

The interplay between genomic surveillance and public health interventions

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BIOCOMPLEXITY INSTITUTE



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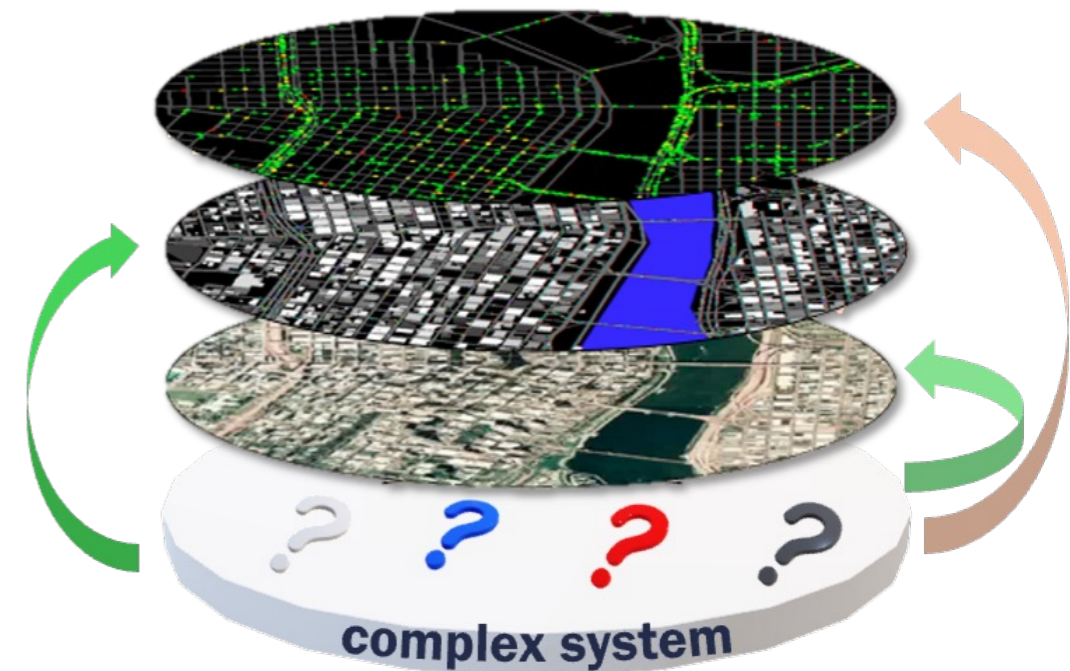
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Biocomplexity Institute's Approach: TRANSDISCIPLINARY TEAM SCIENCE

Can informatics tools be developed to aid design and analysis of actionable policies pertaining to complex STIO (social, informational, technical and organizational) systems, e.g. sustainable habitats, pandemics, global conflicts?

- Undertake problems that cannot be solved by a single faculty member or within the narrow boundaries of a discipline
- Application-driven science and engineering: work on problems that are motivated by real-life applications
- This paradigm produces fundamental advances
- These problems require *sophisticated tools and diverse intellectual resources*, making them impossible to solve within the narrow confines of a single discipline or by a few individuals



Real-time Comprehensive Analytics For Pandemic Response



FEDERAL LEVEL



Participation in CDC-coordinated hubs for forecasting and scenario-based projections



Weekly projections and bespoke modeling for federal partners



STATE & LOCAL LEVEL

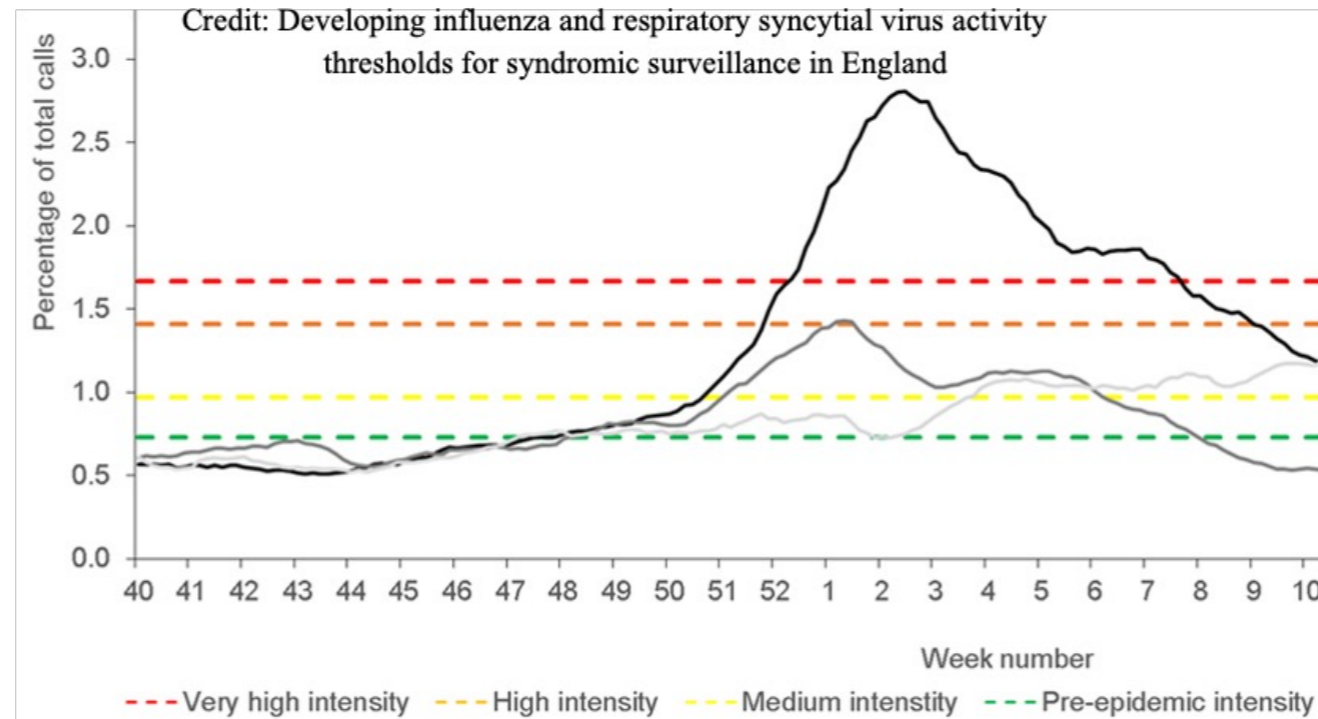


Multiple presentations and data deliveries each week from spring 2020 - present



Challenges of classic disease surveillance systems

Classic surveillance relies on fixed thresholds - a static problem approach



Disease surveillance and response is a dynamic problem and depends on multiple factors:

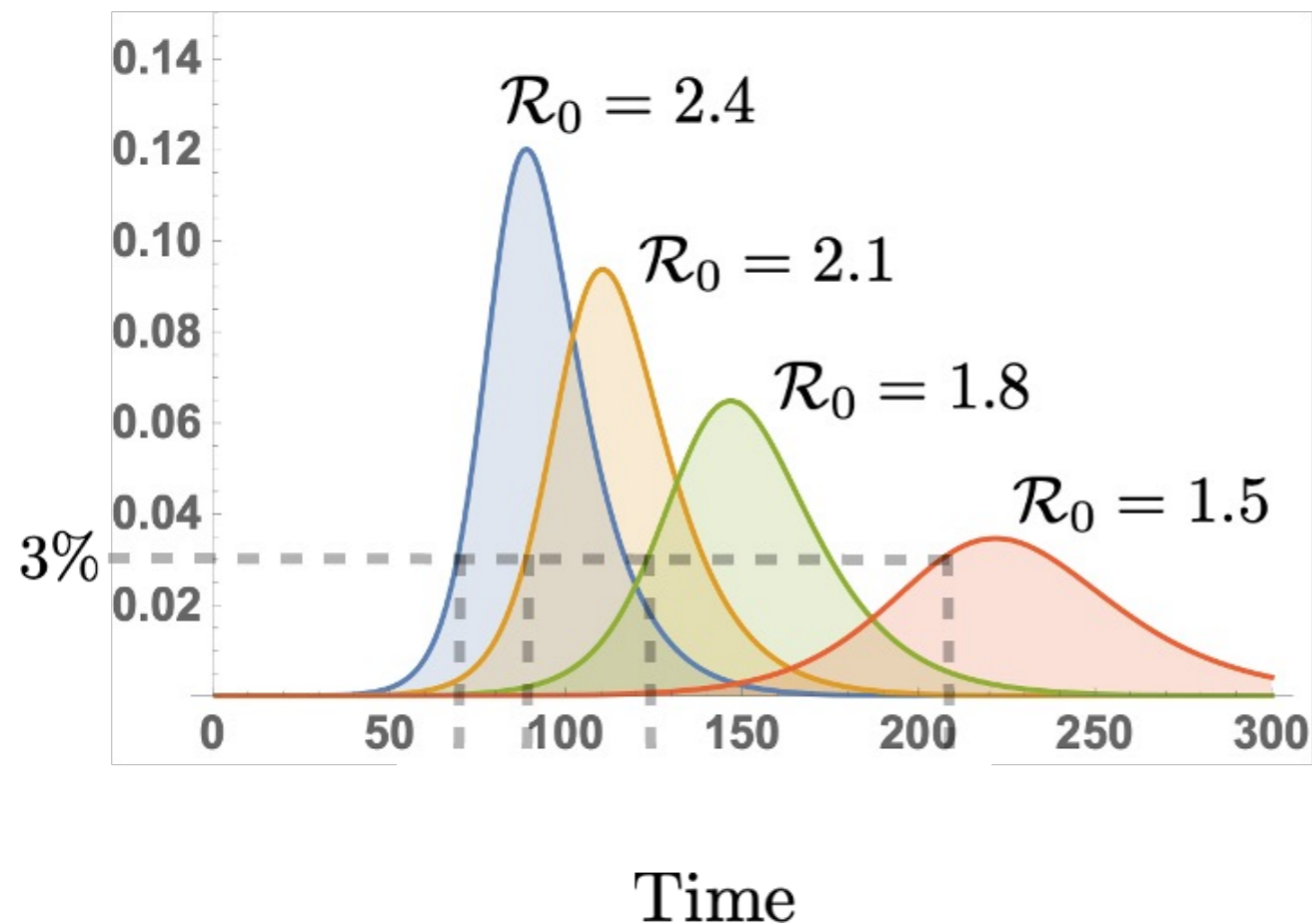
- **Surveillance effort:** number of samples, resource allocation, test accuracy
- **The disease ecology:** pathogen's infectiousness, host immunity, virus evolution
- **Population's structure:** size and density, age distribution, activity patterns

