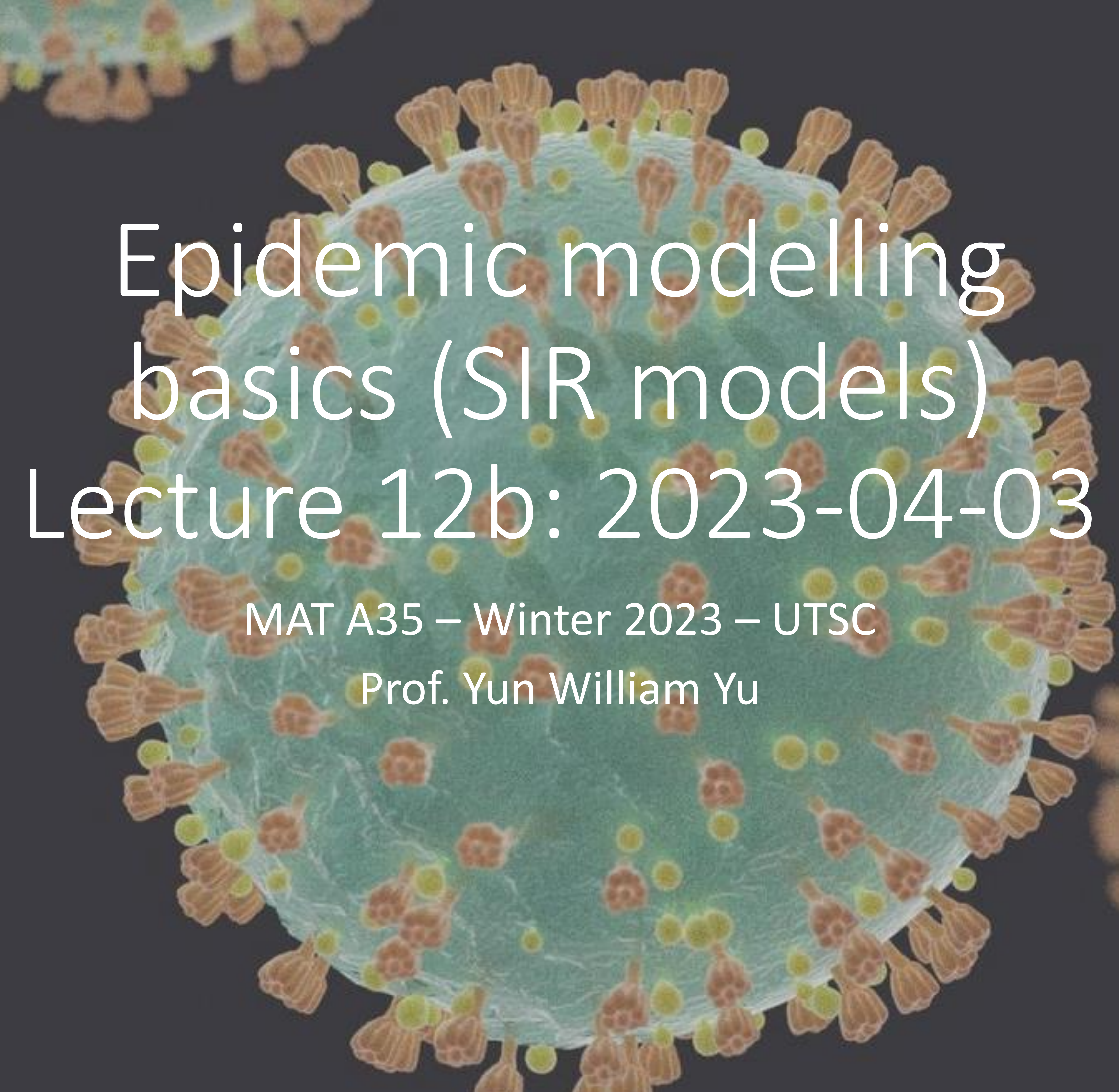


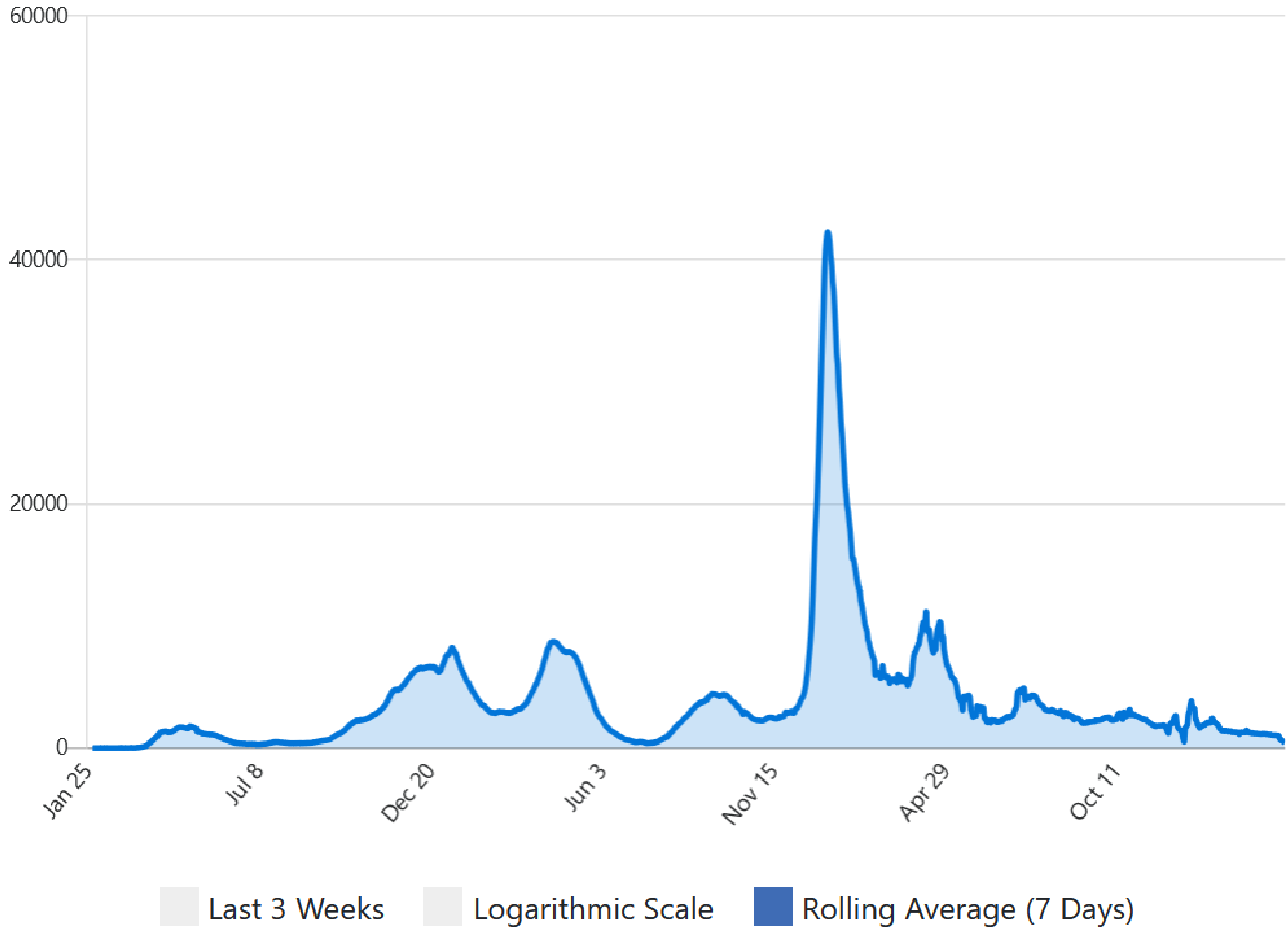
Epidemic modelling basics (SIR models) Lecture 12b: 2023-04-03

