# Spatially explicit model of the efficiency of the lymphocyte diaspora in influenza-infected lung reveals thresholds on chemokine directed migration

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#### Abstract

During the primary immune response, clearance of influenza in the lung requires the homing of activated CD8 T cells from regional lymph nodes to infected foci comprising a small fraction of the total lung. T cell navigation towards infected foci is undirected but emigration into tissue is directed by local cytokine and chemokine production from infected epithelial cells. Weuse a spatially explicit agent-based model (ABM) to examine the efficiency of the recruitment of activated CD8 T cells by local chemokine signaling.

To determine local chemokine production, avian H5N1, seasonal H1N1, and 2009 pandemic influenza strains were used *in vitro* to induce the secretion of CXCL10 (IP-10) and CCL5 (RANTES) in human bronchial epithelial cells. These data were modeled to estimate chemokine production rates. The estimates were combined in the spatial model to test viral variation of T cell recruitment in the lungs.

The spatial nature of the ABM model reveals unique challenges to T cell recruitment not appreciated in standard differential equation models. Infected cells can become isolated in expanding plaques, impeding T cell search, even though the T cells could directionally migrate to low levels of chemokine. This spatially explicit model describes efficient T cell recruitment to small infected foci in the lung, but as foci become large T cell search becomes inefficient, thus emphasizing the importance of enhanced control early in infection. A key limitation imposed on T cells is illustrated by their failure to clear the pandemic H1N1 virus after day 6, when T cells became inefficient in finding infected cells in large foci. We conclude that viral dynamics, rather than chemokine dynamics, dominate the efficiency of T cell recruitment in determining whether the influenza infection is controlled.

### **Author Summary**

Clearance of influenza from the lung depends critically on the efficiency with which T cells travel from the lymph node to the site of infection. However, T cells in the lung must navigate the large branching bronchial network. This maze-like structure is a challenge for T cell search, and the vast majority of T cells exit the lung without encountering infection. Infected cells produce cytokines and chemokines, which signal T cells to exit the vasculature and enter tissue. We developed an agent-based model to quantify how these molecular signals affect search. Our experimental data show that two chemokines, IP-10 and RANTES, are secreted by infected human tracheal epithelial cells. We incorporated this data in our spatially explicit model, which allowed us to identify challenges to the search process not obvious in conventional differential equation models. We ran three sets of simulations, using parameters from seasonal influenza, avian influenza, and 2009 pandemic influenza. The simulated immune response controls and eradicates the first two infections in 10 days but is unable to contain the highly virulent pandemic strain due to the rapidly expanding plaque size. Interventions are unlikely to improve this chemokine-directed T cell search process in the lung but vaccination against more virulent strains will increase the number of searching cells earlier in the response.

#### Introduction

The adaptive immune response induced during acute influenza A infection is an integrated complex of combination of defense mechanisms. Understanding the behavior and interactions of each cellular component may lead to improved vaccines and strategies to control the infection and immunopathology. The innate immune response is critical for early control of viral replication, followed by antibody-mediated viral neutralization [1–3]. Complete resolution of influenza pneumonia depends on antigen-specific CD8+ T effector cells [4,5]. The murine model has been critical in the reductive analysis of each phase and component, permitting temporally explicit descriptions of the induction phase in the regional lymph node [6–8] where recruitment of clonal precursor CD8 T cells is highly efficient [9]. On approximately day 5 after infection, activated CD8 T cells are released from the secondary lymphoid tissue [10] and distributed throughout the body in a process known as the lymphocyte diaspora [11].

CD8 T cells conduct two 'searches' in two different tissue environments, first to encounter antigenloaded dendritic cells in the lymph node, and second to encounter local inflammatory signals in infected lung. Search problems in the lymph node have been simulated using live cell imaging data to provides reliable parameters of cell movement [12]. The T precursor cell must encounter on average hundreds of irrelevant cells before it contacts the loaded DC, but the distance traveled is relatively short, and the direction is likely guided by chemokine signals.