Challenges of classic disease surveillance systems

The basic reproductive number

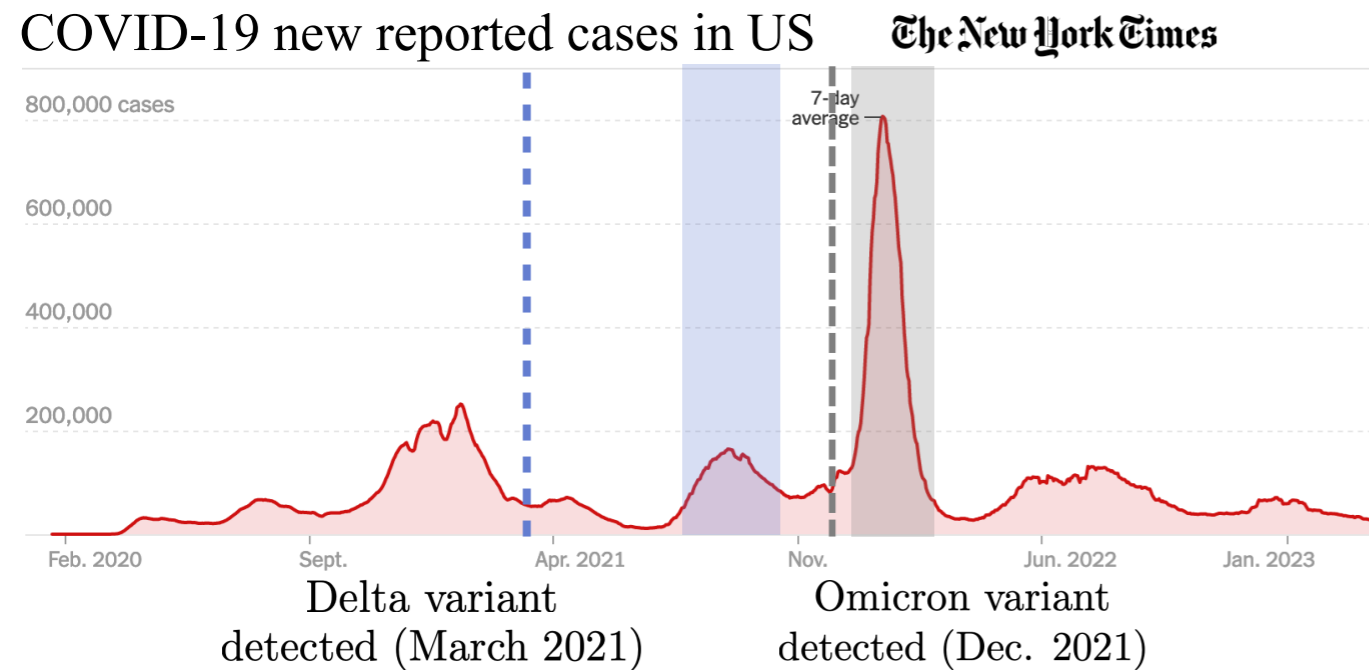
\mathcal{R}_0 determines the outbreak's propagation speed and intensity

Disease prevalence



- Diseases with high \mathcal{R}_0 are detected earlier but there is less time to respond.
- High \mathcal{R}_0 leads to lower marginal benefits of increasing sampling size.

Genomic epidemiology



Variants' dominance time in the US*

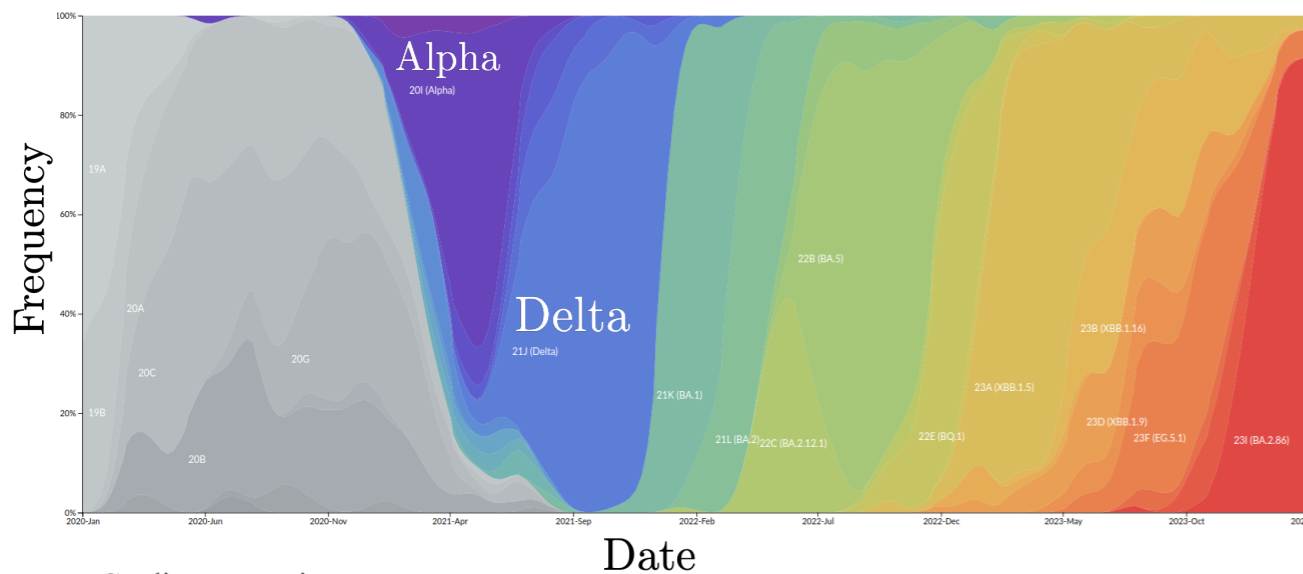
Delta

within 7-13 weeks after detection

Omicron

within 4-6 weeks after detection

COVID-19 variants in North America



Credit: nextstrain.org

How do different variant's importation conditions shape epi-genomic dynamics?

How fast can we detect:

- an outbreak?
- a novel variant?

How do intervention effects depend on the variant's competing dynamics?

*Chan Y, Irvine MA, Prystajecky N, Sbihi H, Taylor M, Joffres Y, et al. Emergence of SARS-CoV-2 Delta Variant and Effect of Nonpharmaceutical Interventions, British Columbia, Canada. *Emerg Infect Dis.* 2023;29(10):1999-2007. <https://doi.org/10.3201/eid2910.230055>

Multi-theory computational modeling framework

- Multi-variant disease dynamics
- Disease and genomic surveillance
- Interventions strategies

Multi-theory computational modeling framework

PNAS

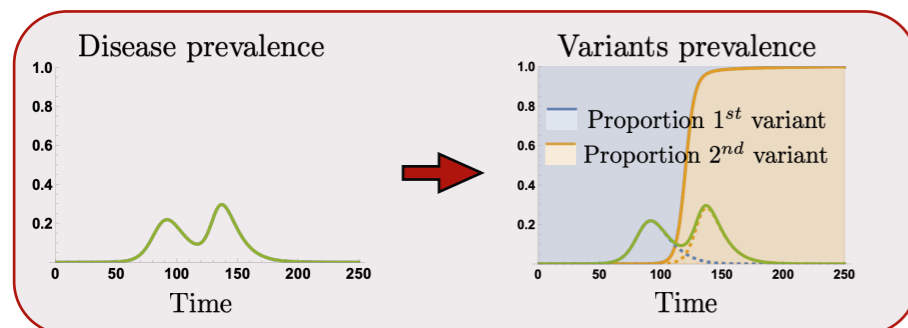
Coupled models of genomic surveillance and evolving pandemics with applications for timely public health interventions

Baltazar Espinoza , Aniruddha Adiga, Srinivasan Venkatramanan, , and Madhav V. Marathe   [Authors Info &](#)

A modeling framework integrating:

- multi-variant disease dynamics
- disease and genomic surveillance
- intervention strategies

Outbreak & variant detection



Testing



Sequencing

Multi-theory computational modeling framework

PNAS

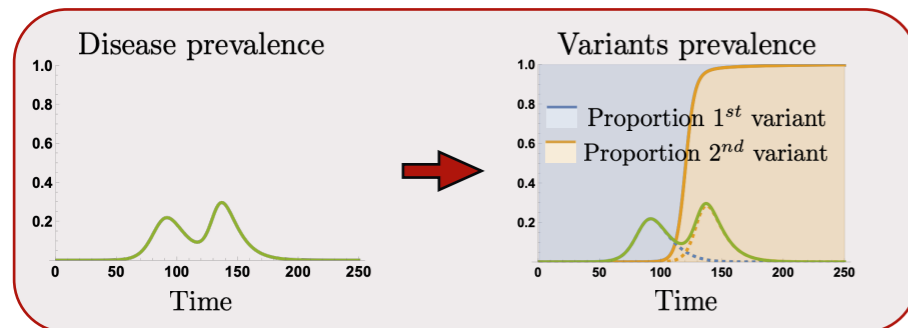
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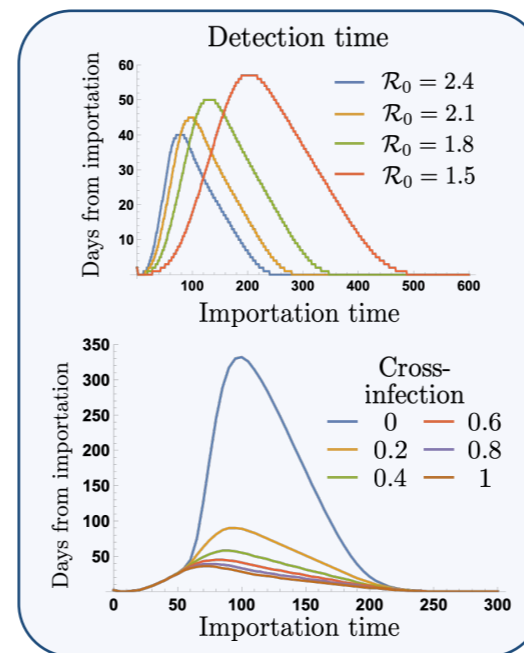
Outbreak & variant detection



 Testing

 Sequencing

Variant characterization



 Cross-infection

 Relative infectiousness

Multi-theory computational modeling framework

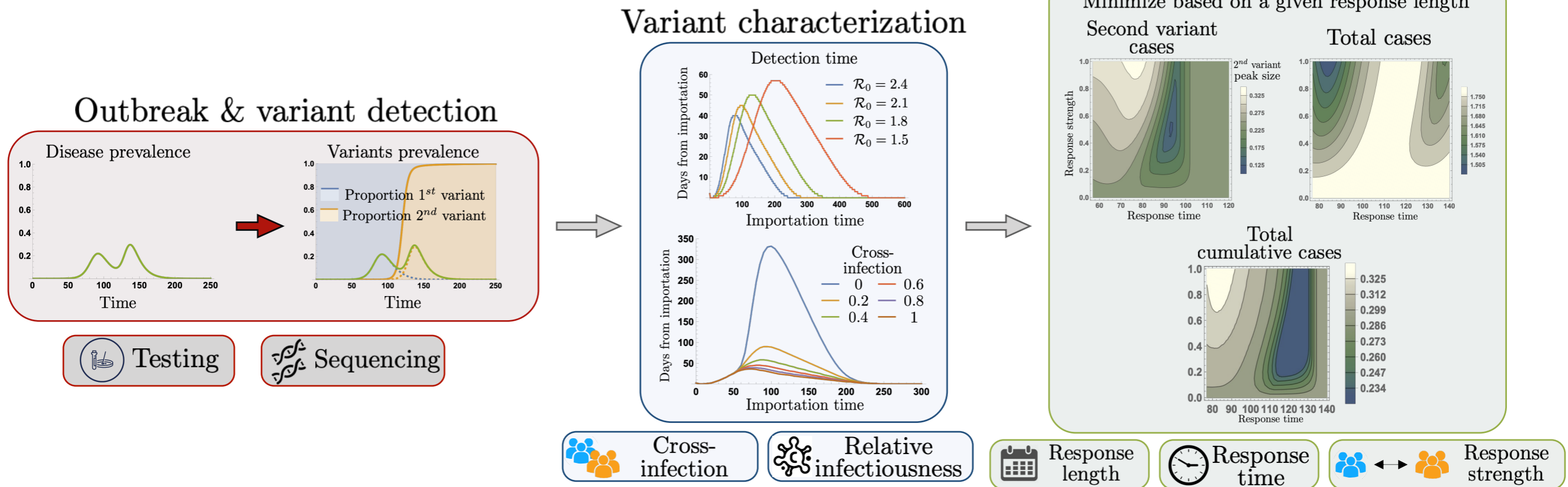
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A modeling framework integrating:

- multi-variant disease dynamics
- disease and genomic surveillance
- intervention strategies



A realistic course of action based on sequencing, analysis, and response

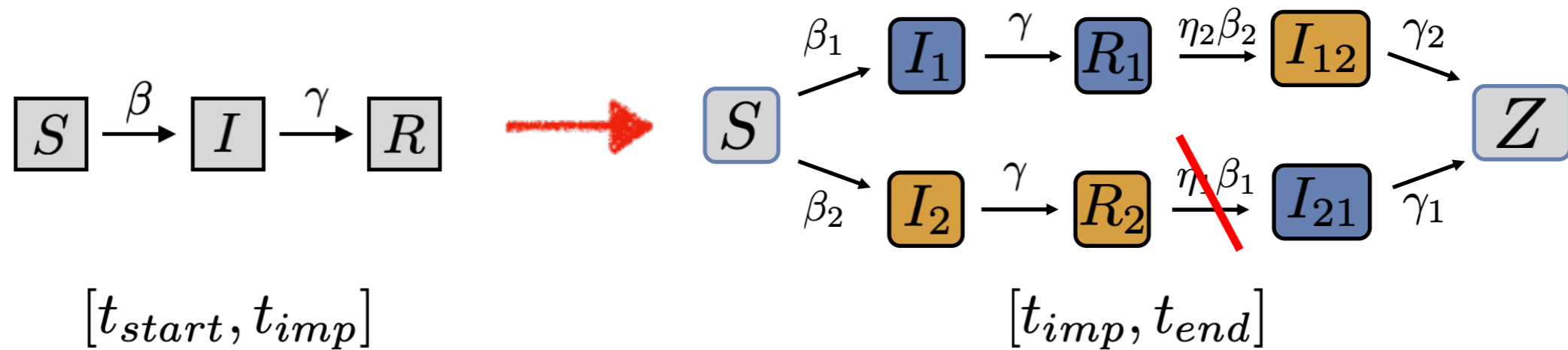
Key policy insights

- Effective responses requires to characterize the genomic landscape
- Depending on the target metric different intervention scenarios lead to better outcomes
- Robust surveillance systems allow to study multiple intervention scenarios
- Sustained interventions suppress potential epidemic revival

Key scientific insights

- Multi-variant competing dynamics inherently limit:
 - Novel variant's detection
 - Intervention effectiveness
- Novel variant's detection could be delayed by:
 - co-circulating variants
 - importation time
- Cross-infection levels do not impact the detection time of early imported variants