MAT A35 – Winter 2023 – UTSC

Prof. Yun William Yu

Image credit: https://commons.wikimedia.org/wiki/File:Coronavirus_SARS-CoV-2.jpg

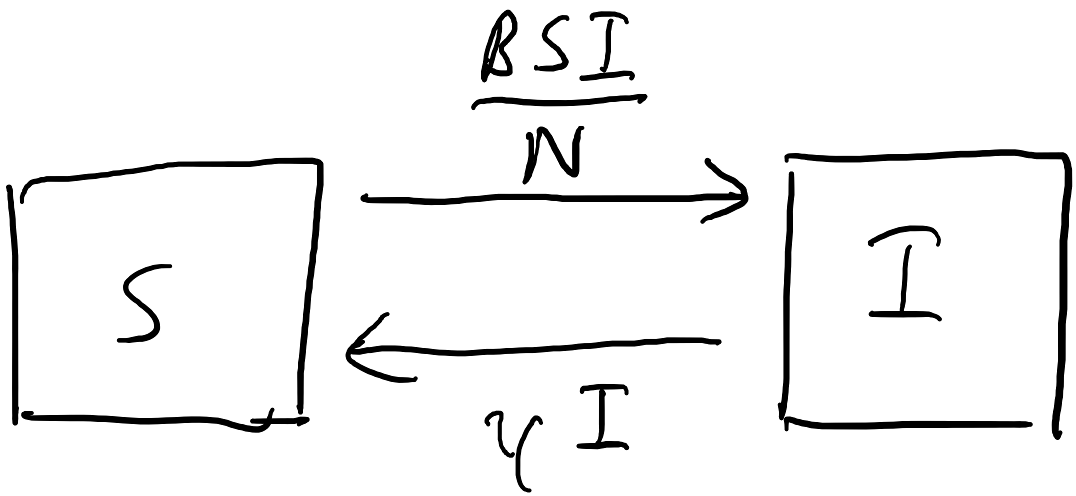
Epidemic curves – Covid19 in Canada



https://covid19tracker.ca/, 7-day rolling average, new cases

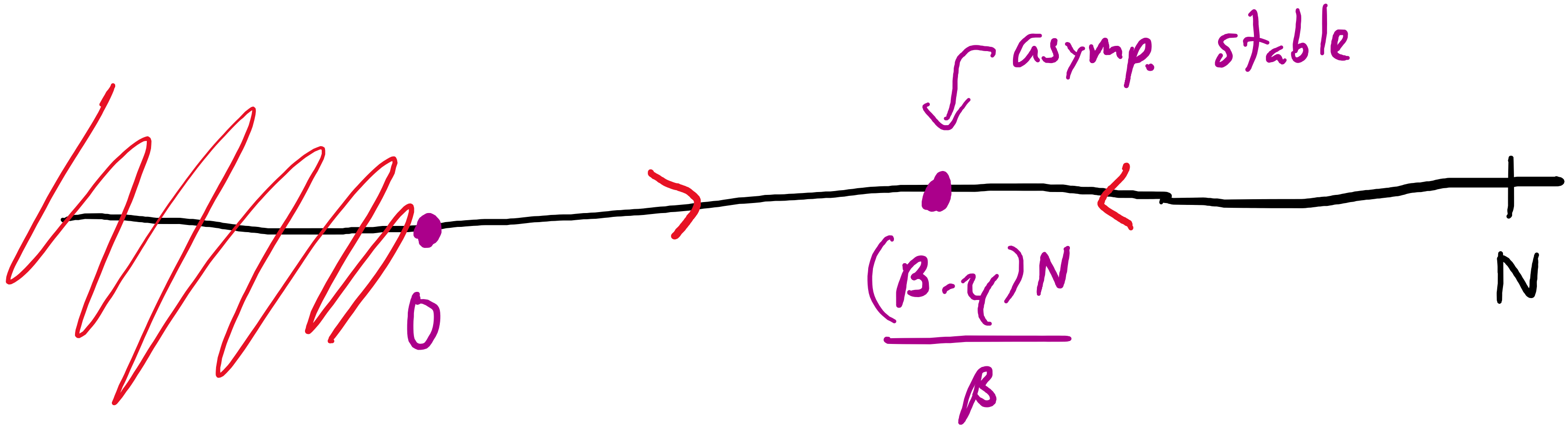
SIS Model

- Modified assumption 1: The infection rate is proportional to the average number of times an infected individual encounters a susceptible individual in the population, assuming random encounters.
- Assumption 2: there is a total fixed population size $N = I(t) + S(t)$, where S is the number of Susceptible individuals.
- Assumption 3: Individuals recover at rate γ, and become susceptible to re-infection.



SIS Model Properties

- In the SIS model, we call the ratio $R_0 = \frac{\beta}{\gamma}$ the *basic reproduction number* of the system because it is the number of secondary infections β caused during the infectious period $\frac{1}{\gamma}$.
