



GOALS OF THE STUDY

- Use model to predict how Mtb interferes with T cells at the molecular level (antigen presentation) and cellular/tissue level (T cell, macrophage movement, interaction)
- Combine molecular and cellular/tissue level models to explore how they interact, look for emergent behavior

T CELLS IN MTB

- Tuberculosis (TB): Infectious disease caused by *Mycobacterium tuberculosis* (Mtb). One-third of the world's population is infected with Mtb, 2 million deaths/year¹
- Host immune cells (macrophages and T cells) form granulomas to contain the infection
- Only ~8% T cells in granulomas respond to Mtb²
- It's unknown why T cell response at site of infection is low



MULTI-SCALE MODELING

Tissue/Cellular Model

- GranSim^{3,4} captures discrete cellular dynamics between immune cells and Mtb leading to tissue scale outcomes
- Comprised of decision-making heuristics via a set of well-described interactions, calibrated using experimental data, and implemented in C++ code with Boost and FFTw libraries.





Figure 2: Example time point in GranSim and different types of agents on the grid.

Intracellular Model

- Chang et al 2005⁵ ODE model captures Mtb-mediated down-regulation of
- MHC II presentation of peptides in macrophages • Comprises 16 non-linear ODEs, solved in MatLab



Hypothesized process affected by Mtb H1: Antigen processing H2: MHC II maturation H3: MHC II peptide loading H4: MHC II transcription

Figure 3: ODE model schematic representing MHC II antigen presentation. Colored circles represent different processes hypothesized to effect antigen presentation by Mtb.

Using Multi-Scale Modeling to explore the effects of Mycobacterium tuberculosis on T cell responsiveness in granulomas

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MULTI-SCALE MODELING

Integrated Multi-Scale Model

- GranSim has the ability to insert ODEs into individual agents • Allows for observation of the interaction between tissue, cellular, and
- molecular scales
- Emergence of behaviors that arise from interactions between agents that would otherwise be impossible to know a priori⁶

Multi-scale model

GranSim: Tissue Scale GranSim: Cellular Scale

ODEs: Molecular Scale



Antigen Presentation ODEs



Figure 4: Model schematics of GranSim and ODEs and how they relate to each other as a multi-scale model

ANALYSIS OF THE MODEL

Latin Hypercube Sampling

- Method for generating a near-random sample of parameter values from a multidimensional distribution
- Used to explore the entire parameter space of a model • Parameters derived from LHS are then used to perform uncertainty analysis



Figure 5: One-dimensional and two-dimensional projections of 3-dimensional Latin hypercube.⁷

Partial Rank Correlation Coefficient

- Measures how the uncertainty of inputs impacts model outputs (sensitivity) Rank transformation allows consideration of non-linear relationships • Allows for inter and intra compartmental sensitivity analysis by varying
- parameters in ODEs, GranSim, or both

Rate constant for antigen processing (kdegAg) Rate constant for degradation of CIITA mRNA (kdegM) Rate of replenishment of MHC-derived self-peptides (kdegMhc) Rate constant for degradation of MHC II mRNA (kdegMrna) Dissociation rate constant of the IFN-g receptor-ligand complex (koffIfng) Dissociation rate constant of peptide-MHC complexes (koffMhc) Association rate constant of the IFN-g receptor-ligand complex (konIfng) Rate constant of MHC II transport from endosomes to surface (kout) Average rate of pinocytic uptake (kpino) Rate constant for antigen degradation (ktoLys) MHC II translation scaling factor (mhcScale) Scaling factor for CIITA transcription (mscale)

Figure 6: Partial rank correlation coefficient sensitivity analysis of ODE model parameters on Mtb peptide-MHC II complexes.

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ANALYSIS

1. MHC II transcription and translation are most influential during first 50 days of infection



2. Mtb antigen uptake and processing are most influential through first 100+ days of infection

- 25, tapers 100+ days post infection
- peptide-MHC II complexes



DISSCUSSION / CONCLUSIONS

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RESULTS OF UNCERTAINTY & SENSITIVITY

• The molecular dynamics change when placed inside agents in GranSim • The effect on Mtb peptide-MHC II complexes is delayed until 30 days post infection in GranSim, when T cells are deployed

• Increases seen most with CIITA transcription, degradation rate of CIITA mRNA, and degradation rate of MHC class II mRNA

Decreases seen with dissociation rate of peptide-MHC complexes

• In GranSim, influence on Mtb peptide-MHC II complexes peaks around day

Increases seen with pinocytic uptake and antigen processing

• Decreases seen with antigen degradation and dissociation rate of Mtb

• Intervention in Mtb infection could possibly consist of finding ways to decrease Mtb antigen degradation and lowering dissociation rates of Mtb peptide-MHC II complexes

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