Multi-variant model

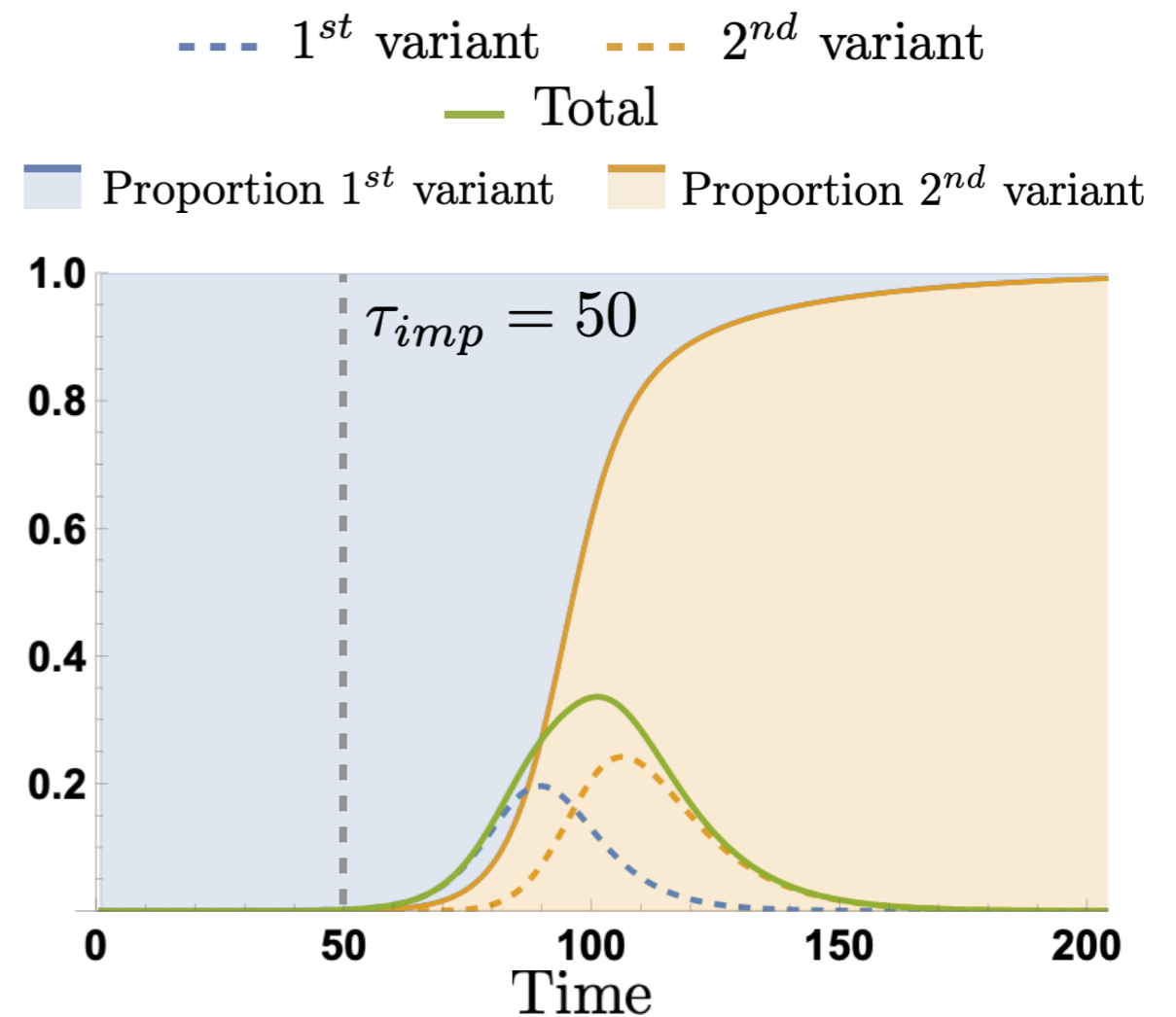
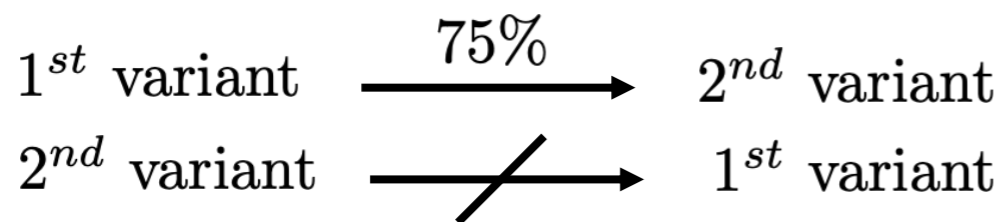


τ_{imp} - novel variant's importation time

New variant's characteristics similar to Delta importation

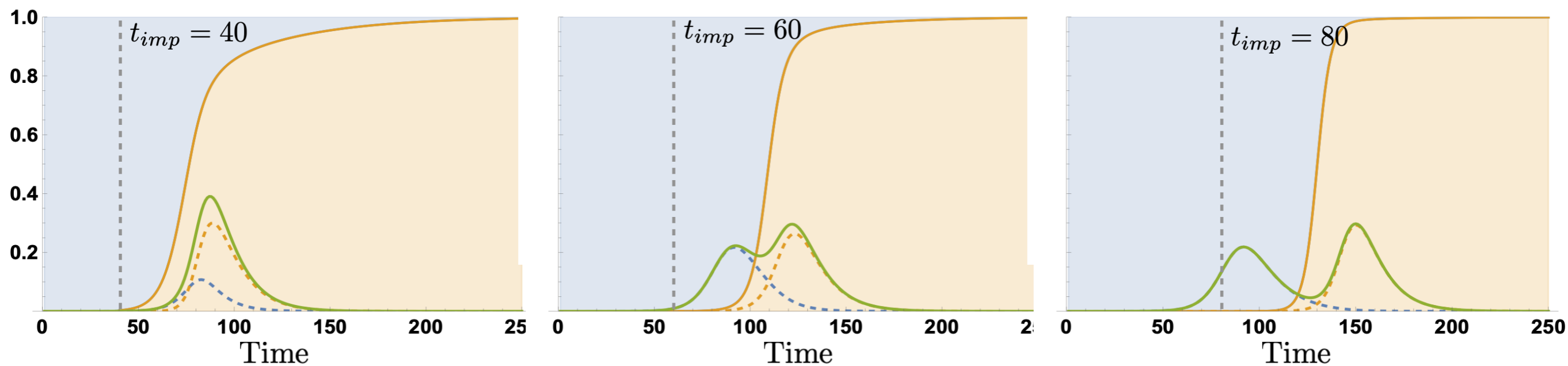
$$\mathcal{R}_0 = 2.4 \quad \beta_2/\beta_1 = 1.6$$

Linear cross-immunity scheme



Variant's competing dynamics

— 1st variant — 2nd variant — Total
■ Proportion 1st variant ■ Proportion 2nd variant



Disease and variants dynamics are highly sensitive to importation conditions:

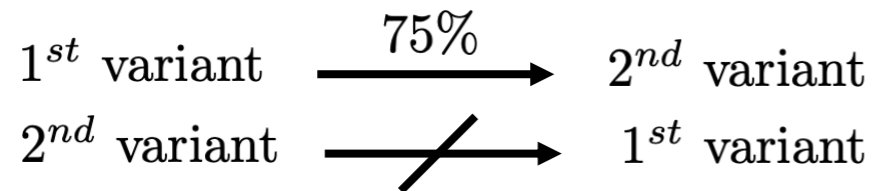
- Importation time
- Relative infectiousness
- Cross-immunity

Novel variant's detection

- Imported variants: population mobility
- Emerging variants: evolutionary processes

Novel variant's detection trade-offs

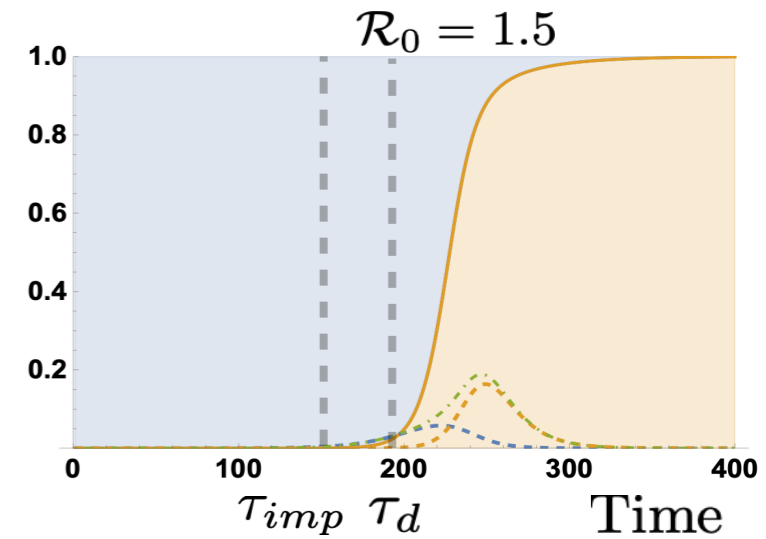
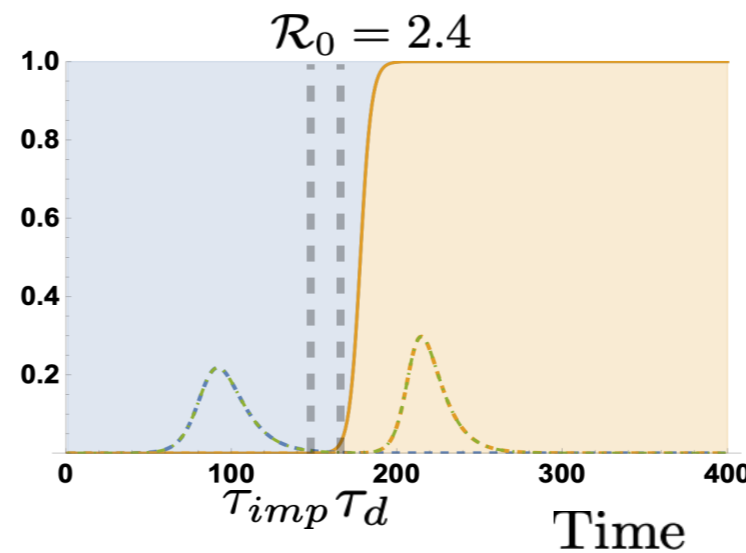
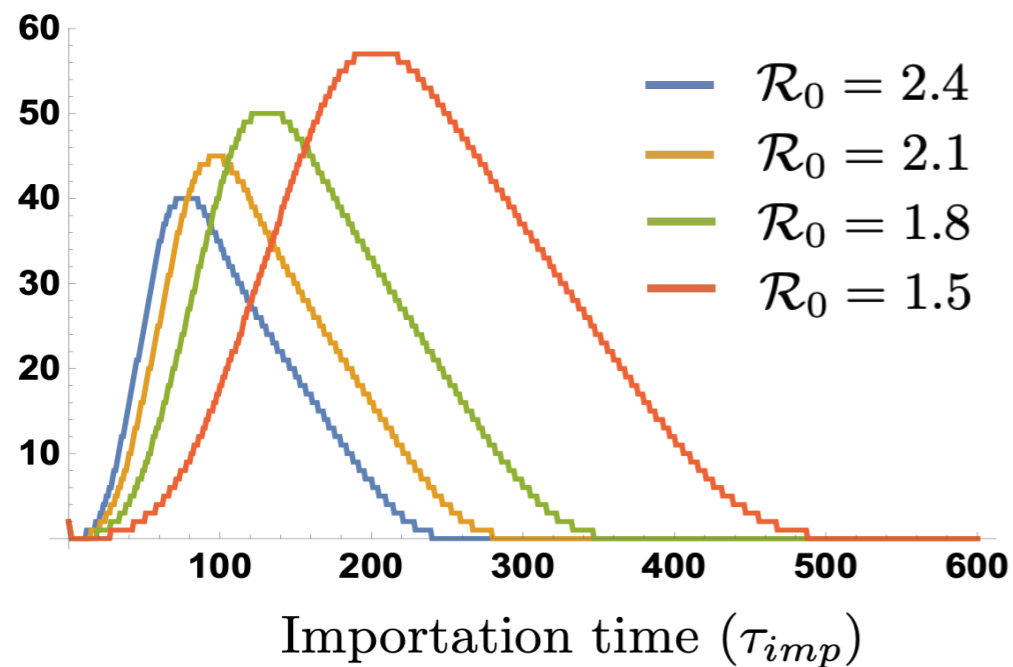
$$\beta_2/\beta_1 = 1.6 \quad \tau_{imp} = 150$$



