

# Strain-space model for Sars-CoV-2

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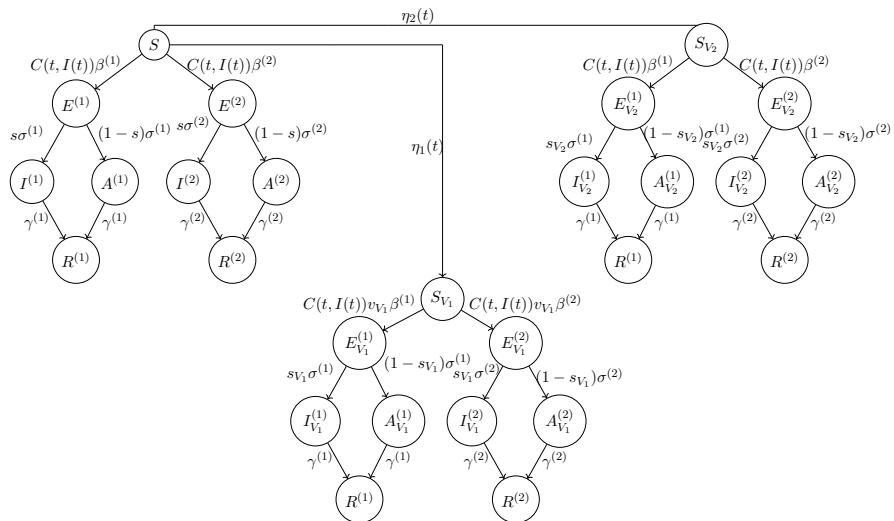
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<sup>4</sup>Simon Fraser University, Burnaby, Canada

<sup>5</sup>Shared Hospital Laboratory, Toronto, Canada

July 6, 2022

# Modeling multiple infections is challenging



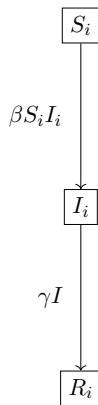
# How to incorporate more data?