The second search, i.e. the process of recruiting activated CD8 effector cells to infected sites in the lung, is different and not as well understood. In contrast to the dense lymph node volume, lung infection dynamics occur on the surface of epithelial cells, the size of the tissue layer being comparable to the size of a tennis court. Initial localization and extravasation may depend on a number of inflammatory signals but viral antigen does not appear to induce recruitment into this tissue [13]. Moreover, the T cells leaving the lymph node are released into the very large systemic vascular 'space' with no initial guiding signals. Effector cells must localize to sites of viral replication [4], but it is not clear whether cells passing through uninfected lung tissue leave through the pulmonary vein or exit the capillary bed and leave the lung through lymphatic channels. Chemotactic proteins play a major role in recruitment in both acute infections and chronic inflammatory diseases [14–18] where recruited T cells mediate either protection or immunopathology. Infected epithelial cells secrete chemokines [19] and contact with CD8 T cells also triggers secretion [20]. Activated CD8 T effector cells upregulate chemotactic protein receptors, particularly CXCR3 and CCR5 [21,22], responding to their primary ligands CXCL10 (IP-10) and CCL5 (RANTES) respectively.

The consequences of chemotactic ligand-receptor interactions, however, are complex [22,23] and variable in the models studied. For example, in the lymphochoriomeningitis virus model the CXCR3 receptor mediates T cell recruitment to infected brain and subsequent immunopathology [24,25], while the CCR5 mediates the opposite effect [26]. In contrast, in the West Nile virus model [27] and the dengue model [28], deficiency of IP-10 reduced T cell recruitment to the brain resulting in higher viral burden and increased mortality. In the herpesvirus model, IP-10 was critical in T cell recruitment and disease control in the HSV-2-infected brain [29,30]. In the parainfluenza virus model, CXCR3 receptor is critical in CD4+ T cell migration to the lung [31]. In the influenza A model, initial studies with chemokine receptor knock-out mice obtained mixed results with respect to changing the course of disease [32,33], concluding that redundancy in chemokine signals may confound interpretation. The CXCR3 receptor mediated T cell localization and spared the increased mortality of CCR5 deficiency, but viral clearance was not altered [34]. The CXCR3 receptor also mediates the balance between effector versus memory cell differentiation among

recruited CD8 T cells in the lung [35].

Although leukocytes exhibit directional behavior to chemokines [36, 37], CD8 T cells have not yet been shown to climb chemokine gradients. Following arrival at the infected site, CD8 T cells secrete cytokines when mononuclear cells are present but in their absence kill infected epithelial cells only by contact cytolysis [38]. Finally the exit of T cells from tissue is tightly regulated to keep effector cells from leaving or dying prematurely [39, 40] but effective life-span is unknown.

There are many mathematical models of the entire adaptive response [41–46], filling gaps with reasonable assumptions of cellular behavior in the absence of experimental support. Mathematical and computational models can examine each event in detail and narrow the range of possible behaviors. Here we examine the CD8 T cell diaspora and recruitment to infected lung in detail. Models using ordinary differential equations are able to predict events with temporal fidelity, but lack the ability to examine the spatial detail [47,48]. Activated CD8 T cells searching for and homing into infected tissue do so in a spatially complex environment. We therefore used agent-based modeling (ABM) to represent the physical environment of the searching T cell. In the model, CD8 T cells travel in vascular and lymphatic channels during part of their search, and respond to localized inflammatory signals to achieve compartmentalized cell-cell interactions.

We ask how small foci of infected tissue, scattered through a very large space of uninfected tissue, can attract and retain limited numbers of activated CD8 T cells. Our model does not consider important features of tissue immunity such as T cell proliferation, resident T cells, and regulatory interactions with other cell phenotypes [49]. The model does not aim to predict control of viral replication but only to compare simulated outcomes among three influenza strains arising from the efficacy of the chemotactic signals.

Previously we developed an agent-based model of the lymphocyte diaspora focusing on the production and diffusion of a generic inflammatory signal without experimental data to calibrate the model [50]. Here we refine that model and incorporate chemotactic protein data from in vitro cultures of human bronchial epithelial cells infected with three strains of influenza virus with different replication rates [51] to focus on two features of searching in the large lung space. First, functional thresholds of the chemotactic proteins were surprisingly low in our model. Second there appears to be a threshold size of the infected focus above which the T cell search becomes inefficient.