τ_{imp} - novel variant's importation time

τ_d - novel variant's detection time

Time from importation to detection

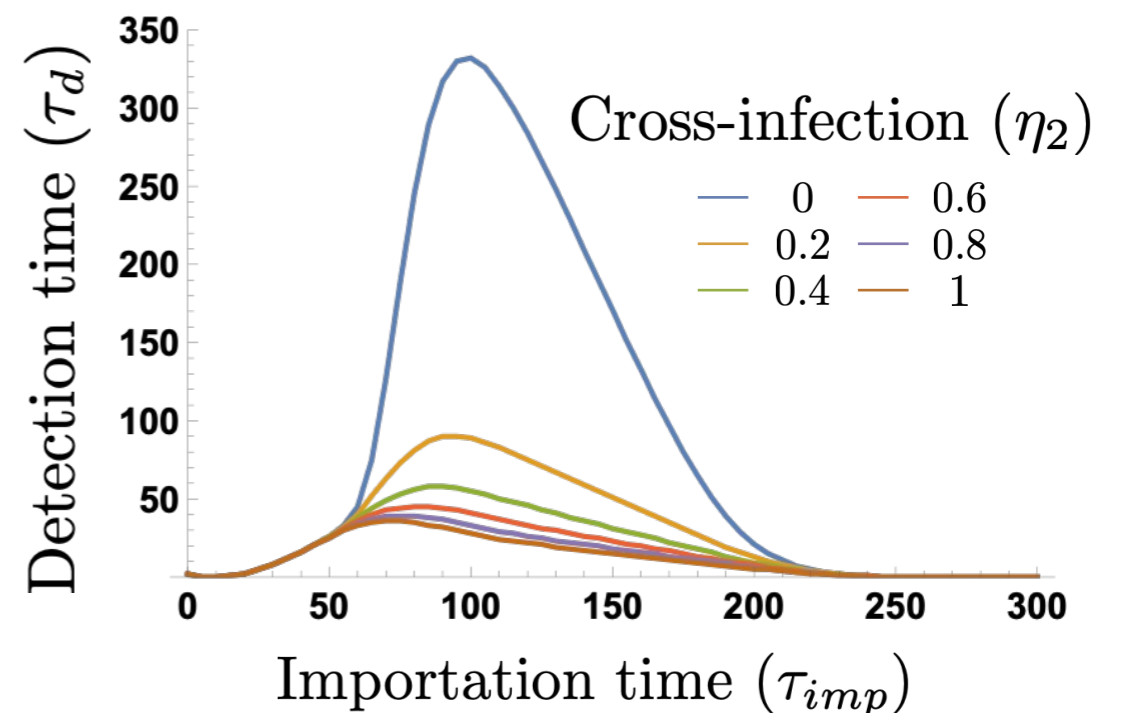
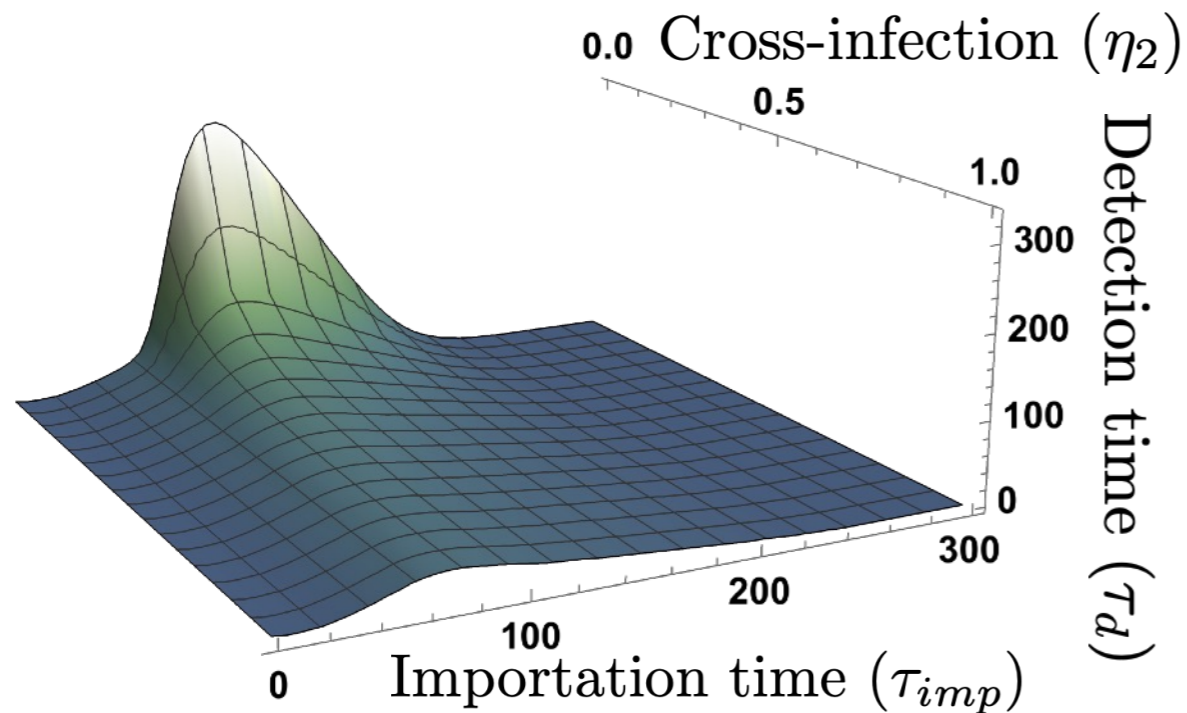


A trade-off between

- the disease's propagation speed (R_0)
- the novel variant's importation time

Novel variant's detection sensitivity

The detection time depends on the varying cross-infection and importation times



- **Detection times are not sensitive to cross-infection for 'early' importation times.**
- **Low cross-infection values delay detection times.**

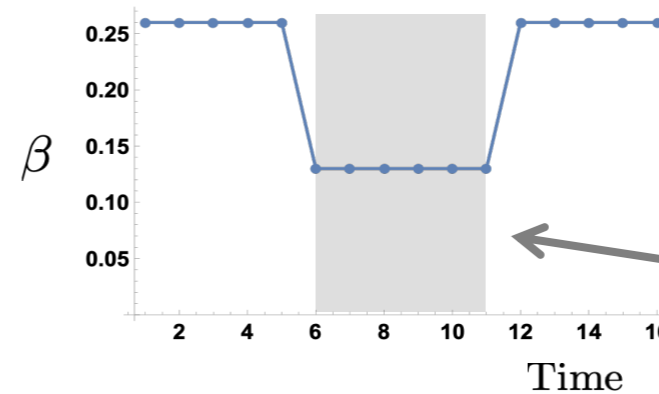
Response after novel variant's detection