- If $R_0 > 1$, the disease persists and we are at the endemic disease equilibrium.



1D systems and complex behavior

- Autonomous 1D ODE systems $\dot{x} = f(x)$ can't have cycles or any doubling back because you'd have to cross a point going the other direction.

Modelling waves

- How can ODE models capture the infection waves that we see in real epidemics?

A: Add more variables
B: Make them nonautonomous
C: Add in births/deaths
D: All of the above
E: None of the above

SIR Model

- Modified assumption 3: Individuals recover at rate γ, and become *Removed* $R(t)$ from the population.
- Modified assumption 2: There is a total fixed population size $N = I(t) + S(t) + R(t)$.

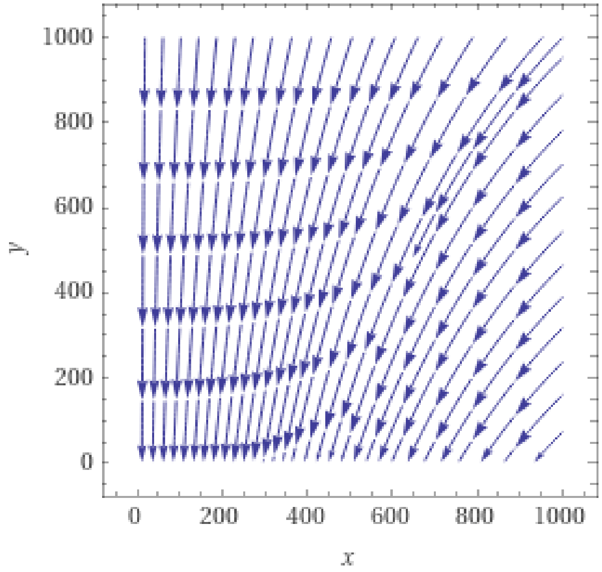
Reduce SIR model to two variables

- $$\begin{cases} \dot{S} = -\frac{\beta}{N} SI \\ \dot{I} = \frac{\beta}{N} SI - \gamma I \\ \dot{R} = \gamma I \end{cases}$$

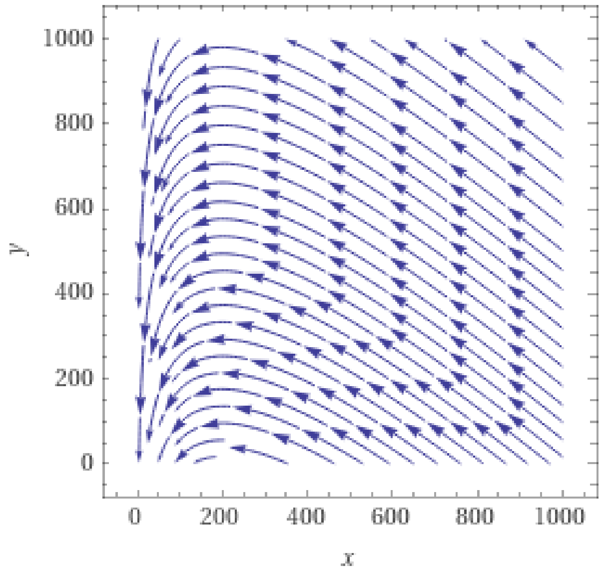
Plots of basic SIR model

WolframAlpha: streamplot {-1/10000 * x * y, x*y/10000 - 0.2*y}, x=0..1000, y=0..1000