### Materials and Methods

#### Computational Modeling

Computational modeling used CyCells [52], a modeling platform for two- or three-dimensional agent-based simulations of the immune response. A simplified model of CD8 T cell activation and recirculation (Fig. 1) was implemented in CyCells (Fig. 2 and 4), and simulations measured efficiency of infection clearance for three different viruses. The lung was represented as a two-dimensional sheet of healthy epithelial cells (S1.4). Vasculature in a mouse is a binary tree with fourteen levels of bifurcation with the origin at the bronchial lymph node complex. Because we assume activated T cells descend at random into branches of the vascular network, T cells are introduced uniformly at random across the entire lung surface (Fig. 1). If T cells detect cytokine signal on the local endothelium they remain in tissue and follow the chemotactic gradient to the site of infection. T cells that do not encounter cytokine recirculate to the lymph node. When a T cell contacts an infected epithelial cell at the site of infection it induces apoptosis. Unimpeded, the infected secretes virus for 16.7 hours and then dies.

The simulation begins when a single cell in the center of the tissue is infected. After the eclipse phase (incubation), the infected cell begins secreting virus and chemokine according to the virus strain and chemokine type as determined by simulations reported in S1. Virus diffuses locally, infecting nearby cells, and continuing the cycle. Chemokine diffuses from secreting cells according to kinetics described below, creating a circle of stimulation around the infected region. After a five day delay to simulate lymph node stimulation and T cell proliferation, activated T cells begin to exit the lymph node at a steady rate and circulate through the vasculature to the tissue. Further details of the CyCells modeling environment are described in S1.3.

Computational limitations force us to implement our model on the scale of a mouse. Parameters affected by this choice include the T cell production rate, T cell circulation time, and the total size of the lung (Table 2). The surface area of a mouse lung is approximately  $100cm^2$ .

#### **Model Definition**

In the model (Fig. 2), epithelial cells are stationary and are described by one of five sequential states: healthy, virus-incubating, virus-expressing, apoptotic, and dead. Healthy cells remain unchanged unless infected by virus. Once infected, the cell transitions from incubating to expressing after a 10 hour

incubation delay. *Expressing* cells secrete virus and chemokine for a fixed 16.7 hours and then die. *Expressing* cells initiate *apoptosis* sooner if they are contacted by activated T cells. Apoptoic cells continue to secrete virus and chemokine and then die after one hour. *Dead* cells take up space and do not regenerate over the course of an infection.

T cells are described by two states. Circulating T cells begin to emerge from the lymph node at five days post infection and arrive at a random location on the lung's surface after six seconds of travel. If chemokine is not encountered, the T cell remains a circulating cell, returning to a new location in the lung after another six seconds. If a circulating cell encounters chemokine, it converts to chemotaxing and begins following the chemotactic gradient to the infected region. Circulating T cells decay exponentially with an average lifespan of three days. Chemotaxing T cells move continuously up the gradient through the two dimensional lung endothelium, inducing apoptosis in the expressing epithelial cells. Chemotaxing cells decay exponentially with an average lifespan of two hours.

The model contains two kinds of particles: virus and chemokine. Both are produced at constant rates by *expressing* epithelial cells. Virus diffuses through the lung tissue, infecting healthy cells at a rate according to the local virus concentration. Chemokine diffuses across the tissue but has no direct effect beyond activating T cells. Both particle types decay exponentially. IgM is modeled by increasing the viral decay rate by a factor of ten after the third day.

Parameters that are consistent between every model are shown in Table 2. Strain-specific values are shown in Table 3. Further details regarding the model definition are included in S1.4.

#### Materials

Chemokine secretion: Epithelial cell culture and supernatant collection was performed as described [51]. Briefly, undifferentiated human tracheal epithelial cells (University of Miami) were cultured for 4 weeks to achieve fully differentiated confluent monolayers on collagen-coated transwell inserts, or commercial differentiated human bronchial epithelial cells (EpiAirway Tissue, MatTek Corp., Ashland, MA) used immediately upon receipt, were infected at an MOI of 0.01 with either seasonal H1N1 virus A/New Caledonia/20/99 (sH1N1), the 2009 H1N1 pandemic strain A/California/04/09 (pH1N1), or avian H5N1 virus A/Hong Kong/483/97 (aH5N1) derived from a fatal human infection. Basal media was collected from previously undisturbed triplicate or quadruplicate wells at 0, 6, 10, 12, 16, 20, 24, 30, 36, 42, 48, and 72 hours after infection, and stored at -80C until assay. Subsequently, apical fluid for virus secretion