Response after novel variant's detection

Interventions temporarily reduce the infection likelihood

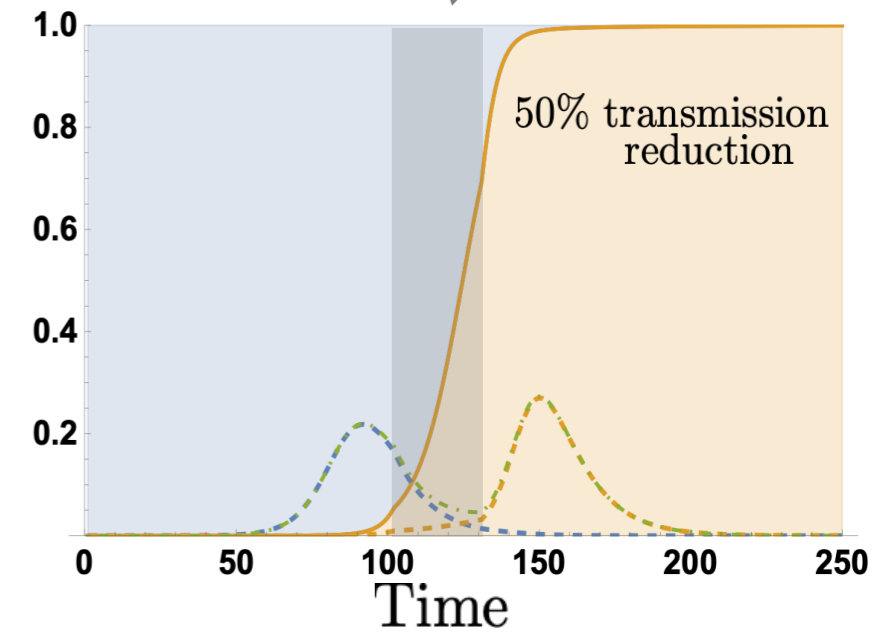
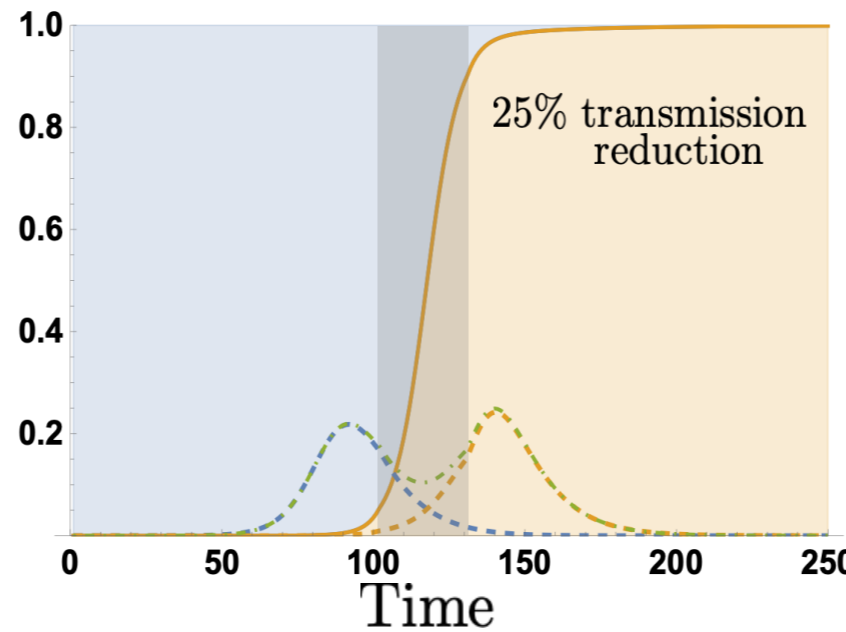
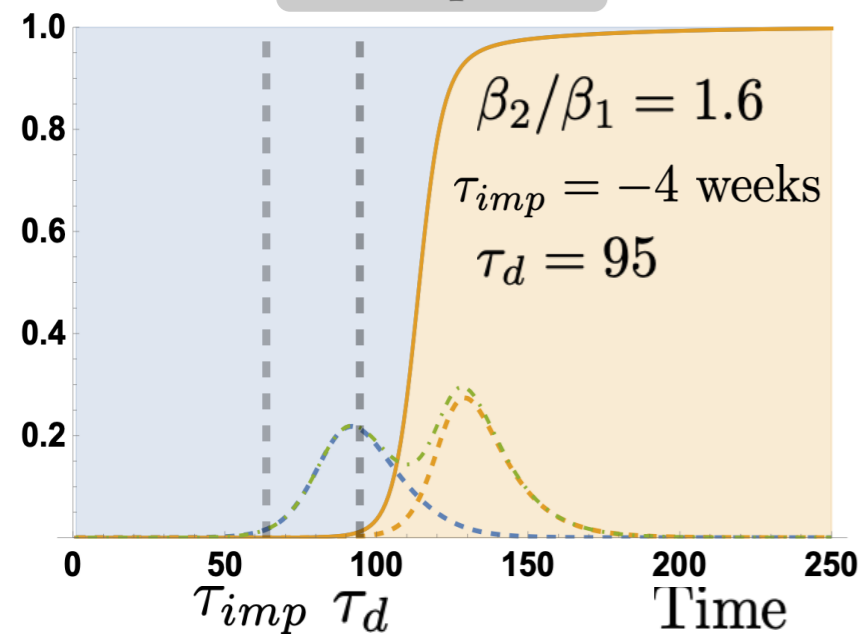
τ_{imp} - novel variant's importation time

τ_d - novel variant's detection time



Intervention period

No response

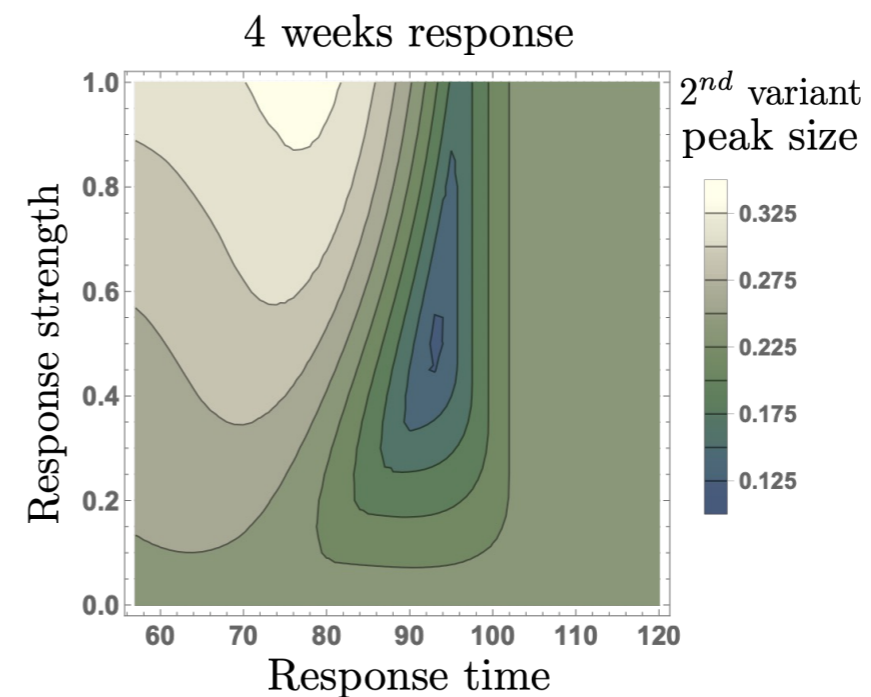
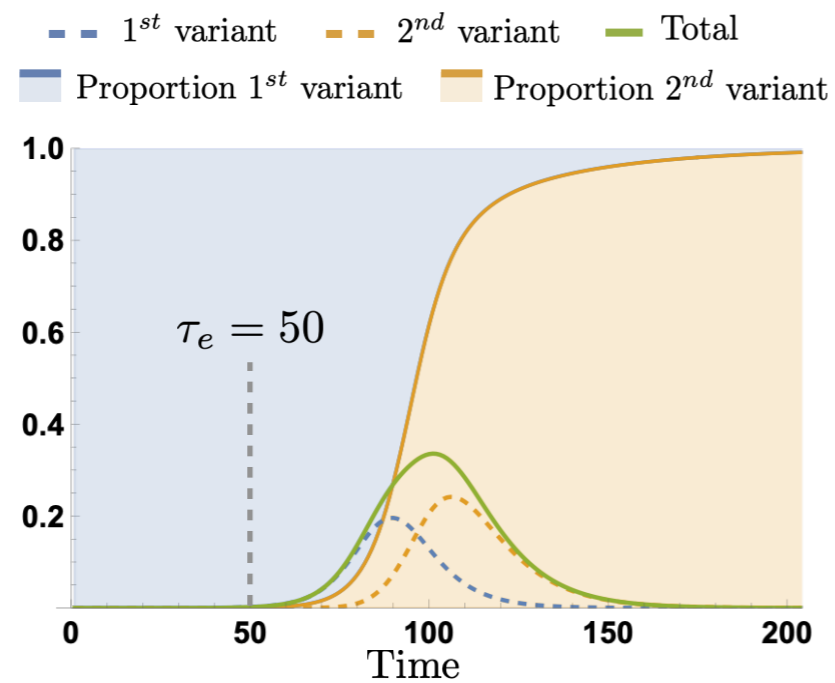
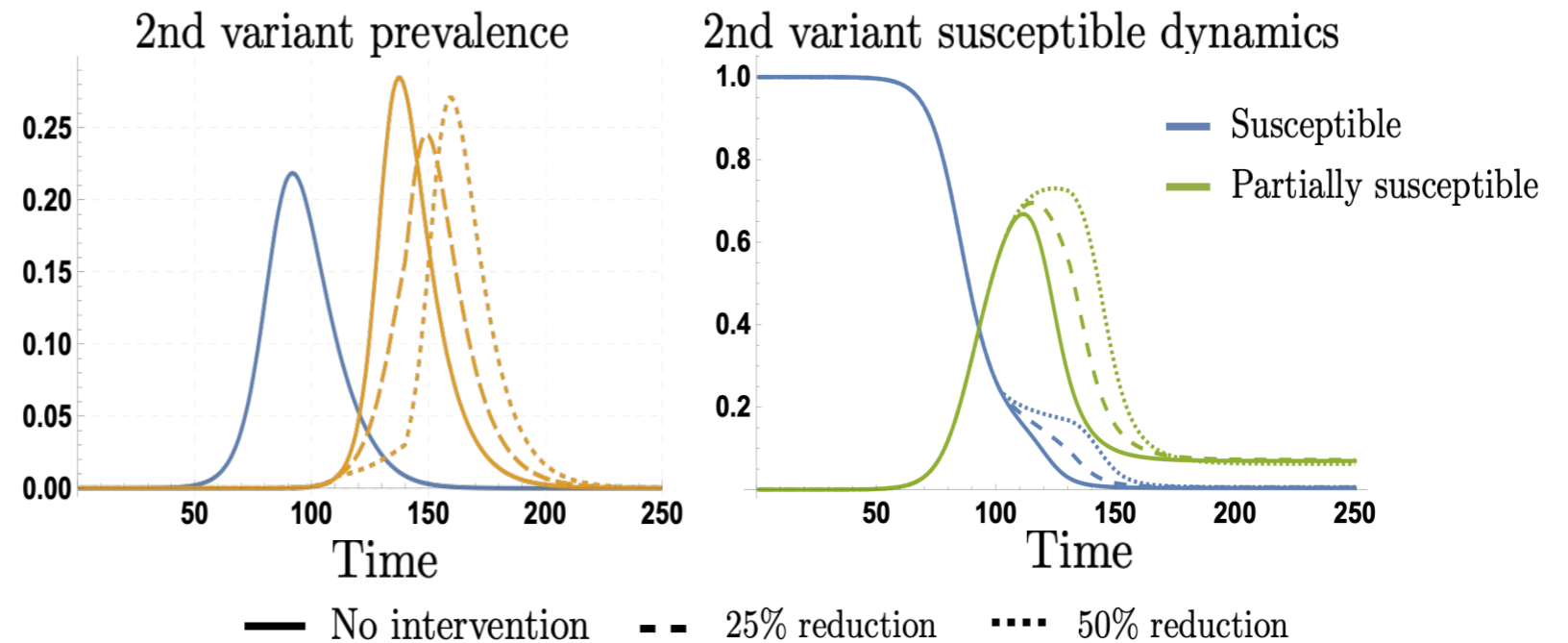


