### Computationally Predicting and Characterizing the Immune Response to Viral Infections

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### **COVID-19 Disease Severity**

Of those who contract SARS-Cov-2, approximately:

- 33% show no symptoms
- 48% show mild symptoms
- 14% show severe symptoms
- 4% show critical symptoms
- 1% die

Factors influencing disease severity:

- Age, sex, comorbidities.
- Past exposure to similar viruses.
- Innate differences in an individual's immune system.



*Our topic of study:* 

Computational prediction of cellular response to pathogens, at molecular scale.

# Cellular Immune Response

First line of defense against viral infections (and cancer), present in all nucleated cells in body.

- Foreign proteins in cells are cleaved into smaller fragments (called peptides).
- These are transported to the cell surface.
- There they can bind to receptor molecules called MHC Class I molecules.
- Once bound, these become targets for the immune system: killer T-cells will come and kill off the infected (or cancerous) cells.



If effective, this cellular response means individual likely will show only mild symptoms. If not, infected cells become factories for the virus, and individual will have full-blown infection.

# **Understanding Cellular Immune Response**

#### Critical for:

- Understanding and predicting disease severity for a novel virus.
- Vaccine development.
- Understanding and predicting impacts of viral mutations.
- Cancer immunotherapy.
- Understanding auto-immune diseases.



Individual has up to 6 different MHC I molecules.

Diversity: 21,000 variants in human population.

Core problem:

Predict whether peptides associated with virus will bind to an individual's MHC I molecules.

# **Prior Work: Predicting MHC Peptide Binding**

Experimental:

• Crystallography, elution, mass spectroscopy, etc.

### Machine Learning & Neural Networks: NetMHC

- Training data: binding strength for known MHC I/peptide pairs.
- Given new peptide, inference based on similarity of amino acid sequence.





Limitation: *Spurious inferences* 

# **Prior Work: Molecular-Level Simulation**

Variety of general-purpose techniques:

- Molecular Dynamics
- Monte Carlo
- Simulated Annealing
- Molecular Docking



#### Limitation:

In principle accurate, but all techniques are computationally intensive. Hours (or days) of computing time per peptide/MHC molecule pair.

Scale: ~1 billion combinations. ~38,000 peptides for SARS-Cov-2 ~21,000 MHC I variants

### **Our Approach**

Simulate Molecular Mechanics

Highly Targeted:

- Start with peptide, *correctly aligned*, inside cleft of MHC molecule.
- Perform moves in *torsional space* to find optimal configuration

Deploy at scale:

- Develop parallel algorithms for GPUs.
- Deploy on cloud-computing infrastructure.



Turns 1 billion days of computing time to 1 million minutes of cloud-computing time.

# **Challenges: Simulating Molecular Mechanics**

Must infer structure:

- For SARS-Cov-2 (and nearly every emerging pathogen), peptide sequences is ewn, so ~38,000 brand new peptides.
- Of ~21,000 MHC I variants, only ~200 structures known from crystallography. *Most know variants are from Western Caucasian demographic*.



Will infer structure starting with similar peptide/MHC complexes and substituting amino acids that differ. *Refold and fit peptide into cleft in one step*.

### Impact

Can easily determine specific variants of MHC I molecules for an individual through HLA typing.

Predict:

- Disease severity for a new pathogen for *different* individuals.
- Disease severity for *different variants* of a virus for *different* individuals.
- Effectiveness of *different* vaccines for *different* variants of a virus for *different* individuals.

Also:

- Characterize good targets for cancer immunotherapy.
- Characterize good targets for auto-immune disease treatment.