

In silico imaging of the dynamics of antigen presentation and cellular activation in a lymph node

Thomas Riggs^a, Adrienne Walts^a, Nicolas Perry^a, Mark J. Miller^c, Joanne Flynn^d, Jennifer J. Linderman^b and Denise Kirschner^a
 Dept of Microbiology and Immunology^a, University of Michigan Medical School; Department of Chemical Engineering^b, University of Michigan, Ann Arbor, MI
 Dept of Pathology and Immunology^c, Washington University School of Medicine, St. Louis, MO
 Dept of Molecular Genetics and Biochemistry^d, University of Pittsburgh School of Medicine, Pittsburgh, PA



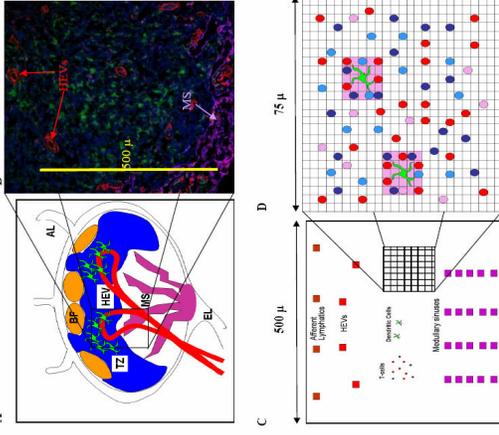
Abstract

The immune system and its processes of antigen presentation in particular, engenders events that occur at multiple length and time scales. Despite a wealth of information in the biological literature regarding each of these scales, no single representation synthesizes this information into a model that incorporates antigen presentation into the overall adaptive immune response. We have begun an approach for integrating information over relevant biological and temporal scales to generate such a representation for MHC class II-mediated antigen presentation. Here we focus on one of the single-scale models to give sufficient details of our approach. In particular, we describe the dynamics of antigen presentation within a LN by individual cellular interactions (e.g., lymphocyte motility, MHC class II-mediated antigen presentation) and how the physical structure of the LN enhances the relevant cell-cell contacts and the by which random motion vs. chemotaxis might enhance stochastic scanning by DCs of many T cells to find a rare cognate match.

Introduction

An appropriate immune response requires a complex interaction of antigens, antigen-presenting cells (APCs) and T-cell populations that modulate, regulate and effectuate an adaptive immune response. Most antigen presentation events occur in secondary lymphoid tissues (e.g., lymph nodes (LN)) and are initiated by the interaction of APCs and T cells specific for the antigen being presented. Our goals are to create a model system of a LN that will recapitulate antigen presentation dynamics observed through current imaging techniques and enable understanding of the anatomical and cellular motion features which contribute to efficient contact between T-cells and DCs and optimal production of non-naïve CD4+ and CD8+ T cells.

Figure 1



Anatomical basis for our model of the T-zone within a lymph node (LN). (A) Schematic of the LN, focusing on the T-zone, where dendritic cells entering from the afferent lymphatics (AL) encounter T cells that are entering via the high endothelial venules (HEV); the latter exit via medullary sinusoids to the efferent lymphatics. (B) Corresponding photomicrograph of the T-zone showing the distance from HEV to MS. (C) Our model lattice representation of the LN volume. (D) Smaller section is shown of two DCs with their surrounding 'sweep area' around each DC. T4 = CD4+ T cells are shown in shades of red, CD8+ are shades of blue, DCs are green.

Methods

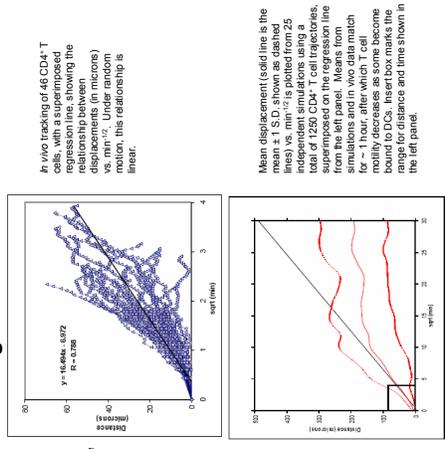
Uncertainty and Sensitivity Analysis

Statistical techniques of Latin hypercube sampling (LHS) and partial rank correlation (PRC) guided our understanding as to how and what extent variability in parameter values affects infection outcome. The t-test and its p-value (based on multiple hypothesis testing) were used to infer statistical significance.