was collected before and after treatment of the monolayer with protease (Pronase, Sigma) to optimize the collection of infectious virus [51]. Quantitative viral culture was performed by standard plaque assay. Quantitative chemokine levels were performed in 30  $\mu$ L aliquots for a panel of chemokines (IL-8, MCP-1, MIP-1 $\alpha$ , MIP-1 $\beta$ , RANTES, IP-10, eotaxin) and cytokines (interferon-gamma, IL-1 $\alpha$ , IL-1 $\beta$ , IL-1RA, IL-2, IL-3, IL-4, IL-6, IL-10, IL-12p40, IL-15, IL-17, TNF $\alpha$ ) (Luminex Assay, Luminex Corp.) and reported as ng/mL basal media sampled from a total volume of 4 mL. Only IP-10, RANTES, and TNF $\alpha$  showed increases in production. TNF activity as not incorporated in the model. Data for other chemokines and cytokines is not shown.

#### Results

#### Chemokine production

To provide estimates of chemokine concentrations and secretion rates present in lung tissue, chemokine levels were measured at 4-6h intervals during the first 48 h of infection in wells containing approximately one million human bronchial epithelial cells (Fig. 3). The dynamic viral loads at these intervals in these cultures infected with seasonal H1N1 virus, pandemic H1N1 virus, and avian H5N1 virus have been reported previously [51]. IP-10 concentration increases were observed by 8h post-infection (p.i.), and RANTES by 16h p.i.. To estimate per-cell production rates, we extended the delay differential equation (DDE) model of Ref. [51] to represent chemokine production from infected cells (Eq. S1). Model fits (Table 3) were computed for three strains using Matlab's nlinfit function (Levenberg-Marquardt algorithm). The resulting chemokine production values were used in the CyCells ABM by combining IP-10 and RANTES into one aggregate rate but also reported separately (Fig. S2). Best-fit expression rates were similar for all strains except for significantly higher RANTES production in aH5N1. There was no positive correlation between viral production and induced chemokine production across the three strains.

#### Stochastic modeling effects

Unlike ordinary differential equation (ODE) and DDE models, which are deterministic, stochastic models such as CyCells can produce different results on different runs (Fig. 4). To test the strength of this effect, we ran each model fifty times using the default parameters given in Table 2.  $R^2$  values were calculated for individual runs versus the average of all runs. The  $R^2$  ( $\pm$  one standard deviation) for avian, seasonal,

and pandemic influenza are:  $(0.957 \pm 0.150)$ ,  $(0.992 \pm 0.013)$ , and  $(0.998 \pm .002)$  respectively (i.e. there was little variation between runs). The three strains show different levels of replication as found in [51].

Overlaying multiple runs on a single plot reveals a growth rate decline at 4 days post infection, reflecting the addition of IgM. Subsequently, for each infection the number of infected cells declines quickly after day five post infection due to the CD8 T cell response. aH5N1 is cleared completely, sH1N1 is more slowly cleared, and pH1N1 clearance at day 6 post infection is halted as replication overtakes clearance.

The rapid production of the pH1N1 virus prevents the immune response from containing the infection (Fig. 4C). Because pH1N1 replicates more rapidly, the number of infected cells outstrips the ability of the T cells to find them. Spatial effects reveal the reasons for this in simulation images and video (Fig. 5, Video S4).

#### Spatial effects

Spatial effects of viral replication and chemokine diffusion play an important role in both the spreading and clearing of infections, imaged as plaque growth over time (Fig. 5). Free virus particles diffuse from virus secreting cells and infect healthy cells. Chemokine produced by infected cells attracts T cells to the infected cells. Although larger size of virus particles causes virus to diffuse much more slowly than chemokine molecules, chemokine decays more quickly. These countervailing effects result in similar spatial profiles for the two particle types (Fig. 5). Until day 4 the plaque is dominated by active (incubating and secreting) cells, whereas there is a smaller percentage of dead cells. Over time, cells in the plaque's interior die, and active cells form a decreasing proportion of the plaque. T cells arrive at day 5 and begin killing the virus-secreting cells. By day 6 many expressing cells have been eliminated and the plaque is dominated by dead cells. Continuous change in plaque size and T cell number is demonstrated for seasonal H1N1 virus (Videos S1-S3).