phylogeny.png

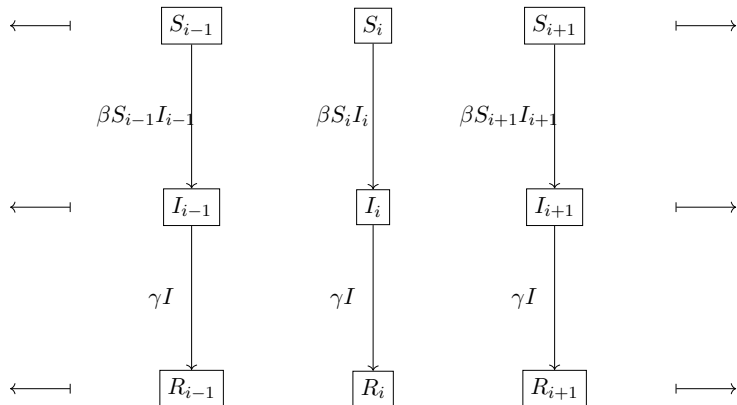
## A useful approximation

`gog_paper.png`

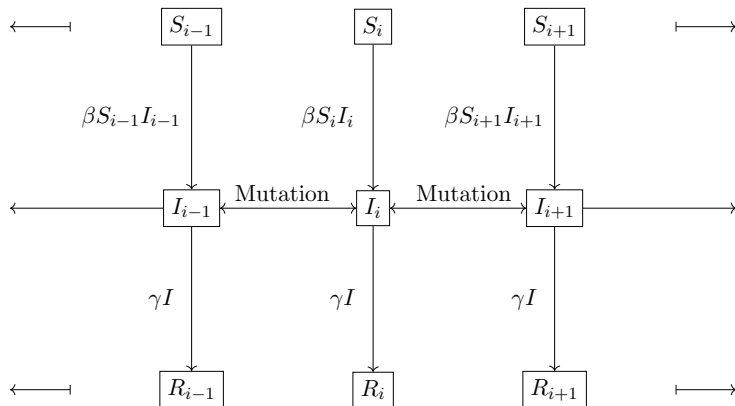
## Begin with simple SIR model



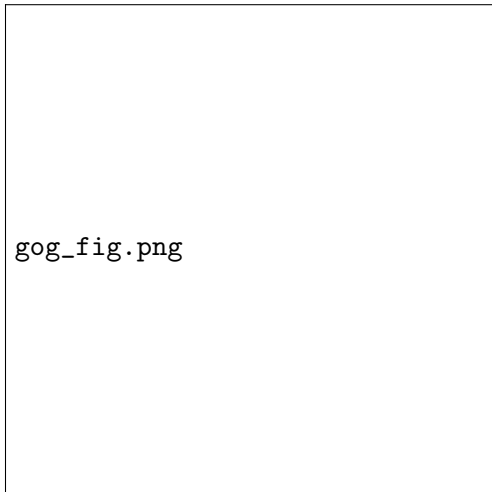
# Organize strains on a lattice



# Strains mutate into adjacent strains



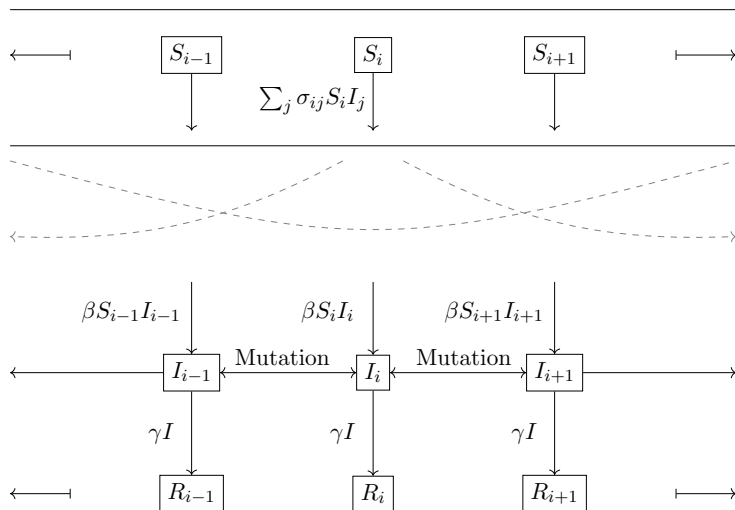
Infections provide cross-immunity to nearby strains



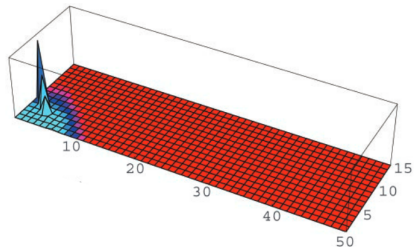
[?]



# Full strain lattice model in 1 dimension

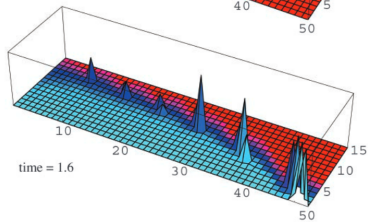
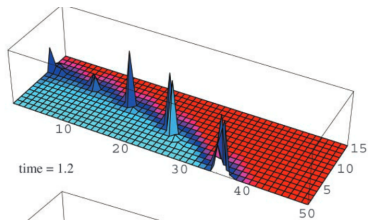
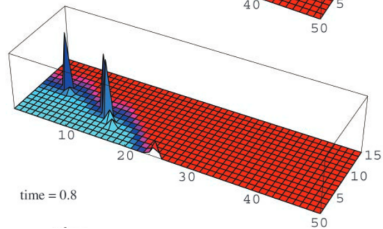
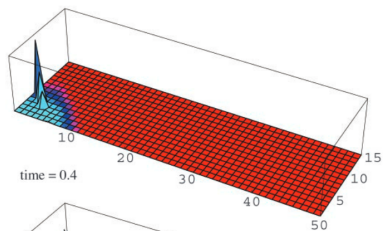


## Extending the strain lattice to 2 dimensions



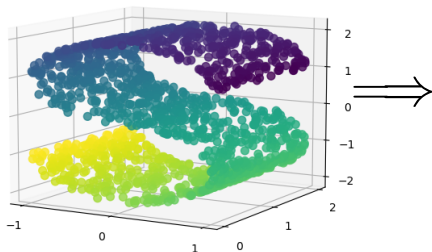
[?]