- Let $\beta = 0.1, \gamma = 0.2, N = 1000$
- $$\begin{cases} \dot{S} = -\frac{0.1}{1000} SI \\ \dot{I} = \frac{0.1}{1000} SI - 0.2I \end{cases}$$



- Let $\beta = 0.1, \gamma = 0.02, N = 1000$
- $$\begin{cases} \dot{S} = -\frac{0.1}{1000} SI \\ \dot{I} = \frac{0.1}{1000} SI - 0.02I \end{cases}$$



Basic SIR model gives a single wave

- Again, $R_0 = \frac{\beta}{\gamma}$.
- If $R_0 < 0$, then no epidemic happens because infected individuals never increase.
- If $R_0 > 1$, then an epidemic with a single wave happens.
- Then higher the R_0, the higher the peak. We can "flatten the curve" by reducing β, the transmission rate.
- https://www.geogebra.org/m/p5ucts38 by Ivan De Winne

Nonautonomous Epidemic model

- What if β is sometimes high, and sometimes low, as a function of time?
- Then $\begin{cases} \dot{S} = -\frac{\beta(t)}{N} SI \\ \dot{I} = \frac{\beta(t)}{N} SI - \gamma I \end{cases}$ is not autonomous.

(autonomous) SIRS Epidemic Model

- Assumption 4: Removed individuals lose immunity at rate ν.

Phase plane analysis of SIRS model

- $$\begin{cases} \dot{S} = -\frac{\beta}{N}SI + \nu(N - I - S) \\ \dot{I} = \frac{\beta}{N}SI - \gamma I \end{cases}$$

Qualitative analysis of equilibria

- $$\begin{cases} \dot{S} = -\frac{\beta}{N} SI + \nu(N - I - S) \\ \dot{I} = \frac{\beta}{N} SI - \gamma I \end{cases}$$
- Equilibria at

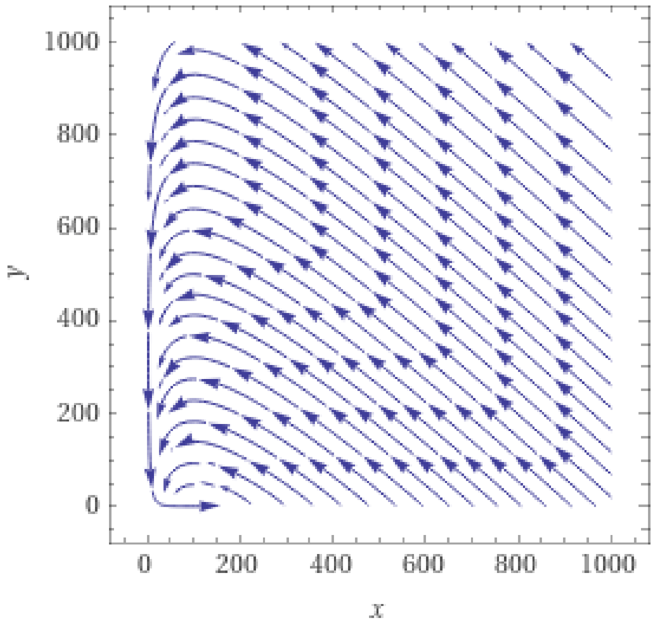
$$\begin{cases} (S, I) = (N, 0) \\ (S, I) = \left(\frac{N\gamma}{\beta}, \frac{\nu N \left(1 - \frac{\gamma}{\beta}\right)}{\gamma + \nu} \right) \end{cases}$$