In the aH5N1 infection, the plaque is dense, allowing T cells to find secreting cells easily, and infection is eliminated. However, secreting cells were not eliminated in either sH1N1 or pH1N1 simulations (Fig. 4). Secreting cells accounted for at most 10% of the active cell population and less than 1% of the total plaque at 6 days p.i. (Fig. 6). T cells still accumulate, but they arrive at a lower rate than the number of infected cells that are added to the plaque. Further, the regions of concentrated chemokine diffuse slowly, causing them to be located behind the expanding infected cell and virus locations. It takes time for newly

secreting cells to produce chemokine while pockets of high chemokine density are slow to decay. Thus, T cells can fail to locate new regions of infected cells within the plaque. Taken together, the delayed response and low proportion of virus secreting cells prevent T cells from completely clearing the infection. Furthermore, increasing chemokine production fails to guide T cells to the actively secreting epithelial cells in the runaway pH1N1 infection (data not shown) suggesting that the effect of high viral production rates dominate increased chemokine production rates.

#### Temporal effects

The notion that T cell can't find infected cells in large plaques was examined in more detail by comparing the ratios of T cells to infected cells in Fig. 6. T cells are unable to kill virus incubating epithelial cells that have not yet presented antigen. At day 5.5, the ratio of antigen presenting secreting epithelial cells to the total plaque size is high (Fig. 5-6). By day 7, this ratio is very small for both the pH1N1 and sH1N1 strains, making it much more difficult for searching T cells to find the secreting cells. The high replication rate of pH1N1 exaggerates this effect (Fig. 6C). Video S4 illustrates this event.

Because there is an interval between first secreted virus and the first arriving T cell, infection control cannot be completed by CD8 T cells alone. An epithelial cell infected with pH1N1 produces new virus at the rate of 5.08e-3 PFU/s [51]. Assuming that secretion continues for one hour after apoptosis is initiated, the best a T cell could do is limit production to 18 new viral particles. Thus, T cells alone cannot contain the pH1N1 infection. In contrast, in a sH1N1 virus infection, T cells would limit a single infected cell to producting 1.4 viral particles in the one-hour window. Avian H5N1 virus-secreting cells produce only 0.2 viral particles in the interval after induced apoptosis. Viral clearance depends on the interaction of the rate of apoptosis, viral decay, viral production, and the viral infection probability.

### Discussion

Our aim was to examine the spatially determined restrictions on the efficiency of the T cell search. Due to the absence of critical data, the inability to image the search in the lung (as was done in the lymph node), and the need for a tractable computation, the model simplified the adaptive immune response. We focused instead on comparing the behavior of thee influenza viruses.

#### Modeling Methodology

The model makes a number of simplifications to and deletions of elements in the innate and adaptive immune response, allowing us to build a tractable model where data was available. Antigen presentation and T cell clonal expansion in secondary lymphoid organs is represented solely by the constant rate of emigration of activated CD8 T cells from regional lymph nodes. Virus-specific strain replication rates are represented as constant rates, and virus clearance is also constant. The contribution of IgM antibody clearing free virus is represented as a constant rate, and the class switch to higher affinity antibody mediated by CD4+ T cells is not represented. All of these rates may in fact be time-variable as indicated by data-fitting models [53]. Immigration and contributions of virus-nonspecific immune cells such as macrophages and/or dendritic cells are not represented in our model. Finally, the proliferation of activated T cells in lung tissue is not represented, but is thought to be crucial to the control of lung infection [41]. Thus, our model is not intended to predict clearance of virus from the lung, but to examine the response features that permit CD8 T cells to sense and contact infected target cells.

The use of an ABM has certain advantages over a spatially homogeneous ODE model. An ODE model assumes that any virus particle is capable of infecting any healthy cell. Figure 5 shows that this is clearly not the case for viral adhesion and entry. In fact, most free virus exists on top of infected cells that are no longer candidates for viral binding and fusion. ODE models account for this discrepancy by lowering rates of infection by a constant amount, but this assumes that the proportion of unsuitable virions will always be the same. This is limiting as can be seen in Figure 5 where the early infection has a higher proportion of virus overlapping healthy cells.