# Extending the strain lattice to 2 dimensions



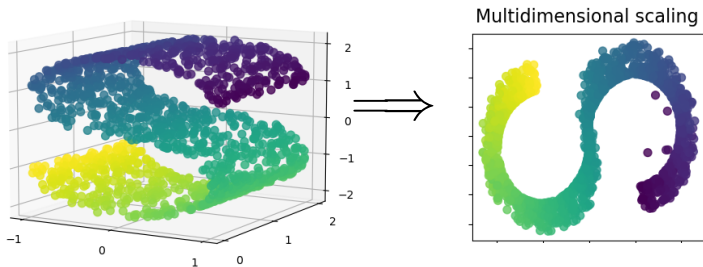
[?]

# Projecting to low dimensions while preserving distances



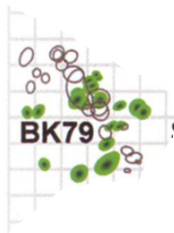
[?]

# Projecting to low dimensions while preserving distances

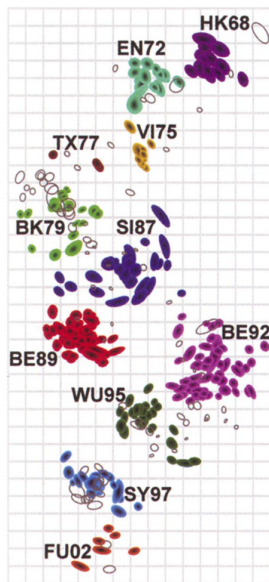


[?]

# Approximating genomic data in 2 dimesions

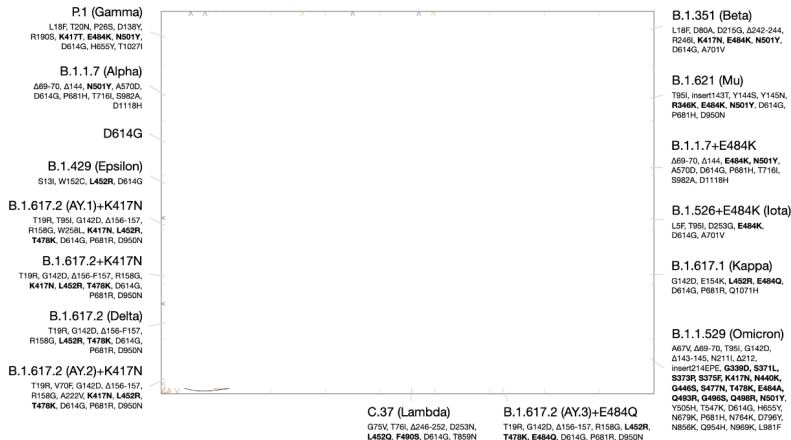


# Approximating genomic data in 2 dimesions



[?, ?]

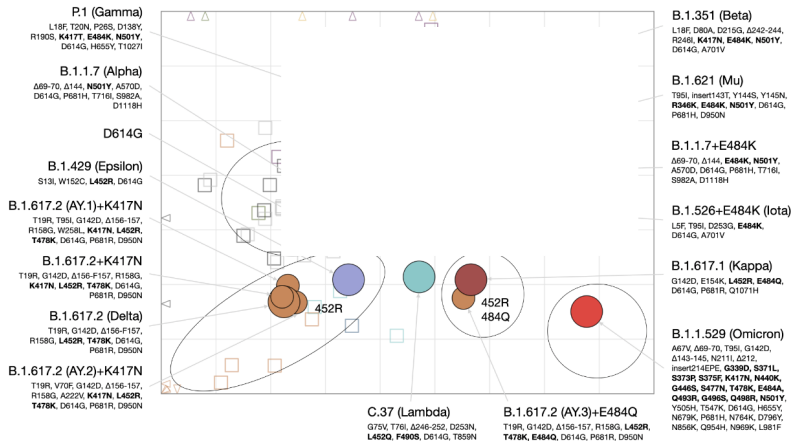
# Mapping SARS-CoV-2 in 2D



[?]