Three epidemiological metrics for interventions:

- Minimize the second variant's peak size
- Minimize the total peak size (both variants)
- Minimize the final epidemic size

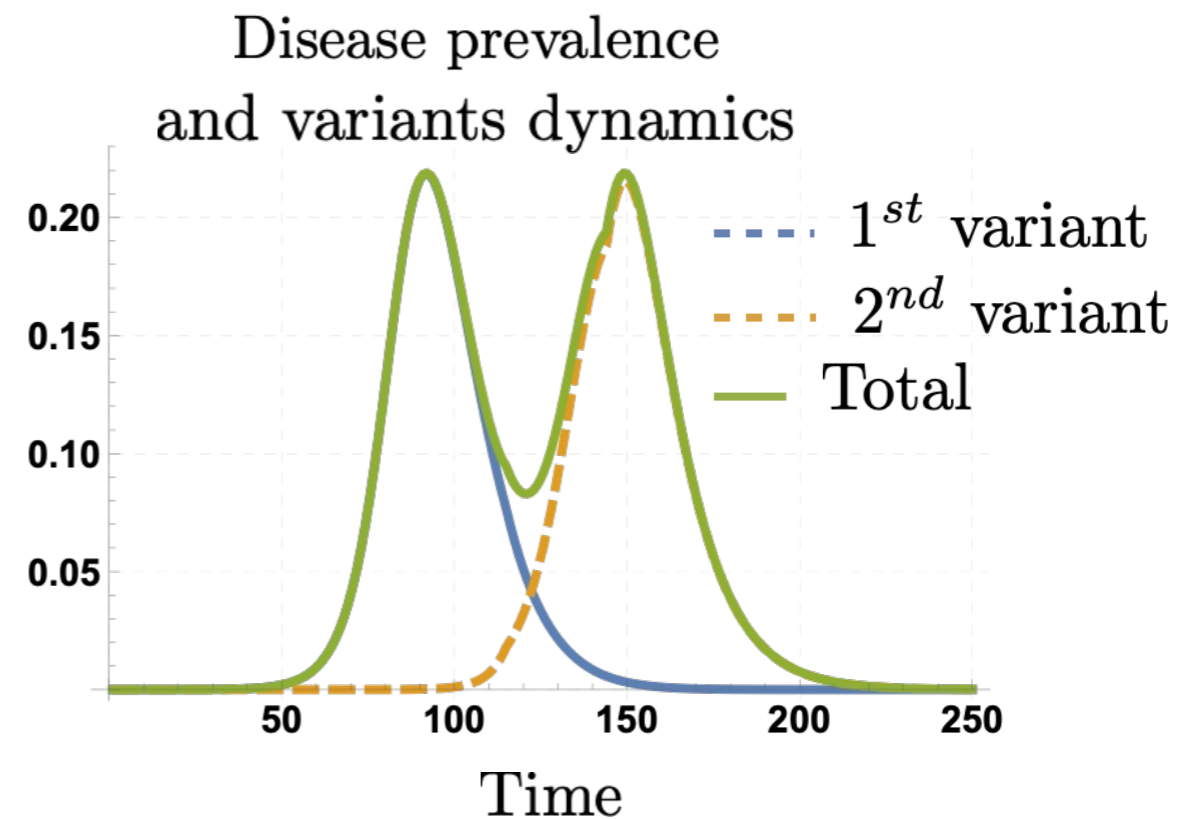
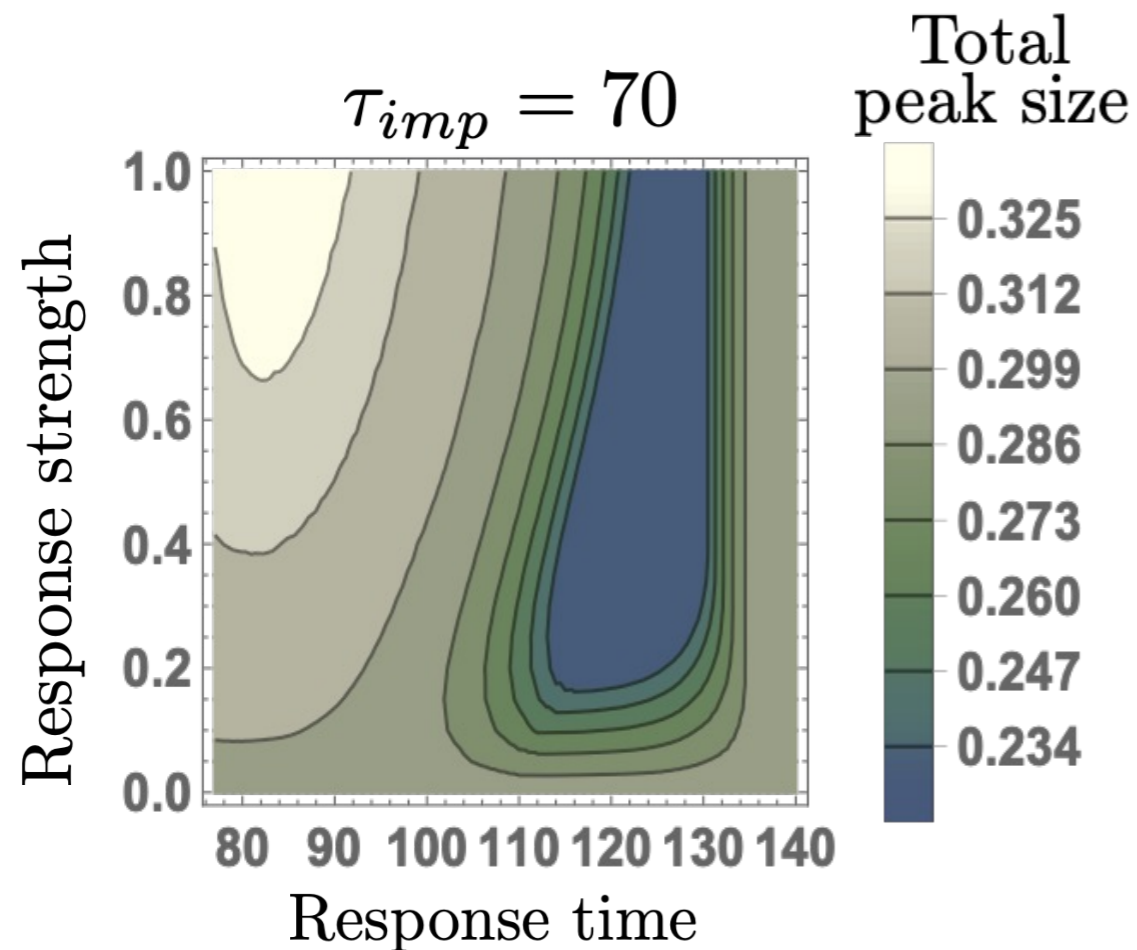
Minimizing the second variant's peak size