Our ABM reveals spatial patterns and dynamics that are absent in differential equation models and difficult or impossible to observe in *in vitro* and *in vivo* systems discussed earlier. Because CD8 T cells find infected cells by presumably climbing a chemokine gradient, we see T cell clustering at local maxima of chemokine concentration, a possible explanation for why T cells do not increase in effectiveness as their numbers increase. We also see that T cell clustering persists after all virus-secreting cells have been eliminated. The local chemokine maxima takes time to diffuse and decay so that T cells can climb the gradient to a new maximum. Finally, we can see that infected cells are more disperse as infection size grows. Because T cells are clustered they cannot cover the increasing plaque effectively. These spatial observations provide explanations for the pH1N1 resurgence that would be obscured without the visualization tools provided by the ABM (*in vivo* pH1N1 infection did not usually exhibit a resurgence

due to features of the immune response not modeled, specifically antibody and local T cell proliferation). The behavior of searching T cells described in this paper can enhance future global immune response models.

#### Chemokine Directed T Cell Search

The actual quantitative chemokine environment in the infected lung is difficult to estimate for a math model. Blood levels documented in virulent influenza infections [54] may not reflect lung tissue concentrations. Dynamic chemokine concentrations secreted by bronchial epithelial cells in vitro depend on infection intensity and cell maturation state [19,51,55,56] but may better approximate tissue levels. Interestingly, the attenuation of the type I interferon response by H5N1 viruses is not associated with attenuation of chemokine secretion in our results and in others [57]. The model did not incorporate the potential contributions from other chemokines such as CXCL8/IL-8 detected in bronchial cell cultures [58,59], nor did the model account for chemokines secreted by immigrant macrophages [60] and amplification of epithelial cell secretion by CD8 T cells [20].

A key determinant in the efficiency of chemokine-directed T cell migration towards virus-secreting epithelial cells is the communication distance, defined by the threshold of sufficient chemokine signal required to induce directed motion of the cell up the chemical gradient [61]. The diameter of this gradient generated by a single cell is a function of production rate, decay rate, protein diffusion and the sensitivity threshold. For the threshold of 100 ng/mL and maximal levels of concentration in tissue of 10,000 ng/mL, we calculated the effective communication distance to be approximately 100 microns in our model by simulating a single chemokine producing cell and observing the radius of the resulting chemokine gradient. This calculation is similar to the distance calculated for generic cytokines secreted by a suspended solitary cell [62].

Future work with spatially explicit modeling should explore the communication distance in more detail, as well as the role of immigrant CD8 T cell proliferation, contribution of resident memory T cells and B cells, and effector cell lifespan.

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Annotation: -; description of the general effect on lung "inflammation" of CCR5 de-

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rate in influenza infected mice but does not significantly decrease the clearance of the

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Key: Fadel2008

Annotation: These data could be interesting if we want to model both CCL5 and

- CXCR3 (receptor for CXCL10 or IP-10) because CXCR3 has regulating properties on CCR5 (receptor for CCL5 or RANTES) dependent and/or independent response.
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## Figure Legends

Figure 1. Model of T cell search. Activated T cells originate in the lymph node and enter the bloodstream after which they randomly navigate through 14 vascular bifurcations of the bronchial network. Upon reaching a capillary, T cells exit into tissue if cytokine signal is present. In the absence of signal, the T cell recirculates either through the lymph network or through the pulmonary vein back to the top of the network.

Figure 2. Visual representation of the model. Healthy epithelial cells infected by virus begin secreting virus after the incubation delay. Activated T cells traverse the bronchial vascular network and may be recruited by inflammatory cytokine. Chemotaxing T cells climb the chemokine gradient and induce apoptosis in infected cells. Solid arrows represent a cell state transition from one behavior to another. Dashed arrows display the mechanism used to induce a transition. Dotted arrows indicate the production of new virus.

Figure 3. Empirical viral and cytokine titers for three strains of influenza: Avian H5N1, Seasonal sH1N1, and Pandemic pH1N1. Viral titer (blue circles) is in PFU/mL, and IP-10 (red triangle) and RANTES (green square) are shown in ng/mL. sH1N1 IP-10 secretion exceeded measurement accuracy above 8500 pg/mL but these three values (empty red triangles) were not included in the model fitting. An extended differential equation model from [51] (see Materials and Methods) was fit to IP-10 and RANTES data. These fits were used to obtain chemokine production values for use in the spatial CyCells model. Human bronchial epithelial cells were infected at an MOI of 0.01 (10,000 virions) with one of the three strains of influenza. Apical fluid for viral secretion and basal media for chemokine secretion was collected at the given time intervals post infection. Viral culture was performed by a standard plaque assay and chemokine levels were measured using 30  $\mu$ l aliots for a panel of 17 chemokines and cytokines (not shown).