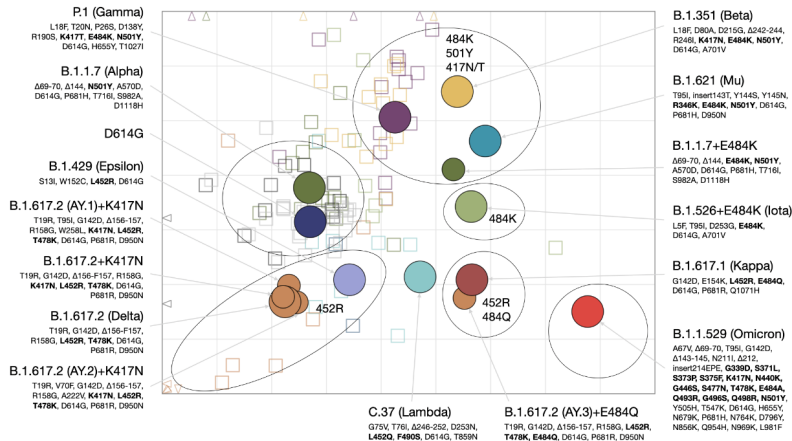


# Mapping SARS-CoV-2 in 2D



[?]

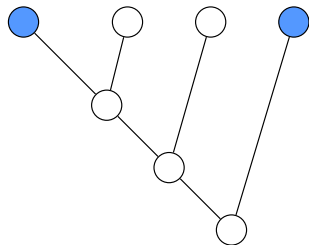
# Mapping SARS-CoV-2 in 2D



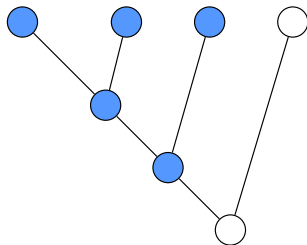
[?]

# Methods of adding more genomes: homoplasious sites

Homoplasious trait

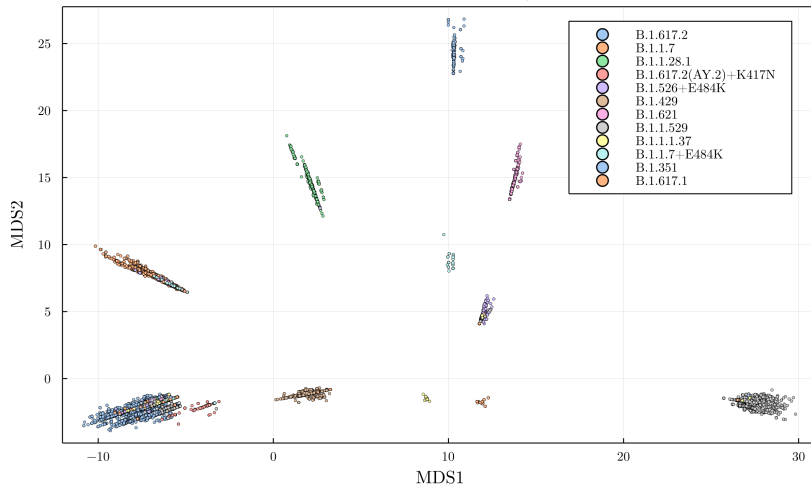


Homologous trait



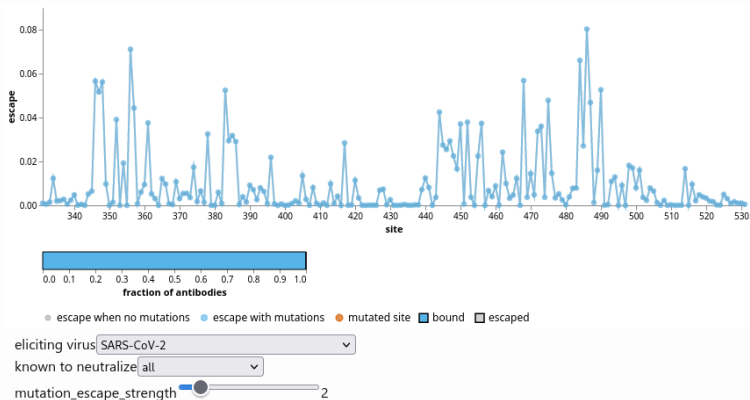
[?]