Example with periodic behavior

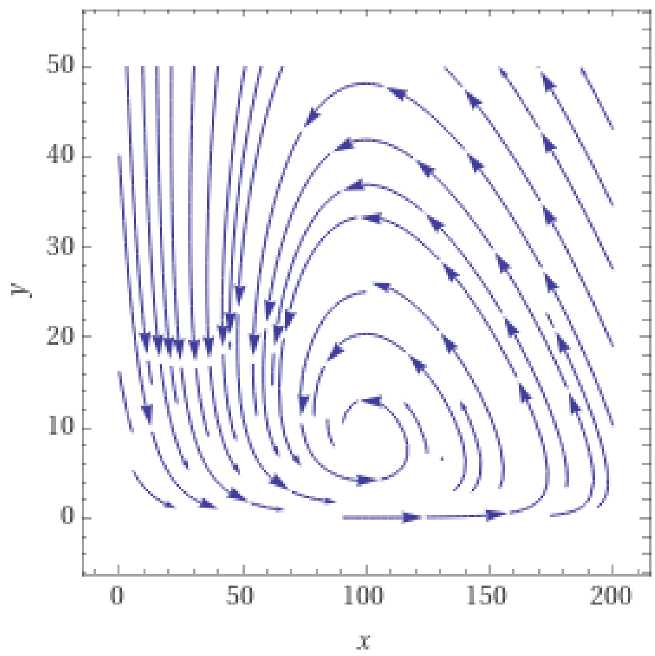
- $\beta = 0.1, \gamma = 0.01, \nu = 0.0001, N = 1000$
- $$\dot{S} = -\frac{\beta}{N}SI + \nu(N - I - S)$$
  $$= -0.0001SI + 0.0001(1000 - S - I)$$
- $$\dot{I} = \frac{\beta}{N}SI - \gamma I = 0.0001SI - 0.01I$$
- Equilibria: $(S, I) = (1000,0)$ or $(S, I) \approx (100, 8.91)$
- Jacobian $J(S, I) =$
  $\begin{bmatrix} -0.0001I - 0.0001 & -0.0001S - 0.0001 \\ 0.0001I & 0.0001S - 0.01 \end{bmatrix}$
- $J(1000,0) = \begin{bmatrix} -0.0001 & -0.1001 \\ 0 & 0.09 \end{bmatrix}$,
  $\lambda_1 = -0.0001, \lambda_2 = 0.09$, saddle pt
- $J(100, 8.91) = \begin{bmatrix} -0.000991 & -0.0101 \\ 0.000891 & 0 \end{bmatrix}$,
  $\lambda_{1,2} \approx -0.0004955 \pm 0.00296i$ so the equilibrium is an attracting spiral