Figure 4. Model results: Time series plots of fifty runs of aH5N1 (A), sH1N1 (B), and pH1N1 (C) infections (gray). Each run took the calculated viral production and chemokine production rates for the three different strains of influenza as input (Table 3) and reported the total number of infected cells, including incubating, virus secreting and apoptotic, but not including dead cells. Therefore the figures approximate the rate of plaque growth over time. IP-10 and RANTES were simulated in each run, except for aH5N1, which produced only RANTES. Each run was initialized identically for each strain save for the random seed. The middle line shows the average while the outer lines show the 95% confidence interval.

Figure 5. Simulated sH1N1 infection. Screenshots from day 4, day 5.5, and day 7. The top row shows the spreading focus of infection through the color coding of individual cells: healthy cells in uninfected tissue (gray), virus-incubating cells (yellow), virus-secreting cells (orange), apoptotic cells (red), dead cells (brown), and T cells arriving at day 5 (green). Free virus and chemokine particles are represented by compartmentalized concentrations of mols/mL and ng/mL. Chemokine shown is an aggregate of total IP-10 and RANTES concentrations. See Videos S1-S3 for an animated visualization of each row.

#### **Tables**

Population	Description	Initial Value
T	Healthy target epithelial cells	1,000,000
$I_1$	Virus-incubating cells	0
$I_2$	Virus-secreting cells	0
V	Free virus particles	10,000
F	Interferon quantity	0
C	Chemokine quantity (ng/mL)	0
Parameter	Description	Value
β	Viral infection rates	
	Avian	$5.3e-7 (PFU \cdot h)^{-1}$
	Seasonal	6.1e-7 $(PFU \cdot h)^{-1}$
	Pandemic	$2.7e-6 (PFU \cdot h)^{-1}$
p	Viral production rates	
	Avian	$0.20 \; (PFU/h)$
	Seasonal	$1.4 \; (PFU/h)$
	Pandemic	$18.3 \; (PFU/h)$
e	Interferon strengths	
	Avian	1.0e-8
	Seasonal	1.6e-6
	Pandemic	3.4e-3
$ au_2$	Interferon production delays	
	Avian	21.5 (h)
	Seasonal	23.6 (h)
	Pandemic	21.0 (h)
δ	Virus-secreting cell decay rate	$0.6  (h^{-1})$
d	Chemokine decay rate	$1.386  (h^{-1})$
$ au_1$	Viral incubation time	10 (h)
$ au_3$	Chemokine production delays	
	IP-10	8 (h)
	RANTES	16 (h)
•		

**Table 1.** All parameters and populations are taken from [51] except for C, d, and  $\tau_3$ . The value chosen for d corresponds to a 30 minute half-life and values for  $\tau_3$  were set from the observed chemokine data (Fig. 3). Interferon (F) is an abstracted unitless quantity and thus e is a unitless multiplier.

Referenced Parameters	Value	Source
Viral Diffusion in Airway	$.0318 \ \mu { m m}^2/{ m s}$	[63]
Viral Decay in Airway	1/day	[46]
Chemokine Diffusion Rate	$.318 \ \mu { m m}^2/{ m s}$	[63]
Incubation Time	10 hours	[51]
Epithelial Cell Radius	$5~\mu\mathrm{m}$	[64]
T cell Radius	$5~\mu\mathrm{m}$	[65]
T cell Production Rate	1257/hour	[41]
T Cell Speed	$6 \ \mu \mathrm{m/min}$	[66]
Blood Circulation Time	6 seconds	[67]
T Cell Sensitivity to Chemokine	100  ng/mL	[68]
Onset of ATC <sup>‡</sup> Lymph Node Exit	Day 5	[50]
IgM Strength	Viral decay of 10/day	[69]
IgM Onset	Day 4	[69]
Estimated Parameters	Value	Footnote
Chemokine Decay Rate	$3.8508 \cdot 10^{-4} / s$	1
Infection Sensitivity Rate	2 hour/virion	2
Expression Time	16.7 hours	$[51]^3$
T Cell Expected Kill Time	10 min	4
Apoptosis Time	60 minutes	$[70]^5$
T Cell Age (at FOI*)	2 hours	6
T Cell Age (in Blood)	3 days	6