# Homoplasic mutations map



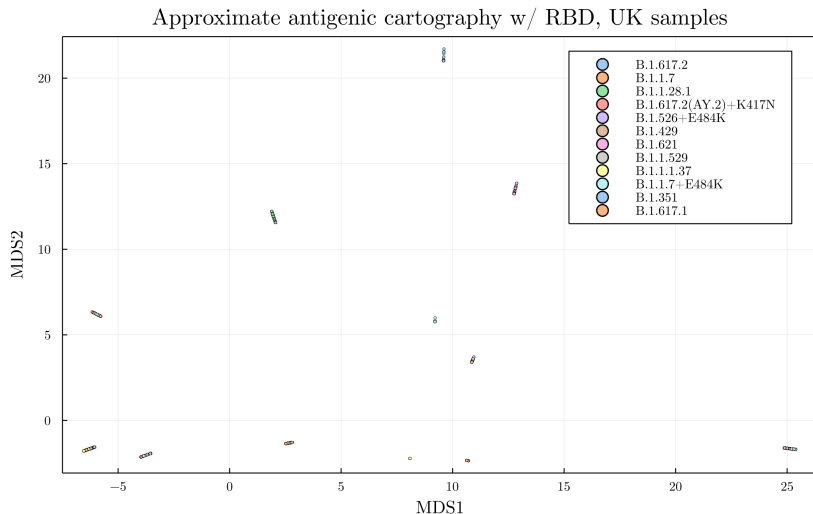
# Methods of adding more genomes: polyclonal antibody binding studies

## Escape calculator for SARS-CoV-2 RBD

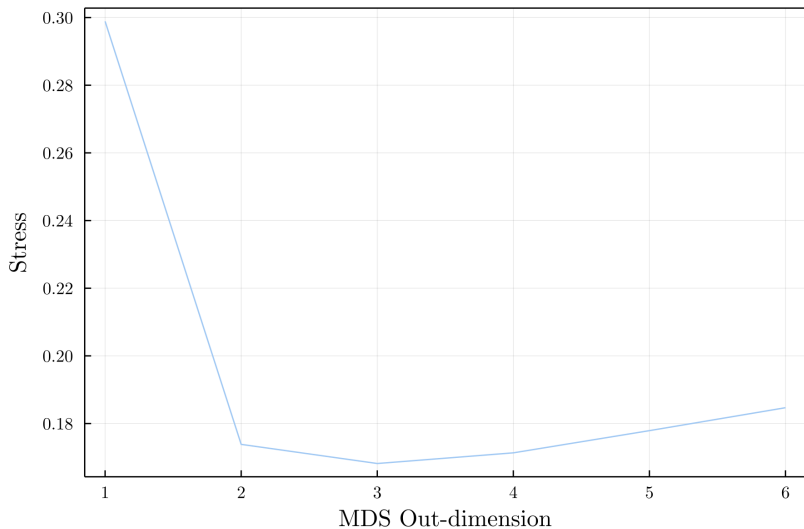


[?]

# Antibody Binding map



# Evaluating the MDS approximation



(animation of kernel approximation)



## Model parameters/variables

Symbol	Description
$N$	Size of variant grid
$S_{ij}$	Population susceptible to variant $(i, j) \in [0, N]^2$
$I_{ij}$	Population infected by variant $(i, j) \in [0, N]^2$
$R_{ij}$	Recovered/Immune to variant $(i, j) \in [0, N]^2$
$\sigma_{ijkl}$	Probability that exposure to variant $(i, j)$ causes immunity to variant $(k, l)$
$\beta_{ij}$	Transmission rate of variant $(i, j)$
$v(t)$	vaccination rate at time $t$
$s(t)$	stringency at $t$
$\xi$	Recovery rate of all strains
$\gamma$	Rate of immunity loss of all strains

Table of symbols for Model 2

# Model Equations

$$\frac{S_{ij}}{dt} = - \sum_{kl} s(t) \beta_{kl} \sigma_{ijkl} S_{ij} I_{kl} + \gamma R_{ij} - V(t) S \quad (1)$$

$$\frac{I_{ij}(t)}{dt} = s(t) \beta_{ij} S_{ij} I_{ij} - \xi I_{ij} + M (-4I_{ij} + I_{i-1,j} + I_{i+1,j} + I_{i,j-1} + I_{i,j+1}) \quad (2)$$

$$\frac{R_{ij}(t)}{dt} = \xi I_{ij} - \gamma R_{ij} + V(t) S \quad (3)$$

Boundary conditions:  $I_{0,j} = 0, I_{j,0} = 0, I_{N,j} = 0, I_{j,N} = 0$

Initial conditions computed from genomic data in GISAID

(gifs of model dynamics)

## Further work

- Estimating a nonlocal diffusion kernel
- Better model fitting
- Predicting vaccine targets



# Mutation homoplasy

gene\_freq.png