Visualization

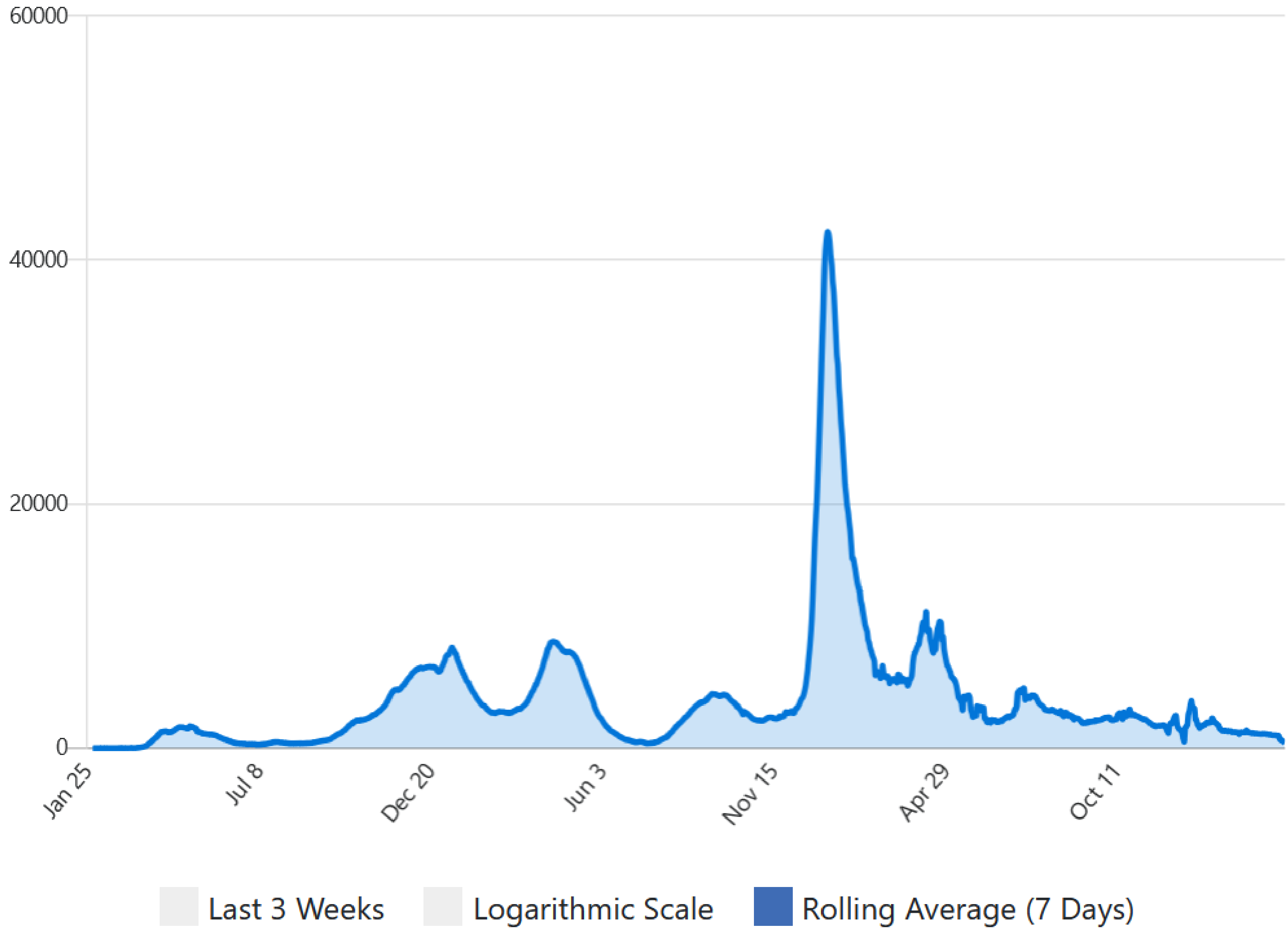
- streamplot {- x * y/10000 + 0.0001*(1000-x-y), x*y/10000 - 0.01*y}, x=0..1000, y=0..1000



- streamplot {- x * y/10000 + 0.0001*(1000-x-y), x*y/10000 - 0.01*y}, x=0..400, y=0..50



Epidemic curves – Covid19 in Canada



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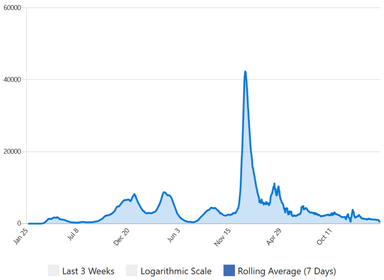
More sophisticated epidemic models

- Vaccinations
- Birth/Death
- Exposed but not Infectious

Modelling summary



- We can build models of complex biological phenomenon like epidemics by looking at behavior we care about and adding more assumptions to reproduce that real behavior in our models.
- Sometimes, we can get similar behavior with two different assumptions, so as mathematical biologists we need to decide what fits better.
- For epidemic modelling, these can include additional variables/compartments, as well as new arrows between compartments.