The second variant's peak size shows non-linear dynamics



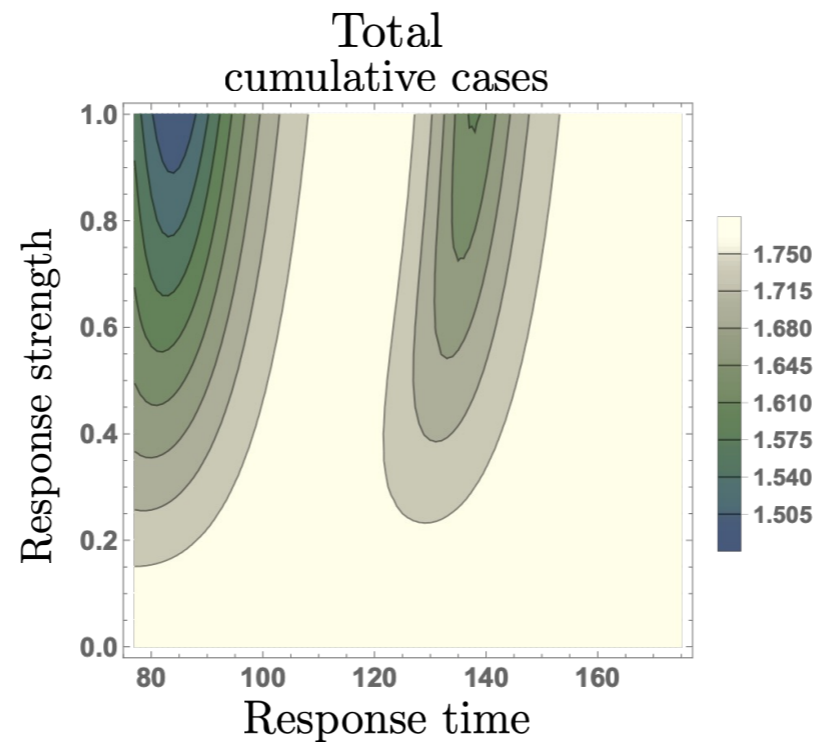
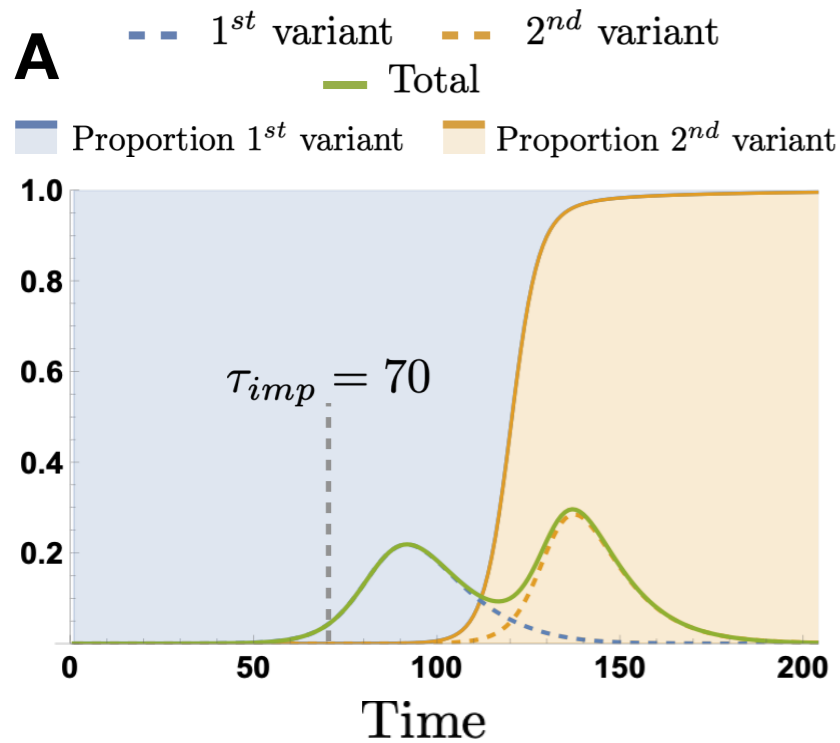
- **Effective responses requires to characterize the genomic landscape**

Minimizing the total disease prevalence

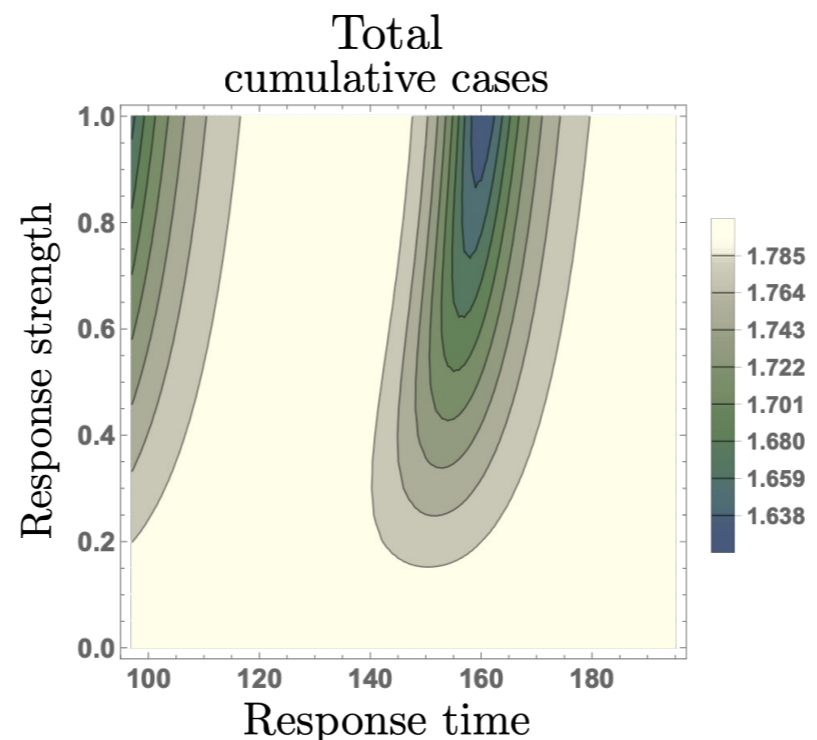
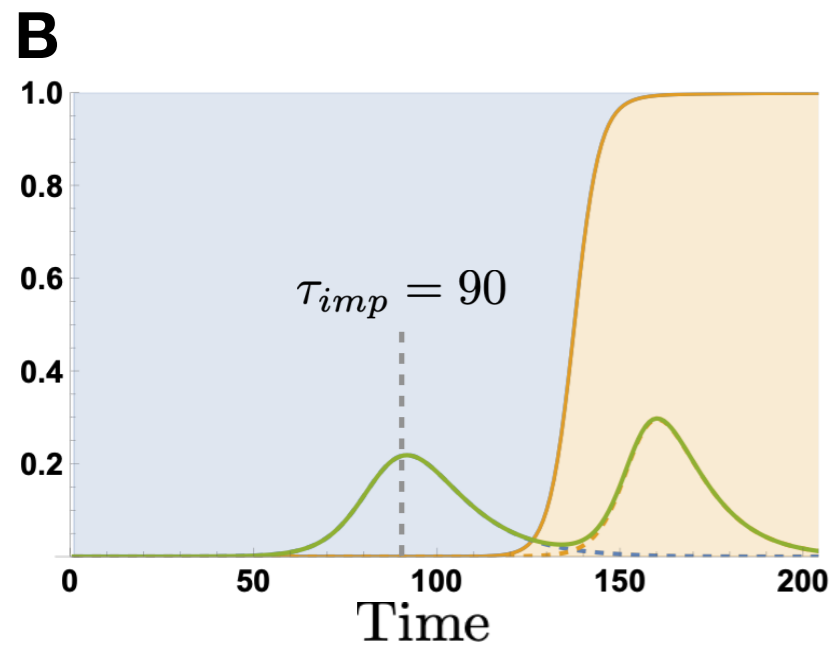


- Early importation leads to minimize the second variant's peak size
- Delayed importation leads to equalize both variant's peak size

Minimizing the final epidemic size



- Early importation:
early and strong interventions
are best



- Delayed importation:
delayed and strong interventions
are best

Limitations of the work

- Homogeneous mixing Susceptible-Infected-Recovered model
 - Complex within host dynamics
 - Spatial disease distribution
- Surveillance is exclusively driven by infectious cases
 - Geographical and temporal surveillance effort distribution and costs
 - Hospitalization and mortality rates must be incorporated
- Centralized simple interventions reducing the effective transmission
 - Pulsated and pharmaceutical interventions
 - Adaptive interventions
- Potential viral evolution

Key insights

- Tracking infection counts alone is not sufficient to assess public health interventions
- Characterization of the genomic landscape is critical to study complex disease dynamics
- Robust surveillance systems provide critical time for planning timely interventions

Ongoing work

- Incorporate adaptive human behavior
- Pathogen's mutation framework
- Multi-variant dynamics
- Complex intervention strategies

Thanks!

Comments and questions

- Model extensions to incorporate zoonotic and vector borne diseases
- Model extension to study potential trade-offs between surveillance and responses across different regions
- How could machine learning leverage the proposed framework?