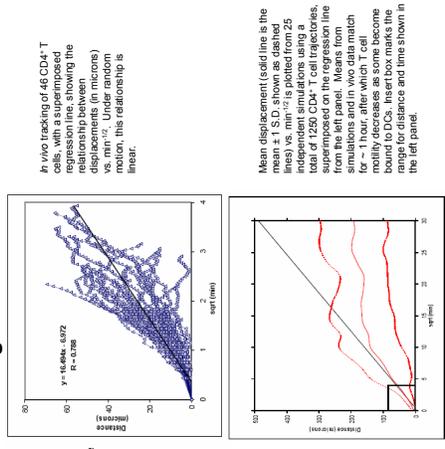
Parameter Range for Uncertainty/Sensitivity analyses

| Parameter | Min. | Max. | Usual |
|---------------------------------------|------|------|-------|
| Probability DC binds cognate CD4 (%) | 50 | 100 | 75 |
| Probability CD4 activated by MDC (%) | 50 | 100 | 95 |
| Prob. recruitment of naïve CD4 (%) | 5 | 11 | 8 |
| Number divisions for activated CD4 | 2 | 6 | 4 |
| Doubling period CD4 (hrs) | 7 | 9 | 8 |
| Effector CD4 lifespan (hrs) | 36 | 60 | 48 |
| Prob. MDC leaves by effector CD4 (%) | 50 | 100 | 95 |
| Prob. non-naïve T cell matures DC (%) | 50 | 100 | 95 |
| MDC lifespan (hrs) | 36 | 60 | 48 |
| LDC lifespan (hrs) (left from dist) | 30 | 72 | 38 |
| Probability LDC binds cognate CD8 (%) | 50 | 100 | 75 |
| Probability DC activated by LDC (%) | 30 | 100 | 95 |
| Prob. recruitment of naïve CD8 (%) | 2 | 6 | 4 |
| CD8 doubling period (hrs) | 7 | 9 | 8 |
| Effector CD8 lifespan (hrs) | 36 | 60 | 48 |
| Probability of straight path (%) | 50 | 95 | 90 |

Matching in vivo cell data to Simulations



Magnitude and time course of CD4+ and CD8+ T cell production



Results

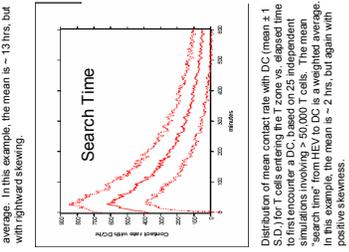
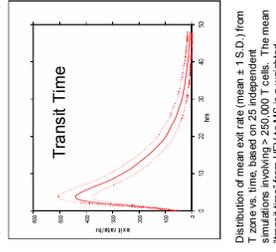


Table 1: Variation in HEV vs. AL Placement Affects T cell search time, transit time, number of contacts per DC and output of non-naïve T cells

| Parameter of HEV vs. AL | HEV same level with AL | AL below HEV |
|------------------------------|------------------------|------------------|
| Transfret time (hrs) | 13.4 ± 0.14 | 13.1 ± 0.17 NS |
| Search time (hrs) | 2.80 ± 0.23 | 2.47 ± 0.17 NS |
| Mean contact rate per T cell | 16.09 ± 2.90 | 11.96 ± 1.70 *** |
| Output of T cells | | 0.78 ± 0.10 *** |
| CD4+ T cells | 15.30 ± 3.00 | 15.00 ± 3.00 NS |
| CD8+ T cells | 11.50 ± 2.00 | 11.60 ± 2.00 NS |
| CD4+ T cells | 210 ± 30 | 224 ± 30 NS |
| CD8+ T cells | 46 ± 15 | 56 ± 15 *** |

All comparisons are vs. original placement (1) of HEV vs. AL. NS = Not Significant. *** = p < 0.05; ** = p < 0.01; * = p < 0.05. HEV to MS was constant. 25 independent simulations were performed. Mean contact rate per T cell means that were each based on 500 DCs, > 250,000 individual transit times and > 50,000 search times.

Conclusions

An agent based model of LN dynamics can simulate the known patterns of cellular motility, velocity and LN transit times and represent the structural anatomy of a LN. The magnitude and temporal sequence of the cellular immune response can be recapitulated. Anatomic placement of HEVs, afferent and efferent lymphatics can be varied in simulations. Agent-based models can capture the stochasticity of encounters between rare cognate T cells and dendritic cells in the LN. Varying degrees of complexity of interaction can be implemented in a rules-based model; hypotheses can be tested in silico to help in design of experiments. Multiple parameters can be varied to assess their association with the immune response. This modeling approach is adaptable to implement new rules of interaction of dendritic cells with T-cells as they become elucidated by experimental data.

Future directions

- Adapt model to immune response agents
- Adapt model to specific infectious agents
- Contrast chemotaxis with random T cell motion
- Test effect of drugs that change recruitment to LN or exit from LN

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