Table 2. Values are taken from experimental literature if possible and from earlier modeling papers if not. If we could not find a value in any literature we estimated parameters as follows. 1) Corresponds to a 30 minute half-life. 2) Epithelial cells are infected at a probabilistic rate such that the expected time for infection in the presence of a single virion is 2 hours. This scales linearly with the number of virions in the cell's vicinity. 3) Chosen as a plausible median time (1,000 minutes) between 6 hours and 24 hours. 4) T cells induce apoptosis in nearby virus-secreting epithelial cells at a probabilistic rate such that the expected time to induce apoptosis is 10 minutes. This rate does not scale with T cell numbers. 5) Calculated for low T cell densities. 6) Chosen to be at the lower end of biologically plausible values because increased T cell counts are shown not to affect the model behavior.

Strain	IP-10 Production	RANTES Production	Viral Production
	$(pg/s \cdot cell)$	$(pg/s \cdot cell)$	$(PFU/s \cdot cell)$
Avian H5N1	2.0e-4	1.3e-5	5.4e-5
	8.4e-5 - 4.2e-4	7.9e-6 - 1.9e-5	4.4e-5 - 3.7e-4
Seasonal H1N1	1.8e-4	8.9e-7	3.8e-4
	1.2e-4 - 3.0e-4	4.8e-7 - 1.6e-6	2.8e-4 - 1.5e-3
Pandemic H1N1	8.7e-5	4.3e-6	5.1e-3
	1.7e-5 - 7.1e-4	5.0 e-7 - 3.5 e-5	2.8e-3 - 5.3e-3

**Table 3.** Strain-specific model parameters. Small text values show 95% confidence intervals resulting from 1,000 bootstrapping runs for each parameter [71]. Bootstrapping for the chemokine values was performed using the original fit of Eq. S1 to the data in Fig. 3 to produce new data sets. Viral production values and confidence intervals are taken from [51].

Figure 6. Simulated infections of aH5N1, sH1N1, and pH1N1. Plotted values: total plaque size (blue), number of virus incubating cells (yellow), number of virus secreting cells (green), total number of T cells (black), and T cells at the focus of infection (FOI) (red). T cells clear secreting and incubating cells in aH5N1, fail to clear incubating cells in sH1N1, and fail to clear either type of infected cell in pH1N1. The number of incubating cells (yellow) after day 5 differs markedly among the three strains indicating that the T cells have differing success at controlling the infection.

Figure S1. Varying T cell sensitivity to chemokine. H5N1 model results use RANTES only, and sH1N1 and pH1N1 use both IP-10 and RANTES. Total number of incubating, secreting and apoptotic cells are plotted for each infection. The sensitivity value specifies the minimum level of chemokine concentration required for T cells to detect it.

**Figure S2. Effects of different chemokine combinations.** A) aH5N1 does not stimulate an IP-10 response. B-C) sH1N1 and pH1N1 show no significant difference between IP-10 alone versus IP-10 and RANTES combined.

**Video S1.** The first of three overlaid videos of a representative seasonal H1N1 infection. This video spans the 10 day infection and shows the cells as they transition from healthy to infected to dead. T cells show half way through the simulation. Healthy cells are gray, virus-incubating cells are yellow, virus-secreting cells are orange, apoptotic cells are red, and T cells are green.

Video S2. The second of three overlaid videos of a representative seasonal H1N1 infection. This video spans the 10 day infection and shows the virus concentration. Notice the volatility when T cells arrive halfway through the simulation. Virus concentration ranges from 1e-13 mols/mL (white) to 1e-27 mols/mL (black). Refer to Figure 5 for the detailed legend.

Video S3. The third of three overlaid videos of a representative seasonal H1N1 infection. This video spans the 10 day infection and shows the chemokine concentration. Notice the volatility when T cells arrive halfway through the simulation. Chemokine concentration ranges from 1e8 ng/mL (white) to 1e-6 ng/mL (black). Refer to Figure 5 for the detailed legend.

Video S4. A closer look at the 2009 pandemic simulation. This video shows the infection from day 6 to day 7 with each frame spanning 60 simulated seconds. Healthy cells are gray, virus-incubating cells are yellow, virus-secreting cells are orange, apoptotic cells are red, and T cells are green. Note the high proportion of virus-secreting cells (orange) early on. As time passes, secreting cells are gradually contained to the point where they become very sparse. T cell clumping often prevents the T cells from quick discovery of new secreting cells.